The Social Integration of International Migrants: Evidence from the Networks of Syrians in Germany *

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Abstract

We use privacy-protected friendship data from Facebook to study the social integration of Syrian migrants in Germany. Our analysis establishes five key findings: (1) Places differ substantially in their propensities to socially integrate migrants. This regional variation in integration outcomes largely reflects causal place-based effects. (2) Spatial variation in migrants' social integration can be decomposed into the rate at which Germans befriend their neighbors in general and the particular rate at which they befriend migrants versus other Germans. We follow the friending behavior of Germans that move across locations to show that both forces are more affected by local institutions and policies than by persistent individual characteristics or preferences of local natives. (3) Integration courses causally affect place-specific equilibrium integration levels by increasing the rate at which Germans befriend Syrian migrants. (4) Social integration helps migrants obtain help from natives across a range of settings such as finding jobs and housing. (5) Natives quasi-randomly exposed to a migrant in high school are more likely to befriend other migrants later in life.

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In 2019, there were 272 million international migrants, comprising 3.5% of the world's population (United Nations, 2019). The challenge of fostering communities that successfully integrate new arrivals with natives has therefore become of increasing importance to policymakers around the globe (e.g., European Commission, 2020; Bundesregierung, 2021). Yet, because of difficulties with measuring social networks using traditional data sources, researchers have long struggled to understand the determinants of the social integration of migrants in their host communities.

In this paper, we use privacy-protected data from Facebook, a global online social networking service, to study the social integration of newly arriving migrants. We focus on individuals who recently migrated from Syria to Germany. Following the start of the Syrian Civil War, millions of Syrians fled their home country, with about 800,000 of them settling in Germany since 2014. The social and economic integration of these migrants has been a dominant political issue in Germany in the years since, with policymakers attempting to facilitate this integration through a variety of programs. In 2018 alone, for example, the German government spent €2bn on integration courses that teach migrants the German language and provide information on the country's culture and legal system.

While prior work studies the *economic* integration of Syrians in Germany—with a special focus on assessing attempts to bring them into the labor force—data challenges have hindered empirical studies of the *social* integration of these migrants. How common are social interactions between Syrian migrants and Germans?[1] How does this frequency differ across demographics and locations? Which Germans form social ties with Syrians? Can local policies affect this? And does social integration affect migrants' economic outcomes? Our unique data and research design allow us to answer these questions.

We begin by identifying Syrian migrants as Facebook users who currently live in Germany, but who specified a hometown or high school in Syria in their Facebook profiles, or who previously had a predicted home region in Syria.[2] This methodology generates spatial variation in Syrian migrant population shares across German counties (*Kreise*) that closely resembles German administrative data. We also construct a group of users we call "German natives" based on self-reported profile information, home region predictions, and German language usage.[3] We use these data to measure Syrian migrants' social integration along three key dimensions: (i) friendships between migrants and German natives; (ii) migrants' German language usage; and (iii) migrants' participation in local social groups.

Syrian migrant users have five local German native friends on average, and 30% of them produce German-language content such as posts or comments on Facebook. Controlling for Facebook usage patterns, younger and male migrants have higher levels of social integration than others. Our measures of social interactions in the Facebook data strongly correlate with individual responses to a recent Facebook survey asking Syrian migrants about their interactions with German natives, suggesting that they do, in fact,0 represent real-world integration outcomes.

[1] While there is no single definition of social integration, the concept is often defined by the frequency of interactions of individuals of different groups (e.g., Phillips et al., 2019). This conceptualization of "social integration" is distinct from that of *assimilation* (Berry, 1997), which is defined in terms of cultural identity, and is not the focus of our work.

[2] Estimated home region is determined by a person's information on Facebook, including the stated city on their Facebook profile, and device and connection information (see also Herdağdelen et al., 2016; Chi et al., 2019).

[3] We describe these criteria in detail in Appendix B. When constructing both the "Syrian migrant" and "German native" samples, we define the sample of users on the basis of past location and language usage. Broadly speaking, our sample of Syrians comprises users who appear to have lived in Syria and who now live in Germany, while our sample of Germans comprises users who appear to have lived in Germany for an extended period and exclusively or primarily use the German language.

We structure our results around five key lessons from this migration episode. We next briefly describe each lesson and our empirical evidence. In the paper, we provide further details.

Lesson I: *Places differ in their propensities to integrate migrants. The substantial spatial variation in Syrian migrant integration outcomes in Germany largely reflects causal placed-based effects*.

We document sizable spatial heterogeneity in Syrian migrants' social integration across the 401 German counties: an average Syrian migrant living in a 90th percentile county has more than twice as many native German friends as an average Syrian migrant living in a 10th percentile county. These spatial patterns are highly correlated across our three measures of social integration. We show that these measures pick up true differences in integration levels rather than sampling variation or differences in Facebook usage across space; for example, we show that average observed integration outcomes align with external survey measures of integration available at higher levels of geographic aggregation (complementing our own individual-level survey evidence described above).

The German government assigns migrants to locations to ensure dispersion throughout the country, suggesting the observed spatial differences might at least be partly driven by causal place-based effects rather than primarily being the result of migrants with higher integration propensities selecting to live in certain regions. We further test this hypothesis with a mover research design that follows the (relatively few) Syrian migrants who move across German counties. We find that these movers' social integration patterns quickly adjust from those of their origin counties toward those of their destination counties. This variation allows us to estimate that most of the observed regional differences in migrants' social integration are indeed due to causal place-based factors rather than migrant characteristics, consistent with prior work exploiting the random assignment of refugees in other countries (e.g., Auer, Egger and Kunz, 2022; Edin, Fredriksson and Åslund, 2003; Beaman, 2012; Damm, 2014).

Lesson II: *Spatial variation in migrant social integration can be decomposed into the rate at which natives befriend their neighbors in general and the particular rate at which they befriend migrants. Both forces vary across space, and both are largely influenced by local equilibria rather than spatial differences in immutable native preferences.*

Data challenges and the lack of random assignment of natives to locations have precluded prior attempts to estimate the importance of immutable native preferences in explaining variation in migrants' integration outcomes. Our unique panel data on the characteristics and behaviors of Germans who befriend Syrians allows us to make progress on this important question.

We begin by showing that the level of Syrians' social integration in a location can be decomposed into two forces: (i) the rate at which local Germans befriend their neighbors in general (their *general friendliness*), and (ii) Germans' particular friending behavior towards migrants, given by their relative propensity of befriending local Syrians versus other locals (*relative friending*). Put simply, if Germans in a given location are more likely to befriend *all* of their neighbors, including their German ones, they are also more likely to befriend newly arriving migrants. All else equal, this aids migrants' social integration, even if the level of general friendliness is unlikely to be strongly affected by migrants' behavior or integration policies. In addition, Syrian migrants will be more socially integrated when Germans befriend them at rates more similar to those at which they befriend local Germans. We show that both general friendliness and relative friending vary across locations, with differences in relative friending

explaining about two-thirds of the spatial variation in the social integration of Syrian migrants.

We next explore whether spatial differences in relative friending and general friendliness are driven by immutable preferences of the populations of local natives (e.g., if Germans in some regions happen to have a persistently friendlier disposition towards foreigners) or by place-specific factors that would shift the same Germans' friending behavior if they were to move. Our analysis shows that place-specific factors such as local policies, institutions, or social equilibria play a dominant role in explaining Germans' social behaviors towards migrants, and thus the spatial variation in migrant integration outcomes.

To document this, we follow the friending behavior of Germans who move across locations. Native movers adjust their general friendliness about two-thirds of the way to that of comparable destination natives within a year of moving; their relative friending adjusts almost fully to that of destination natives. These findings highlight that Syrian migrants' lack of integration in some locations is not primarily the result of immutable preferences or beliefs of the native locals. Instead, our results show that the probability of the same two individuals—the same German and the same Syrian—becoming friends varies substantially with the institutional frameworks or the prevailing social equilibria across locations.

Lesson III: *Integration courses can causally affect place-specific equilibrium integration levels. The availability of these courses for Syrian migrants shifted the relative rates of German-Syrian friendships.*

To understand the factors that shape regional variation in social integration, we next explore the relationships of both general friendliness and relative friending with county-level characteristics. These correlations can be informative about the mechanisms that drive migrants' integration outcomes even as they capture equilibrium relationships that complicate assigning a direction of causality. As we show, the correlational analysis can also help identify factors that merit further causal study.

We highlight three findings. First, similar to ethnographic work on integration in smaller European towns and cities (Gauci, 2020), we find that migrants' social integration decreases with population density. This is driven both by Germans in cities being less likely to befriend any of their neighbors—consistent with work exploring the "loneliness of cities" (Hammoud et al., 2021)—and by Germans in cities being particularly unlikely to befriend migrant neighbors, consistent with work showing that social segregation increases in group size (Chetty et al., 2022*b*). Second, the 'relative friending' component of integration decreases with a county's Syrian population share in 2019, but increases with the share that was Syrian in 2010. Earlier migrants may boost relative friending by supporting new arrivals and positively shaping local natives' views, whereas a large influx of migrants simultaneously may lead to the formation of migrant cliques and fewer migrant-native connections.[4] Third, in counties with more completed integration courses per Syrian migrant, relative friending is higher, consistent with these courses potentially shifting equilibrium friending behaviors in a location.

Language and integration courses are among the few direct tools available to policymakers for fostering migrant social integration and have been a key component of German government policy. Motivated by our correlational result, we use an instrumental variables approach to study whether the provision of these courses had a *causal* effect on integration outcomes, contributing to a literature that

[4]This finding speaks to the "ethnic enclaves" literature that finds migrant networks support integration in some settings and hinder it in others (e.g. Lazear, 1999; Edin, Fredriksson and Åslund, 2003; Cutler, Glaeser and Vigdor, 2008; Beaman, 2012; Sale, 2021; Martén, Hainmueller and Hangartner, 2019).

has studied various government policies intended to integrate minority groups.[5] Our instrument, the local availability of teachers qualified to teach these courses who were unemployed at the start of the Syrian migration wave, is correlated with the completion of integration courses, even after controlling for the overall unemployment rate and other county characteristics. This aligns with anecdotal evidence that the unavailability of qualified teachers substantially limited the government's ability to offer integration courses. We estimate that a 10% increase in 2015-19 integration course completion per Syrian (driven by higher course availability) raised friending integration by 18%. This effect comes entirely from raising Germans' rate of befriending Syrians in particular (i.e., by raising relative friending); as expected, Germans' general friendliness is unaffected by migrants' completion of integration courses.

Lesson IV: *Social integration appears to be an important channel of positive effects on migrants' labor market, housing, and education outcomes.*

While social integration is itself an important outcome for policymakers, social connections may also impact other aspects of migrants' well-being. For instance, a German native friend might help a migrant find employment or housing, assist with schooling, or provide guidance in accessing public services. Correlationally, the share of Syrian migrants employed increases with a county's friending integration, consistent with such a positive impact, but also with reverse causality.

To better understand the observed relationships between social connections and economic outcomes, we analyze responses to a short user survey fielded by Facebook that asked migrants about how native friends had impacted their experiences in Germany. We find that the number of local native friends is highly correlated with migrants' likelihood of reporting that such friends helped them find a job, secure housing, complete schoolwork, and navigate the bureaucracy. For example, a one standard deviation increase in local native friends corresponds to a 12.6% increase in the probability of reporting to have received job-finding assistance from a native German. While these results are correlational, the questions focus directly on *causal mechanisms* through which native friends help migrants, strongly supporting the notion that social integration positively affects other outcomes. These findings add to a literature on refugees' economic integration in high-income countries (see Becker and Ferrara, 2019; Brell, Dustmann and Preston, 2020, for overviews), highlighting social ties as an important determinant.

Lesson V: *Natives exposed to a migrant in high school are more likely to befriend other migrants later in life. Connections directly facilitated by the first migrant do not fully explain this effect.*

In the final section of the paper, we return to the determinants of natives' persistent friending behaviors and study the longer-term effects of exposure to Syrian migrants on subsequent friending patterns. Specifically, we use fluctuations in the presence of Syrian migrants across high school cohorts as a quasi-random source of variation of exposure to such migrants. We find that exposure to Syrian migrants in high school leads to higher probabilities of German natives befriending Syrians even outside the high school setting, consistent with the contact hypothesis, which outlines the circumstances in which social contact between members of different groups can help to reduce prejudice and animosity (Allport, Clark

[5] See e.g., Abdelgadir and Fouka (2020); Abramitzky, Boustan and Eriksson (2020); Arendt et al. (2024, 2023); Bandiera et al. (2019); Battisti, Giesing and Laurentsyeva (2019); Fouka (2020); Heller and Slungaard Mumma (2023); Lleras-Muney and Shertzer (2015); Kanas and Kosyakova (2022); Emeriau et al. (Forthcoming).

and Pettigrew, 1954; Bursztyn et al., 2024; Boisjoly et al., 2006; Carrell, Hoekstra and West, 2019; Paluck, Green and Green, 2019; Rao, 2019; Corno, La Ferrara and Burns, 2022).

Contribution to Literature. Each of our five lessons offers new insights into the determinants and effects of migrants' social integration, a topic that has long been studied in social science research (e.g., Srole, 1956; Coleman, 1988; Putnam, 1995*a*; Alesina, Baqir and Easterly, 1999). Within this literature, our work relates most closely to studies that use surveys or assimilation-related measures to proxy for migrants' social integration. Laurentsyeva and Venturini (2017) provide a recent overview of this literature (see also Niehues, Rother and Siegert, 2021; Schmidt, Jacobsen and Krieger, 2020; Cheung and Phillimore, 2014). In contrast to these studies, we are able to directly measure key elements of migrants' social integration in large-scale data, allowing us to explore granular spatial variation in integration outcomes. Our unique panel data on the friending behaviors of both Germans and Syrians allows us to obtain a more holistic view on social integration, which, by its nature, depends on the behaviors of both migrants and natives. In particular, our ability to study the friending behavior of natives (and not just migrants) enables us to generate novel insights on the determinants of this integration.

We also add to a literature that uses experimental and quasi-experimental methods to study the causal effects of local environments on a variety of economic, social, and health outcomes (see Chyn and Katz, 2021, for a review). We believe we are the first to use a mover-based research design to study the effects of place on migrants' social integration, adding to existing evidence that is observational or relies on quasi-random refugee settlements (e.g. Åslund and Rooth, 2007; Damm, 2014; Braun and Dwenger, 2017; Aksoy, Poutvaara and Schikora, 2023; Jaschke, Sardoschau and Tabellini, 2021; Sale, 2021). We also introduce the use of movers to study the effect of places on *native* rates of befriending migrants, highlighting that place-based effects are not primarily picking up fixed preferences of local natives.

The remainder of this paper is structured as follows. In Section 1 we describe our data, sample, and outcomes of interest. We also document overall patterns of social integration and the relationship of individual-level migrant and native characteristics with friending outcomes. In Section 2 we generate regional measures of social integration and use movers to study the extent to which they reflect place-based effects. Sections 3 and 4 focus on the roles of natives and local institutions, exploring the forces that make migrants more likely to integrate in one place versus another. Section 5 studies the effects of social integration on other real outcomes. Section 6 looks at how quasi-random exposure to migrants shapes natives' long-term behavior. We conclude in Section 7.

1 Data and Descriptive Statistics

We work with privacy-protected data from the online social networking site Facebook. In March 2021, Facebook had over 2.8 billion monthly active users, including 423 million in Europe (Facebook, 2021). Facebook is used widely by Syrian migrants in Germany to share information and communicate with friends and family in Syria and elsewhere (Scheibe, Zimmer and Stock, 2019). Many individuals opened their Facebook accounts prior to arriving in Germany, while others likely created accounts during their migration, as Facebook was frequently cited as a tool used by refugees fleeing to Europe to share information (Dekker et al., 2018; Mall et al., 2015; Ritscher, 2016; Mustafa and Lamb, 2017).

Establishing a "friendship" connection on Facebook requires the consent of both parties, and a per-

son can have at most 5,000 connections. As a result, Facebook connections are usually between individuals who interact in person (Jones et al., 2013). Facebook networks thus resemble real-world social networks more closely than networks on other online platforms where uni-directional links to non-acquaintances (e.g., celebrities) are common. As a result, prior studies have used Facebook data to explore the relationship between social connections and many economic and social outcomes such as trade flows, patent citations, travel flows, disease transmission, bank lending, social capital, social program participation, investment decisions, product adoption decisions, housing choices, migration decisions, and beliefs and behaviors related to public health (Bailey et al., 2018*a*,*b*, 2019, 2020*a*,*b*, 2021, 2022, 2024; Chetty et al., 2022*a*,*b*; Kuchler, Russel and Stroebel, 2021; Kuchler et al., 2022; Koenen and Johnston, 2024; Rehbein and Rother, Forthcoming; Wilson, 2019).

1.1 Sample Construction and Measures of Social Integration

We construct our primary sample from a sub-population of Facebook users who had active accounts in October 2021, were 18 or older, lived in Germany, and had 25 or more friends. Each user is predicted to live in one of 401 German districts (*Kreis*, *Landkreis*, or *Stadtkreis*), with an average population of just over 200,000.[6] We refer to these geographies as "counties."

Syrian Migrant & German Native Samples. For many of our analyses, we use two sub-samples.

1. *Syrian Migrant Sample:* We construct a set of users who specify a Syrian hometown or high school in their Facebook profile, or who previously had a predicted home region in Syria. There are about 350,000 such users, which we refer to as "Syrian migrants" (see footnotes 2 and 3 for details). In Appendix Figures A1 - A3, we compare the demographics and locations of our sample against the full corresponding population using administrative data from the Federal Statistical Office of Germany. Syrian migrant population shares across counties and age groups in our data closely correspond to those in the administrative data, highlighting that we observe Syrian and non-Syrian users at similar rates across demographics (though we somewhat over-sample male Syrians relative to their true population shares). For example, we find population-weighted correlations between county $\times$ age $\times$ gender shares in the Facebook sample versus actual population of 0.97.

2. *German Native Sample:* We also construct a group of users, which we refer to as "German natives", who meet the criteria described in Appendix B based on self-reported profile information, home region predictions, and German language usage. We identify 18 million such users. The median county has 34,063 German native users; the 10th-90th percentile range is 17,057 to 74,651 German native users. Appendix Figure A4 benchmarks this sample against administrative data. The share of users in the primary Facebook sample that are natives is somewhat lower than the true population share, a result of our relatively strict assignment criteria. The German population shares in our data are also consistent with administrative data sources across county and gender, with population-weighted correlations of 0.94.

[6] These locations are assigned based on user information and activity on Facebook, including their self-reported profile information, and device and connection information.

Measures of Migrants' Social Integration. We capture the social integration of Syrian migrants using three primary measures (see Appendix C for detailed definitions):

1. The number of native German friends a Syrian migrant user has in the same or a bordering county;
2. An indicator for whether the Syrian migrant user produces content such as Facebook posts and comments in German; and
3. How many local native Facebook groups (e.g., for local sport clubs or cultural societies) a Syrian migrant user joins.

1.2 Sample Summary Statistics

Panel (a) of Table 1 summarizes the Syrian migrant sample. The median Syrian migrant user is 31 years old, with a 10th-90th percentile range of 22 to 48 years. The sample is 32% female, somewhat lower than 40% in the administrative data. The median number of Facebook friends and groups joined is 226 and 56, respectively. The median user in the Syrian migrant sample first used Facebook in Germany 23 quarters ago. About 8% of Syrian migrants list a German college on their profile.

Syrian migrant users have five native local friends on average.[7] This magnitude is broadly consistent with data from the German Socio-Economic Panel (SOEP), a longitudinal survey of German households. In the 2016 wave of the *IAB-BAMF-SOEP Survey of Refugees in Germany* the average recent Syrian migrant in Germany reported to have "regular contact" with 6.2 German acquaintances.[8] By contrast, Syrian migrant users have 15 Facebook friendships with other Syrian migrants in the same location. About 30% of Syrian migrant users produce content on Facebook in German. At the median and 90th percentiles, Syrian migrant users are members of zero and two local native groups, respectively.

Appendix Figure A5 presents binned scatter plots showing relationships between our three primary integration outcomes—local native friends, German content production, and local native groups—at the individual level. There are strong positive relationships, both with and without controls for individual-level demographics and Facebook usage, providing evidence that our measures are capturing related and strongly correlated aspects of social integration (also see Appendix Tables A2 and A3).

Panel (b) of Table 1 summarizes the German native sample. The median user is 38 years old, with a 10th-90th percentile range of 24 to 60 years. The sample is 52% female and 33% of users list a German college on their profile. The median German native has a total of 181 Facebook friends, 79 local native friends, and 0.1 local Syrian migrant friends (users at the 99th percentile have two local Syrian migrant friends), highlighting that most German native users are not Facebook friends with a single Syrian migrant. German natives are members of four local native groups on average.

[7] Friendship requests between natives and Syrians are initiated at essentially equal rates by each group. On average, Syrians sent the friend request in 50.01% of their friendships with native local Germans.

[8] The exact question asked by the SOEP is: "How many German people have you met since your arrival in Germany with whom you have regular contact?" The average responses reported in the text is based on responses from 1,095 survey respondents. If the roughly 1/3 of adult German natives we capture on Facebook were randomly selected, we might expect migrants in the SOEP to have on average $5.03 \times 3 \approx 15$ native friends. That the survey measure is somewhat lower may reflect differences in the survey timing (2016 vs 2021); respondents narrowly interpreting "regular contact" or failing to recall connections; and/or a higher propensity of migrants with Facebook accounts to friend natives with Facebook accounts.

Table 1: Syrian Migrant and German Native Sample Summary Characteristics

Panel (a): Syrian Migrant Sample

	Mean	SD	P10	P25	P50	P75	P90	P99	Survey Mean
Age	32.90	10.26	22	25	31	38	48	66	39.97
Female (0/100)	32.07	46.68	0	0	0	100	100	100	25.79
DE College (0/100)	7.92	27.00	0	0	0	0	0	100	6.29
N Friends	347.89	385.84	62	117	226	423	751	2431	527.27
N Groups	104.55	137.09	8	22	56	129	256	831	192.39
Qs Since 1st on FB in DE	20.30	8.04	7	15	23	25	28	36	30.51
N Local Native Friends	5.03	12.24	0	0	1	4	13	87	5.43
N Local Syrian Friends	14.99	17.43	1	4	9	20	36	103	20.56
Produces DE Content (0/100)	30.40	46.00	0	0	0	100	100	100	29.42
N Local Native Groups	0.55	1.41	0	0	0	0	2	9	1.58

Panel (b): German Native Sample

	Mean	SD	P10	P25	P50	P75	P90	P99
Age	40.23	13.79	24	29	38	51	60	77
Female (0/100)	51.74	49.97	0	0	100	100	100	100
DE College (0/100)	32.93	47.00	0	0	0	100	100	100
N Friends	253.72	243.28	51	93	181	327	535	1535
N Groups	25.22	34.52	2	6	14	30	59	231
Qs Since 1st on FB in DE	31.87	8.26	18	33	36	36	36	36
N Local Native Friends	122.52	128.88	12	32	79	168	295	687
N Local Syrian Friends	0.09	0.34	0	0	0	0	0	2
Produces DE Content (0/100)	100.00	0.00	100	100	100	100	100	100
N Local Native Groups	3.98	4.92	0	1	2	5	10	26

Note: Table presents summary statistics describing users in our samples. Panel (a) shows users in the Syrian migrant sample. Panel (b) shows users in the German native sample. Each measure is winsorized at the 99% level. Section 1.1 describes the sample construction. Appendix C provides more information on how individual-level outcomes are defined. Appendix Table A1 provides additional summary statistics. The final column of panel (a) shows summaries of survey respondents, as described in Section 1.3. The demographics in this column are as of May 2024, rather than October 2021, as described in footnote 10.

1.3 Survey Validation of Observational Integration Measures

To ensure our social integration measures capture meaningful in-person interactions, we analyze responses to a short user survey conducted by Facebook in May 2024. The survey targeted users in our Syrian migrant sample through a post on their News Feed. All questions were translated to the user's preferred language on Facebook. Appendix N provides screenshots of the survey in English, German, and Arabic. In total, 3,413 individuals responded to the survey.[9] The final column of Panel (a) of Table 1 shows the mean demographics of survey respondents. They generally align with the overall Facebook sample, with similar average friendships with local natives.[10] In the survey, respondents were asked

[9] Of these, 3,332 finished the survey. Most, but not all, finishers answered every question. We use the broadest sample of respondents available for each question, but have verified our facts do not change using narrower samples of users that answer every question.

[10] Note also that the survey sample demographics are as of May 2024 whereas the primary sample demographics are as of October 2021, which contributes to some of the observed differences between samples, for example in terms of average age.

about their social interactions with native Germans broadly, as well as specific questions about visiting, hosting, dining, and playing sports with natives.[11]

Table 2: Survey Responses vs Measured Friending Integration at Individual Level

	Been Invited to Native Home		Invited Native to Own Home		Visited Restaurant with Native		Played Sports with Native		Freq. of Native Social Interactions	
N Local Native Friends	0.823***	0.779***	0.658***	0.628***	0.884***	0.797***	0.609***	0.396***	0.007***	0.006***
	(0.086)	(0.105)	(0.088)	(0.105)	(0.087)	(0.107)	(0.083)	(0.116)	(0.002)	(0.002)
Control Covariates		X		X		X		X		X
N	2,987	2,940	2,987	2,940	2,987	2,940	2,987	2,940	3,328	3,286
Sample Mean	42.85	42.69	54.6	54.69	44.69	44.8	33.38	33.37	4.053	4.051

Note: Table shows results of individual-level regressions of survey responses on the number of local native friends. The outcomes in columns 1-8 are responses to "Which of the following interactions with Germans have you had in the past year?" The sub-questions were: "I have been invited to a German friend's home (for a dinner, a birthday party, etc.)" (columns 1-2); "I have invited a German friend to my home (for a dinner, a birthday party, etc.)" (columns 3-4); "I have gone to a restaurant, cafe, or bar with German friends" (columns 5-6); "I have played sports with German friends" (columns 7-8). Columns 9-10 show agreement with the statement "I have many social interactions with Germans in the city I live in" on a scale from 1 (strongly disagree) to 5 (strongly agree). Columns 2, 4, 6, 8, and 10 include (i) controls for age and gender; (ii) fixed effects for the number of quarters on Facebook in their current county and the number of quarters since arrival in Germany (we use a single dummy value for those for which we do not observe arrival); (iii) three linear controls for measures of Facebook usage: log(0.5 + minutes on FB in the last 28 days), log(91 - days on Facebook out of the last 90), log(1081 - days on Facebook out of the last 1080); (iv) county fixed effects; and (v) controls for each user's total number of friends outside Germany, total number of non-local/native groups joined, and total amount of content produced in the last year. Standard errors are clustered by county. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01).

Columns 1 to 8 of Table 2 show that our Facebook measure of friending integration is a significant predictor of each specific real-world interaction with German natives. For example, columns 1 and 2 show that a one standard deviation increase in a Syrian migrant's number of local native friends corresponds to roughly a 22% increase in the probability of having been invited to a native's home in the past year. Columns 9 and 10 show that our measure also predicts self-reported levels of agreement with the statement "I have many social interactions with Germans in the city I live in." These strong relationships provide evidence that our measure picks up real-world integration patterns.

1.4 Migrant and Native Characteristics and Integration

In the following, we first analyze the relationship between Syrian migrant characteristics and integration outcomes. We then study the relationship between native characteristics and migrant friendships.

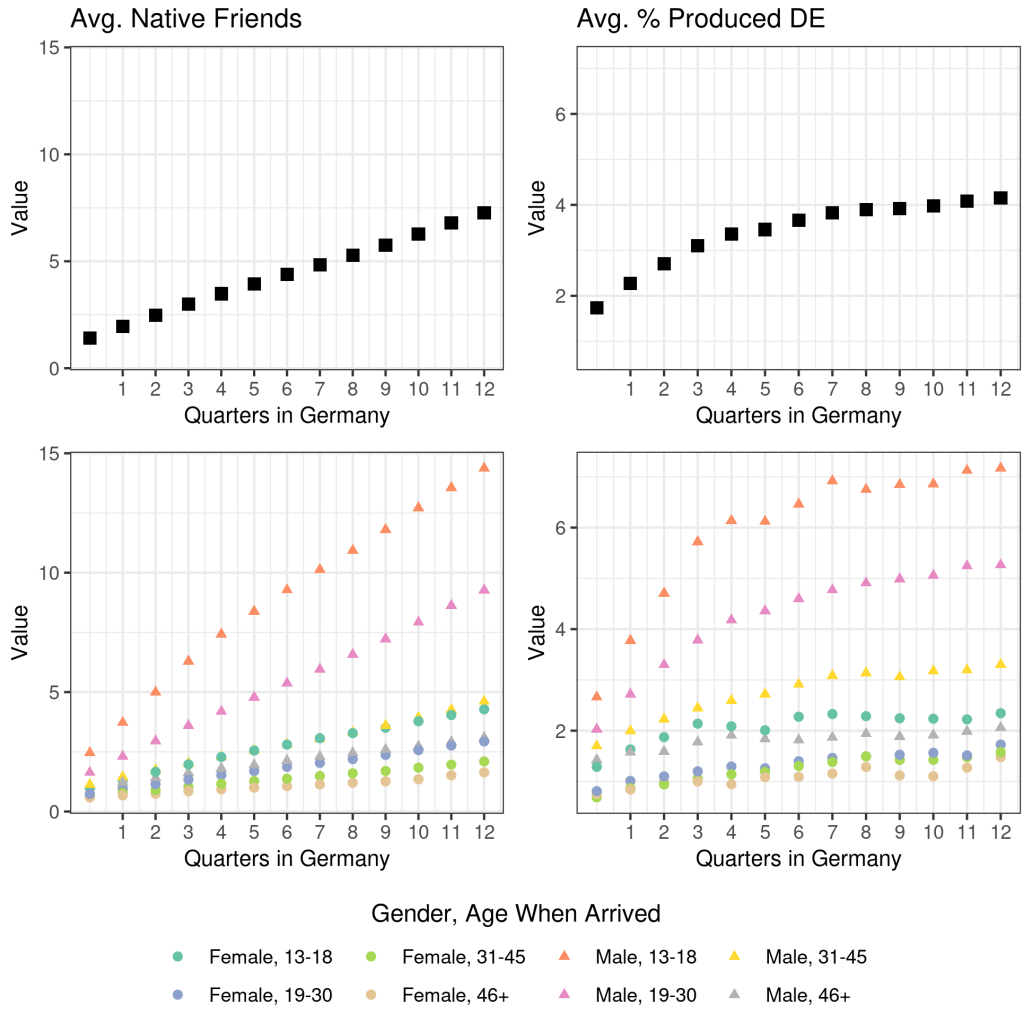
Migrant characteristics and integration outcomes. Figure 1 shows the relationship between individual-level characteristics and integration outcomes for a cohort of Syrian migrants with an "observed arrival" in 2015-2016.[12] Migrants become increasingly socially integrated as they spend more time in Germany. For example, after their first quarter in Germany, Syrian migrant users on average had 1.4 native friends and produced 1.7% of their Facebook content in German; three years later, these numbers were 7.3

[11] To align survey responses with our Facebook measure of natives, the survey instructions stated: "In the following we are going to ask you several questions about your interactions with the German population. By this, we mean individuals who have lived in Germany most of their lives."

[12] These are Syrian migrant users who first used Facebook *outside* Germany, then began using Facebook inside Germany in 2015 or 2016. Appendix Figure A6 reproduces this plot with additional integration measures.

friends and 4.2% of content, respectively. The bottom row of Figure 1 shows considerable heterogeneity in the degree of integration across age and gender groups, with younger and male migrants integrating more quickly than older and female migrants. In Appendix D, we further explore these heterogeneities in integration outcomes across individuals, using a multivariate regression model which allows us to include various controls, including controls for Facebook usage patterns, as well as state and even family fixed effects. The demographic patterns shown in Figure 1 remain: female and older migrants have fewer local native friends than male and younger migrants.[13] We also show that the demographic differences in integration outcomes across individuals align with those in the SOEP survey.

Figure 1: Integration Over Time For 2015-16 Arrival Cohort



Note: Figures show the average values, by quarter since they arrived in Germany, of integration measures for users in the Syrian migrant sample with an observed arrival in 2015 or 2016. The measures are total native friends (left column) and the share of content produced in German (right column). Appendix C provides more details on each measure. The top row shows overall trends. In the bottom row each observation's shape and color represents a gender-by-age group.

[13] Appendix Table A4 also presents multivariate regression results for our key language- and group-based measures of social integration, and Appendix Table A5 uses a different variation of our friend-based integration measures. Across all measures, we find highly consistent relationships between demographic characteristics and the social integration of Syrian migrants.

Native characteristics and friendships with migrants. Our data allows us to not only observe the social integration of migrants, but also the characteristics of the natives that interact with and befriend migrants. We analyze these in detail in Appendix J and summarize our findings here.

Overall, younger and male German natives have more Syrian migrant friends than older and female natives. Because Syrian migrants in Germany are more likely to be young and male than the average German native, one possible explanation for this finding is that homophily plays a strong role in shaping which natives befriend Syrian migrants. Put differently, younger German natives might be more likely to connect with younger Syrian migrants because younger people are more likely to connect in general, rather than because of a particularly friendly behavior toward migrants among younger versus older Germans. Consistent with such an interpretation, we show that it is, in fact, older and female natives that are more likely than others to join pro-immigration groups on Facebook, conditional on the relevant patterns of Facebook usage. In other words, it is not necessarily those who are most vocally supporting immigration (measured by supporting pro-immigration groups) that are most likely to befriend Syrian migrants and thereby directly foster their integration.

In Table 3, we explore the extent to which friendship links to Syrians disproportionately come from a small number of Germans that one might call "super integrators." Overall, 71% of all friendships between Germans and Syrians are to Germans with three or fewer Syrian friends and only 0.04% of Germans have more than 10 local Syrian friends. While there are some Germans with more than 50 local Syrian friends—which could include Germans working directly with refugees—they account for only 1.6% of all friendships that Syrians have with Germans.

We conclude that friendships between Syrians and Germans are not overwhelmingly driven by Germans with a large number of Syrian friends. Instead, most Syrian-German friendships are with Germans with few other Syrian friends. The role of possible "super integrators" seems limited.

Table 3: Concentration of Friendships Between Syrian Migrants and German Natives

			Average Age		**Share Male**		**Total Friends**	
Number of Migrant Friends	Share of Natives	Share Friendships to Migrants	Native	Migrant Friends	Native	Migrant Friends	Native	Migrant Friends
0	93.96%	0%	43.1	-	0.474	-	262	-
1	4.47%	44.6%	36.4	32.8	0.512	0.865	493	886
2-3	1.19%	26.8%	35.4	31.8	0.524	0.879	644	915
4-5	0.21%	9.0%	35.7	31.6	0.528	0.882	777	927
6-10	0.12%	8.7%	36.8	32.0	0.531	0.872	861	929
11-20	0.03%	5.5%	38.3	32.6	0.548	0.859	965	937
21-50	0.01%	3.7%	39.7	33.3	0.555	0.849	1119	956
51-100	0.002%	1.0%	42.9	33.6	0.601	0.845	1516	994
100+	0.0004%	0.6%	41.2	34.4	0.58	0.854	1981	1087

Note: Table shows summary statistics on Germans natives with various numbers of connections to local Syrian migrants. For example, the second row shows that about 4.5% of Germans have a single Syrian friend. These friendships make up 44.6% of all friendships between migrants and Germans. On average, Germans with 1 Syrian friend are 36.4 years old, and have 493 total Facebook friends. Their Syrian friends are, on average, 32.8 years old, and have 886 Facebook friends.

2 Determinants of Migrant Integration: The Effect of Place

In this section, we explore the determinants of social integration by asking "do places differ in their propensity to integrate migrants?" This question is important for understanding the extent to which local conditions (e.g., local native preferences or institutional factors) affect migrants' social integration.

If migrants were randomly assigned locations to live without the ability to move, differences in their average outcomes by location would reflect causal effects of place. The setting in Germany does not feature such random assignment of migrants to locations. In particular, while the *number* of asylum seekers dispersed to locations within Germany is determined by a formula based on local population and tax revenues (the *Königsteiner Schlüssel*),[14] it remains possible that the *composition* of migrants by place is non-random. To overcome this challenge, we use a movers design that leverages differential changes in the same migrant's friending across locations to explore the extent to which regional variation reflects causal effects of location. We describe this design, and its limitations, in greater detail below.

County-Level Estimates. We begin by estimating county-level averages of our measures of Syrian migrants' social integration. Figure 2 maps the resulting county-level measures of friending integration, while Appendix Figures A9 and A10 show analogous maps for our language-based and group-based measures of integration. Syrian migrants in a 90th percentile county make more than twice as many local native friends on average as Syrian migrants in a 10th percentile county (7.9 vs. 3.9). Consistent with anecdotal evidence in Nawras (2017), the social integration of migrants tends to be highest in rural areas: migrants living in counties along the southern border, in Rhineland-Palatinate (along the western border), in Lower Saxony (in the northwest), and in Mecklenburg-Western Pomerania (near the Baltic Sea in the northeast) each have particularly high levels of social integration. By contrast, many mid-sized cities such as Ansbach, Kaiserslautern, and Cottbus rank among the bottom 20% of places in terms of the integration of migrants living there. Migrants living in larger cities, including Berlin, Munich, and Cologne, often have intermediate levels of social integration. Interestingly, there do not appear to be systematic differences between East and West Germany, despite their histories as distinct countries.

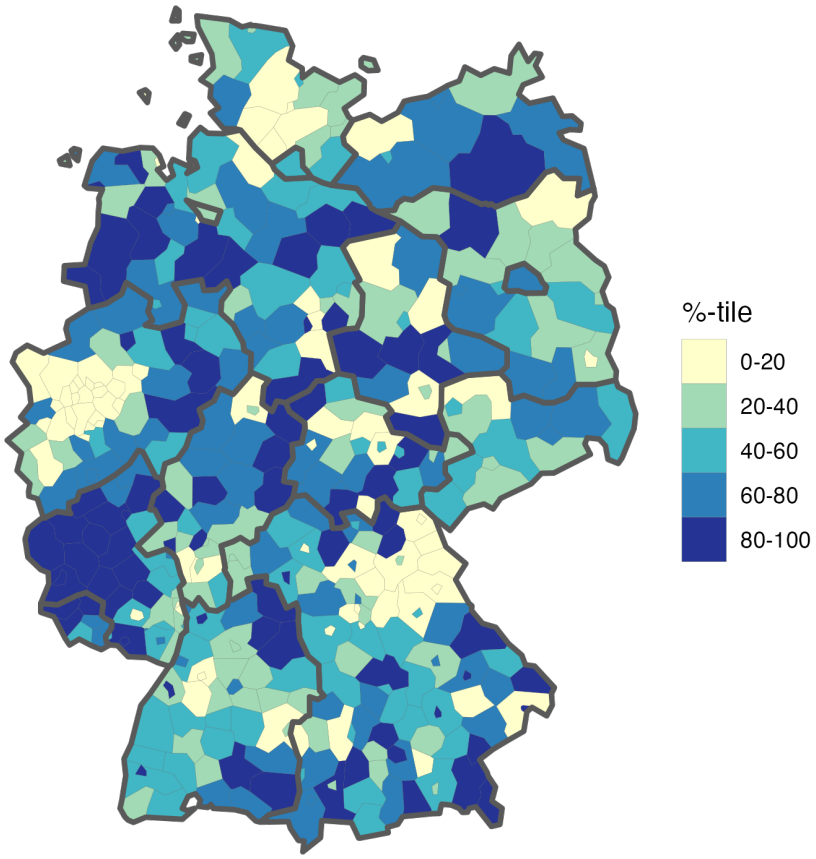
Panel A of Table 4 shows population-weighted county-level correlations between our various integration measures. The different integration outcomes are positively correlated across counties: those counties where Syrian migrants make more German friends are also the counties where they are more likely to use the German language and more likely to participate in local social groups.

County-Level Estimates: Validation. We next confirm that the differences in integration outcomes shown in Figure 2 reflect true differences in integration, expanding on the individual-level survey evidence presented in Section 1.3.

First, Appendix E shows that the county-level estimates of integration have high reliability, suggesting that the observed differences in integration do not arise from sampling error. For example, we find that if we randomly split the individual-level data into two halves and estimate the county-level average of native friending in each half, the two estimates have a correlation of 0.94.

[14] In Appendix Section F, we compare the distribution of refugees across places to the official assignment key and find that the two line up very closely, indicating that the assignment key has been followed relatively strictly even during these years of increased migration.

Figure 2: Regional Estimates of Integration – Friending to Native Locals



Note: Figure shows county-level estimates of Syrian migrant integration based on the average number of local native friends among Syrian migrants in each county (residualized on regional patterns of German natives' Facebook usage). Colors correspond to measure ventiles. Darker areas indicate the counties with the highest integration levels. The county-level estimates are reported in Appendix O.

Second, one might be concerned that differences in our county-level measures of social integration might reflect spatial variation in Facebook usage. While we find no spatial differences in Facebook usage among Syrian migrants, there are small spatial differences in Facebook usage patterns of German natives which could influence some measures of Syrian migrants' integration. For example, in a region where a smaller share of German natives uses Facebook, it might incorrectly look as if local Syrians were relatively less well-integrated according to the "local native friends" measure. To account for such concerns, we always residualize the observed average integration outcomes on county-level measures of the intensive and extensive Facebook usage of German natives. However, given the small magnitude of regional differences in natives' Facebook usage patterns, results are essentially the same when using unresidualized integration measures.[15]

Finally, we validate our regional measures of the social integration of migrants by comparing them to the average number of native acquaintances made by Syrian migrants in Germany as reported in

[15]Due to Facebook business restrictions, we are unable to publicly characterize the spatial distribution of natives' Facebook usage patterns. We verify that the high reliability estimates documented above are not driven by usage differences: in Appendix Table A12, we show the split-sample reliability before and after residualizing is similar (0.96 vs 0.94, for friending).

Table 4: Correlation Between County-Level Outcomes

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Panel A: Baseline Integration Measures							
(1) SY Migrants - N Local Native Friends	1.00						
(2) SY Migrants - Produced Content in DE	0.59	1.00					
(3) SY Migrants - N Local Native Groups	0.25	0.49	1.00				
(4) SY Migrants - N Local SY Friends	-0.03	-0.51	-0.41	1.00			
Panel B: Decomposition of Integration Measures							
(5) General Friendliness	0.62	0.29	-0.04	0.11	1.00		
(6) Relative Friending	0.73	0.51	0.40	-0.16	-0.05	1.00	
Panel C: Labor Market Integration Measure							
(7) Share Syrians in Employment or Training	0.45	0.59	0.13	-0.36	0.29	0.33	1.00

Note: Table presents correlations across county-level estimates. Panel A shows the regional averages of Syrian migrants after residualizing on measures of local German natives' intensive and extensive Facebook usage (see Section 2). Panel B shows the regional decomposition measures described in Section 3.1. Row 5 is general friendliness, generated as the regional average of German natives' local native friends after residualizing on local patterns of Facebook usage. Row 6 is relative friending, generated as the quotient from dividing the measure in row 1 by the measure in row 5. Panel C shows an external county-level measure of the share of all Syrians that are employed or in training programs according to data from the federal employment agency (see Appendix A16). Correlations are weighted by the number of Syrian migrant users in each county. Appendix Table A6 presents analogous signal correlations, which remove noise due to sampling error from the correlations.

the SOEP (see Section 1). This survey data is only available at less granular geographic levels, so we compare the two data sources at the state (and state-by-age group) levels. Despite different definitions of friendships and small sample sizes in the SOEP data, the regional measures of social integration are correlated with $\rho \approx 0.5$ across the two data sets, providing further evidence that our Facebook-based measures are picking up true variation in migrants' social integration (see Appendix Figure A8).

2.1 Evidence of Causal Place-Based Effects

The observed regional variation in integration outcomes of Syrian migrants could be explained by at least two forces. A first possibility is that places have causal effects on integration, either because of characteristics of the German natives living there, or because of institutional forces in the location. A second possibility is that there exist systematic differences in characteristics of Syrian migrants by place that shape their propensity to integrate—for example, if migrants with knowledge of the German language are more likely to live in certain areas. In this section, we provide evidence that the observed regional differences largely reflect causal place-based effects on integration.

Before turning to our movers design, we can directly rule out that *observable* Syrian migrant demographics are driving the regional differences in average integration outcomes. For example, regressing migrants' age, gender, and number of quarters since arriving in Germany on county fixed effects results in R^2s of 0.005, 0.003, and 0.005, respectively, highlighting that these characteristics vary little across counties. This finding is consistent with the fact that regional integration measures with and without individual-level observable controls are highly correlated (see Appendix Figure A7).[16]

[16] It is also consistent with the fact that adding county fixed effects in column 2 of Table A11 had little effect on the demographic coefficients relative to estimates in column 1.

Migrant Movers Design. Despite the lack of evidence for selection on observables and adherence to the Königsteiner Schlüssel, one might still worry that selection on unobservable characteristics explains the regional variation in integration. For example, while restrictions exist on asylum seekers' movements after settlement, these limitations are less severe for individuals who arrived prior to August 2016 or who have been in Germany for more than three years (see Hilbig and Riaz, 2022).

We next exploit such migrant moves to separate the role of place-based and non-place-based factors. Specifically, we focus on Syrian migrants who move between non-neighboring German counties, and study changes in the moving migrants' propensity to befriend local natives. This approach builds on recent work using similar designs to study place-based effects in different contexts (e.g., Card, Heining and Kline, 2013; Finkelstein, Gentzkow and Williams, 2016, 2021; Chetty and Hendren, 2018*a*,*b*).[17]

To see the intuition behind this research design, consider a Syrian migrant who moves from Ansbach, where Syrian migrants generally make few native German friends, to Saarlouis, where they make more native German friends. If the differences in Syrians' friending behavior between Ansbach and Saarlouis were due to (unobservable) characteristics of the Syrians in those places, we would expect the moving migrant's likelihood of befriending local natives to remain largely unchanged after the move. By contrast, if the observed geographic differences in Syrian migrants' social integration were primarily due to a causal effect of place, we would expect the moving migrant's likelihood of befriending native locals to increase by the average difference in this likelihood across the two locations. The *within-migrant* magnitude of the change in the rate of befriending local Germans around a move thus captures the importance of each explanation.

To study migrant movers, we construct a sample of Syrian migrants who were in one county for at least four consecutive quarters followed by a different, non-neighboring county for at least six consecutive quarters. We allow a user to be included for multiple moves so long as each move meets these criteria. Our sample includes 33,772 moves and 31,721 unique movers.[18]

Figure 3 plots Syrian migrants' probabilities of befriending local natives around a move, where quarter = 0 is the first quarter we observe the migrant in their new location. Counties are grouped into terciles of the integration measure mapped in Figure 2. Panels (a) and (b) focus on users who lived in a bottom and top tercile county prior to moving, respectively. In each panel, the lines correspond to individuals who move to counties in different integration terciles. The vertical axis plots the probability that a migrant makes at least one local German friend in a given quarter, a flow measure of social integration that allows us to study changes in the rate of integration around a move. To avoid picking up possible differences in natives' Facebook usage across locations, we residualize this flow measure of friending on measures of German natives' Facebook usage in the same location-quarter.[19]

[17] Our movers design uses panel data, as in Finkelstein, Gentzkow and Williams (2016). This design requires weaker assumptions than cross-sectional movers designs such as Chetty and Hendren (2018*a*), Chetty and Hendren (2018*b*), and Finkelstein, Gentzkow and Williams (2021). We provide more detail on the identifying model and assumptions in Appendix G.

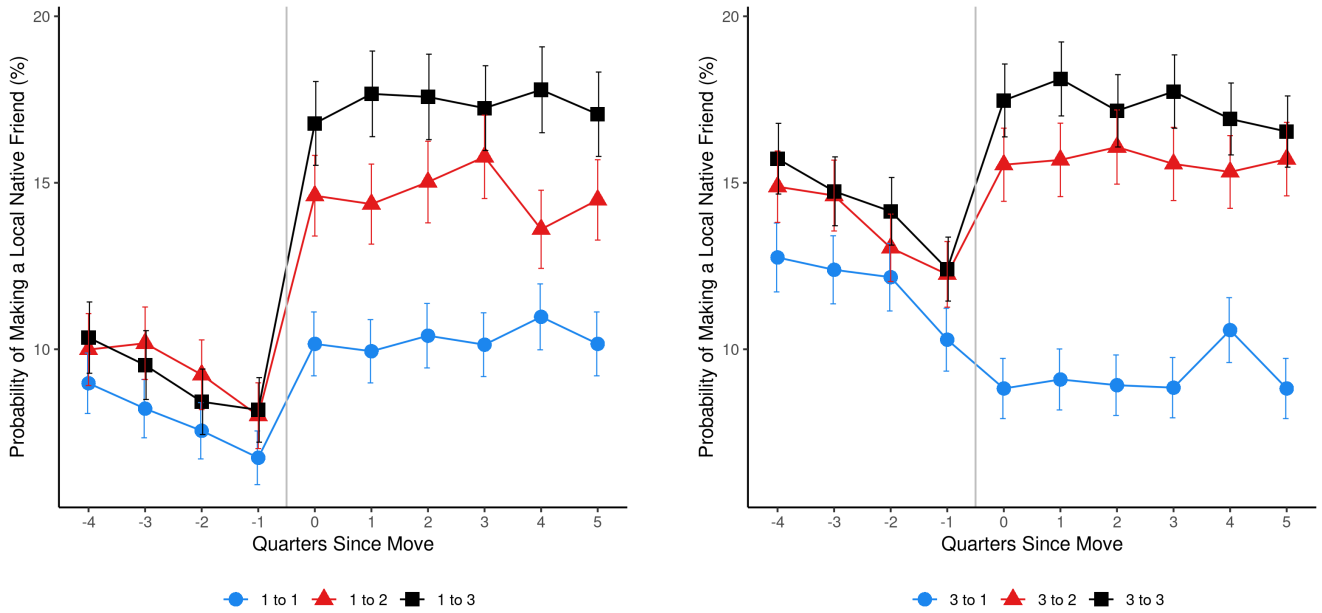
[18] Appendix Figure A11 shows that the number of moves between counties observed in the Facebook data is highly correlated with the number of moves observed in administrative data. Appendix Figure A14 shows that migrants do not systematically move to destinations with higher levels of social integration. Because our design is identified with *within-migrant* variation, even if such differential patterns of moving by local social integration existed they would not confound our results.

[19] In addition to overall usage, our design could be partially confounded by a differential representativeness of the Germans we pick up on Facebook (in terms of their propensity to befriend migrants) across space. The regional correlations of native behaviors with external survey data presented in Appendix L provide evidence that this concern is limited.

Figure 3: Change in Syrian Migrants' Friending of Local Natives Around a Move

(a) Moving From Bottom Integration Tercile

(b) Moving From Top Integration Tercile



Note: Figures show the quarterly probability that a moving Syrian migrant befriends at least one local German native, relative to the timing of the migrant's move. The population is Syrian migrant users who moved between non-neighboring counties and were in the first and second county for 4+ and 6+ consecutive quarters, respectively. Counties are grouped into terciles (weighted by the number of Syrian migrant users) of the regional friending-based measures of integration in Figure 2. Panels (a) and (b) limit to users who move from a county in the bottom and top tercile of integration, respectively. The different lines show movers to counties in each of the three terciles of social integration. The individual-level outcomes are residualized by the regional measures of Facebook usage described in Section 2. Bars display 95% confidence intervals of the estimates.

In both panels, the likelihood of migrants making new local German friends is decreasing prior to the move, consistent with individuals investing less effort in making new friends ahead of an anticipated move. There is little variation in the pre-move rate of making local German friends across the destination terciles, suggesting that individuals moving to a high-integration place behaved similarly prior to the move compared to individuals moving to a low-integration place.

Following the move, the probabilities of making local German friends vary systematically by the movers' destinations, with higher probabilities for individuals moving to places with higher overall social integration levels. These pattern exists in both panels, which we interpret as evidence for symmetric place-based effects. In general, there is also an additive increase in the rate of making local friends following a move, independent of integration levels in the origin and destination, consistent with all movers building new local networks following a move.[20]

In Appendix G we formally outline and estimate a simple model in which a migrant's rate of befriending local natives is determined by the sum of place-based effects—which we allow to vary across time and with observable migrant characteristics—and other *unobservable* individual-level factors. Since only place-based factors change around a move, this model allows us to estimate the share of regional variation in the social integration of migrants that can be attributed to place-based effects.

[20] In Figure A13, we show that both the probability of incoming and the probability of outgoing Facebook friendship requests follow similar patterns around a move.

The results suggest that differences in social integration across regions are largely due to causal place-based effects. Specifically, we find that nearly three-quarters of the observed regional variation in Syrian migrants' friendship formation with local natives is directly attributable to place-based effects that occur within the first year after their move. The results are not driven by any particular demographic group and are fully symmetric, with moves to low-integration places leading to declines in the rates of making native friends of the same magnitude as moves to high-integration places increase that rate. We summarize our results in Lesson I.

Lesson I: *Places differ in their propensities to integrate migrants. The substantial spatial variation in Syrian migrant integration outcomes in Germany largely reflects causal placed-based effects.*

3 Place-Based Effects: Immutable Native Preferences vs. Local Equilibria

Given the evidence for causal place-based effects, we now explore the role that immutable preferences of local natives play in determining these place-based effects (see also Khatua, Zagheni and Weber, 2023). We decompose local native behaviors into the rate at which they befriend their neighbors in general and the particular rate at which they befriend Syrian migrants versus other Germans. We then ask, "To what extent do persistent native characteristics (e.g., attitudes toward neighbors or migrants) versus the structure of local institutions or social equilibria shape each force?"

If some German natives were randomly assigned locations and could not move, the extent to which their behaviors matched the average behaviors in their location would reflect the extent to which local equilibria (instead of immutable preferences) shaped friending behaviors. In the absence of such an experiment, we again study variations in within-individual behavior around a move, now focusing on native movers. We describe the design and its potential limitations in this context below.

3.1 Decomposing Migrants' Integration: General Friendliness and Relative Friending

We distinguish two forces that can contribute to regional variation in migrants' social integration.

The first force, which we call *general friendliness*, is the overall rate at which natives in a location befriend others in their community: if local natives in a given location are more likely to befriend any neighbor, they might also be more likely to befriend their Syrian migrant neighbors.

The second force, which we call *relative friending*, is the relative probability of a German native befriending a given local Syrian migrant versus a given local German native. When natives befriend migrants and other natives with similar likelihoods, social integration becomes easier for migrants.

Our unique data allow us to measure these two components separately, and thus improve our understanding of the causal effects of place documented in Section 2. We define a county's general friendliness as German natives' average number of local German friends. Relative friending in a county is defined as migrants' average number of local German friends divided by the county's general friendliness. General friendliness and relative friending thus determine friending integration multiplicatively:

$$\underbrace{NLocalFriends_j^{SY\to DE}}_{Friending\ Integration} = \underbrace{NLocalFriends_j^{DE\to DE}}_{General\ Friendliness} \times \underbrace{\frac{NLocalFriends_j^{SY\to DE}}{NLocalFriends_j^{DE\to DE}}}_{Relative\ Friending}. \tag{1}$$

The variables $NLocalFriends_j^{DE \to DE}$ and $NLocalFriends_j^{SY \to DE}$ correspond to the average number of local native friends among native and Syrian migrant users in county j, respectively, after residualizing on regional patterns of Facebook usage in the native population as before.

Intuitively, relative friending captures how much harder it is for a Syrian migrant to make a local native friend than it is for a native German to make that friend. To build intuition for its determinants, it is possible to re-write county-level relative friending as a function of only natives' friending behaviors. We do so using the fact that, within a county, the total number of friendships from local migrants to local Germans must equal the total number of friendships from local Germans to local migrants:

$$Rel.\ Friending = \frac{NLocalFriends_j^{SY \to DE}}{NLocalFriends_j^{DE \to DE}} = \frac{NLocalFriends_j^{DE \to SY}}{NLocalFriends_j^{DE \to DE}} \times \frac{NGer_j}{NSyr_j} = \frac{\frac{NLocalFriends_j^{DE \to SY}}{NLocalFriends_j^{DE \to DE}}}{\frac{NSyr_j}{NGer_j}}. \quad (2)$$

Here, $NGer_j$ and $NSyr_j$ are the numbers of German native and Syrian migrant Facebook users local to county j, respectively. $NLocalFriends_j^{DE \to SY}$ is the average number of local Syrian friends of German natives in county j. Relative friending will thus be equal to one if German natives befriend local Syrian migrants and other local German natives in proportion to their population shares.

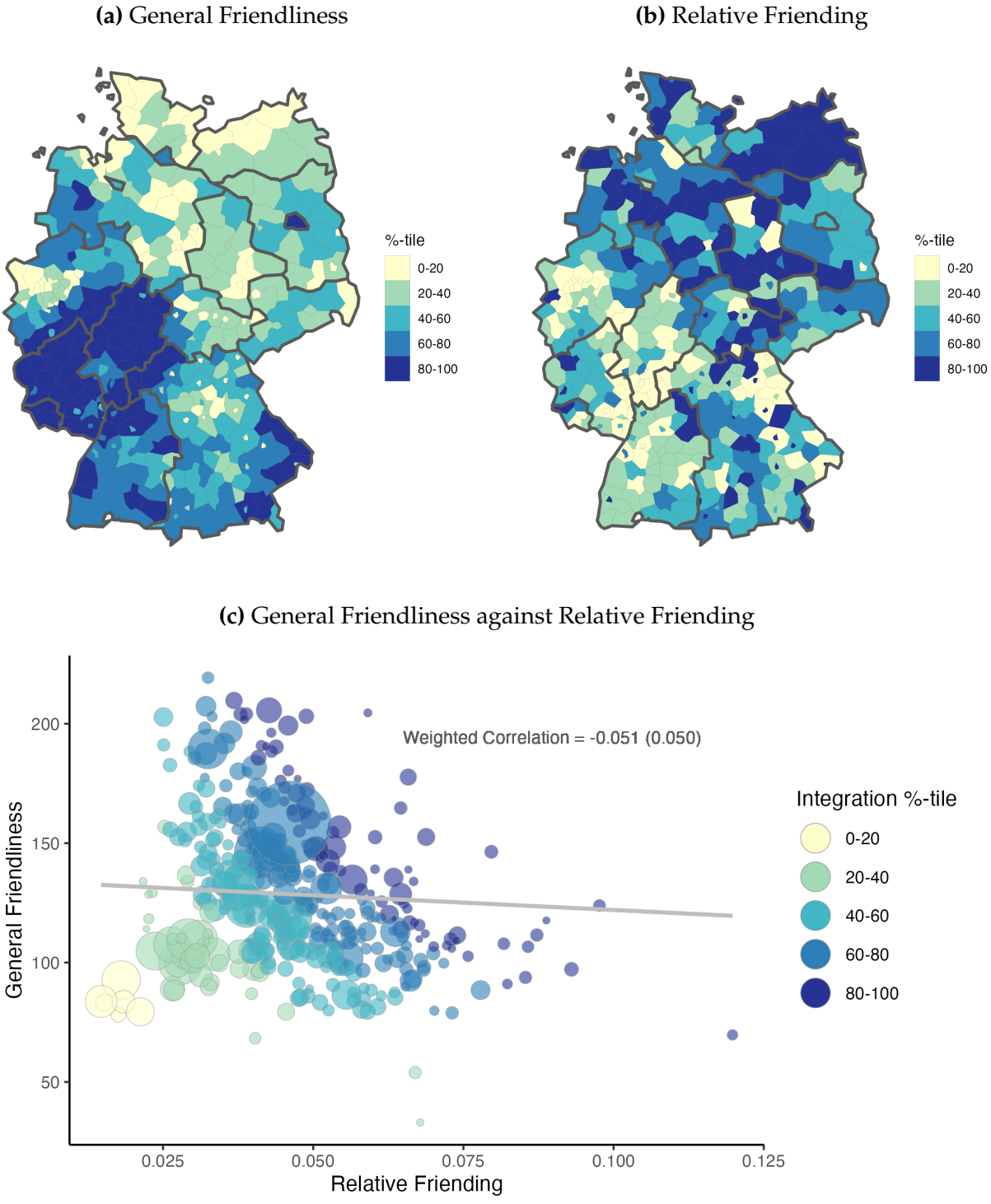
Panels (a) and (b) of Figure 4 map general friendliness and relative friending by county, while Panel (c) shows their across-county correlation, with different colors representing different overall integration levels of Syrian migrants. General friendliness is higher in Western states and lower in Northern Germany, while relative friending is generally higher in Northern Germany. The industrial areas in the Ruhr area of North Rhine-Westphalia—including the cities of Duisburg, Oberhausen, Bottrop, and Gelsenkirchen—as well as parts of upper Franconia in northern Bavaria have low general friendliness and low relative friending; migrants have the lowest integration levels in these places. Overall, general friendliness and relative friending are weakly negatively correlated across counties, with a weighted correlation of -0.05.[21]

To quantify the relative importance of general friendliness and relative friending in explaining county-level differences in integration, in columns 1 and 2 of Table 5 we separately regress the log of overall friending integration on the log of each component. The R^2 estimates of 0.41 and 0.66 for general friendliness and relative friending, respectively, suggest that differences in relative friending explain 50% more of the geographic variation in integration than differences in general friendliness do (see Appendix I for related analyses).

For some policy questions, it is not necessarily central to determine whether good integration outcomes in a given place are driven by high general friendliness or high relative friending. For instance, a policymaker interested in simply assessing the potential of different regions to socially integrate migrants—perhaps because they are interested in determining where to settle new refugees—may be

[21] In Appendix Figure A15 we plot our regional measures of integration and relative friending of Syrian migrants against analogous measures for migrants from countries that had many asylum seekers in Germany as of 2020. There exist strong positive relationships, suggesting local factors shape social integration similarly across migrant groups. Intuitively, the correlation is higher for the friending integration measures compared to relative friending measures, consistent with local natives' general friendliness playing an important role in shaping the former. At the same time, the variation in both plots shows there exist place-by-migrant-group-specific forces that affect integration outcomes.

Figure 4: Regional Estimates of General Friendliness and Relative Friending



Note: Panel (a) shows county-level estimates of general friendliness, the average number of local native friends among natives in each county (residualized on Facebook usage). Panel (b) shows county-level estimates of relative friending, given by the ratio of the overall friending integration measures and general friendliness (see equation 1, also residualized on Facebook usage). Colors correspond to measure ventiles. Darker colors indicate counties with higher values of general friendliness and relative friending, respectively. Panel (c) shows a county-level scatter plot of relative friending against general friendliness. The size of the bubbles corresponds to the number of Syrian migrants in the county. Darker colors indicate counties with the highest friending integration levels (mapped in Figure 2). The county-level estimates are reported in Appendix O.

Table 5: County-Level Relationship Between Integration Measures

	Friending Integration		Language		Employment / Training	
General Friendliness	1.098***			0.183***		0.558***
	(0.13)			(0.07)		(0.08)
Relative Friending		1.056***		0.255***		0.459***
		(0.07)		(0.03)		(0.06)
Friending Integration			0.228***		0.494***	
			(0.04)		(0.05)	
N	401	401	401	401	385	385
R-Squared	0.408	0.664	0.367	0.374	0.353	0.356

Note: Table shows results from multivariate regressions exploring the county-level relationship of integration measures with general friendliness and relative friending. In every specification, the outcomes and all controls are measured in logs. The outcomes are friending integration (columns 1 and 2), the share of Syrian migrants on Facebook who produce German content (columns 3 and 4), and the share of Syrians employed or in training programs (columns 5 and 6) according to data from the federal employment agency (see Appendix A16). Regressions are weighted by the number of Syrian migrants in the Facebook data. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01).

indifferent to which of the components drive this integration. Indeed, columns 3 to 6 of Table 5 show that both components of social integration have strong and similarly-sized positive associations with language- and economic-based measures of integration that policymakers may care about.

However, the distinction between general friendliness and relative friending can be important in other settings. Consider a policymaker seeking to improve a location's integration outcomes. While targeted policies might reduce the gap between natives' rate of befriending migrants versus other locals (i.e., relative friending), increasing the overall friending rate of natures (i.e., general friendliness) is likely more challenging. In addition, since general friendliness and relative friending shape integration multiplicatively, interventions that raise relative friending will increase integration most where general friendliness is high. Observing each component separately therefore allows policymakers to most effectively target interventions that maximize the overall social integration of migrants.

General Friendliness Validation. A potential concern with our measure of general friendliness is that it may partially capture local social norms about Facebook usage rather than real-world friending behaviors. While we control for county-level measures of overall Facebook usage (as described in Section 2), it remains possible that there are social norms around sending and accepting Facebook friendship requests that differ by place. To explore this concern, Appendix L benchmarks our regional general friendliness measures against external survey measures of social activity and trust. General friendliness is strongly correlated with the survey responses, suggesting that our measure captures real-world behaviors of German natives (consistent with prior evidence that our data captured the true levels of social integration of Syrian migrants).

3.2 Separating Between Immutable Preferences and Place-Effects

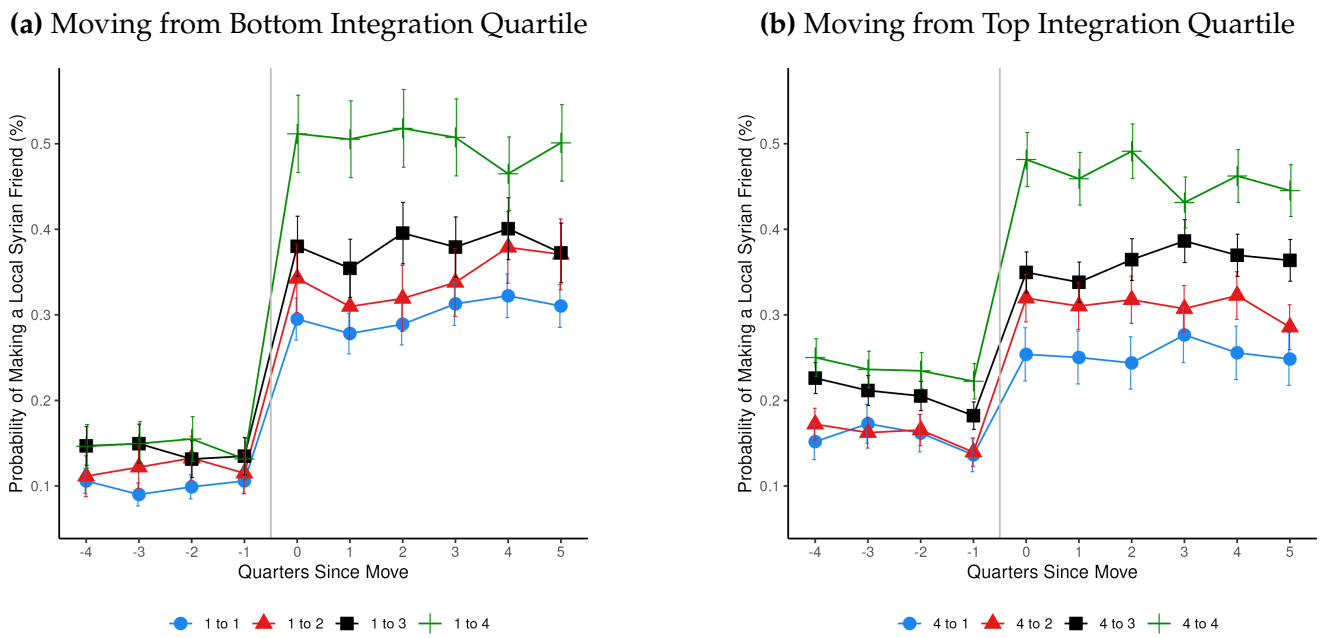
We next ask what role persistent native characteristics (e.g., attitudes toward neighbors or migrants) versus place-based effects (e.g., the structure of local institutions or social equilibria) play in shaping

general friendliness and relative friending. To do this, we use a movers design that explores changes in *natives'* friending patterns as they move between places with different relative friending and general friendliness. When place-based effects dominate fixed individual characteristics in determining local friending patterns, the native movers' friending behaviors should adjust substantially towards those of natives in the place they move to.

We focus on users who moved between two non-neighboring counties and who lived in the origin county for at least four consecutive quarters and the destination county for at least six consecutive quarters. We focus on moves since Q1 2017, when the substantial number of Syrians in Germany allows us to obtain more precise measures of friending.

Figure 5 shows event studies analogous to those in Figure 3. In both panels, the overall likelihood of German natives making new Syrian migrant friends in the post-move period is higher than in the pre-move period, consistent with natives increasing their general rate of friendship formation after a move. The probability of making local migrant friends in the post-period varies systematically by the movers' destination, with higher probabilities for individuals moving to places with higher overall social integration levels.[22]

Figure 5: Change in Natives' Friending of Local Syrian Migrants Around a Move



Note: Figures show the quarterly probability that a moving German native befriends at least one local Syrian migrant, relative to the timing of the native's move. The population is German native users who moved between non-neighboring counties and were in the first and second county for 4+ and 6+ consecutive quarters, respectively. Counties are grouped into quartiles (weighted by the number of German native users) of the regional friending-based measures of integration in Figure 2. Panels (a) and (b) limit to users who move from a county in the bottom and top quartile of integration, respectively. The different lines show movers to counties in each of the four quartiles of social integration. The individual-level outcomes are residualized by the regional measures of Facebook usage described in Section 2. Bars display 95% confidence intervals of the estimates.

[22] Appendix Figure A16 shows that natives do not systematically move to destinations with higher or lower levels of social integration. Because our design is identified with *within-native* variation, even if such differential patterns of moving by local social integration existed they would not confound our results.

We next estimate a specification that compares *changes* in the rates at which movers make friends in the year before vs. after their move to *differences* in the average friending rates of otherwise similar non-movers in each location. Appendix G provides additional details and a formal discussion of the underlying identifying assumptions. Specifically, the outcome variable $y^{\Delta}_{i,t}$ is the change in *yearly general friendliness* or *yearly relative friending* around a move. Yearly general friendliness is the number of local native friends a user makes in a given year. Yearly relative friending is the ratio of local Syrian migrant friends to local native friends made by a German native in a given year, compared to the relative population shares of Syrian migrants and German natives in that location (i.e., an annualized version of the "ratio of ratios" introduced in equation 2). $x^{\Delta}_{i,t}$ is the difference in the corresponding averages between native stayers of in mover's origin and destination, calculated at the same time and for the same gender $\times$ age group. Appendix Table A9 summarizes the sample of native movers and matched non-movers. We then estimate:

$$y^{\Delta}_{i,t} = \alpha_0 + \alpha_1 x^{\Delta}_{i,t} + \xi_t + \epsilon_{i,t}, \tag{3}$$

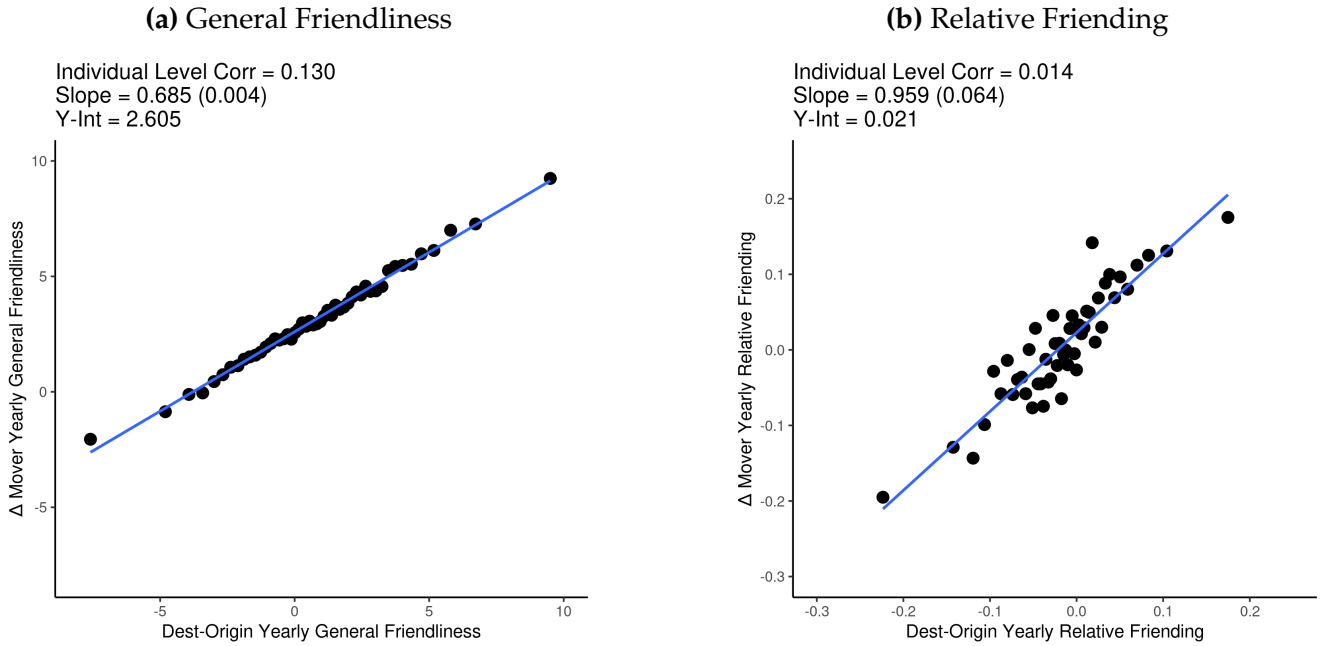
where ξ_T are quarter-of-move fixed effects. The slope α_1 provides an answer to the following question: "within a year of moving to a new place, to what extent does a moving native's friending behavior adjust to that of observably similar destination non-movers?"[23] An α_1 close to 1 suggests native movers' behavior completely adjusts, whereas an α_1 close to 0 suggests it does not adjust at all.

Figure 6 shows conditional binned scatter plots of $y^{\Delta}_{i,t}$ against $x^{\Delta}_{i,t}$, with slopes corresponding to α_1 (Appendix Table A10 provides the underlying regressions, as well as robustness specifications).[24] Panels (a) and (b) show plots for general friendliness and relative friending, respectively. In both panels, the relationship is linear and symmetric around zero, providing evidence of additive place-based effects. In Panel (a), the slope estimate suggests that, within a year of moving to a new place, a native will adjust their general friending 69% of the way to the level of comparable destination natives. In Panel (b), our estimates suggest that movers' relative friending will adjust nearly fully to that of their destination, though the estimates are somewhat less precise, since few natives make any Syrian migrant friends. Both panels thus provide evidence that institutional factors and local policies play important roles in shaping natives' friending behaviors. The fact that relative friending adjusts almost fully suggests that spatial differences in time-invariant individual-level characteristics such as attitudes towards migrants

[23] This interpretation is intentionally narrower than that in Section 2.1, where we interpreted α_1 as the share of across-region variation in integration that is explained by place-based effects. In particular, whereas regional differences in the observables for which we allow flexibility (gender, age, and arrival cohort) were essentially non-existent for Syrian migrants, regional differences in native demographics do have the potential to shape overall variation in our measures. For example, since older people are less likely to befriend Syrian migrants, regions with older populations on average may have lower levels of integration. Since we match movers to stayers with similar observables, our estimates will not capture variation in friending patterns across space that is due to the age of the native population. (Though we will show in Section 4 that relative to other factors, the quantitative importance of these county-level differences in natives' gender and age is small).

[24] One challenge with our estimation is that we only observe a sample estimate of each mover's $x^{\Delta}_{i,t}$, denoted by $\hat{x}^{\Delta}_{i,t}$. Measurement error in the true differences in friending probabilities of non-movers across locations would thus lead to attenuation bias in α_1. To account for this sampling error, when estimating equation 3, we randomly split the individual-level data of the friending behavior of non-movers used to construct $\hat{x}^{\Delta}_{i,t}$ into two sub-samples and instrument for the value constructed in one sub-sample with the value constructed in the other sub-sample (see Appendix E for details).

Figure 6: Δ Native Mover Behaviors vs. Matched Non-Movers



Note: Figures show binned scatter plots describing the change in the friending behavior of German natives before and after a move within Germany. The population is German native users who moved between non-neighboring counties and were in the first and second county for at least 4 consecutive quarters each. In both panels, the y-axis displays $y^{\Delta}_{i,t}$, an individual level change in movers' behavior the year before vs. after the move, and the x-axis displays $\hat{x}^{\Delta}_{i,t}$, the difference in average outcomes for comparable non-movers at the same time. In panel (a), the outcome is the change in the number of local German native friends made (*yearly general friendliness*) between the years. In panel (b), the outcome is the change in the ratio of the number of local Syrian migrant vs. local native friends, divided by the ratio of the number of local Syrian migrants vs. natives in the Facebook data (*yearly relative friending*) between the years. Panel (b) excludes users who make no local native friends in either the year before or after the move. In both panels we match each mover to a set of non-movers who match on gender and age buckets (18-29, 30-44, 45+). We include observations for which there are at least 1,000 non-movers in both the origin and destination match groups. Both panels include quarter-of-move fixed effects. We correct for sampling error in the x-axis measures by randomly splitting the individual-level non-mover data into two halves and instrumenting for one set of averages with the other. See Appendix E for more information this procedure. Standard errors are shown in parentheses.

play only a small role in explaining regional variation in Syrian migrants' social integration outcomes.[25] Lesson II summarizes this result.

Lesson II: *Spatial variation in migrant social integration can be decomposed into the rate at which natives befriend their neighbors in general and the particular rate at which they befriend migrants. Both forces vary across space, and both are largely influenced by local equilibria rather than spatial differences in immutable native preferences.*

[25] A number of works studying place effects in the U.S. find that new places exert stronger effects on younger individuals (Kling, Liebman and Katz, 2007; Chetty, Hendren and Katz, 2016; Chyn, 2018). Consistent with this, Appendix Figure A17 shows that younger native movers adjust their general friendliness and relative friending substantially more than older native movers. One possible reason for the stronger adjustment by younger movers is that places have cumulative effects, a force that would lead our large estimates of place-based effects to *understate* the full role of places on individuals' behaviors. In Section 6, we explore the potential role of such lasting effects by analyzing whether contact between migrants and natives in one setting has lasting effects on natives' friending behavior in other settings.

4 Place-Based Integration Outcomes: The Effect of Policy

Given our results in Sections 2 and 3 on the importance of place-based effects beyond any immutable preferences of the local populations, we next explore the determinants of local equilibrium integration outcomes and the extent to which these are shaped by local policies. We first show salient correlations between integration outcomes and a number of regional characteristics. Motivated by these analyses, we ask "Can the provision of integration courses improve social integration?"

If the availability of integration courses varied randomly across counties, differences in average migrant integration outcomes by the availability of courses would reflect a causal effect of these courses. In the absence of such random variation, we use an instrumental variables approach to provide evidence for a causal effect of integration courses on social integration outcomes. Our instrument, which leverages quasi-random variation in the presence of qualified teachers across counties, is described below.

4.1 Correlational Analyses

Table 6 presents multivariate regression analyses that explore how various county-level characteristics correlate with social integration, general friendliness, relative friending, and language integration.[26] Appendix M describes each measure in detail, and Appendix Figure A18 presents univariate county-level correlations between these and several additional county-level measures and social integration outcomes. To help with the interpretation of magnitudes, we use the log-form for the dependent and some of the explanatory variables, but the presented relationships are similar with raw magnitudes.

Demographics & Urbanity. While Syrians tend to be less socially integrated in places with an older population unconditionally (Appendix Figure A18), this relationship weakens significantly in the multivariate regressions in Table 6. In contrast, in both univariate and multivariate analyses, migrants are better integrated in less densely populated areas. The results in Table 6 show that this is driven by both relative friending and general friendliness being lower in urban areas. These trends are consistent with research finding that rural areas have higher levels of social capital and lower levels of social isolation relative to more densely populated urban areas (Putnam, 1995*b*; Rupasingha, Goetz and Freshwater, 2006; The Social Capital Project, 2018; Henning-Smith, Moscovice and Kozhimannil, 2019).

Economic Conditions. Some prior works have explored the feedback between social and economic integration. For example, Laurentsyeva and Venturini (2017) discuss the possibility that employment contributes to the social integration of migrants and Cheung and Phillimore (2014) use survey data to highlight the importance of language proficiency for employment. Table 6 shows that while there is no strong relationship between the average income level in a county and migrants' social integration, integration does appear to be higher in areas with lower unemployment rates, in particular when comparing counties within states. For instance, controlling for state fixed effects, we find that a 1% higher unemployment rate is associated with a 0.29% lower level of social integration, an effect that is largely driven by lower relative friending rather than lower general friendliness.

[26] In Table 6, we weight all relationships by the county's Syrian migrant sample size, except when we look at general friendliness as the outcome variable, in which case we weight by the county's German native sample size.

Table 6: County-level Multivariate Relationships with Friending Integration

	Friending Integration		General Friendliness		Relative Friending		Language	
Average Age	-0.032	-0.034*	-0.034***	-0.034***	0.015	0.003	-0.005	-0.011*
	(0.02)	(0.02)	(0.01)	(0.01)	(0.02)	(0.02)	(0.01)	(0.01)
Log Pop. Density 2018	-0.098*	-0.136***	-0.029	-0.071***	-0.066**	-0.058**	-0.034**	-0.016
	(0.05)	(0.03)	(0.03)	(0.02)	(0.03)	(0.03)	(0.01)	(0.01)
Log Average Income (in EUR)	-0.198	0.140	0.168	0.097	-0.296	0.054	0.070	0.035
	(0.26)	(0.18)	(0.14)	(0.10)	(0.20)	(0.15)	(0.08)	(0.06)
Log % Unemployed	-0.056	-0.291***	-0.108***	-0.065*	0.015	-0.209***	-0.129***	-0.032
	(0.09)	(0.09)	(0.04)	(0.04)	(0.07)	(0.07)	(0.03)	(0.03)
Vote Share AFD European Elections 2014	-8.953***	-6.167***	-1.939**	-1.039	-6.917***	-5.091***	-0.569	-1.289**
	(2.64)	(1.92)	(0.85)	(0.69)	(2.29)	(1.55)	(0.68)	(0.65)
Number of ProAsyl Groups per Pop	4.778*	4.286***	-1.381	-0.341	4.876***	3.167**	3.557***	1.558**
	(2.55)	(1.40)	(1.22)	(0.76)	(1.69)	(1.29)	(0.85)	(0.62)
Log Fraction of Syrians 2010	0.105***	0.150***	0.025***	0.030***	0.067***	0.114***	0.019**	0.043***
	(0.02)	(0.02)	(0.01)	(0.01)	(0.02)	(0.02)	(0.01)	(0.01)
Log Fraction of Syrians 2019	-0.239***	-0.135***	-0.048*	-0.065***	-0.117**	-0.060	-0.044*	-0.103***
	(0.08)	(0.05)	(0.03)	(0.02)	(0.05)	(0.04)	(0.02)	(0.02)
Log Int. Courses Completed 2015-19 per Syrian	0.235***	0.200***	0.005	-0.013	0.222***	0.202***	0.076***	0.052***
	(0.05)	(0.04)	(0.02)	(0.02)	(0.04)	(0.03)	(0.02)	(0.01)
State FE		x		x		x		x
R-squared	0.487	0.709	0.261	0.665	0.330	0.633	0.519	0.668
N	390	390	390	390	390	390	390	390

Note: Table presents results from regressions of various county-level measures on the logs of friending integration (columns 1 and 2), general friendliness (columns 3 and 4), relative friending (columns 5 and 6), and language (columns 7 and 8). The regional measures are average age, log 2018 population density; log average income, log employment rate; the vote share for the Alternative for Germany, demeaned by state, pro-immigration groups per capita; log of the shares of the population that were Syrian in 2010 and 2019, and log of the numbers of integration courses completed from 2015-2019 per Syrian. For more information on each measure see Appendix Table A16. Regressions are weighted by the number of Syrian migrants in the Facebook data in columns 1-2 and 5-8. Regressions in columns 3 and 4 are weighted by the number of natives in the Facebook data. Standard errors are shown in parentheses. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01)

Attitudes Towards Migrants. We explore correlations of integration outcomes with two measures of local attitudes towards migrants: (i) the vote share for Alternative for Germany or AfD, a political party in favor of limiting immigration, in the 2014 EU Election (predating the main influx of Syrian migrants);[27] and (ii) the number of pro-immigration groups per capita. Support for the AfD has a strong negative relationship with social integration and relative friending: a one percentage point increase in AfD vote share relative to state-level averages is associated with a decrease in social integration of nearly 9% and in relative friending of 6.9%. Pro-immigration groups are independent organizations that offer a wide range of services to migrants, including help filing for asylum status, medical attention, and the provision of child care. We study groups registered with *ProAsyl*, a widely-known pro-immigration organization in Germany. In both univariate and multivariate analyses, we find places with relatively more pro-immigration groups per capita tend to have higher levels of social integration. Table 6 shows this is driven entirely by variation in relative friending rather than general friendliness.

Concentration of Migrants. Several researchers have studied the relationship between local co-ethnic populations and the economic integration of migrants. For example, Edin, Fredriksson and Åslund (2003) and Damm (2009) find a positive effect on earnings for refugees living in areas with more co-ethnic individuals (so-called "ethnic enclaves"), while Cutler, Glaeser and Vigdor (2008) find negative

[27] Because political parties in Germany are differentially important across states, and often run with varying policy positions by state, in Table 6 we always demean AfD vote share by state.

effects if the community has low levels of average education. Our results suggest that newly-arriving migrants do make fewer native friends in places with more *recent* Syrian migrants. However, we see that social integration generally increases with the share of the population that was Syrian in 2010, largely through effects on relative friending. We find similar results when looking at the extent of German language usage. These patterns are consistent with earlier migrants providing important information or connections with natives to aid the social integration of new arrivals. It is also possible that local natives more exposed to Syrian migrants in 2010 became more friendly toward Syrians in the future, a notion we explore at the individual level in Section 6. On the other hand, large communities of migrants arriving at the same time appear to facilitate fewer migrant-native connections.

Integration Courses. The German government has invested heavily in efforts to integrate recent migrants (see, e.g., Bundesregierung, 2021). Integration courses, which are intended to teach migrants the German language and other relevant information, are "at the core of the government's integration measures" (BAMF, 2015). Indeed, 1.13 million individuals participated in these courses between 2015-2019 (BAMF, 2021). In both the univariate and multivariate analyses, we find strong positive relationships between a county's social integration outcomes and the number of integration courses completed per Syrian between 2015 and 2019. The effect appears to be entirely driven by a relationship between integration course completion and relative friending. While these results are not causal, they are consistent with integration courses supporting the integration efforts of Syrian migrants. To isolate a possible causal effect of integration courses, we next use an instrumental variables approach that leverages exogenous variation in course availability across regions.

4.2 Causal Effect of Integration Policy: Integration Courses

Unlike many regional characteristics related to social integration, such as population density, policymakers can and do influence the offering of integration courses. In this section, we therefore study the *causal* effects of integration courses on integration outcomes using an instrumental variables (IV) approach that exploits the effect of quasi-random variation in the presence of qualified teachers across counties on the availability—and in turn completion—of integration courses. This IV approach is necessary to identify causal effects, since prior work has noted that integration courses are offered more frequently in urban areas with a high share of foreigners, attributes that themselves affect migrants' social integration (Kanas and Kosyakova, 2022).

The German government required individuals teaching integration courses to either have a college degree in teaching German as a second language or, with a degree in a different pedagogical field, significant experience teaching German as a second language (BAMF, 2018). Due to these very specific requirements, integration courses were generally taught by the small group of previously unemployed teachers with these qualifications. Indeed, in a widely-televised 2016 interview, the federal government's coordinator of refugee policy (*Flüchtlingskoordinator*) called on unemployed teachers to meet the rapid demand for integration course instructors (Tagesschau, 2016). The unemployment rate of qualified teachers in a given county at the start of the major influx of migrants thus likely affected the availability of integration courses in that county. We test this story using county-level data on 2014 teacher unemployment from the Federal Employment Agency. These data allow us to distinguish between four types

of teachers: general, vocational, driving or sports, and other. "Other" teachers are primarily adult educators, often focused on non-native populations, and are much more likely than the other groups of teachers to meet the requirements to teach integration courses. Therefore, if local teacher unemployment affects integration course availability, it should do so primarily through this particular set of teachers.

Table 7: Integration Courses and Teacher Unemployment Rates

	Log Integration Courses per Syrian 2015-19			
Log Unemp. General Schools Teachers 2014 per Syrian	0.103*			
	(0.06)			
Log Unemp. Vocat. School Teachers 2014 per Syrian		0.089*		
		(0.05)		
Log Unemp. Driving and Sports Teachers 2014 per Syrian			0.056	
			(0.06)	
Log Unemp. Other School Teachers 2014 per Syrian				0.234***
				(0.05)
Control Covariates	x	x	x	x
Control Log General Unemployment Rate	x	x	x	x
F-statistic	3.44	4.43	1.17	23.56
N	390	387	388	390
R-Squared	0.398	0.401	0.394	0.427

Note: Table presents results from county-level regressions between various 2014 teacher unemployment rates per Syrian and integration course completion. The outcome is the log of the number of integration courses completed per Syrian between 2015 and 2019. In all regressions we control linearly for the log of the share of the population unemployed, the number of unemployed people per Syrian (as of 2014) as well as average age, log population density, log average income and log number of open training positions per applicant. Regressions are weighted by the total number of Syrians in each county as of 2019. Standard errors are shown in parentheses. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01)

Table 7 presents results that are highly consistent with the presence of qualified teachers driving the availability, and eventual completion, of integration courses. Columns 1-3 show that, after controlling for the general unemployment rate and other county-level covariates, there are at most weak relationships between integration course completion and unemployed general, vocational, and driving or sports teachers per Syrian. By contrast, column 4 shows a positive and highly significant relationship for "other" teachers: a 10% increase in the number of unemployed "other" teachers per Syrian as of 2014 corresponds to a 2.3% increase in integration course completion per Syrian. With an F-statistic of over 23, this first-stage relationship is remarkably strong given the limited number of counties.

While this evidence supports the notion that teacher unemployment meaningfully affects the completion of integration courses, for the measure to serve as a valid instrument it must also satisfy the exclusion restriction. Namely, teacher unemployment must not affect social integration other than through its effect on the availability of integration courses. To mitigate concerns that our results are driven by general economic conditions or other confounders that might affect integration, we always include a rich set of county-level controls in our regressions: general unemployment rates, the number of unemployed people per Syrian, average age, population density, average incomes, and open training posi-

tions.[28] Moreover, our use of 2014 teacher unemployment, before the large influx of migrants, allows us to rule out stories in which reverse causality violates the exclusion restriction.

Table 8: IV Estimates - Measures of Integration and Integration Courses

	Integration	General Friendliness	Relative Friending	Language	Employ. / Training
Log Integration Courses per Syrian	1.792***	0.154	1.542***	0.342***	0.902***
	(0.34)	(0.17)	(0.27)	(0.07)	(0.15)
Control Covariates	x	x	x	x	x
Control Log General Unemployment Rate	x	x	x	x	x
N	390	390	390	390	384

Note: Table presents results from county-level IV regressions of various measures related to integration on the completion of integration courses. We instrument for integration courses with the 2014 total number of unemployed "other" per Syrian. In both stages of our estimation we include the same controls as in Table 7. The outcomes are overall friending integration (column 1), general friendliness (column 2), relative friending (columns 3), the share of Syrian migrant Facebook users producing content in German (column 4), and the share of all Syrians employed or in training programs (column 5). All dependent variables are specified in logs. Regressions are weighted by the total number of Syrians as of 2019 except when the outcome variable is general friendliness in which case we weight by the number of German natives in the Facebook data. Standard errors are shown in parentheses. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01)

Table 8 presents results from our IV regressions. Column 1 suggests that a 10% increase in completed integration courses per Syrian raises the social integration of Syrians by nearly 18%. Quantitatively, this means that moving a migrant from a 25th percentile to a 75th percentile county in terms of the relevant teacher unemployment would result in them having about 1.7 more native friends.

This IV estimate is substantially larger than the OLS estimates in columns 1 and 2 of Table 6. At least two forces contribute to this. First, our IV strategy corrects for possible downward bias due to omitted variables in the OLS estimates. Such downward bias can occur, for example, if integration courses were specifically targeted toward areas with low integration levels. We find supporting evidence that this is indeed the case: on average, courses tend to be concentrated in urban areas and places with a greater total immigrant share, both factors that are negatively correlated with integration, as discussed in Section 4. Second, the IV identifies a Local Average Treatment Effect (LATE), rather than an Average Treatment Effect (ATE). If the marginal integration course participant (aided by expanded course supply) had higher-than-average returns from integration courses, the LATE would exceed the ATE. There are good reasons to think that the marginal course participant did indeed benefit more from the course. For example, women were less likely to participate in integration courses when those courses were in short supply, but they also achieved substantially higher performance in both language and civic tests administered at the end of the course (Tissot et al., 2019; Tissot, 2021). While both LATE and ATE estimates are relevant for different applications, the LATE from our IV strategy is likely to be of particular interest for policymakers, whose primary tool to increase the completion of integration courses is to make them more easily accessible. Our LATE provides an estimate of the marginal effectiveness of such relaxations of supply constraints.

Columns 2 and 3 of Table 8 present IV estimates of the effect of integration courses on general friend-

[28] Our controls differ from the variables used in Table 6, since we refrain from controlling for covariates that are potentially endogenous to our outcome of interest, such as the share of Syrians in 2019 or the number of pro-immigration groups.

liness and relative friending—the two factors driving migrant integration. Because friending behavior among natives should not be impacted by integration courses, integration courses should affect overall integration only through relative friending. Highly consistent with this story, we find significant effects for relative friending, but not for general friendliness. We summarize our results as follows:

Lesson III: *Integration courses can causally affect place-specific equilibrium integration. The availability of these courses for Syrian migrants shifted the relative rates of German-Syrian friendships.*

Columns 4 and 5 measure the causal effect of integration courses on language and economic integration. In particular, our outcomes are the share of Syrian migrant Facebook users producing content in German (in column 4) and the share of all Syrians employed or in training programs (in column 5). For both of these outcomes, we find highly significant and positive effects of integration courses. The IV estimates suggest that a 10% increase in integration course completion increases language integration by just under 2% and the rate of Syrians in employment or training by about 9%. The latter result is suggestive of causal ties between social integration and economic outcomes, which we explore in the next section.

5 The Effects of Social Integration

In the prior sections, we explored the determinants of social integration. While social integration is itself an important outcome for many policymakers, we next ask whether social integration directly affects other outcomes across the following domains: the labor market, housing, education, and health.

If German friends were assigned randomly to migrants, differences in migrant outcomes would reflect the causal effects of social connections. Conceptually, such an experiment is difficult to imagine. Instead, to understand the observed relationships between social connections and economic outcomes, we survey migrants about the ways their native friends have impacted their experiences in Germany.

Before describing our survey results, recall two prior findings that support a relationship between social integration and other outcomes. First, columns 5 and 6 of Table 5 show that a county's friending integration, general friendliness, and relative friending are all positively correlated with the share of Syrian migrants employed or in training programs. This correlation is consistent with a causal effect of social integration, but might also partially reflect a reverse effect of economic integration on social integration. Second, column 5 of Table 8 shows a positive *causal* effect of integration courses on Syrian migrant employment. However, the observed effect might be due to a direct effect of integration courses on employment outcomes rather than an effect mediated through social integration (e.g., if the course provides job-seeking support).

To address these potential confounders, we analyze responses to the short user survey described in Section 1.3.[29] The survey asked migrants whether they had "a German friend or acquaintance" who helped them or a member of their family in various specific ways. The relatively high average outcome levels show that migrants frequently receive help from Germans across a range of setting. For example,

[29] All questions were translated to the user's preferred language on Facebook. Appendix N provides screenshots of the survey in English, German, and Arabic.

48% of migrants report having received help from a native friend with finding a job, and about 55% report having received help navigating the German bureaucracy.

In Table 9, we document that migrants with more native German Facebook friends are more likely to report having received various help from natives. Column 1 shows that a one standard deviation increase in a Syrian migrant's number of local native friends corresponds to a 12.6% (6.1 percentage-point) increase in the probability a native German helped them find a job. In column 2, a strong relationship remains even with controls for age, gender, county, and measures of Facebook usage. While this result is correlational, it is consistent with the notion that social integration *as measured by Facebook* positively affects migrants' labor market integration (the causal framing of the question already provides direct evidence that native friends frequently help Syrian migrants find jobs).

Table 9: Outcomes vs Measured Friending Integration at Individual Level

	Native Friend Helped Find Job		Native Friend Helped Find Housing		Native Friend Helped with School		Native Friend Helped with Bureaucracy		Native Friend Helped with Healthcare	
N Local Native Friends	0.497***	0.347***	0.370***	0.375***	0.159*	0.191*	0.444***	0.375***	0.131	0.084
	(0.095)	(0.109)	(0.096)	(0.121)	(0.084)	(0.102)	(0.095)	(0.109)	(0.090)	(0.112)
Control Covariates		X		X		X		X		X
N	2,738	2,687	2,738	2,687	2,738	2,687	2,738	2,687	2,738	2,687
Sample Mean	48.32	48.49	47.59	47.45	26	25.98	54.67	54.45	32.58	32.45

Note: Table shows results of individual-level regressions of survey responses about real outcomes on the number of local native friends. The outcomes in all columns are responses to the question "Do you have German friends or acquaintances that have helped you or a member of your family? If so, please select all the ways in which they have helped." The sub-questions were: "Finding a job" (columns 1-2); "Finding an apartment or place to live" (columns 3-4); "Completing school work" (columns 5-6); "Navigating the bureaucracy (filling out official documents, identifying the right people to speak to, etc.)" (columns 7-8); "Navigating the healthcare system (finding doctors, scheduling appointments, etc.)" (columns 9-10). Columns 2, 4, 6, 8, and 10 include the same controls used in 2, including for age, gender, county, and measures of Facebook usage. Standard errors in these columns are clustered by county. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01).

Columns 3-8 show similarly strong relationships across a variety of additional outcomes. Columns 4, 6, and 8 suggest that, with controls, a one standard deviation increase in local native friends corresponds to an increase in the probability that a native helped find housing by 9.7%, helped with school work by 9.0%, and helped navigate the bureaucracy by 8.4%. Each of these are important factors that shape the overall well-being of migrants, consistent with normatively positive effects of social integration on other outcomes.[30] While there is a positive relationship between native friending and the probability of receiving healthcare help, it is insignificant, possibly because migrants are relatively young on average. We summarize our results in Lesson IV.

Lesson IV: *Social integration appears to be an important channel of positive effects on migrants labor market, housing, and education outcomes.*

[30] We also asked migrants directly about whether they were satisfied with life in Germany. There is a positive but insignificant relationship with local native friending. The lack of a stronger relationship may largely be because overall levels of satisfaction were high (an average of 4.26 of 5) with little variation.

6 Exposure and Native Behaviors Toward Migrants

While Lessons I-III provide insights into the across-region variation in average integration outcomes, there is also substantial within-region variation in the friending behaviors of natives (see Appendix D). In this section, we explore one potential determinant of these differences, asking "Do differences in natives' exposure to Syrian migrants early in life affect their later attitudes and behaviors?" We focus on studying the long-run effects of high school interactions between natives and migrants.

If migrants were randomly assigned to high schools, differences in average native later-in-life attitudes and behaviors by the presence of migrants in their school would reflect a causal effect. To approximate such a research design, we exploit Germany's strict age cutoffs for school entry to provide variation in natives' exposure to migrants.

Sample Construction. To generate our sample for this analysis, we first subset our German native and Syrian migrant samples to those with a birth date between 1995 and 1999. These individuals were roughly 15 to 19 years old in 2014, at the start of the major influx of Syrian migrants. We observe 26,000 Syrian migrant users and 2.2 million German native users in that age range. We match individuals to their high schools using self-reports and friend-based imputations (see Appendix K). We assign 63.2% of individuals within this age group to a high school. We then sort individuals into cohorts within a school using the German system of age cutoffs for school entry. In Germany, children are eligible to enroll in school for the first time if they have turned six by a certain date that varies by state. Though students are allowed to enroll earlier or to defer enrollment at the advice of a pediatrician, the vast majority of students comply with the entry time suggested by the cutoff date (Schwandt and Wuppermann, 2016).

Research Design. Since students are disproportionately exposed to individuals in their own grade (relative to individuals in the years above and below them), variation in cohort composition can generate exogenous differences in the social networks formed by the members of each grade. Similar sources of variation in exposure and network composition have been utilized in other studies (e.g. Chetty et al., 2022*b*; Billings, Chyn and Haggag, 2021; Sacerdote, 2011). Because Syrian students are relatively uncommon in the German school system overall, we focus on how German natives are affected by having at least one Syrian migrant in their cohort. In particular, we focus on adjacent cohorts within a school where one cohort contains at least one Syrian migrant and the other does not. For instance, if the only Syrian who attends Marie Curie Gymnasium is in the class of 2016, we will study natives who fall on either side of the cutoff that divides the 2015 and 2016 cohorts.[31] We estimate equations of the form:

$$Y_i = \alpha_1 SyrianInCohort_i + \xi_{t,L} + \gamma_s + \epsilon_{i,t}. \tag{4}$$

[31] Conceptually, we could also study Germans around the assignment cutoff for the 2016 and 2017 cohorts. However, since many Syrians enter the German school system with low levels of German proficiency, some are assigned to a cohort younger than would be suggested by the assignment rule (though we find that most Syrians have a plurality of their friends in the cohort they would be assigned into under the allocation rules used for Germans). As a result, if we use this second design (where the Syrian is supposed to be in the older cohort), we will swap the treatment and control groups of Germans when the Syrian is assigned to a *younger* cohort. We also exclude pairs of years where there is a cohort without Syrians that is flanked by cohorts with Syrians. Since Syrians from the older cohort are sometimes misassigned, these configurations can lead us to inadvertently compare two cohorts that both contain Syrians, which would attenuate our results.

Here, Y_i is the number of friends of a given type that user i has today, and $SyrianInCohort$ is an indicator variable set to one if a user has at least one Syrian in their assigned school cohort. We also include birth year-by-county fixed effects $\xi_{t,L}$ to address concerns that locations where a particular cohort is more likely to encounter a Syrian in high school might also be locations where one is more likely to encounter Syrians in other settings (such as sport clubs). Finally, we include school fixed effects, γ_s. Under the assumption that it is random whether a student's birth date places them into a cohort with a Syrian or into an adjacent cohort without one, α_1 identifies the effect of the additional exposure via placement into a cohort containing a Syrian. In some specifications, we include an interaction term, $SyrianInCohort_i \times CohortSize_i$, where $CohortSize_i$ is the number of students in that cohort, normalized to have mean 0 and standard deviation 1. This interaction term allows us to examine how the effects of exposure differ according to the size of the cohort.

Effects of Exposure. In Table 10, we quantify the effects of being in a cohort that includes a Syrian migrant. The first column presents baseline results: students placed into a cohort containing a Syrian have 0.02 more Syrian friends by age 21, an increase of around 40% relative to the 0.054 Syrian friends that Germans in the adjacent cohort have on average. In the second column, we interact the treatment term with the cohort size. We find that treated students in a cohort that is one standard deviation larger make one-third fewer additional Syrian friends than treated students in a smaller cohort.

Table 10: Impacts of High School Exposure on Friendship

	Syrian Friends		Syrian Friends (Excluding Classmates)		Syrian Friends (Excluding Syrian Classmates and their Friends)	
Syrian in Cohort	0.020***	0.020***	0.005***	0.005***	0.005***	0.005***
	(0.002)	(0.002)	(0.002)	(0.002)	(0.001)	(0.001)
Syrian in Cohort x Standardized Cohort Size		-0.007***		-0.003***		-0.003***
		(0.001)		(0.001)		(0.001)
School FE	X	X	X	X	X	X
Birth Year x County FE	X	X	X	X	X	X
N	115,625	115,625	115,625	115,625	115,625	115,625
Mean in Control Cohort	0.054	0.054	0.029	0.029	0.027	0.027

Note: Table presents results from regressions of the form outlined in Equation 4. The sample includes Germans who were assigned to a high school cohort pair where the younger cohort contains a Syrian and the older cohort does not. The treatment years include students who entered kindergarten between 2001 and 2004, while students in the paired control cohorts entered kindergarten between 2002 and 2005. In columns 1-2, we include all Syrian friends that a user makes; in columns 3-4, we only include Syrian friends who did not attend the user's high school; and in columns 5-6 we only include Syrian friends who did not attend the user's high school and who did not have a prior friendship with a Syrian that attended the user's high school. In all columns, we include only Syrian friends made in the first 21 years of a person's life, in order to avoid mechanically calculating larger treatment effects for older users. All users in our sample have already turned 21. In all columns, we cluster standard errors at the school and cohort level. *(p<0.10), **(p<0.05), ***(p<0.01)

We next turn our attention to the mechanisms through which these friendships can be formed. Broadly speaking, there are three possible mechanisms. First, and most trivially, German natives can befriend the Syrian in their cohort. Second, the Syrian can play a direct role in mediating connections between native Germans and other Syrians by introducing previously disconnected individuals across groups. Third, the presence of the Syrian can play a role in shaping the preferences of native Germans for contact with other Syrians. This last mechanism could play a role in future network formation if

stereotypes about individuals outside one's own group affect friendship formation.

In columns 3 and 4, we repeat the regressions in columns 1 and 2, but now include only Syrian friends who did not attend the German's high school in our outcome measure. This allows us to isolate friends made through the second and third mechanisms above. We find that Germans in the treated cohorts make 0.005 more friends of this type, about 17% more than the average number of such friends in the control group. As in column 2, these effects are larger for students whose cohorts are smaller. These friendships outside of one's school comprise about one-quarter of the overall effect of exposure.

In columns 5 and 6, we exclude from the dependent variable both Syrians who attended the German's high school (as in columns 3 and 4) as well as any friends of those Syrians. The estimate is similar to column 3, indicating that many of the new friendships were made in new social contexts and do not correspond to connections directly facilitated by the Syrians in one's school. We summarize this finding, which is consistent with evidence that quasi-random exposure to migrants shifts natives' attitudes found in other settings (see Bursztyn et al., 2024), as follows.

Lesson V: *Natives exposed to a migrant in high school are more likely to befriend other migrants later in life. Connections directly facilitated by the first migrant do not fully explain this effect.*

7 Conclusion

The challenge of successfully integrating immigrants into new communities has become of central importance for policymakers around the world. In the coming decades, climate change could displace as many as one billion individuals, increasing the flow of international migrants and further raising the importance of these challenges (Kamal, 2017). However, due to the difficulty of measuring social networks using traditional data sources, understanding the drivers and effects of migrants' social integration has historically proven challenging. Are there environments where newly arriving migrants are relatively better integrated, and why? What can governments do to foster the social integration of migrants?

We use data from Facebook to draw five lessons from the experience of Syrian migrants in Germany. First, we document sizable spatial variation in the social integration of Syrian migrants in Germany. We show this variation is largely driven by causal place-based factors rather than unobserved migrant characteristics. Second, we show that regional variation in migrants' social integration outcomes is shaped by both the rate at which local natives befriend other locals in general (*general friendliness*) and the relative rate at which they form friendships with Syrian migrants in particular (*relative friending*). Natives' friending behavior adjusts substantially along both margins when they move between locations, suggesting that local institutions and environments are more important than fixed individual preferences of natives in determining whether a native makes migrant friends (although both play some role).

We then describe several characteristics of communities where migrants are better integrated. For example, our results suggest that large numbers of migrants arriving at the same time may lead to fewer migrant-native connections, but when migrants arrive in a place with many *earlier arriving* migrants they make more native connections. Similarly, our third lesson is that integration courses have a substantial positive *causal* effect on relative friending. This finding highlights that integration outcomes are not immutable, but can be shaped by government policies.

While social integration is itself an important outcome for policymakers, social connections may

also impact other aspects of migrants' well-being. Using responses to a short Facebook user survey, we document our fourth lesson: social integration has positive causal effects on migrants labor market, housing, and education outcomes. Finally, our fifth lesson is that natives exposed to a migrant in high school are more likely to befriend other migrants later in life, consistent with a long literature on the contact hypothesis (Allport, Clark and Pettigrew, 1954).

While our paper does not make concrete policy suggestions, several of our insights are relevant for policymakers. For example, the fact that regions differ in their ability to integrate migrants—combined with the fact that migrants do not seem to disproportionately move to regions with better integration outcomes—suggests that a better understanding of where migrants have the best chance of socially integrating could become an important consideration in policymakers' decisions of where to locate newly arriving migrants. Similarly, the positive causal effects of integration courses on equilibrium integration outcomes suggest that policymakers might consider expanding the use of such programs.

We hope that the increasing availability of data sources similar to the ones used in this paper—as well as other digital trace data discussed in Kuchler and Stroebel (2023)—will help researchers better understand the forces that shape social integration and help policymakers develop programs that effectively foster interconnected communities. For example, while our work has focused primarily on studying the relationship between Syrian migrants and native Germans, future work should additionally consider the determinants and effects of the relationships between migrants, including between migrants from different origin countries. In particular, it would be interesting to study the extent to which such relationships are complements or substitutes to friendships between migrants and natives. In addition, given the global scale of the Facebook platform, we are hoping to construct measures of the "relative friending" between native-migrant pairs that cover a much wider range of host and origin countries. This would allow researchers to better understand which migrants find it easier to integrate socially in what host communities.

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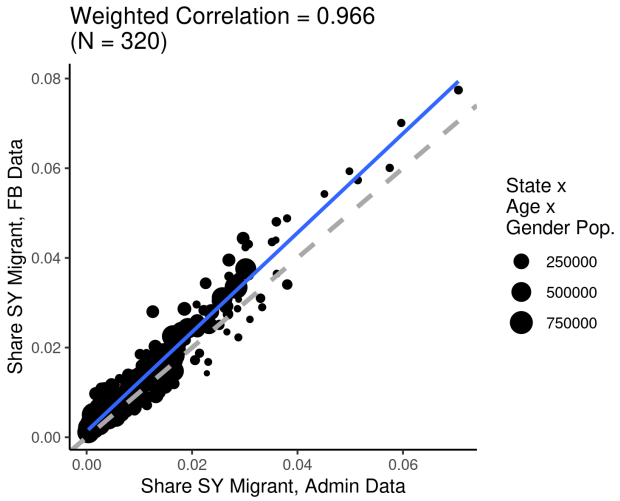
Appendices

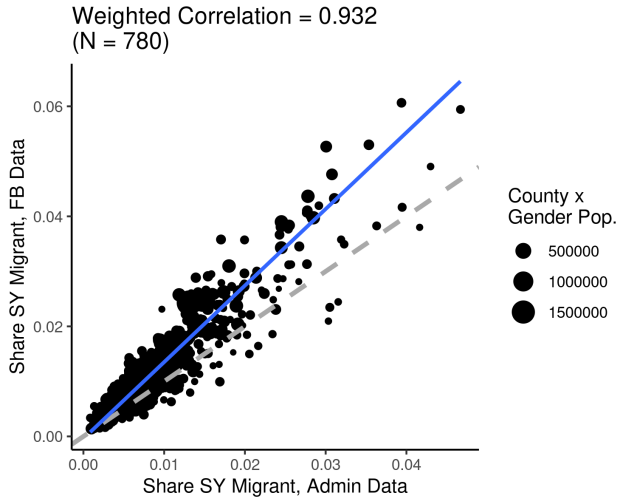
A Additional Figures and Tables

Figure A1: Syrian Migrant Sample vs. Admin Data

(a) By State × Age × Gender



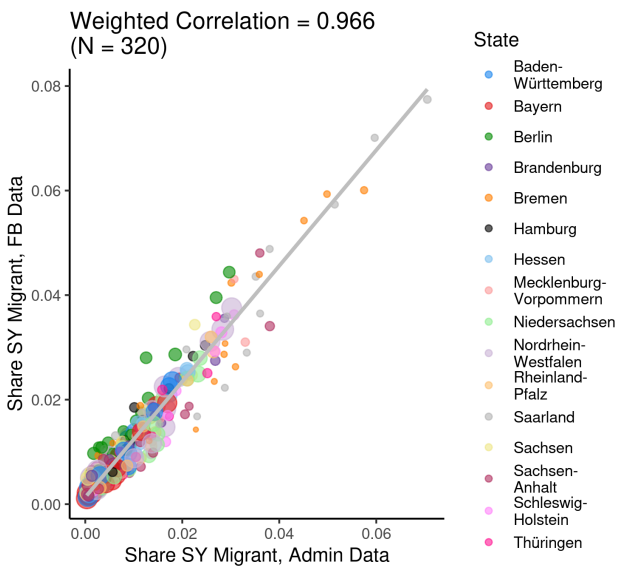
(b) By County × Gender



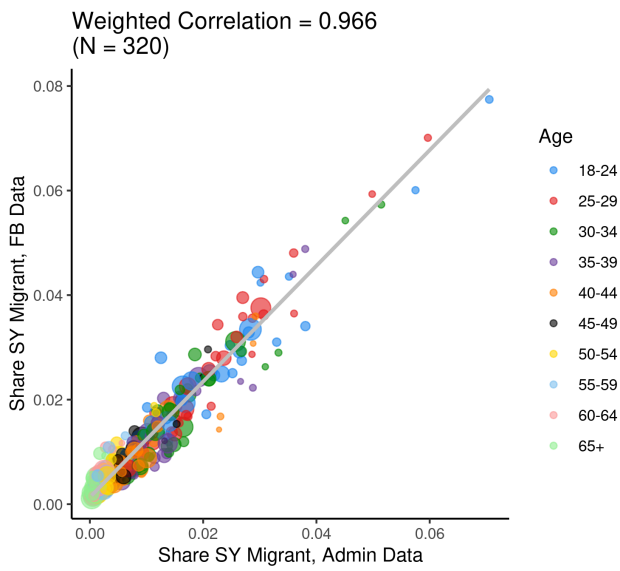
Note: Figures show the shares of the primary sample of Facebook users that are also in the Syrian migrant sample (on the y-axis), against shares of the population that are Syrian from administrative data (on the x-axis). The size of each dot is proportional to the true population it represents. The solid blue lines are from weighted linear regressions. The dashed grey line is the line $y = x$. Panel (a) plots these shares by state × age × gender. The age groups are 18-24, 25-29, 30-24, 35-39, 40-44, 45-49, 50-54, 55-50, 60-64, and 65+. There are 16 states X 10 age groups X 2 genders = 320 observations. Panel (b) plots these shares by county × gender. Admin data is unavailable for 11 counties. There are 390 counties X 2 genders = 780 observations.

Figure A2: Syrian Migrant Sample vs. Admin Data

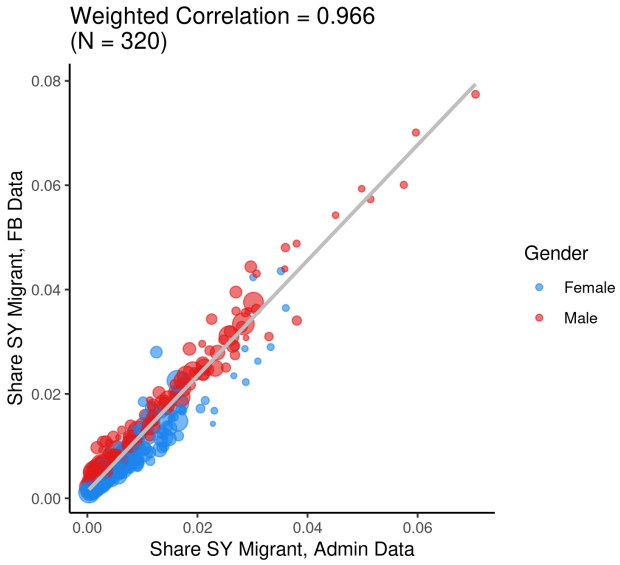
(a) By State × Age × Gender – Color by State



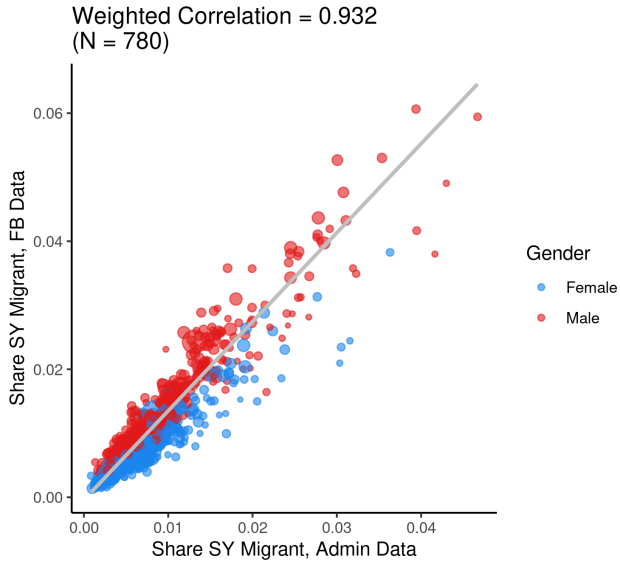
(b) By State × Age × Gender – Color by Age



(c) By State × Age × Gender – Color by Gender

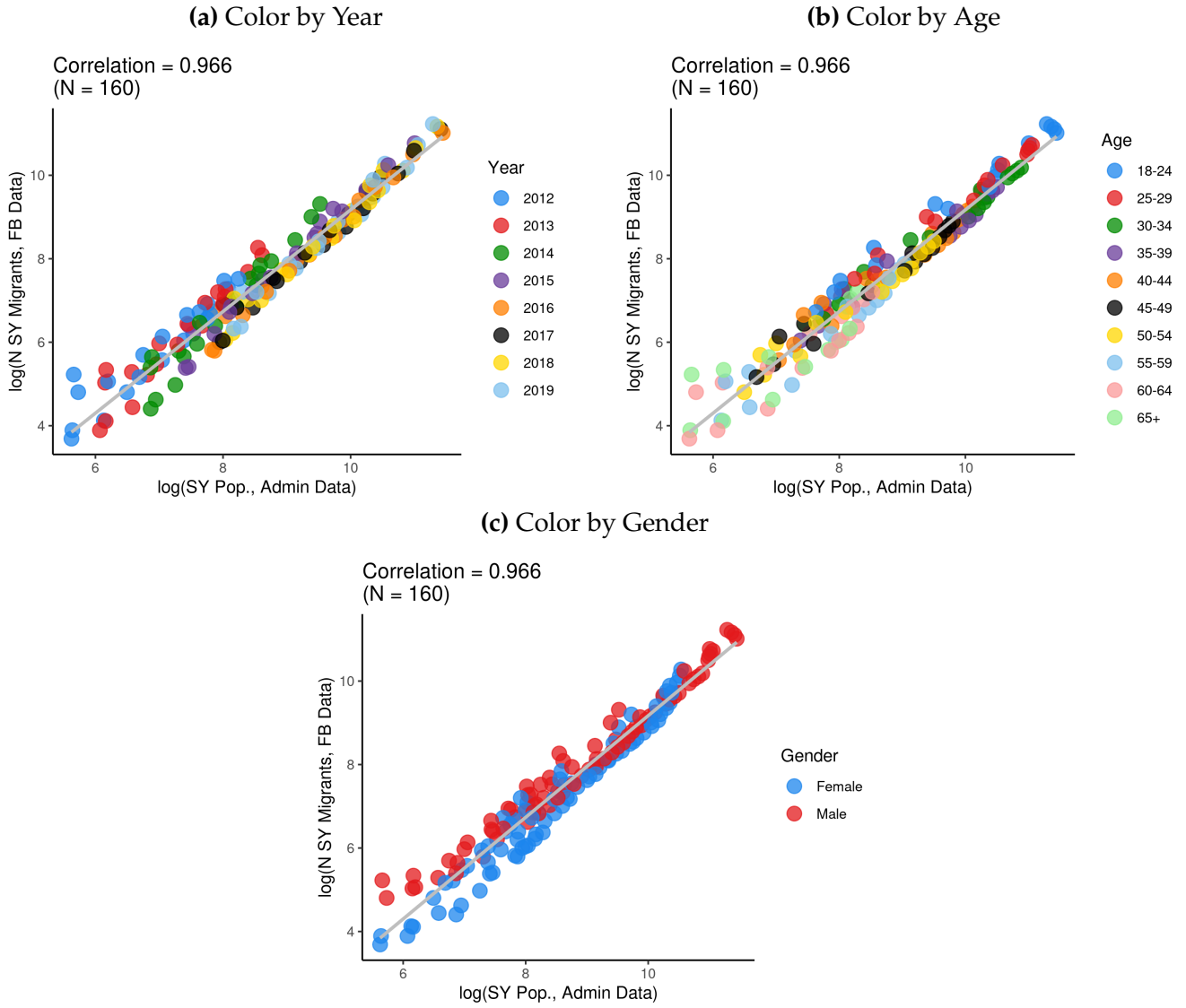


(d) By County × Gender – Color by Gender



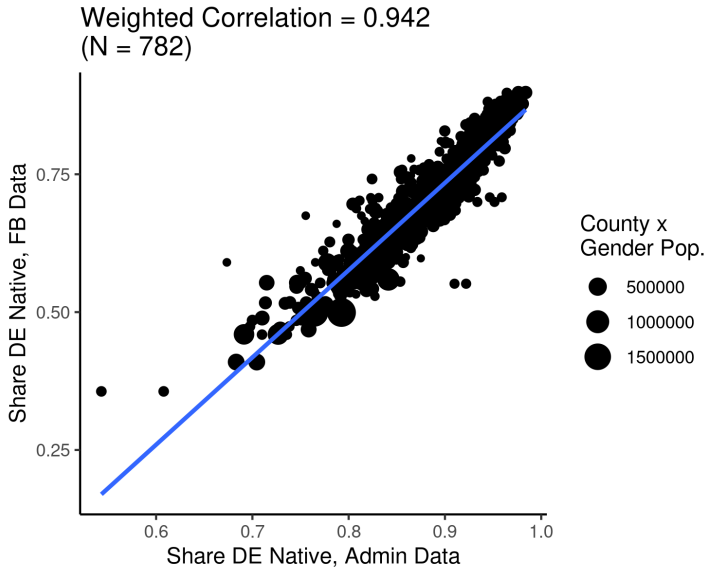
Note: Figures show the shares of the primary sample of Facebook users that are also in the Syrian migrant sample (on the y-axis), against shares of the population that are Syrian from administrative data (on the x-axis). The size of each dot is proportional to the size of the population it represents. The solid grey lines are from weighted linear regressions. Panels (a), (b), and (c) plot these shares by state, age, and gender. The age groups are 18-24, 25-29, 30-34, 35-39, 40-44, 45-49, 50-54, 55-50, 60-64, and 65+. There are 16 states × 10 age groups × 2 genders = 320 observations. Panel (d) plots these shares by county and gender. Administrative data is unavailable for 11 counties. There are 390 counties × 2 genders = 780 observations. Panel (a) colors observations by state; panel (b) colors by age; and panels (c) and (d) color by gender.

Figure A3: Syrian Migrant Sample vs. Admin Data – By Age × Gender × Year



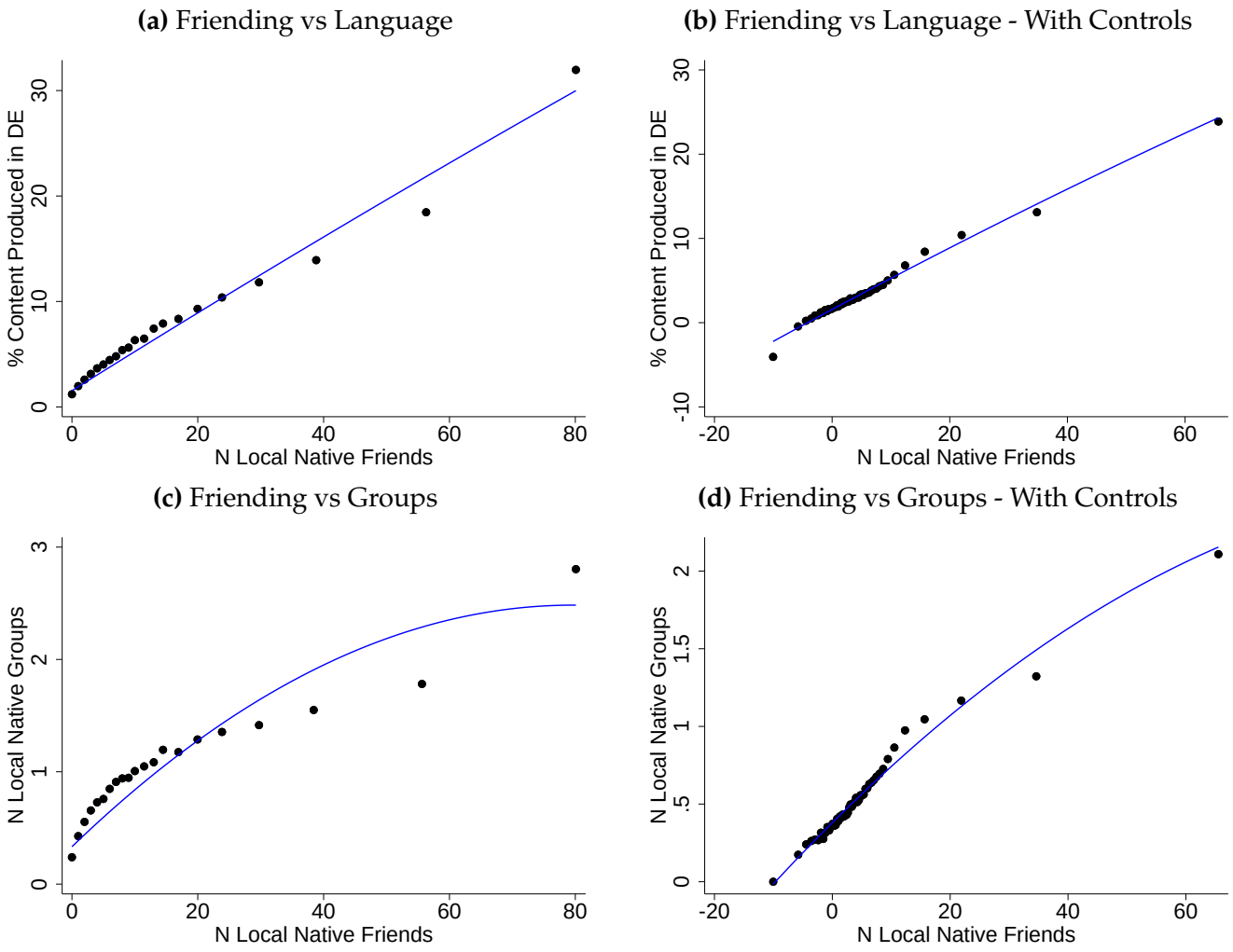
Note: Figure shows the number of users in our Syrian migrant sample using Facebook in Germany by the end of each year from 2012 to 2019 (on the y-axis), against analogous measures of Syrian migrant population from German administrative data (on the x-axis). Each observation is an age by gender by year group. The age groups are the same as those used in Figure A1. Both axes are transformed by the natural logarithm. The solid grey line is from a linear regression. Observations are colored by year in panel (a), age in panel (b), and gender in panel (c).

Figure A4: Native German Sample vs Admin Data



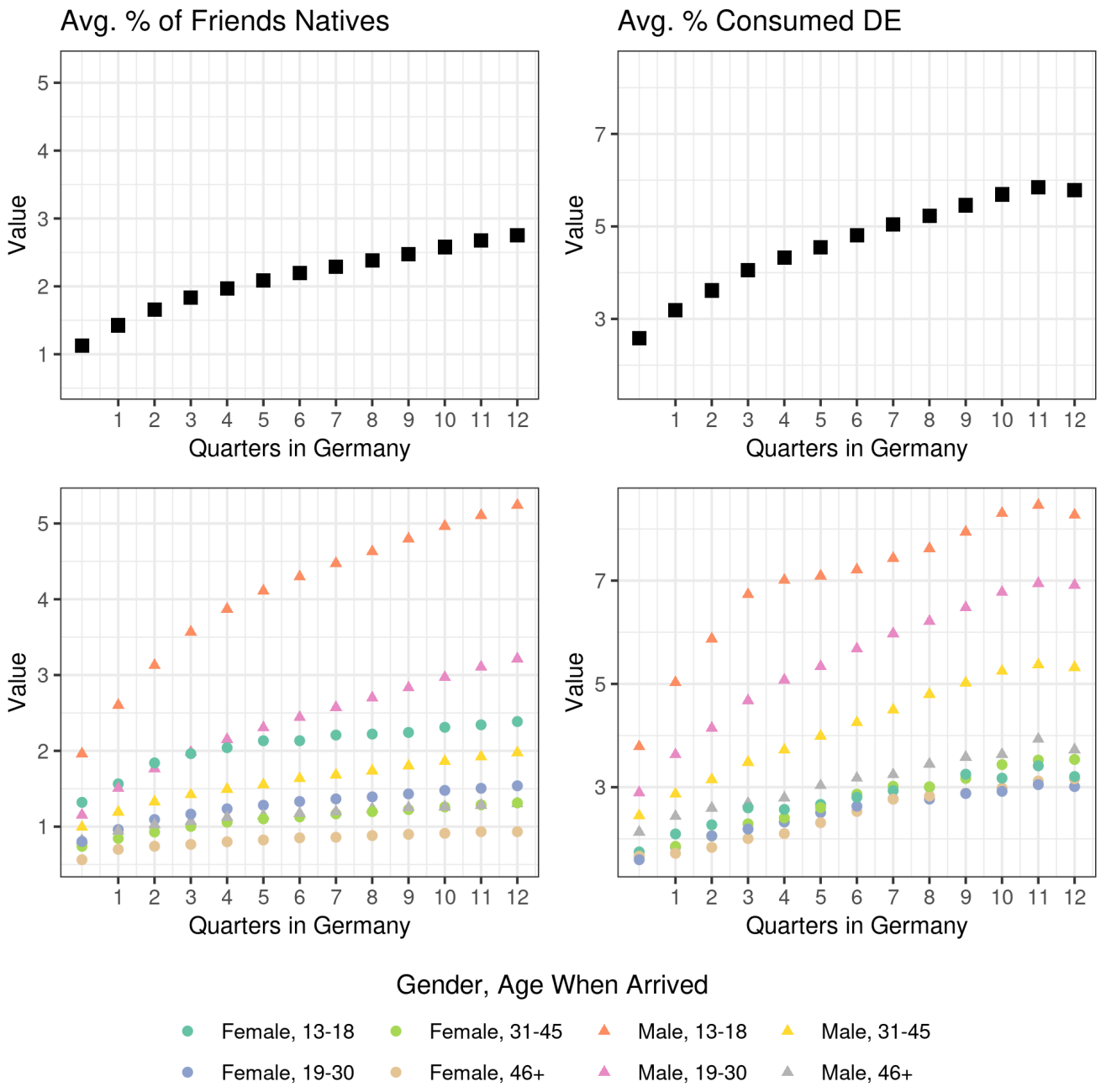
Note: Figure shows the shares of the primary sample of Facebook users that are also in the German native sample (on the y-axis), against shares of the population that are native from administrative data (on the x-axis). Each observation is a county by gender group. The size of each dot is proportional to the true population it represents. The solid blue lines are from weighted linear regressions. Admin data is unavailable for 10 counties. There are 391 counties $\times$ 2 genders = 782 observations.

Figure A5: Relationship Between Integration Outcomes, Individual Level



Note: Figures show binned scatter plots of individual Syrian migrants' number of local German native friends on the x-axis, against their share of content produced in German in panels (a) and (b), and the number of local native groups they are in panels (c) and (d). Appendix C provides more details on each measure. The measures in panels (b) and (d) are first residualized on the individual-level controls used in column 3 of Table A11. Lines are fit from quadratic regressions.

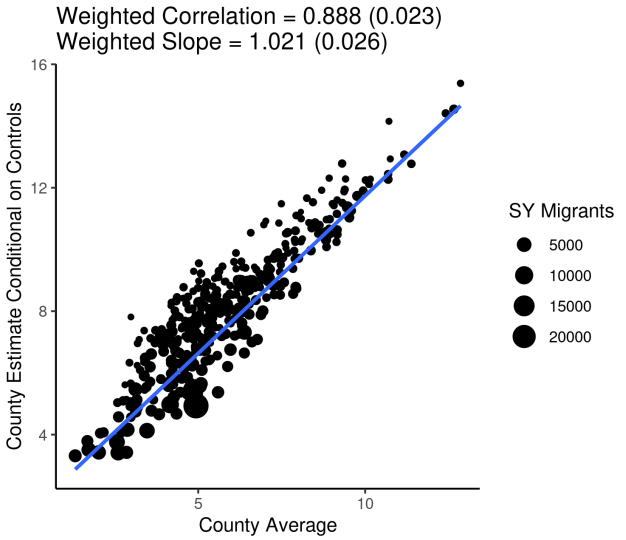
Figure A6: Integration Over Time For 2015-16 Cohort — Additional Measures



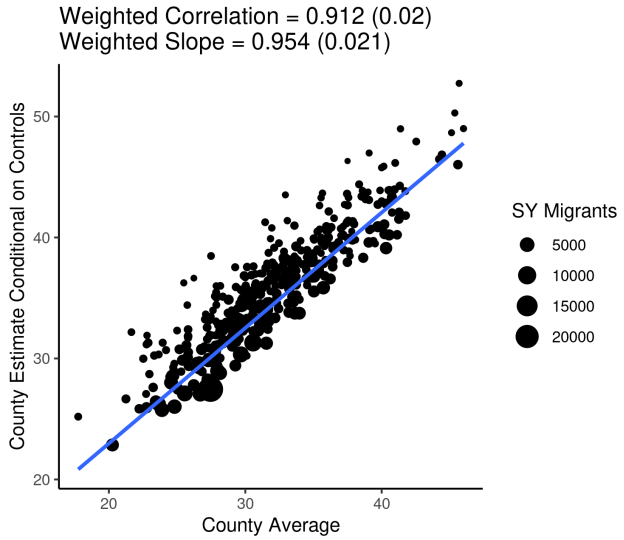
Note: Figures show the average values, by quarter, of integration measures for users in the Syrian migrant sample with an observed arrival in 2015 or 2016. The measures are share of friends native (left column) and the share of content consumed in German (right column). Appendix C provides more details on each measure. The top row shows overall trends. In the bottom row each observation's shape and color represents a gender-by-age group.

Figure A7: Regional Estimates With and Without Controls

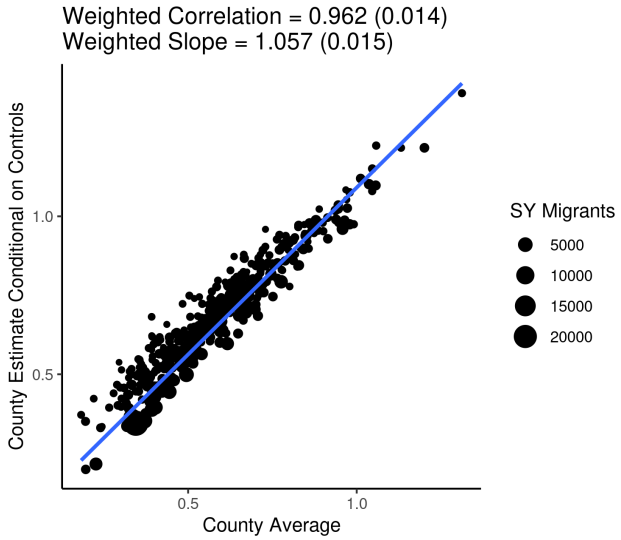
(a) Friending



(b) Language

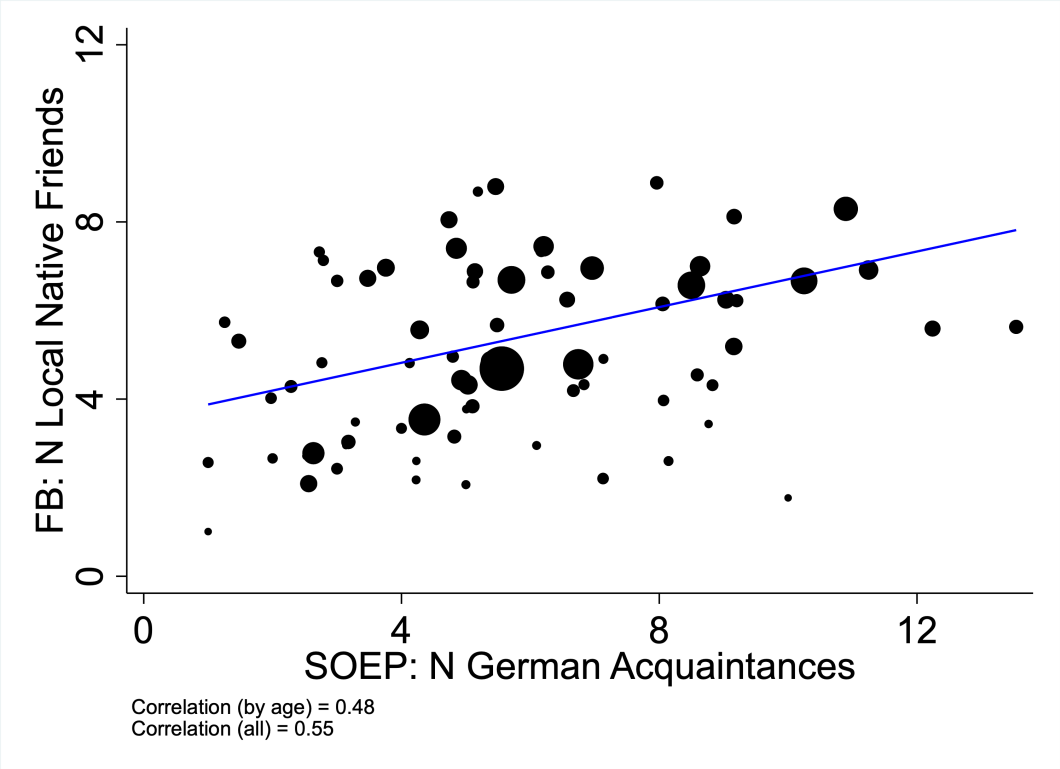


(c) Groups



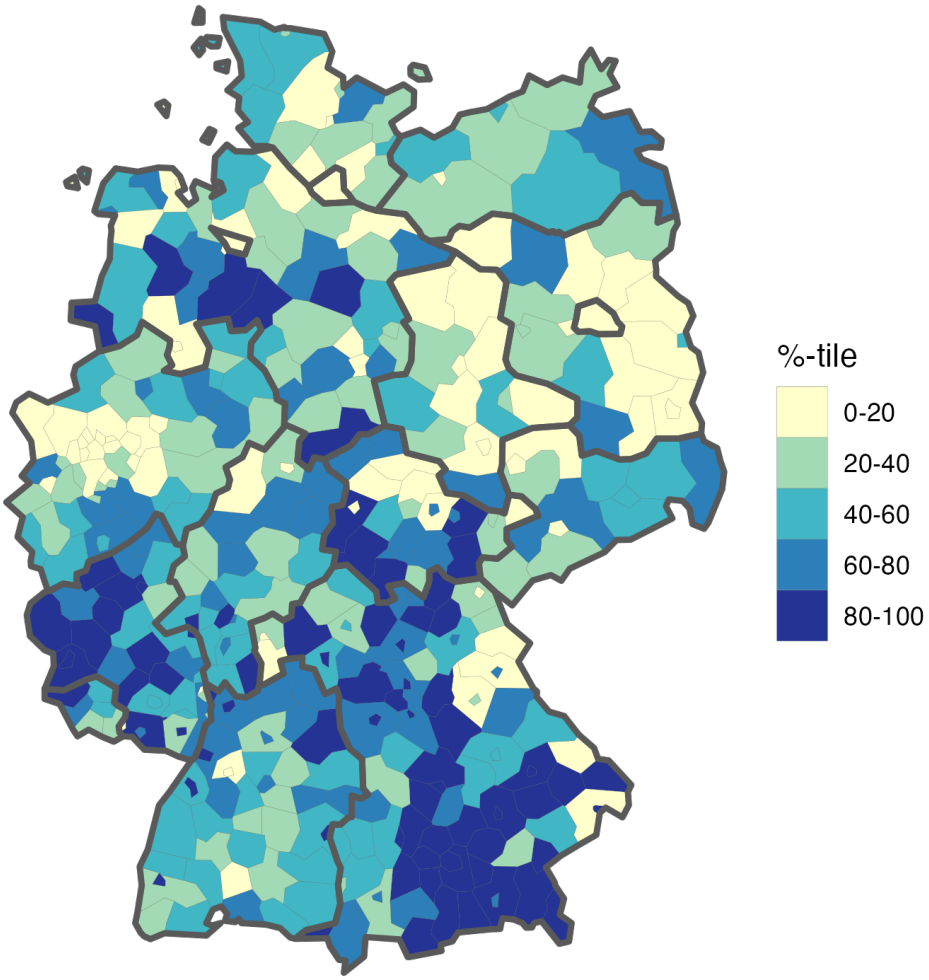
Note: Figures show the relationship between county averages of integration outcomes among Syrian migrants vs county-level fixed effect estimates constructed from versions of equation 5. The outcomes are a user's number of local German native friends in panel (a), whether the user produces content in German in panel (b), and the number of local native groups a user is in in panel (c). Appendix C provides more details on each measures. The controls in the fixed effect regressions are those used in column 3 of Table A11.

Figure A8: Comparing Regional Estimates of Integration - Facebook vs. SOEP



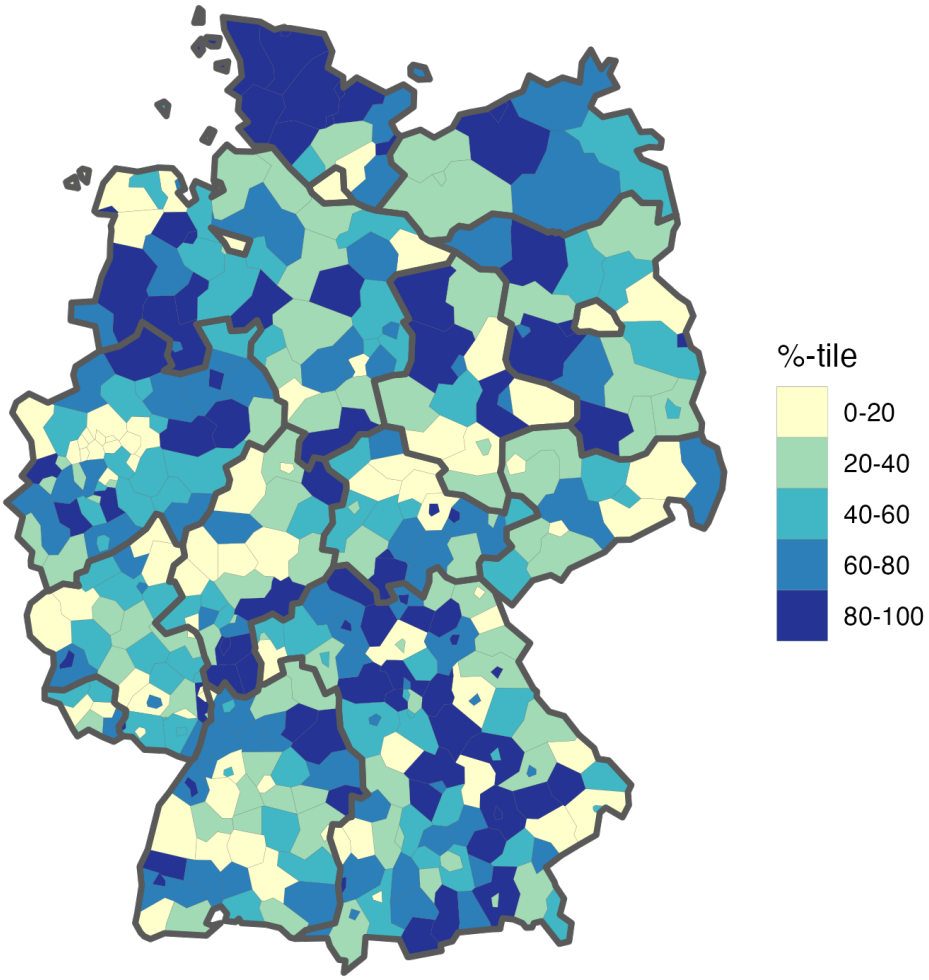
Note: Figure compares estimates of social integration based on our Facebook sample with the average number of acquaintances made by recent Syrian migrants in Germany in the SOEP data. The SOEP question is "How many German people have you met since your arrival in Germany with whom you have regular contact?". Each observation in the Figure is a state-by-age-group combination. The size of each dot corresponds to the number of Syrian migrants in the Facebook data. At the bottom of the figure, we report two correlations. The first is a correlation at the state by age-group level, i.e., the same level of aggregation as shown in the plot. The second is a correlation estimated at the state-level, i.e., we further aggregate observations to the state-level and then correlate the two data sources. Both correlations are weighted by the number of Syrian migrants in our Facebook sample.

Figure A9: Regional Estimates of Integration - German Language Usage



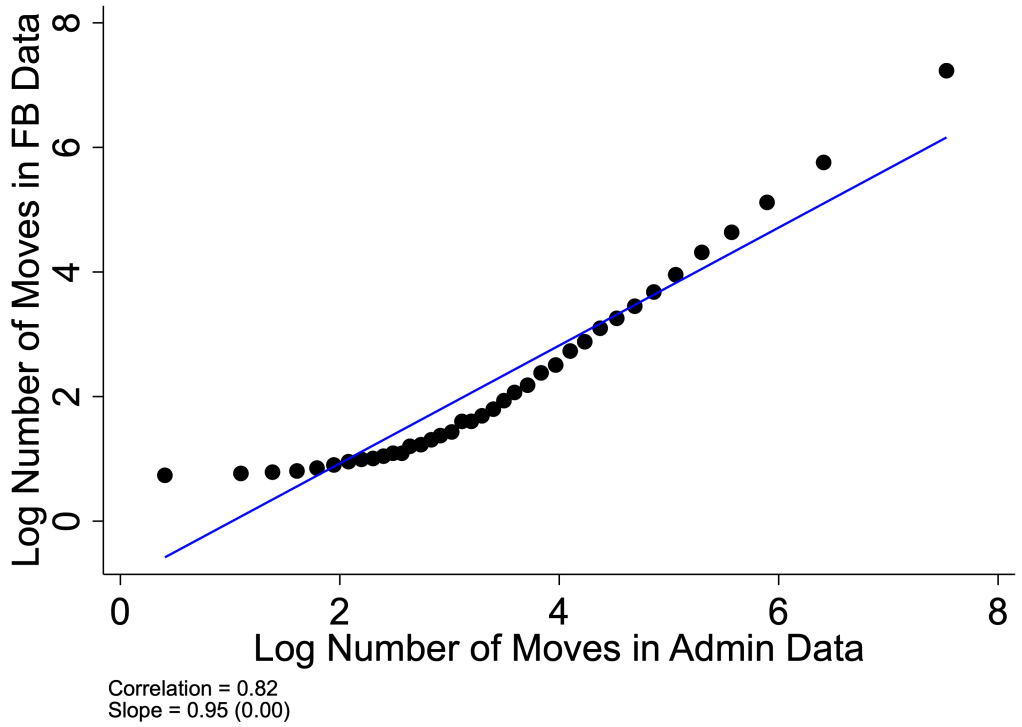
Note: Figure shows county-level estimates of Syrian migrant integration based on the share that produce content in the German language (residualized on regional patterns of Facebook usage). Darker areas indicate the highest integration counties.

Figure A10: Regional Estimates of Integration - Local Native Group Joining



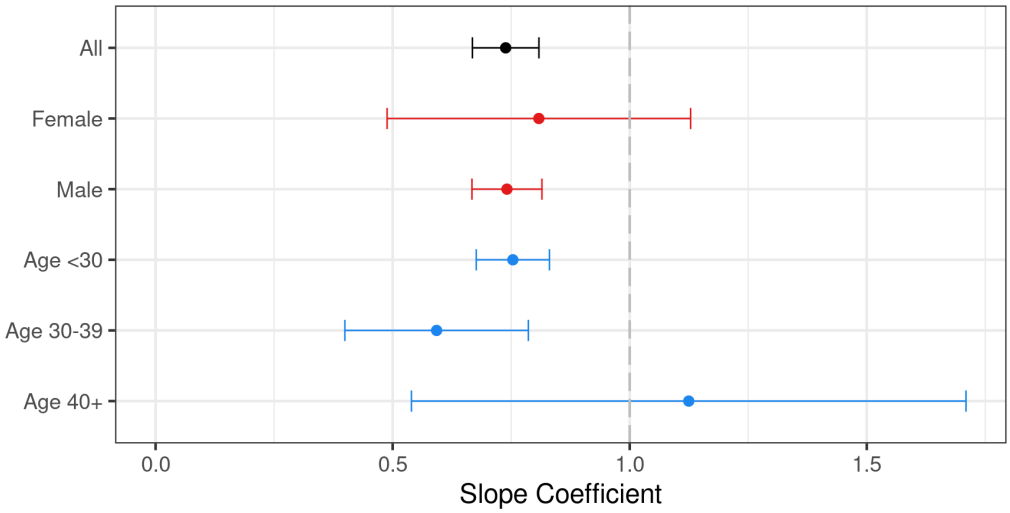
Note: Figure shows county-level estimates of Syrian migrant integration based on the average number of native local groups joined (residualized on regional patterns of Facebook usage. This includes the average number of total groups natives in the region have joined, allowing us to account for variation driven by differential usage of the groups feature in general). Colors correspond to measure ventiles. Darker areas indicate the highest integration counties.

Figure A11: Comparing Movers in Facebook and Administrative Data



Note: Figure compares the number of moves between counties made by all individuals (i.e., including natives, migrants, and others) between the ages of 18-64 in 2016 and 2017 in Facebook and administrative data. We obtained the administrative data from the German Statistical Office. Each observation in this analysis is a county-to-county combination. The Figure is a binned scatter plot with 40 equally sized bins. The Figure is weighted by the the total number of individuals living in origin and destination county.

Figure A12: Syrian Migrant Movers - Slope by Demographics

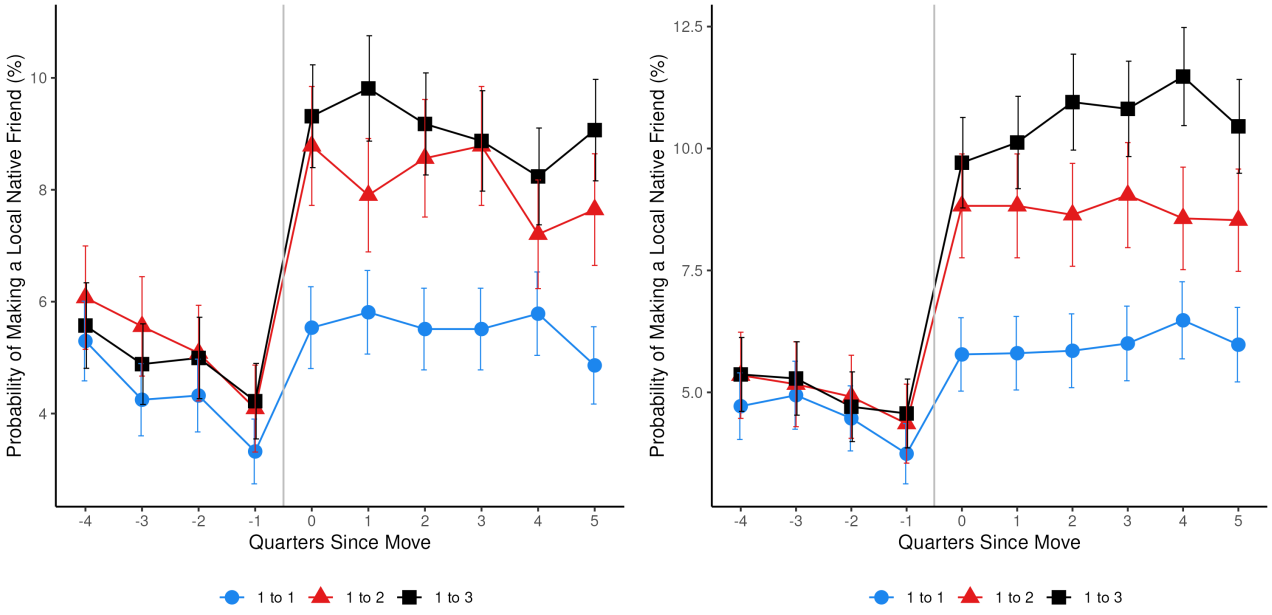


Note: Figure shows slopes corresponding to versions of Figure A19 over certain sub-samples. The coefficient in black corresponds to the slope using the full sample of Syrian migrant movers; the coefficients in red use samples of only one gender; and the coefficients in blue use samples of only one age group. Bars display 95% confidence intervals. The sample sizes used to generate each coefficient are (from top to bottom) 32,853, 6,144, 26,709, 20,796, 8,623, and 3,434.

Figure A13: Change in Syrian Migrants' Friending of Local Natives Around a Move—Split by Friendship Initiator

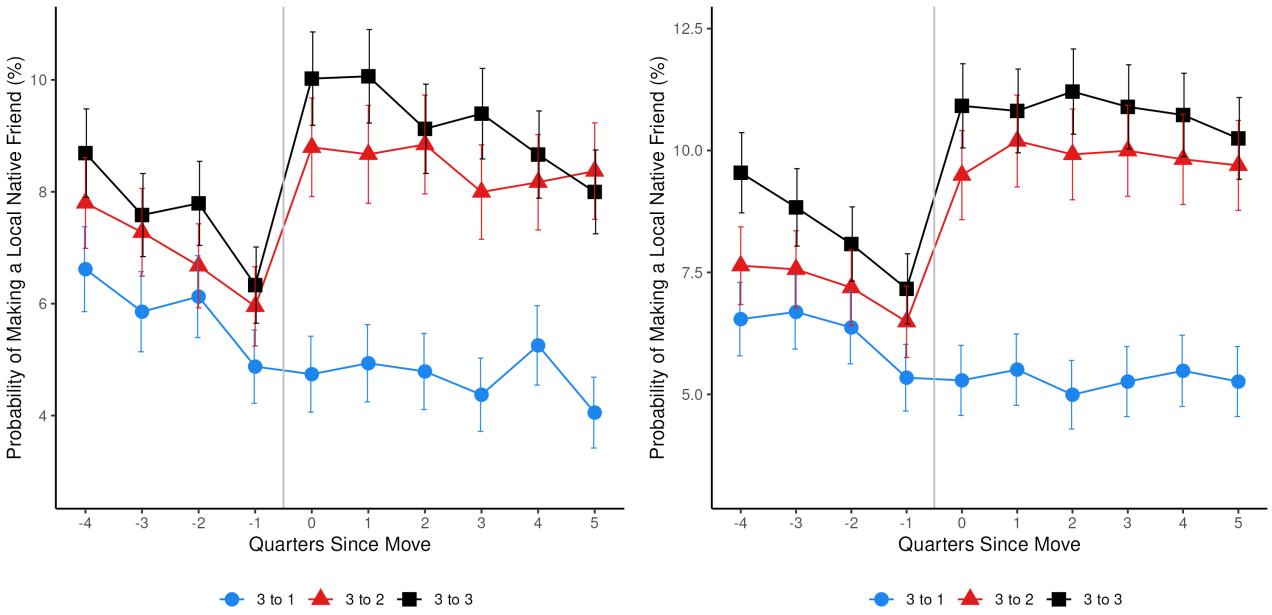
(a) Moving From Bottom Integration Tercile (Only Friendships Initiated by Syrian Migrants)

(b) Moving From Bottom Integration Tercile (Only Friendships Initiated by Native Germans)



(c) Moving From Top Integration Tercile (Only Friendships Initiated by Syrian Migrants)

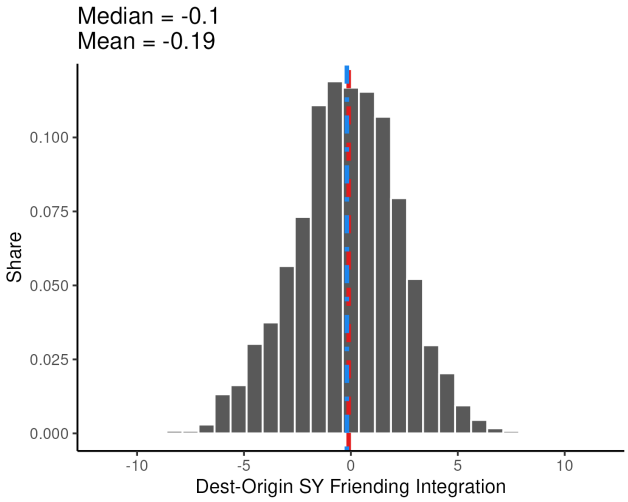
(d) Moving From Top Integration Tercile (Only Friendships Initiated by Native Germans)



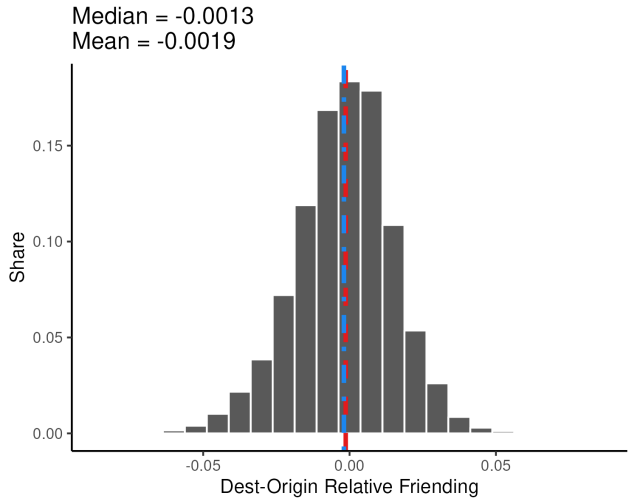
Note: This figure reproduces the analyses presented in Figure 3. Panels (a) and (b) disaggregate the results of panel (a) of Figure 3, splitting the friendships formed into two groups according to whether it was the Syrian migrant or the local German native who sent the friendship request on Facebook. Panels (c) and (d) repeat the same exercise for panel (b) of Figure 3.

Figure A14: Distribution of Syrian Migrant Moves

(a) Distribution of Δ Friending Integration



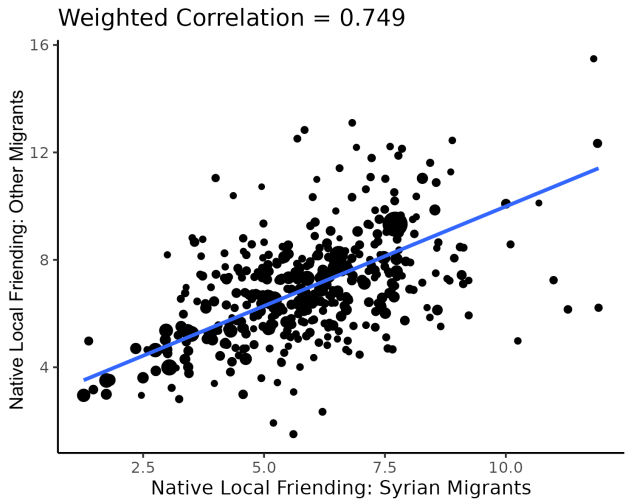
(b) Distribution of Δ Relative Friending



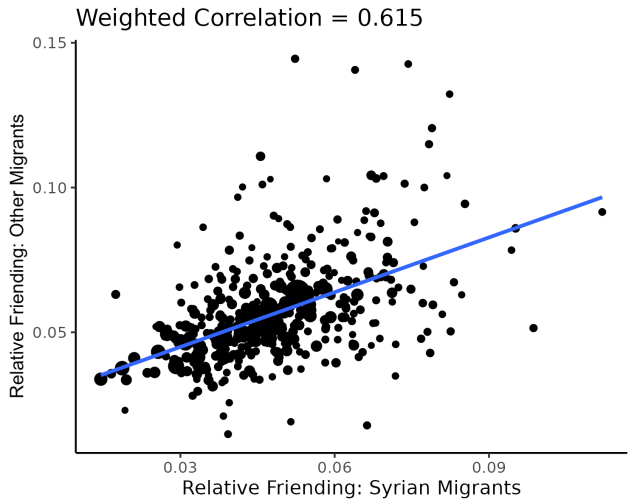
Note: Figures show, for Syrian migrant movers, the distribution of destination minus origin regional friending-based measures of Syrian migration integration. Panel (a) shows the distribution of the measure in Figure 2. Panel (b) shows the distribution of relative friending in Figure 4. The red and blue lines show the median and mean, respectively.

Figure A15: Social Integration Across Counties: Syrian Migrants vs Other Migrants

(a) Friending Integration



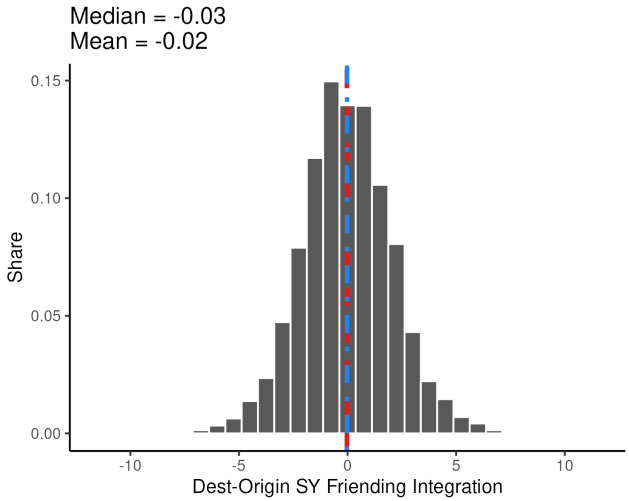
(b) Relative Friending



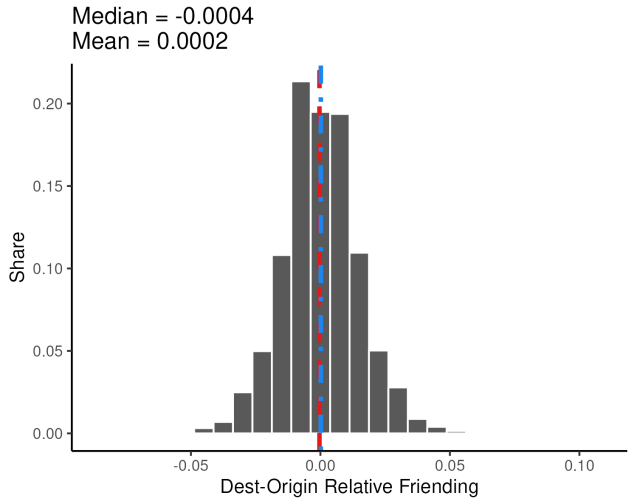
Note: Figure compares estimates of friending integration (panel a) and relative friending (panel b) across counties. Measures on the x-axis are calculated for Syrian migrants. Measures on the y-axis are calculated for users from one of the five countries with the most asylum applicants in Germany in 2020 other than Syria: Turkey, Afghanistan, Iraq, Nigeria, and Iran.

Figure A16: Distribution of German Native Moves

(a) Distribution of Δ Friending Integration



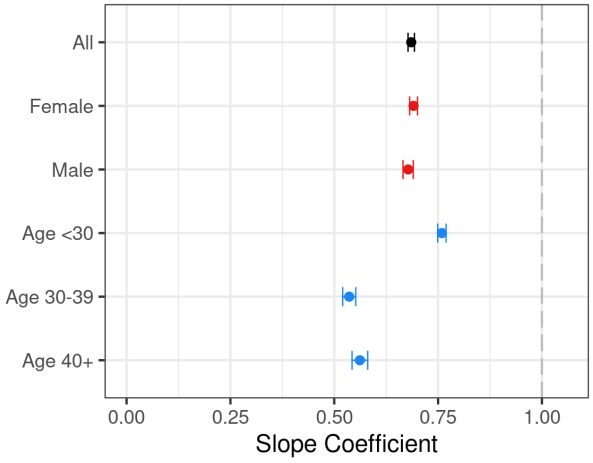
(b) Distribution of Δ Relative Friending



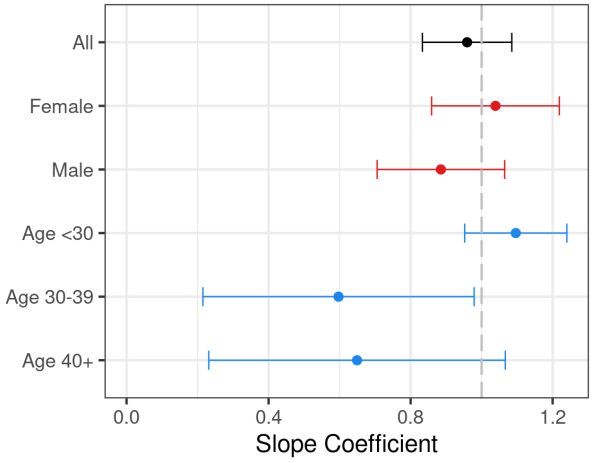
Note: Figures show, for German native movers, the distribution of destination minus origin regional friending-based measures of Syrian migration integration. Panel (a) shows the distribution of the measure in Figure 2. Panel (b) shows the distribution of relative friending in Figure 4. The red and blue lines show the median and mean, respectively.

Figure A17: Δ Native Mover Behaviors vs. Matched Non-Movers - Slope By Demographics

(a) General Friendliness

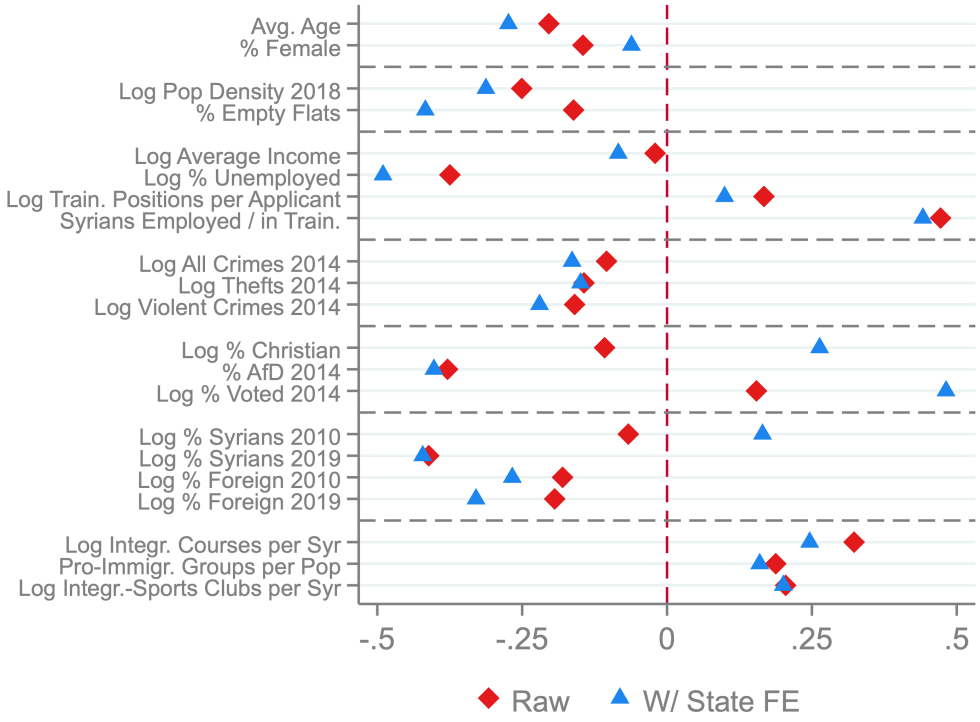


(b) Relative Friending



Note: Figures show slopes corresponding to versions of the respective panels in Figure 6. The coefficients in black are the slopes using the full sample of German native movers; the coefficients in red use samples of only one gender; and the coefficients in blue use samples of only one age group. Bars display 95% confidence intervals.

Figure A18: County-Level Univariate Correlations with Friending Integration - Long Version



Note: Figure presents correlations between our county-level measure of social integration and various other regional measures. Social integration is based on Syrian migrants number of native local friends (Figure 2). Correlations are weighted by the size of the Syrian migrant sample in each county. Red diamonds depict raw, univariate correlations and blue triangles depict correlations after controlling for state fixed effects. For more information on each measure, see Appendix Table A16.

Table A1: Syrian Migrant and German Native Sample Summaries - Additional Measures

Panel (a): Syrian Migrant Sample

	Mean	SD	P10	P25	P50	P75	P90	P99
N Native Friends	9.09	20.54	0	0	2	8	24	151
N Top 50 Native Friends	1.02	2.46	0	0	0	1	3	16
% of Friends Native	3.04	6.19	0.00	0.00	0.80	2.99	8.19	40.25
N Local Other Refugee Country Friends	2.04	3.63	0	0	1	2	6	21
N Local Recent Other Refugee Country Friends	1.04	1.87	0	0	0	1	3	10
% Content Produced in DE	3.39	9.89	0.00	0.00	0.00	2.31	8.48	70.00
% Content Consumed in DE	3.48	8.64	0.00	0.00	0.00	2.91	9.09	60.00
Consumes DE Content (0/100)	41.81	49.32	0	0	0	100	100	100
Account in DE	14.90	35.61	0	0	0	0	100	100
% Groups Local Native	0.88	3.55	0.00	0.00	0.00	0.00	2.22	15.38
Avg. % Native in DE Groups	31.09	30.21	0.15	0.52	25.06	56.44	77.84	92.91

Panel (b): German Native Sample

	Mean	SD	P10	P25	P50	P75	P90	P99
N Native Friends	204.73	189.58	40	74	148	269	443	1151
N Top 50 Native Friends	36.87	8.76	25	33	39	43	46	49
% of Friends Native	82.09	14.70	63.75	77.84	86.67	91.61	94.52	98.16
N Local Other Refugee Country Friends	1.12	2.58	0	0	0	1	3	17
N Local Recent Other Refugee Country Friends	0.05	0.22	0	0	0	0	0	1
% Content Produced in DE	94.49	9.70	81.19	92.90	100.00	100.00	100.00	100.00
% Content Consumed in DE	88.60	16.55	65.84	84.06	95.90	100.00	100.00	100.00
Consumes DE Content (0/100)	97.69	15.02	100	100	100	100	100	100
Account in DE	98.61	11.69	100	100	100	100	100	100
% Groups Local Native	22.07	22.34	0.00	4.55	16.67	33.33	50.00	100.00
Avg. % Native in DE Groups	90.42	5.88	83.52	88.16	91.70	94.15	95.95	100.00

Note: Table presents summary statistics describing users in our Facebook samples. Panel (a) shows users in the Syrian migrant sample. Panel (b) shows users in the German native sample. Each measure is winsorized at the 99% level. Section 1.1 describes sample construction. Appendix C provides more information on how individual-level outcomes are defined.

Table A2: Correlation Between Integration Outcomes, Individual Level

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	(13)	(14)	(15)
(1) N Local Native Friends	1.00														
(2) N Native Friends	0.64	1.00													
(3) N Top 50 Native Friends	0.61	0.54	1.00												
(4) % of Friends Native	0.69	0.61	0.88	1.00											
(5) N Local SY Friends	0.29	0.14	0.01	0.02	1.00										
(6) N Local Other Refugee Country Friends	0.47	0.30	0.19	0.23	0.54	1.00									
(7) N Local Recent Other Refugee Country Friends	0.28	0.15	0.06	0.07	0.53	0.85	1.00								
(8) % Content Produced in DE	0.45	0.40	0.65	0.67	-0.02	0.17	0.03	1.00							
(9) % Content Consumed in DE	0.46	0.40	0.67	0.68	-0.01	0.18	0.05	0.80	1.00						
(10) Produces DE Content	0.24	0.19	0.27	0.31	0.04	0.11	0.03	0.33	0.33	1.00					
(11) Consumes DE Content	0.37	0.25	0.37	0.40	0.09	0.19	0.12	0.52	0.45	0.27	1.00				
(12) Account in DE	0.32	0.21	0.32	0.34	0.11	0.19	0.13	0.34	0.47	0.25	0.57	1.00			
(13) N Local Native Groups	0.29	0.25	0.25	0.27	0.12	0.14	0.09	0.23	0.25	0.14	0.26	0.24	1.00		
(14) % Groups Local Native	0.33	0.26	0.37	0.40	0.03	0.13	0.05	0.36	0.37	0.19	0.27	0.24	0.61	1.00	
(15) Avg. % Native in DE Groups	0.32	0.23	0.32	0.36	0.01	0.14	0.08	0.33	0.35	0.26	0.38	0.36	0.43	0.47	1.00

Note: Table presents correlations at the user level across outcome measures for the Syrian migrant sample. Each measure is winsorized at the 99% level. Appendix C provides more information on how outcomes are defined.

Table A3: Correlation Between Integration Outcomes, Individual Level - With Controls

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	(13)	(14)	(15)
(1) N Local Native Friends	1.00														
(2) N Native Friends	0.61	1.00													
(3) N Top 50 Native Friends	0.60	0.54	1.00												
(4) % of Friends Native	0.69	0.62	0.86	1.00											
(5) N Local SY Friends	0.20	0.07	0.02	0.03	1.00										
(6) N Local Other Refugee Country Friends	0.39	0.24	0.17	0.20	0.46	1.00									
(7) N Local Recent Other Refugee Country Friends	0.20	0.09	0.05	0.06	0.45	0.83	1.00								
(8) % Content Produced in DE	0.43	0.38	0.61	0.63	-0.01	0.15	0.03	1.00							
(9) % Content Consumed in DE	0.44	0.39	0.63	0.63	0.00	0.16	0.04	0.77	1.00						
(10) Produces DE Content	0.19	0.15	0.21	0.24	0.02	0.06	-0.00	0.27	0.27	1.00					
(11) Consumes DE Content	0.31	0.21	0.32	0.33	0.02	0.12	0.05	0.48	0.40	0.21	1.00				
(12) Account in DE	0.25	0.17	0.26	0.27	0.04	0.12	0.07	0.28	0.42	0.19	0.51	1.00			
(13) N Local Native Groups	0.25	0.22	0.24	0.26	0.03	0.08	0.03	0.23	0.25	0.13	0.22	0.18	1.00		
(14) % Groups Local Native	0.28	0.23	0.33	0.36	0.02	0.09	0.02	0.31	0.32	0.14	0.22	0.18	0.63	1.00	
(15) Avg. % Native in DE Groups	0.23	0.17	0.26	0.29	-0.04	0.05	-0.00	0.27	0.29	0.19	0.30	0.27	0.43	0.42	1.00

Note: Table presents correlations at the user level across outcome measures for the Syrian migrant sample. Each measure is first winsorized at the 99% level. Appendix C provides more information on how outcomes are defined. Before constructing the correlations, each measure is residualized on the individual-level controls used in column 3 of Table A11.

Table A4: Syrian Migrant Integration by Demographics - Language and Groups

	Produces Content in German (0/100)				N Local Native Groups			
Age 25 - 34	-2.407***	-2.241***	-2.275***	-3.312***	0.167***	0.171***	0.136***	0.140***
	(0.204)	(0.203)	(0.203)	(0.596)	(0.006)	(0.006)	(0.006)	(0.019)
Age 35 - 44	-7.133***	-7.161***	-6.875***	-6.615***	-0.002***	-0.007***	0.039*	0.072**
	(0.238)	(0.237)	(0.237)	(0.733)	(0.007)	(0.007)	(0.007)	(0.023)
Age 45 - 54	-13.651***	-13.798***	-12.553***	-16.243***	-0.184***	-0.189***	-0.064***	-0.070***
	(0.306)	(0.305)	(0.307)	(0.854)	(0.010)	(0.010)	(0.009)	(0.027)
Age 55+	-18.045***	-18.134***	-16.451***	-24.395***	-0.298***	-0.300***	-0.088***	-0.228***
	(0.382)	(0.380)	(0.384)	(1.116)	(0.012)	(0.012)	(0.012)	(0.035)
Female	-15.767***	-15.560***	-16.725***	-18.765***	-0.202***	-0.200***	-0.372***	-0.447***
	(0.164)	(0.164)	(0.173)	(0.418)	(0.005)	(0.005)	(0.005)	(0.013)
Household Member in DE 1+ Year Prior	-2.420***	-2.298***	-2.113***		-0.057***	-0.058***	-0.060***	
	(0.384)	(0.383)	(0.382)		(0.012)	(0.012)	(0.012)	
Non-Household Family in DE 1+ Year Prior	3.418***	3.451***	4.045***		0.023***	0.025***	0.030***	
	(0.347)	(0.345)	(0.345)		(0.011)	(0.011)	(0.010)	
Quarters Since DE FEs	X	X	X	X	X	X	X	X
Prev Quarters in NUTS3 FEs	X	X	X	X	X	X	X	X
Personal Usage Controls	X	X	X	X	X	X	X	X
County FEs		X	X	X		X	X	X
Log (1 + Total Outside Germany Friends)			X	X			X	X
Log (1 + Total Other Groups)			X	X			X	X
Log (1 + Total Content Produced Past Year)			X	X			X	X
Household FE				X				X
N	349,072	349,072	349,072	84,216	349,072	349,072	349,072	84,216
R-Squared	0.098	0.108	0.113	0.590	0.059	0.076	0.133	0.606
Sample Mean	30.401	30.401	30.401	27.215	0.545	0.545	0.545	0.574

Note: Table shows results from regressing various measures on language- and groups-based measures of integration. Each observation in every column is a user in the Syrian migrant Facebook sample. Columns 1 and 5 include controls for age and gender, as well as fixed effects for the number of quarters on Facebook in their current county and the number of quarters since arrival in Germany. For the latter fixed effect, we use a single dummy value for those for which we do not observe arrival, but obtain nearly identical results if we instead drop these users. We also include dummies for whether the user has another Syrian migrant household member or non-household family member in Germany more than year prior to their arrival. For all users not in the "observe arrival timing" sample, these two dummies are set to 0. Columns 2 and 6 add county fixed effects. Columns 3 and 7 add controls for each user's total number of friends outside Germany, total number of non-local/native groups joined, and total amount of content produced in the last year. Columns 4 and 8 add a household fixed effect, limiting to households for which we observe more than one Syrian migrant. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01).

Table A5: Syrian Migrant Integration by Demographics - Other Measures

	N Native Friends	N Top 50 Native Friends	% of Friends Native	% Content Produced in DE	% Content Consumed in DE	Account in DE	% Groups Local Native	Avg. % Native in DE Groups
Age 25 - 34	-0.894***	0.004***	-0.467***	0.076**	0.078***	-2.683***	0.197***	-0.136***
	(0.184)	(0.014)	(0.032)	(0.044)	(0.038)	(0.160)	(0.010)	(0.160)
Age 35 - 44	-4.728***	-0.263***	-1.446***	-0.694***	-0.749***	-7.099***	0.043	-4.347***
	(0.216)	(0.016)	(0.038)	(0.051)	(0.044)	(0.187)	(0.012)	(0.187)
Age 45 - 54	-6.928***	-0.454***	-1.927***	-1.245***	-1.298***	-7.676***	-0.164***	-6.940***
	(0.279)	(0.021)	(0.049)	(0.066)	(0.057)	(0.241)	(0.015)	(0.254)
Age 55+	-8.157***	-0.421***	-1.862***	-1.221***	-1.327***	-6.151***	-0.350***	-7.334***
	(0.349)	(0.026)	(0.061)	(0.083)	(0.072)	(0.302)	(0.019)	(0.360)
Female	-7.188***	-0.787***	-2.334***	-2.339***	-2.154***	-5.377***	-0.485***	-11.601***
	(0.157)	(0.012)	(0.027)	(0.037)	(0.032)	(0.136)	(0.009)	(0.137)
Household Member in DE 1+ Year Prior	-0.610	-0.030	0.013	0.146	-0.057	0.182	-0.014	-0.875***
	(0.347)	(0.026)	(0.061)	(0.082)	(0.071)	(0.300)	(0.019)	(0.295)
Non-Household Family in DE 1+ Year Prior	0.667***	0.075***	0.360***	0.535***	0.404***	3.659***	0.098***	2.649***
	(0.314)	(0.023)	(0.055)	(0.074)	(0.064)	(0.271)	(0.017)	(0.257)
Quarters Since DE FEs	X	X	X	X	X	X	X	X
Prev Quarters in County FEs	X	X	X	X	X	X	X	X
Personal Usage Controls	X	X	X	X	X	X	X	X
County FEs	X	X	X	X	X	X	X	X
Log (1 + Total Outside Germany Friends)	X	X	X	X	X	X	X	X
Log (1 + Total Other Groups)	X	X	X	X	X	X	X	X
Log (1 + Total Content Produced Past Year)	X	X	X	X	X	X	X	X
N	349,072	349,072	349,072	345,814	346,367	349,072	345,162	237,563
R-Squared	0.064	0.111	0.163	0.121	0.125	0.083	0.077	0.171
Sample Mean	10.592	1.101	3.221	3.388	3.474	14.896	0.754	31.091

Note: Table shows results from regressing various measures on outcomes for Syrian migrants in the Facebook sample. All columns include controls for age, gender, time spent on Facebook, number of friends outside Germany, total number of non-local/native groups joined, and total amount of content produced in the last year. They include fixed effects for county, the number of quarters since arrival in Germany (with a single dummy for those for which we do not observe arrival) and the number of quarters on Facebook in their current county. They also include dummies for whether the user has another Syrian migrant household member or non-household family member in Germany more than year prior to their arrival. Column 8 limits to migrants who are members of at least one group of majority users in Germany. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01).

Table A6: Signal Correlation Between Outcomes, Regional Level

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Panel A: Baseline Integration Measures							
(1) SY Migrants - N Local Native Friends	X						
(2) SY Migrants - Produced Content in DE	0.65	X					
(3) SY Migrants - N Local Native Groups	0.27	0.55	X				
(4) SY Migrants - N Local SY Friends	-0.04	-0.55	-0.42	X			
Panel B: Decomposition of Integration Measures							
(5) General Friendliness	0.64	0.31	-0.04	0.11	X		
(6) Relative Friending	0.77	0.56	0.43	-0.16	-0.05	X	
Panel C: Labor Market Integration Measure							
(7) Share Syrians in Employment or Training	0.46	0.63	0.14	-0.36	0.29	0.34	X

Note: Table presents signal-adjusted correlations between county-level estimates. The outcomes in panel (a) are the regional averages of Syrian migrants after residualizing on local German natives' Facebook usage, as described in Section 2. The outcomes in panel (b) are the regional decomposition measures described in Section 3.1. Row 5 is general friendliness, generated as a regional average of German natives after residualizing on local German natives' Facebook usage. Row 6 is relative friending, generated as the quotient from dividing the measure in row 1 by the measure in row 5. The outcome in panel C is an external county-level measure of the share of all Syrians that are employed or in training programs as described in Section 4.2. Correlations are weighted by the number of Syrian migrant users in each county. Our methodology for adjusting correlations to remove sampling error is described in Appendix E.

Table A7: Syrian Migrant Mover and Comparable Non-Mover Sample Summaries

	All		To Below Median Place		To Above Median Place	
	Movers	Matched	Movers	Matched	Movers	Matched
% Female	18.70	18.70	19.54	19.54	17.95	17.95
Avg Age	27.97	27.49	27.98	27.51	27.97	27.47
Avg Qs in DE	6.47	6.42	6.54	6.50	6.40	6.36
Avg Friends Made (total in year)	44.72	43.97	44.78	44.07	44.66	43.87
% of Qs Produ in DE	45.77	45.01	44.31	44.01	47.09	45.90
% of Qs Makes Native Local Friend	11.80	17.18	10.51	16.72	12.96	17.60

Note: Table presents summary statistics describing the movers underlying Figure A19 and their matched non-movers in their origin. Movers are matched to non-movers on county, time, age group (18-29, 30-39, 40+), gender, and the year we first observed the user on Facebook in Germany. To be in the final sample, a mover must be matched to five or more non-movers in both the origin and destination. Measures are constructed using the movers' information in the year prior to the move and their matched users in the origin location and time. Matched non-mover summaries are generated by first constructing measures within each mover's set of matched movers, then averaging across these measures. "Avg Friends Made" is constructed from summing quarterly measures that are winsorized at the 99% level across all migrant user-by-quarter observations. "% of Qs Makes Native Local Friend" is residualized by local natives' Facebook usage.

Table A8: Δ Migrant Mover Friending Integration vs. Matched Non-Movers: Robustness

	Change Quarterly Prob of Making Native Local Friend			
Dest-Origin Quarterly Prob of SY Making Native Local Friend	0.738***		0.758***	0.724***
	(0.036)		(0.051)	(0.053)
Origin Quarterly Prob of SY Making Native Local Friend		-0.712***		
		(0.037)		
Dest Quarterly Prob of SY Making Native Local Friend		0.773***		
		(0.037)		
Quarter FEs	X	X	X	X
Origin County FEs			X	
Dest County FEs				X
N	32,853	32,853	32,849	32,845
Sample Mean	0.934	0.934	0.933	0.938

Note: Table shows results from regressions exploring the change in friending of Syrian migrants to German natives, before and after a move within Germany. Column 1 corresponds to the relationship depicted in Figure A19. Column 2 regresses each component of the difference in the right-hand side measure in column 1 separately on the outcome. Columns 3 and 4 repeat column 1 with origin and destination fixed effects, respectively. We correct for sampling error in the right-hand side measures by randomly splitting the individual-level non-mover data into two halves and instrumenting for one set of averages with the other. See Appendix E for more information this procedure. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01).

Table A9: Native Mover and Comparable Non-Mover Sample Summaries

Panel A: Yearly General Friendliness Sample

	All		To Below Median Place		To Above Median Place	
	Movers	Matched	Movers	Matched	Movers	Matched
% Female	51.95	51.95	51.74	51.74	52.07	52.07
Avg Age	33.70	33.34	34.21	33.87	33.39	33.03
Avg Friends Made (total in year)	21.22	20.11	19.71	19.68	22.12	20.36
Yearly General Friendliness	5.33	9.74	4.81	9.49	5.63	9.89

Panel B: Yearly Relative Friending Sample

	All		To Below Median Place		To Above Median Place	
	Movers	Matched	Movers	Matched	Movers	Matched
% Female	52.75	52.75	52.48	52.48	52.90	52.90
Avg Age	31.90	31.86	32.35	32.35	31.65	31.58
Avg Friends Made (total in year)	28.19	20.70	26.41	20.20	29.21	20.99
Yearly Relative Friending	0.20	0.23	0.17	0.22	0.21	0.23

Note: Table presents summary statistics describing the users underlying Figure 6. Panels (a) and (b) show summaries for movers and matched non-movers in panels (a) and (b) of Figure 6, respectively. Measures are constructed using movers' information in the year prior to the move and their matched users in the origin location and time. Matched non-mover summaries are generated by first constructing measures within each mover's set of matched movers, then averaging across these measures. "Avg Friends Made" is constructed from summing quarterly measures winsorized at the 99% level across all native user-by-quarter observations. The final outcome in each panel is residualized by local natives' Facebook usage.

Table A10: Change in Native Mover SY Migrant Friending vs Matched Non-Movers

	Change in Mover Yearly General Friendliness				Change in Mover Yearly Relative Friending			
Dest-Origin Yearly General Friendliness	0.685***		0.711***	0.602***				
	(0.004)		(0.005)	(0.005)				
Origin Yearly General Friendliness		-0.636***						
		(0.005)						
Dest Yearly General Friendliness		0.739***						
		(0.005)						
Dest-Origin Yearly Relative Friending					0.959***		0.926***	0.988***
					(0.064)		(0.094)	(0.086)
Origin Yearly Relative Friending						-0.988***		
						(0.071)		
Dest Yearly Relative Friending						0.926***		
						(0.071)		
Quarter FEs	X	X	X	X	X	X	X	X
Origin County FEs			X				X	
Dest County FEs				X				X
N	1,771,041	1,771,041	1,771,041	1,771,041	1,096,874	1,096,874	1,096,874	1,096,874
Sample Mean	3.160	3.160	3.160	3.160	0.005	0.005	0.005	0.005

Note: Table shows results from regressions exploring the change in friending of natives, before and after a move within Germany. Columns 1 and 5 correspond to the relationships depicted in panels (a) and (b) of Figure 6. Columns 2 and 6 regress each component of the difference in the right-hand side measure in columns 1 and 5 separately on the outcome. Columns 3 and 7 repeat columns 1 and 5 with origin fixed effects; columns 4 and 8 repeat columns 1 and 5 with destination fixed effects. We correct for sampling error in the right-hand side measures by randomly splitting the individual-level non-mover data into two halves and instrumenting for one set of averages with the other. See Appendix E for more information this procedure. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01).

B Construction of "Native German" Sample

For many of our analyses we use a sample of Facebook users, which we refer to as "German natives", that meet *both* criteria 1 and 2 described below (as well as the primary sample inclusion criteria described in Section 1.1). Our methodology is not intended to proxy for citizenship status or ethnicity; rather it generates a sample of users who generally use the German language and—according to self-reported profile information and home region predictions—appear to have lived in Germany for a substantial amount of time. This will include, for example, individuals of Syrian descent who report a German hometown and primarily use the German language on Facebook. For more details, see footnote 3.

- **Criteria 1:** The user meets *one* of the following
  - The user produces $\geq$ 75% of their content in German
  - The user produces $\geq$ 50% of their content in German, AND lists a German hometown or high school on their profile
- **Criteria 2:** The user meets *all* of the following
  - Does not list a hometown in a "top migration country"
  - Does not list a high school in a "top migration country..
  - Did not first have a predicted home region in a "top migration country

The top migration countries are the 15 countries outside of the European Union and within Eastern Europe, the Middle East, or Africa with the most foreign nationals in Germany.

C Individual-Level Outcomes

We consider three dimensions of social integration of Syrian migrants: friendship, language, and participation within local groups. Within each dimension, we construct a number of measures, though we focus on a primary measure within each dimension, which is noted in **bold**.

1. Friendship Measures
   (a) ***N Local Native Friends:*** **The number of friends a user has in the same county or a bordering county that are in the German native sample.**
   (b) *N Native Friends:* The number of friends a user has in the German native sample.
   (c) *N Top 50 Native Friends:* The number of a user's closest 50 friends that are in the German native sample.
   (d) *% of Friends Native:* The percent a user's total friends that are in the German native sample.
2. Language Measures
   (a) *% Content Produced in DE*: The share of content a user produces (e.g., in posts, comments) that is in German. "Half-life" of 30 days (i.e., a post 30 days ago is weighted as half a post today).
   (b) *% Content Consumed in DE*: The share of the content a user engages with by using the "react" and "comment" features that is in German. 1 comment = 7 reactions. "Half-life" of 30 days.
   (c) ***Produces Any DE Content*** **: An indicator for "% Content Produced in DE" is >1%.**
   (d) *Consumes Any DE Content*: An indicator for "% Content Consumed in DE" is >1%.
   (e) *Account in DE*: Whether a user selected German as their language in their account settings.
3. Local Group Participation Measures
   (a) ***N Local Native Groups*****: The number of groups a user is in that have 5 - 5,000 users; $\geq$ 90% of users in Germany and $\geq$ 75% of users in one NUTS2 region; and $\geq$ 50% of users in the German native sample.**
   (b) *% Groups Local Native*: The share of groups a user is in that match the criteria in "N Local Native Groups."
   (c) *Avg. % Native in DE Groups*: Among groups a user is in which have > 90% of users in Germany, the average share of users that are German natives.

We also observe the following additional measures at the individual level:

- *N Local Syrian Friends*: The number of friends a user has in the same county or a bordering county that are in the Syrian migrant sample
- *N Local Other Refugee Country Friends*: The number of friends a user has in the same or bordering county that are migrants (determined by hometown, high school, or past usage) from one of the five countries with the most asylum applicants in Germany in 2020 other than Syria: Turkey, Afghanistan, Iraq, Nigeria, and Iran.

- *N Local Recent Other Refugee Country Friends*: The number of friends a user has matching the "N Local Other Refugee Country Friends" criteria with observed arrival in Germany 2015 or later. As described in Section 1.1, users with an "observed arrival timing" are those who first used Facebook outside of Germany.

D Syrian Migrant Integration by Demographics

We explore the heterogeneity in integration outcomes by demographics formally using the the following multivariate regression model:

$$Y_{i,j} = \alpha_0 + \alpha_1 Z_i + \psi_{j(i)} + \epsilon_i. \tag{5}$$

For the results in columns 1-4 of Table A11, $Y_{i,j}$ is the number of native local friends of individual i has. All specifications include various controls Z_i for the amount of time users spend on Facebook, ensuring that differences in observed integration outcomes are not driven by variation in the intensity of Facebook usage. We also include fixed effects for the user's number of quarters since arrival in Germany and the number of quarters living in their current county.

In column 1, Z_i also includes dummies for age, gender, and whether the user has another Syrian migrant household member or non-household family member who was in Germany more than a year prior to their arrival.[32] Consistent with the univariate patterns in Figure 1, we find that younger and male Syrians befriend disproportionately many local German natives. All else equal, a female Syrian migrant has 3.7 fewer local native friends than a male does. Similarly, a Syrian migrant aged 55 or older has 4.6 fewer native local friends than a comparable individual under the age of 25. Column 1 also shows that, while migrants with a family member who arrived earlier in Germany *outside* of the household have more local native friends, individuals with an earlier arriving Syrian migrant *inside* their household have fewer local native friends. This result adds to prior findings that connections to other migrants support integration in some settings and hinder it in others (e.g., Lazear, 1999; Edin, Fredriksson and Åslund, 2003; Cutler, Glaeser and Vigdor, 2008; Damm, 2009; Beaman, 2012; Martén, Hainmueller and Hangartner, 2019). In our context, the results suggest that somewhat-distant familial connections might provide support and guidance to help the social integration of newly arriving migrants, whereas the presence of close household connections might reduce the need to form connections with local natives.

Column 2 adds fixed effects for the Syrian migrants' current county of residence, $\psi_{j(i)}$, to the regression. The R^2 increases by 21% from 0.132 to 0.160, consistent with the presence of important regional differences in the social integration of Syrian migrants. The coefficients on the demographic characteristics in Z_i are largely unaffected by the addition of county fixed effects, suggesting there is a little selection based on these characteristics into more or less integrated places.

Column 3 adds controls for each user's total number of friends outside Germany, total number of groups joined, and total amount of recent content produced. These controls absorb additional variation in individuals' Facebook usage patterns beyond those in column 1, but could also remove variation in the true sociability of individuals that might influence their ability and desire to socially integrate with natives. While most coefficients remain largely unchanged, the gender coefficient falls somewhat in absolute terms, from -3.6 to -3.2. A possible interpretation is that Syrian migrant men generally have larger social networks, but, even conditional on overall network size, also make more German friends.

In column 4 of Table A11 we add household fixed effects while dropping individuals without additional household members from the sample. Even within the same household, and conditional on

[32] Family and household information is determined through self-reports and model-based imputations. Similar data are used in Bailey et al. (2022) and Chetty et al. (2022*a*,*b*).

Table A11: Syrian Migrant Integration by Demographics - Friending to Natives

	Facebook Sample N Local Native Friends				SOEP Sample N German Acquaintances	
Age 25 - 34	-1.012***	-0.894***	-0.873***	-1.148***	-0.839*	-1.089**
	(0.053)	(0.052)	(0.052)	(0.129)	(0.47)	(0.47)
Age 35 - 44	-2.963***	-3.019***	-2.941***	-2.375***	-1.116*	-1.070*
	(0.062)	(0.061)	(0.061)	(0.158)	(0.58)	(0.58)
Age 45 - 54	-4.012***	-4.102***	-4.147***	-4.765***	-2.362***	-2.238***
	(0.080)	(0.079)	(0.079)	(0.184)	(0.78)	(0.77)
Age 55+	-4.548***	-4.531***	-4.586***	-7.226***	-3.378***	-3.594***
	(0.100)	(0.098)	(0.099)	(0.241)	(1.24)	(1.23)
Female	-3.676***	-3.610***	-3.225***	-3.267***	-1.421***	-1.512***
	(0.043)	(0.042)	(0.045)	(0.090)	(0.47)	(0.48)
Household Member in DE 1+ Year Prior	-0.377***	-0.290**	-0.352***			
	(0.100)	(0.099)	(0.099)			
Non-Household Family in DE 1+ Year Prior	0.524***	0.621***	0.421***			
	(0.091)	(0.089)	(0.089)			
Quarters Since DE FEs	X	X	X	X	X	X
Prev Quarters in NUTS3 FEs	X	X	X	X		
Personal Usage Controls	X	X	X	X		
County / State FEs		X	X	X		X
Log (1 + Total Outside Germany Friends)			X	X		
Log (1 + Total Other Groups)			X	X		
Log (1 + Total Content Produced Past Year)			X	X		
Household FE				X		
N	349,072	349,072	349,072	84,216	1,095	1,095
R-Squared	0.132	0.160	0.165	0.658	0.048	0.093
Sample Mean	5.029	5.029	5.029	4.195	6.232	6.232

Note: Table explores variation in migrants' social integration. Each observation in columns 1-4 is a user in the Syrian migrant Facebook sample. Column 1 includes (i) controls for age and gender; (ii) fixed effects for the number of quarters on Facebook in their current county and the number of quarters since arrival in Germany (we use a single dummy value for those for which we do not observe arrival, but obtain nearly identical results if we instead drop these users); (iii) dummies for whether the user has another Syrian migrant household member or non-household family member in Germany prior to their arrival. (For all users not in the "observe arrival timing" sample, these two dummies are set to 0); and (iv) the following measures of the Facebook usage intensity: linear controls for log(0.5 + minutes on FB in the last 28 days), log(91 - days on Facebook out of the last 90), log(1081 - days on Facebook out of the last 1080). Column 2 adds county fixed effects. Column 3 adds controls for each user's total number of friends outside Germany, total number of non-local/native groups joined, and total amount of content produced in the last year. Column 4 adds a household fixed effect, limiting to households for which we observe more than one Syrian migrant. Columns 5 and 6 use data from the Socio-Economic Panel in 2016. The dependent variable in these columns is the number of new acquaintances made in Germany (see footnote 8). Each observation is a recent migrant from Syria living in Germany as of the date of the survey. Both columns 5 and 6 include controls for the number of quarters in Germany. Column 6 also controls for state fixed-effects. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01).

general Facebook usage patterns, younger and male Syrian migrants are better socially integrated.

Appendix Table A4 presents results analogous to column 1-4 of Table A11 for our key language- and group-based measures of social integration, and Table A5 presents results analogous to column 3 of Table A11 for a number of other outcomes. Across all measures, we find highly consistent relationships between age, gender, and family connections and the social integration of Syrian migrants.

One concern with this analysis may be that, despite our strict controls for Facebook usage and the consistency of our results across outcome, the observed differences in integration outcomes across demographic groups may still be driven by patterns of Facebook usage, rather than reflecting true demographic variation in social integration. To address this concern, we also look at related outcomes in the Socio-Economic Panel data, namely the number of native acquaintances made in Germany among a sample of recent Syrian migrants. In 2016, the SOEP administered a survey specifically targeted at recent migrants to Germany. We focus on the 1,095 Syrian migrants in the data that are 18+ years old.

Columns 5 and 6 show that the patterns of friending across demographics in the SOEP data mirror those we observe in the Facebook data in columns 1-4. Female and older migrants have fewer local acquaintances than male and younger migrants, respectively, on average. This holds with state fixed effects in column 6. Indeed, even the coefficient estimates using the Facebook and SOEP data are generally quite similar. We interpret this as reassuring as it shows that the patterns of social integration we identify in the Facebook data align closely with available survey evidence. The Facebook data, however, is much larger and more detailed, allowing us to more precisely explore the spatial variation in integration and to better understand the determinants of this variation.

E Assessing the Reliability of Regional Estimates

A potential concern with our regional estimates of integration outcomes is that the differences we observe might be due to sampling error, instead of capturing actual differences in the parameters of interest. In this appendix we explore this concern and describe the methods used to address it.[33]

To assess the degree to which our variation is driven by sampling error, we seek an estimate of:

$$r = \frac{Var(\delta_j)}{Var(\delta_j) + Var(\epsilon_j)} \tag{6}$$

Here δ_j is the true (un-observable) parameter for county j, $Var(\delta_j)$ is the variance of that parameter across all counties, and $Var(\epsilon_j)$ is the variance due to sampling error (noise) when we measure our estimate $Var(\hat{\delta}_j)$, such that $Var(\hat{\delta}_j) = Var(\delta_j) + Var(\epsilon_j)$. Our outcome of interest is the reliability, r.

We estimate r in two ways: (i) a "split sample" estimate generated by randomly splitting the individual-level data in half (within counties) and comparing the resulting estimates; and (ii) a "standard error-based" estimate generated by comparing the magnitudes of the standard error squared of each estimate with the variance of the estimates across counties.

Formally, our "split sample" estimates are given by:

$$\hat{r} = Corr(\hat{\delta_j^1}, \hat{\delta_j^2}) \cdot \frac{\sqrt{Var(\hat{\delta_j^1}) Var(\hat{\delta_j^2})}}{Var(\hat{\delta}_j)} \tag{7}$$

Where $\hat{\delta}_j$ is the county-level estimate of δ in county j, the average of individual-level measures across users in the county; $Var(\hat{\delta_j^1})$ and $Var(\hat{\delta_j^2})$ are the population-weighted variances of these measures in the first and second split samples; $Var(\hat{\delta}_j)$ is the population-weighted variance in the full sample; and $Corr(\hat{\delta_j^1}, \hat{\delta_j^2})$ is the population-weighted correlation.

Our "standard error-based" estimates are given by:

$$\hat{r} = \frac{Var(\hat{\delta}_j) - E[s^2_{\hat{\delta}_j}]}{Var(\hat{\delta}_j)} \tag{8}$$

Where $s_{\hat{\delta}_j}$ is the standard error of the county level average $\hat{\delta}_j$ for county j.

The first two columns of Appendix Table A12 show that the reliability of each of our regional averages is around 0.9 or above regardless of the method used. This suggests that 90% or more of the variance in a given regional measure reflects true latent differences rather than sampling error.

As noted in Section 2, there are moderate differences in the Facebook usage of natives across space (largely at the intensive margin) which could affect the raw regional averages we measure. To account for this, our estimates in Figure 2 and Appendix Figures A9 and A10 are constructed after residualizing by differences in natives' Facebook usage. Column 3 of Appendix Table A12 shows split-sample reliability estimates using $\hat{\delta_j^1}$ and $\hat{\delta_j^2}$ that have been residualized in this same manner. The reliability estimates are largely unchanged, suggesting they are not driven by regional differences in usage.

[33] The methods described in this appendix are similar to procedures used in Chetty and Hendren (2018*b*), Chetty et al. (2022*a*), and Chetty et al. (2022*b*).

Table A12: Reliability of County-Level Measures, Syrian Migrant Sample

	Reliability		
	Split-Sample	SE-Based	Split-Sample, Usage Control
N Local Native Friends	0.962	0.961	0.938
Produced Any DE Content	0.909	0.901	0.883
N Local Native Groups	0.948	0.946	0.934
N Local Syrian Friends	0.989	0.989	0.989

Note: Table shows the reliability of county-level measures. In columns 1 and 2 the measures are averages across Syrian migrant users. In column 3 these measures are residualized on extensive and intensive measures of local natives' Facebook usage, as described in Section 2. Reliability is defined by equation 6. The split sample reliability estimates are generated using equation 7. The standard error-based reliability estimates are generated using equation 8.

In Section 3.1, we construct regional measures of *general friendliness* using the German native sample. The sample size for these measures is very large and, accordingly, the reliability estimates using both methods is greater than 0.995. Therefore, essentially all of the sampling error present in our measures of *relative friending* (generated by dividing the Syrian migrant integration outcomes by general friendliness) is driven by the Syrian migrant integration outcomes.

In Table 4 we correlate regional measures against each other across counties. In these cases, the correlations between the estimates may understate the true correlations between parameters because of noise introduced by the sampling error. To recover estimates of the correlation between the true parameters we calculate:

$$\hat{Corr}(\psi_j, \mu_j) = Corr(\hat{\psi}_j, \hat{\mu}_j)\sqrt{\frac{1}{\hat{r}_\psi}}\sqrt{\frac{1}{\hat{r}_\mu}}. \tag{9}$$

Where $Corr(\hat{\psi}_j, \hat{\mu}_j)$ is the correlation between estimates $\hat{\psi}_j$ and $\hat{\mu}_j$ (of parameters ψ_j and μ_j) across all counties j, and $\hat{r}_\psi$ are $\hat{r}_\mu$ are their reliability estimates from equation 8. We present these "signal correlations" in Appendix Table A6.

In Section 2.1 and 3.2, we use certain regional (and region-by-demographics) measures as right-hand side variables in our movers specifications. The sampling error in these estimates will attenuate their regression coefficients. To see this, take the simple regression $Y = \beta \cdot X + \omega$ where we observe $\hat{X}$, an estimate of X with independent sampling error ϵ. Then when estimating $Y = \hat{\beta} \cdot \hat{X} + \nu$ we have:

$$\begin{aligned} \hat{\beta} &= \frac{Cov(Y, \hat{X})}{Var(\hat{X})} \\ &= \frac{Cov(Y, X+\epsilon)}{Var(X+\epsilon)} \\ &= \frac{Cov(Y, X)}{Var(X) + Var(\epsilon)} < \frac{Cov(Y, X)}{Var(X)} = \beta. \end{aligned} \tag{10}$$

To account for this, in our movers analyses we first randomly split the individual-level data used to construct the relevant right-hand side measures in two halves. We then instrument for the value con-

structed by one half with the other. To see the intuition behind this procedure, let $\hat{X}_1$ and $\hat{X}_2$ be the split sample estimates. Then the first stage of a two-stage least squares estimate is given by $\hat{X}_1 = \phi_1 \cdot \hat{X}_2 + \nu_1$, where $\phi_1 = \hat{r} = \frac{Var(X)}{Var(X)+Var(\epsilon_2)}$. The reduced form is given by $Y = \phi_2 \cdot \hat{X}_2 + \nu_2$, where $\phi_2 = \frac{Cov(Y,X)}{Var(X)+Var(\epsilon_2)}$. Then the resulting estimate is:

$$\hat{\beta} = \frac{\phi_2}{\phi_1} = \phi_2 \cdot \frac{1}{\hat{r}} \approx \frac{Cov(Y,X)}{Var(X)} = \beta. \tag{11}$$

F Königsteiner Schlüssel and the Assignment of Refugees to Place

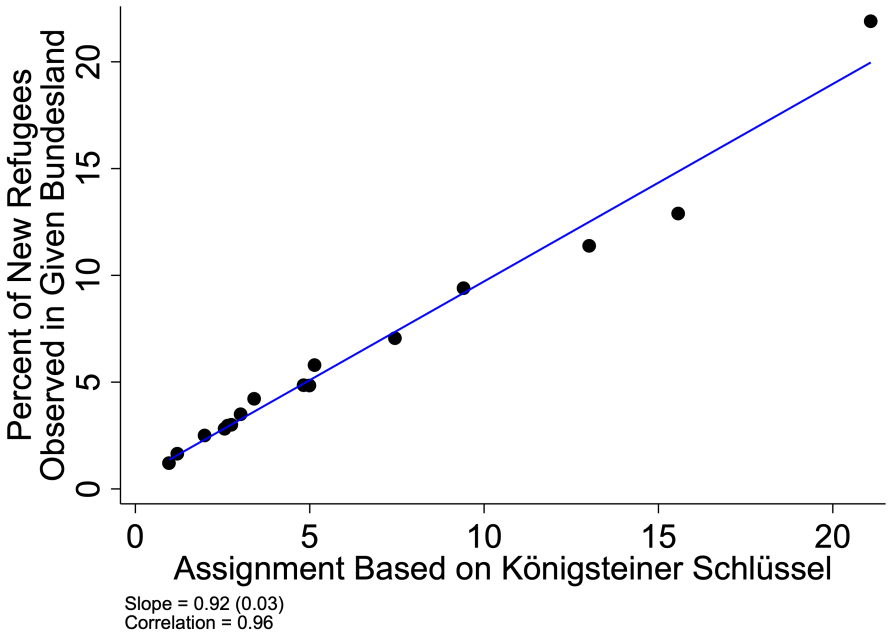
In this section, we attempt to compare the official refugee allocation rule—the so-called Königsteiner Schlüssel—to observed administrative data on refugee assignment.

The Königsteiner Key is an allocation rule which was designed in the 1940s to assign refugees to the sixteen different German states. It takes as input a state's population and tax income and weights these two factors with 1/3 and 2/3, respectively (Deutscher Bundestag, 2020). The key is updated annually, but given the slow-moving nature of its inputs, it is stable over time.

To infer to what extent the key has been abided to during the time period of interest for our study, we compare the 2019 assignment key (for data availability reasons) to the percentage of the total number of refugees that live in a given state and have been in Germany for less than 1 year, for each year from 2015 to 2019. The latter measure is intended to approximate for new-arrivals in the absence of direct data on this and the data for this approximate measure is obtained from the German Statistical Office.

Figure A13 shows the result of our comparison. The correlation of 0.96 and a slope of 0.92 indicates that the observed assignment lines up very closely with the official assignment rule. We find this reassuring, as it suggests that despite the large influx of migrants during these year, refugee assignment largely followed the official assignment key. While we believe this is strong suggestive evidence that, to adhere to this rule, assignment to places was somewhat random, it remains possible that the composition of migrants by place is non-random.

Table A13: Comparison Königsteiner Key and Assignment of Refugees to Place



Note: Figure compares assignment of recent refugees to place with the official assignment key, i.e. the Königsteiner Schlüssel from 2019. The Königsteiner Schlüssel is compromised of a state's total population and a state's tax income where the former is weighted with one third and the latter is weighted two thirds. Assignment of recent refugees is approximated by the percentage of the total number of refugees that live in a given state and have been in Germany for less than 1 year, for each year from 2015 to 2019. The data comes from the German Statistical Office.

G Identifying Place Based Effects with Movers

To quantify the contribution of place-based effects to the spatial variation in migrants' integration outcomes, we propose a simple model in which the rate of friendships between migrants and a local natives is determined by the sum of place-based effects—which we allow to vary across time and with observable migrant characteristics—and other *unobservable* individual-level factors of the individuals involved. Since only place-based factors change around a move, this model allows us to estimate the share of regional variation in the social integration of migrants that can be attributed to place-based effects. We describe here the friending model and identifying assumptions in the context of the migrant mover design from Section 2.1. These features carry over to the native mover design in Section 3.2.

Friending model. We consider the following basic model of friending between migrants and locals which is similar to Finkelstein, Gentzkow and Williams (2016). We let each individual's friending outcome be the sum of their county's effect ($\text{PlaceEffect}^{(p)}$) and their personal individual effect (IndivEffect_i). Let $\text{AvgIndivEffect}^{(p)}$ be the average of IndivEffects for individuals in county p. Then the difference between the average outcomes, x, in two regions, (2) and (1), is the sum of differences between the place-based effect and the average of individual-effects.

$$x^{(2)} - x^{(1)} = (\text{PlaceEffect}^{(2)} - \text{PlaceEffect}^{(1)}) + (\text{AvgIndivEffect}^{(2)} - \text{AvgIndivEffect}^{(1)}). \tag{12}$$

We want to know the share of $x^{(2)} - x^{(1)}$ that is due to place-based effects, formally:

$$\frac{\text{PlaceEffect}^{(2)} - \text{PlaceEffect}^{(1)}}{(\text{PlaceEffect}^{(2)} - \text{PlaceEffect}^{(1)}) + (\text{AvgIndivEffect}^{(2)} - \text{AvgIndivEffect}^{(1)})}. \tag{13}$$

We cannot observe any of these parameters directly. At the individual level, however, we know that when a mover moves from (1) to (2), only the place-based factors should change. Her individual level effects are constant, so any change in friending outcomes must be driven by place based effects. So for mover i who moves from (1) to (2) at time t:

$$y^{\Delta}_{i,t} = (\text{PlaceEffect}^{(2)} - \text{PlaceEffect}^{(1)}). \tag{14}$$

Where $y^{\Delta}_{i,t}$ is the change in outcome before and after the move for mover i. Then α, below, is equivalent to equation 13, our outcome of interest.

$$y^{\Delta}_{i,t} = \alpha \cdot (x^{(2)} - x^{(1)}). \tag{15}$$

In addition to this baseline logic, we allow for separate place effects across certain observable demographics such as age and gender, as well as time since moving to Germany. The *AvgIndivEffect* is then the average of the remaining unobservable individual effects. When estimating α we remove the variation in $y^{\Delta}_{i,t}$ explained by overall time trends (e.g., if throughout Germany Syrian migrants make more native friends over time) by adding quarter of move fixed effects, ζ_t.

Taking model to the data. We bring this model to the data by comparing the rate at which movers make friends in the year before and after their move to the difference in the average friending rates of otherwise similar non-movers in each location.[34] Focusing on migrant movers (rather than on native movers as in section 3.2), for each user i moving in quarter t, the outcome of interest is the change in the quarterly probability of making at least one local German friend, $y^{\Delta}_{i,t}$, defined as:

$$y^{\Delta}_{i,t} = 0.25 \left[\sum_{\tau=t}^{t+3} Y_{i,\tau} - \sum_{\tau=t-4}^{t-1} Y_{i,\tau} \right]. \tag{16}$$

Here, $Y_{i,t}$ is an indicator for whether Syrian migrant i makes at least one local German friend in quarter t. Similar to before, we residualize each side of the difference on regional measures of natives' Facebook usage. To compare $y^{\Delta}_{i,t}$ to differences in the average integration rates of observably similar non-movers in each place, we construct sets of users who match each mover on the important determinants of social integration in Section 1.4: gender, age group, and time spent in Germany. Formally, for user i moving in quarter t, we let $O(i,t)$ and $D(i,t)$ be the sets of similar non-movers in the origin at time $t-4$ and in the destination at time t, respectively. We then define the differences in their average outcomes, $x^{\Delta}_{i,t}$, as:

$$x^{\Delta}_{i,t} = 0.25 \left[\frac{1}{|D(i,t)|} \sum_{j \in D(i,t)} \sum_{\tau=t}^{t+3} Y_{j,\tau} - \frac{1}{|O(i,t)|} \sum_{j \in O(i,t)} \sum_{\tau=t-4}^{t-1} Y_{j,\tau} \right]. \tag{17}$$

The set cardinalities $|O(i,t)|$ and $|D(i,t)|$ are the number of non-movers in the matched comparison groups for each mover. Intuitively, $x^{\Delta}_{i,t}$ is the difference in the average quarterly probability of a non-mover migrant making a native local friend between the destination location in the year after the move and the origin location in the year before the move. Time-specific measures allow for changes in the differences between regions over time. Again, we residualize each side of the difference on regional measures of natives' Facebook usage. We then estimate:

$$y^{\Delta}_{i,t} = \alpha_0 + \alpha_1 x^{\Delta}_{i,t} + \xi_t + \epsilon_{i,t}, \tag{18}$$

where slope α_1 is our outcome of interest. An estimate of α_1 close to 1 would suggest that, within the first year of moving, migrant movers' friending behavior fully adjusts to the level of local non-movers' friending behavior. An α_1 close to 0 would suggest that migrants do not adjust their friending rates systematically toward the level of local non-movers. Because migrant observables do not differ significantly across space, under the relatively weak identification assumptions discussed below, α_1 estimates the share of the observed differences in the social integration of migrants across locations that are due to causal place-based effects rather than unobservable individual characteristics. The quarter of move fixed effect, ξ_T, remove variation in overall time trends in the rates of befriending local natives.

One challenge with our estimation is that we only observe a sample estimate of each mover's $x^{\Delta}_{i,t}$, denoted by $\hat{x}^{\Delta}_{i,t}$. Measurement error in the true differences in friending probabilities of non-movers across locations would thus lead to attenuation bias in α_1. To account for this sampling error, when

[34] In this analysis we limit to movers who were in their origin and destination counties for four or more consecutive quarters each, less stringent than the prior analysis which required six quarters in the destination. In addition, we only include observations for which there are at least five "matched" non-movers in both the origin and destination.

estimating equation 18, we randomly split the individual-level data of the friending behavior of non-movers used to construct $\hat{x}_{i,t}^{\Delta}$ into two sub-samples and instrument for the value constructed in one sub-sample with the value constructed in the other sub-sample (see Appendix E for details).

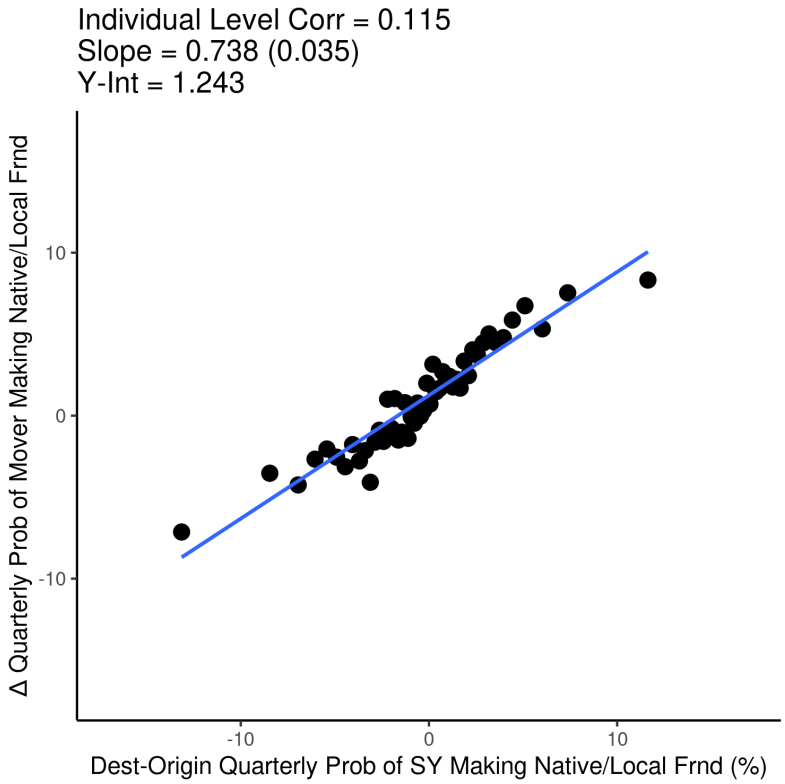
Identification Assumptions. Our interpretation of α_1 relies on the identifying assumption that place-based effects are additive and additively separable from any unobservable individual-level factors. This additivity allows us to aggregate the level of within-migrant differences across migrants to identify α. It implies, for example, that a move from place A to place B should have the same effect as a move from place B to place A. This is supported by Figure 3, as well as the results in Figure A19 and Table A8. Additive separability also implies that migrants' friending rates between locations will vary by the same *absolute amount* across unobservables. (The model does, however, allow for non-additive relationships between our key observables—gender, age, and time in Germany—and migrants' friending rates). Our identification also relies on there being no systematic shocks to unobservable factors that coincide exactly with the move quarter and affect native friending differentially by origin and destination.

These identifying assumptions are relatively weak and allow for movers to differ from non-movers on observable and unobservable characteristics, and for these differences to correlate with origin and destination characteristics. For example, our model allows for "better integrating migrants" to be more likely to move to "better places." Intuitively, this is because our estimates come from *within-migrant* differences in integration over time, and "better" integrating migrants will make more friends both before and after the move. This differs from designs used in papers such as Chetty and Hendren (2018*a*) and Chetty and Hendren (2018*b*). These papers, which rely on cross-sectional outcomes, use within-family designs to rule out selection effects. Our data allow us to measure the outcome in the panel context (as in Finkelstein, Gentzkow and Williams, 2016), mitigating these concerns.

Our research design allows the level of movers' pre-move friending within an origin county to correlate with destination friending levels due to differences in individual characteristics. Movers' native friending around a move can also differ from the trends of non-movers. This could occur if, as suggested by Figure 3, all movers make fewer local connections in anticipation of a move or more connections immediately after a move. Each of these would increase α_0, but leave α_1 unaffected. Our model would be affected if these downward trends in movers' propensity to make friends before relocating differed systematically by the integration levels in the movers' destinations.[35] Figure 3 provides evidence that such differential trends do not exist. As an additional test, in Figure A13, we decompose our results from Figure 3 into friendships initiated by the mover and those initiated by the Germans in their destination. We find that, following a move, both migrant-initiated and *native-initiated friendships* change in the predicted direction. This provides more evidence that our results are not driven by changes in migrant friending preferences around the time of the move that correlate with the characteristics of the destination.

[35] Put differently, our model allows for migrants' individual characteristics to change around a move so long as they do not differ systematically by destination location. For example, our estimates of α_1 would be biased upward if movers to better places became differentially less sociable before a move.

Figure A19: Δ Syrian Migrant Mover Friending Integration vs. Matched Non-Movers



Note: Figure shows a binned scatter plot describing the change in the friending of Syrian migrants to German natives before and after a move within Germany. The population is Syrian migrant users who moved between two non-neighboring counties and were in the first and second county for 4+ consecutive quarters each. The y-axis displays $y^{\Delta}_{i,t}$, movers' change in the quarterly probability of making a native local friend the year before to after the move. The x-axis displays $\hat{x}^{\Delta}_{i,t}$, the difference in average outcomes for comparable non-movers at the same time. We match each mover to a set of non-movers who lived in the origin location a year before the move and to a set who lived in the destination location at the move. In addition we also match movers to non-movers of the same gender and age bucket (18-29, 30-39, 40+), and whom we first observed on Facebook in Germany in the same year. We include observations for which there is at least 5 non-movers in both the origin and destination match group. We control for quarter of move fixed effects. We correct for sampling error in the x-axis measures by randomly splitting the individual-level non-mover data into two halves and instrumenting for one set of averages with the other. See Appendix E for more information this procedure. Standard errors are shown in parentheses. Appendix Table A8 presents formal regression results on the relationships in this figure.

Results for Migrant Movers. Figure A19 displays a binned scatter plot of $y^{\Delta}_{i,t}$ against $x^{\Delta}_{i,t}$, with the slope corresponding to α_1 in equation 18.[36] The relationship is symmetric around zero and linear, consistent with additive effects of place. The fact that the scatter plot is horizontally centered around zero also suggests that, conditional on demographics, migrants do not systematically move to places with higher or lower levels of integration. The slope estimate is 0.738: nearly three quarters of the observed regional variation in Syrian migrants' friendship formation with local natives is directly attributable to place-based effects that occur within the first year of after their move, rather than individual characteristics. In Appendix Figure A12 we plot the slope estimates separately for samples of users that are male, female, younger than 30 years old, 30 to 39 years old, and over 40 years old. For each group, the estimates are similar, suggesting our results are not driven by any particular demographic group of Syrian migrants.

[36] Appendix Table A7 summarizes the sample of movers and the corresponding matched sample of otherwise similar non-movers in the origin location.

While this section focuses on measures of social integration based on migrants' friending patterns, Appendix H explores our language-based measure of integration. Whereas our prior analysis could use panel data on quarterly friending rates, our language outcome—whether the user produces content in German—is only observable at high quality in the cross section. We thus study how a mover's language use *today* is shaped by the set of places they have lived, following similar analyses in Chetty and Hendren (2018*a*) and Finkelstein, Gentzkow and Williams (2021). Our results suggest that place-based effects drive much of the cross-sectional variation in Syrian migrants' German language usage.

The prior results have documented that when Syrian migrants move between German counties, their social integration patterns quickly adjust from those of their origin towards those of their destination county. Our results thus show that most of the observed regional differences in social integration are explained by the effect of places—either due to institutional factors associated with the location, or due to local native characteristics—rather than by the characteristics of the migrants. In this context, it is important to note that a mover design will not even capture the full extent to which individual integration is shaped by place-based effects. For example, Syrian migrants who learn the German language in high-integration places (possibly in local integration courses) might then use these skills to make German friends more quickly after moving to a low-integration place. This effect might be considered "place-based" in the sense that it is shaped by features of the mover's origin location, but will not be captured by our estimates. To the extent that such additional long-term place-based effects are important, our estimates of α_1 will even *understate* the extent to which places truly shape migration outcomes.

H Cross-Sectional Analysis of Movers and German Language Usage

We assess the degree to which selection drives our regional estimates of German language integration using a cross-sectional movers design. This follows similar designs in Chetty and Hendren (2018*a*) and Finkelstein, Gentzkow and Williams (2021), and differs from the design used in Sections 2.1 and 3.2 which utilize panel data on movers' friending. In particular, we model German language usage as a linear combination of the outcomes of non-movers in each of the mover's locations. Then, using the same mover criteria as in Figure A19, we estimate:

$$y_i = \alpha_0 + \alpha_1 \sum_p q(i,p) * x_{p,d(i)} + \kappa_{d(i)} + \epsilon_i \tag{19}$$

Here, y_i is an indicator for whether individual i produces German content on Facebook and $q(i,p)$ is the share of their quarters in Germany spent in place p. The notation $d(i)$ represents a set of demographics used to match movers to similarly situated non-movers. $x_{p,d}$ is the share of users in place p and demographic group d that produces German content, and $\kappa_{d(i)}$ are demographic group fixed effects, which remove variation driven by the demographic matching from our slope estimates. In our strictest specifications, we also add fixed effects for users' first and current county in Germany.

Table A14: Syrian Migrant Mover Language Integration vs Weighted Average of Places

	Produces Content in German (0/100)				
Predicted Prob. Of Using German (Weighted Avg. of Places Lived)	0.863***	0.857***	0.863***	0.813***	0.816***
	(0.037)	(0.043)	(0.058)	(0.042)	(0.058)
FEs	Cohort	Cohort	Cohort	Cohort X Curr. Cnty.	Cohort X Curr. Cnty. X First Cnty.
Sample		< 75% in Max County	< 60% in Max County		
N	23,249	18,233	10,172	23,069	14,474
Sample Mean	38.075	37.959	38.252	38.099	36.977

Note: Table shows results for comparisons between the German language usage of Syrian migrants who moved between counties and their predicted language usage based on the outcomes of non-movers in the places they lived. For each location, movers are matched non-movers by age, gender, and the first year they used Facebook in Germany (cohort). Column 1 shows our baseline specification from equation 19, which includes cohort fixed effects. Column 2 limits to only users who spent < 75% of their quarters in Germany in one county. Column 3 limits to those who spent < 60%. Column 4 repeats column 1 with cohort-by-current county fixed effects; column 5 repeats column 1 with cohort-by-current county-by-first county in Germany fixed effects. We correct for sampling error in the right-hand side measures by randomly splitting the individual-level non-mover data into two halves and instrumenting for one set of averages with the other. See Appendix E for more information this procedure. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01).

In contrast to equation 18, our unit of observation is a mover, not a move, and we use movers' location for every quarter they have been in Germany. As in our panel analyses, we cannot observe $x_{p,c(i)}$, but instead account for sampling error by constructing estimates $\hat{x}_{p,c(i)}$ from random halves of the data and instrumenting for one with the other. We also again relax the assumption of fully additive-seperability between individual-level factors and place-based effects by matching movers to similarly situated non-

movers on gender, age group, and year of arrival in Germany. This allows for non-additive interactions with these demographics. We enforce that each mover must have 20 matched non-movers.[37]

Table A14 presents results from our analysis. In column 1, an estimate of α_1 close to 1 would suggest that a Syrian migrant's likelihood of using German on Facebook is close to the averages of migrants in each location they have lived, weighted by the amount of time they lived in each location. The resulting slope estimate of 0.86 shows that this is the case. While this evidence is consistent with places having an *effect* on migrants' German language integration, it does not rule out alternative explanations. For example, it is possible that our sample includes many users who have spent a long time in a single location, and that the right hand side weighted averages are often dominated by a single region. If this were the case, our estimates could be largely driven by movers behaving similarly to local non-movers in general, rather than by place-based effects in particular. Columns 2 and 3 provide evidence that this story does not drive our overall results, as our estimates of α_1 remain similar when limiting our sample to users who spent <75% or <60% of their time in Germany in one county, respectively.

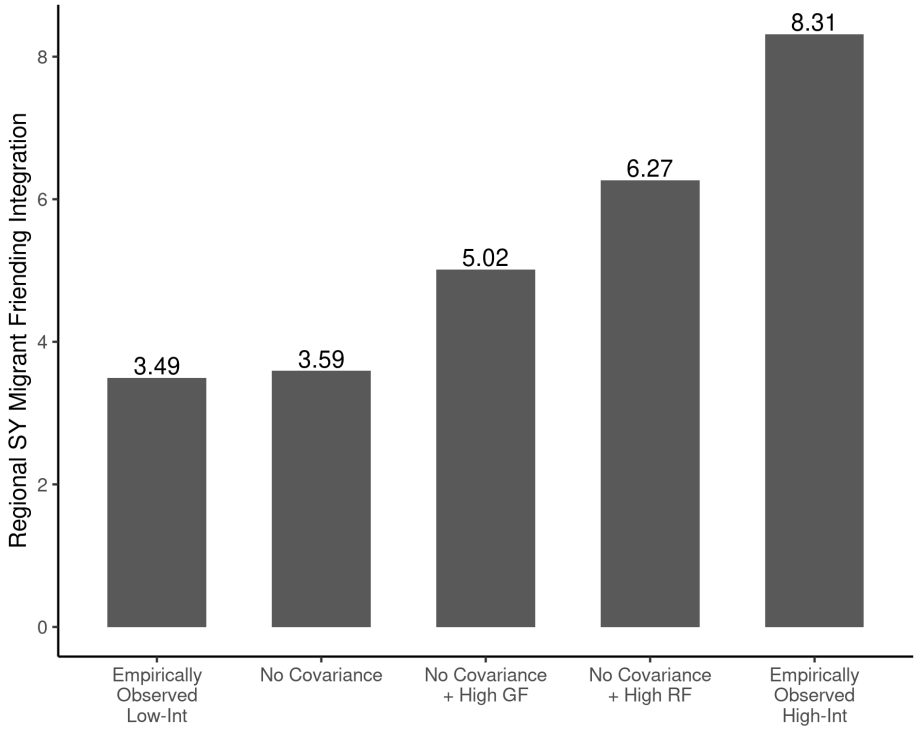
In column 4 we take another approach to testing whether our results are indicative of causal effects of place. In particular, we control for each user's current county, thereby identifying our slope estimates from variation in the user's origin counties. The slope estimate decrease slightly, but remains around 0.81. This suggests that much of the variation in language outcomes amongst movers across regions today is determined by where they *originally* lived in Germany, providing evidence against selection effects. In the final column, we control for both first county and final county fixed effects. Our identification, therefore, comes from the amount of *time* users' spend in each particular place. The slope estimates remains at 0.82, providing more evidence that a migrant's probability of using the German language scales linearly in proportion to the time they spend in high- and low-integration places.

[37] This threshold is higher than the five user minimum in Section 2.1. Our sample in this analysis, however, will remain larger because we (mechanically) do not enforce temporal matching.

I Decomposition of High- vs Low-Integration Regional Differences

In Figure A20, we conduct counterfactual exercises to explore the degree to which each of our two components explain the differences between counties with high- and low-friending integration. This follows a similar exercise in Chetty et al. (2022*b*). The first and fifth bars show the average integration of migrants in top and bottom quintile counties, respectively. Syrian migrants in top quintile counties make 8.31 native local friends on average, versus 3.49 in bottom quintile counties. In the second bar we multiply the bottom quintile averages of general friendliness and relative friending, thereby removing any within-quintile covariance. Doing so somewhat increases the value from the first bar, consistent with the small negative correlation between the two components in Table 4. The third and fourth bars replace the bottom-quintile averages of general friendliness and relative friending with the corresponding top-quintile averages, respectively. We view this as a counterfactual in which we hold one of the two integration components of low-integration regions fixed and adjust the other to the levels of high-integration regions. We interpret the difference between the second and fourth bars (2.68), compared to the second and third bars (1.43), as relative friending explaining about 1.9x as much of the difference between high and low-integration places as general friendliness.

Figure A20: Decomposition of Difference Between High- and Low-Integration Regions



Note: Figure shows how much of the difference between high and low friending integration counties is driven by general friendliness versus relative friending. The first and fifth bars show the average friending integration of Syrian migrants in top and bottom quintile counties, respectively. The second bar replaces each county observation from the first bar with the bottom quintile averages of general friendliness and relative friending. The third and fourth bars replace the bottom-quintile averages of general friendliness and relative friending with the corresponding top-quintile averages, respectively.

J Individual-level Correlates of Natives Behavior Towards Migrants

This appendix explores the relationship between observable native characteristics and behaviors toward Syrian migrants. In particular we focus on their (i) friending of local Syrian migrants; (ii) general friendliness; (iii) relative friending; and (iv) joining of pro-immigration organizations on Facebook.

Table A15: Natives - Measures of Friending

	N Local SY Friends		General Friendliness		Relative Friending		In Pro Imm. Group (0/100)	
Age 25 - 34	-0.073***	-0.073***	-19.097***	-14.407***	-0.059***	-0.061***	0.359***	0.146***
	(0.000)	(0.000)	(0.098)	(0.092)	(0.001)	(0.001)	(0.018)	(0.018)
Age 35 - 44	-0.116***	-0.114***	-55.586***	-52.328***	-0.081***	-0.080***	0.951***	0.858***
	(0.000)	(0.000)	(0.103)	(0.097)	(0.001)	(0.001)	(0.018)	(0.018)
Age 45 - 54	-0.132***	-0.131***	-62.533***	-62.415***	-0.098***	-0.095***	1.116***	1.152***
	(0.000)	(0.000)	(0.108)	(0.102)	(0.001)	(0.001)	(0.019)	(0.019)
Age 55+	-0.139***	-0.141***	-82.666***	-84.728***	-0.098***	-0.095***	2.105***	2.157***
	(0.000)	(0.000)	(0.108)	(0.102)	(0.001)	(0.001)	(0.020)	(0.020)
Female	-0.015***	-0.015***	-19.519***	-18.725***	-0.008***	-0.009***	0.882***	0.843***
	(0.000)	(0.000)	(0.056)	(0.053)	(0.001)	(0.001)	(0.010)	(0.010)
Has College	0.006***	0.006***	4.131***	7.619***	-0.000	-0.002***	1.931***	1.788***
	(0.000)	(0.000)	(0.060)	(0.056)	(0.001)	(0.001)	(0.011)	(0.011)
Prev Quarters in NUTS3 FEs	X	X	X	X	X	X	X	X
Personal Usage Controls	X	X	X	X	X	X	X	X
County FEs		X		X		X		X
N	17,768,822	17,768,822	17,768,822	17,768,822	17,515,164	17,515,164	17,768,141	17,768,141
R-Squared	0.020	0.031	0.170	0.263	0.001	0.002	0.035	0.042
Sample Mean	0.086	0.086	122.510	122.510	0.074	0.074	4.835	4.835

Note: Table shows results from regressing various outcomes on the demographics of users in the German native Facebook sample. The outcome is their number of local friends in the Syrian migrant sample in columns 1 and 2; their number of local friends in the German native sample in columns 3 and 4; their relative friending to Syrians and Germans defined by equation 2 in columns 5 and 6; and the number of groups registered with *ProAsyl* they are in in columns 7 and 8. Columns 1, 3, 5, and 7 include controls for age, gender, and whether they list a college on Facebook, as well as fixed effects the number of quarters on Facebook in their current county. They also include linear controls for log(0.5 + minutes on FB in the last 28 days), log(91 - days on Facebook out of the last 90), log(1081 - days on Facebook out of the last 1080). Columns 2, 4, 6, and 8 add county fixed effects. In columns 7 and 8 the personal usage controls also include fixed effects for each number of Facebook groups a user is in. Significance levels: *(p<0.10), **(p<0.05), ***(p<0.01).

Equation 5 is our multivariate regression of interest. Each observation is a German native user. In all specifications we include controls for the amount of time each user spends on Facebook and for the number of quarters they have been on Facebook in their current county. In certain specifications we also include county fixed effects. Y_i represents measures of the four outcomes listed above. Friending of local Syrian migrants is measured by the user's number of local Syrian migrant friends. Individual-level general friendliness is measured by the user's number of local native friends. We construct individual-level relative friending by replacing each term in the numerator of equation 2—$NLocalFriends_c^{DE \to SY}$ and $NLocalFriends_c^{DE \to DE}$—with its individual-level analog.[38] We identify pro-immigration Facebook pages and groups using a combination of string, url, and manual matching. Our outcome measure is

[38] A user must have at least one local native friend for this individual-level measure. The county-level average of this measure will equal the county-level measure in equation 2 if each observation in the former is weighted by the user's number of local native friends.

whether a user "likes" one of these page or is in one of these groups. In total, we identify 8,171 groups and pages, and measure 2.1 million user-page or user-group connections.

Table A15 presents results. Columns 1 and 2 show that younger natives and male natives are more likely to befriend migrants than older and female natives, respectively. Columns 3 and 4 show that these patterns are driven in part by general friendliness: a native being younger, male, or college educated is associated with having a larger network of local native friends. Columns 5 and 6 show that our individual-level measure of relative friending is also higher for younger and male German natives, while it is somewhat lower for college educated Germans compared to college educated Germans. Because Syrian migrants in Germany are more likely to be young and male than the average German native (see Table 1), one possible explanation for this finding is that homophily plays a strong role in shaping which natives befriend Syrian migrants. For example, younger German natives might be more likely to connect with younger Syrian migrants because younger people in general are more likely to connect, rather than because of particular behaviors toward migrants.

Columns 7 and 8 show that older, female, and college-educated natives are more likely than others to join pro-immigration groups on Facebook, conditional on Facebook usage. (For these analyses we include fixed effects for each number of total Facebook groups as user is in, holding constant a user's overall propensity to join Facebook groups. Our results remain qualitatively unchanged without this control). These are *opposite* the relationships presented for relative friending in columns 5 and 6, suggesting that is not necessarily those who are most supportive of pro-immigration groups that are most likely to disproportionately befriend Syrian migrants. This is again consistent with a story in which homophily, above specific attitudes or behaviors toward migrants, contribute to the demographic differences we observe in prior columns.

K High School Matching Procedure

We assign users to high schools using a three-step process. On Facebook, users can provide the high school that they attended in their profile. Some of these high schools (such as "Hogwarts" and "the School of Hard Knocks") are obviously incorrect, so we begin by filtering out such schools. We are left with a list of plausible high school names, which we then need to disambiguate, since many high schools share the same name. For this, we use a listing of high schools from the websites of German state governments (see DatenSchule Project.) For each user in our sample, we are able to observe the counties in which they lived during high school age. We use this information and their self-reported high school name to match them to a high school in the administrative data. To do this, we make use of a fuzzy string matching algorithm, applied to the list of high schools that are in the regions in which they lived between the ages of 13 and 18.[39] Using this methodology, we are able to match 1.2 million of the 2.2 million users to high schools from the administrative data.

In the second step, we consider the users who report a high school that we are unable to find in the administrative data. In some cases, simple misspellings or inconsistencies in the school's name prevent a match from being formed between the two data sets. In other cases, these discrepancies are due to variations in states' criteria for including schools in the lists provided on their websites (e.g., states differ in their inclusion of vocational high schools in the lists we use). For this reason, we create a listing of school names that are reported by 50 or more users in a single county, but which are not included in the administrative data. We allow users to be assigned to these well-attested schools as we would any other. We call these schools the "non-canonical schools", and include them in all regressions, though our results are robust to excluding them. This process adds another 81 thousand users to our sample. For users who attend a school which we cannot find in the administrative data, and which appears in the self-reported data fewer than 50 times in the same county, we discard their self-reported school.

Finally, for users without a validated self-reported high school, we attempt to impute the school they attended using information on their social network. Intuitively, this approach takes advantage of the fact that most users will attend the same school as their friends who live in the same area and are the same age. To do this, we find the modal high school among a user's friends in the county they live in (as well as counties bordering it) and who are no more than 3 years different in age from the user. If this modal high school is attended by at least 10 friends, and there are at least 5 times as many friends attending this high school as the next most common school, we assign the user to this high school. We repeat this process 10 times, adding 137 thousand more users to our sample.[40]

We are able to assign 63% of native users to high schools using this methodology. In the cohorts we use for our regression, the median cohort has 31 students, with an inter-quartile range of 15 to 52 students. The match rate is lower (24%) for Syrian migrant students, since they have relatively few local friends and are less likely to list a high school on their profile. Any mistakes we make in assigning Syrians to high schools are likely to bias our analyses away from finding an effect of exposure.

[39] If we are unable to find a high school that matches in one of the regions that they lived in, we consider the regions that neighbor the regions the user lived in.

[40] To get a sense for the predictive power of the above imputation methodology, we can examine how accurate it is in determining the high school attended by users who self-report the school they attended. The imputation method is able to assign a school to 25% of such users, agreeing with the self-reported school in more than 90% of cases.

L Validating General Friendliness Against External Surveys

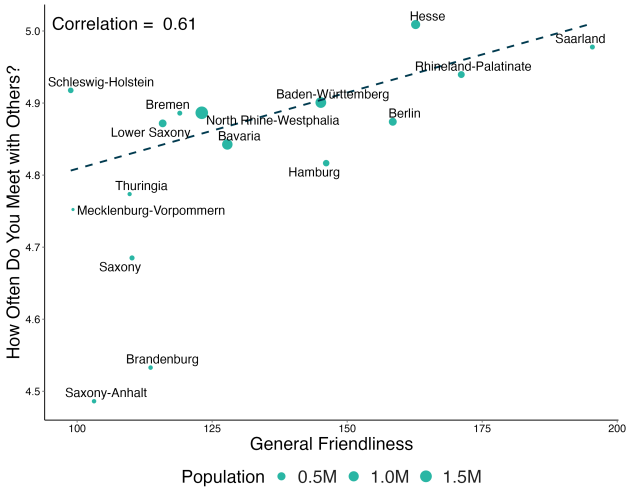
In this appendix, we assess the degree to which regional differences in general friendliness—given by the number of Facebook friendships that German natives have with other local German natives—reflects true variation in sociability versus just variation in regional Facebook usage patterns. As discussed in the paper, regional variation in observed Facebook usage patterns of German natives are small. For example, there is not much variation in the share of the German population that is on Facebook, or the time spent on Facebook by those that are active. Nevertheless, one might be concerned that our measures of general friendliness are predominantly picking up variation in social norms, for example related to how well I must know a person before sending them a Facebook friend request. To assess this concern, we benchmark our measures of general friendliness to related measures of sociability observed in two external surveys, the European Social Survey and the European Values Survey.

European Social Survey (ESS). We analyze how often people meet socially and take part in social activities using two questions from the European Social Survey (European Research Infrastructure Consortium, 2020, 2021). The first question captures the frequency of social meetings: "How often do you meet socially with friends, relatives, or work colleagues?" Respondents could answer: never (0), less than once a month (1), once a month (2), several times a month (3), once a week (4), several times a week (5), or every day (6). The second question captures participation in social activities: "Compared to Other People Your Age, How Often Do You Take Part in Social Activities?" Respondents could answer: much less than most (1), less than most (2), about the same (3), more than most (4), or much more than most (5). In our analysis, we pool responses from rounds 8 and 9 of the ESS, conducted between 2016 and 2017 as well as between 2018 and 2019, respectively. Figure A21 plots state-level measures of general friendliness against average survey responses (the ESS does not provide respondent locations at a more disaggregated level). Panel (a) shows a strong positive correlation between general friendliness and the average frequency of social meetings. Panel (b) shows a positive correlation between general friendliness and the frequency of participating in social activities.

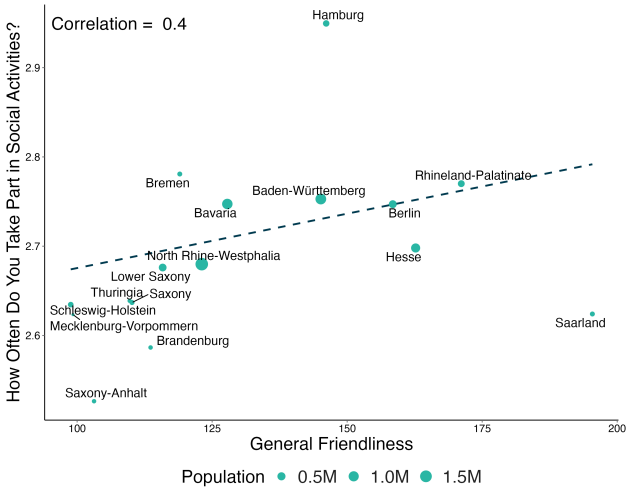
European Value Survey (EVS). The European Values Survey (EVS, 2022*a*,*b*) attempts to measure how trusting people are of one another in a region. Respondents were asked, "Could you tell me whether you trust people you meet for the first time completely, somewhat, not very much, or not at all?" We study responses from wave five of the EVS, conducted in Germany between 2017 and 2018. We measure average trust at both the NUTS2 and NUTS3-level. Panels (c) and (d) show a positive correlation between what percentage of people generally trust strangers—measured as the percentage who responded "Trust Completely" or "Trust Somewhat."—and general friending. These surveys provide reasonable evidence that friending activity on Facebook reflects true friending behavior.

Figure A21: General Friendliness Measured on Facebook Validated Against Survey Responses

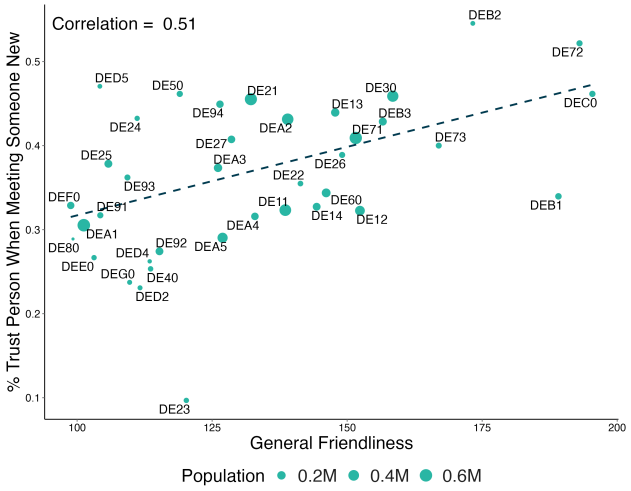
(a) Frequency of Social Meetings



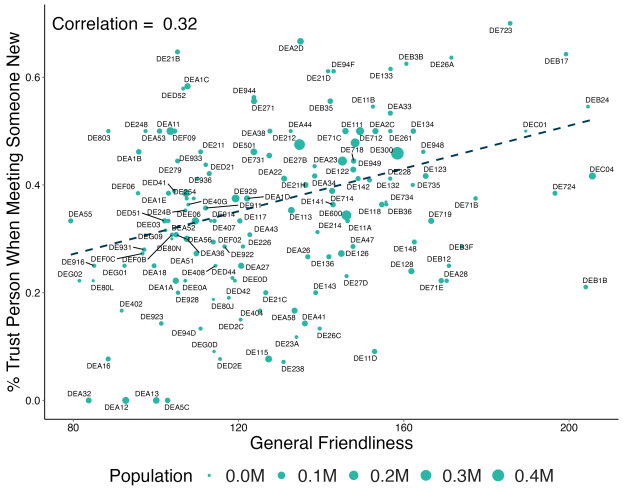
(b) Frequency of Participation in Social Activities



(c) Percentage who Trust Strangers (NUTS2)



(d) Percentage who Trust Strangers (NUTS3)



Note: Figure shows constructed measures of general friendliness benchmarked against survey data from the European Social Survey (ESS) and the European Values Survey (EVS). All panels show general friendliness on the x-axis. Panel (a) plots the average coded response to "How often do you meet socially with friends, relatives, or work colleagues?" Responses are coded as follows: never (0), less than once a month (1), once a month (2), several times a month (3), once a week (4), several times a week (5), or every day (6). Panel (b) plots the average coded response to "How often do you take part in social activities?" Responses are coded as follows: much less than most (1), less than most (2), about the same (3), more than most (4), or much more than most (5). Panel (c) plots the percentage of people who "trust somewhat" or "trust completely" people they meet for the first time by NUTS2 region. Panel (d) plots the percentage of people who "trust somewhat" or "trust completely" people they meet for the first time by NUTS3 region (counties). All panels size points by population. Lines of best fit are weighted by population.

M Data Description of County-Level Covariates

Table A16: Data Description of County-Level Covariates

Variable	Description	Data Source
Average Age	Average age of populaton, 2014	German Statistical Office
% Female Age	Share of population that is female, 2014	German Statistical Office
Pop. Density 2018	Population density, 2018.	Regionalatlas Deutschland
% Empty Flats	Share of flats that are vacant, 2017	Thünen-Landatlas
Average Income	Average income, 2018	Statistische Ämter des Bundes und der Länder (Federal and state statistical offices)
% Unemployed	Unemployment rate, 2014	Bundesagentur für Arbeit (Federal Employment Agency)
Train. Positions per Applicant	Number of training positions (Lehrstellen) per applicant (Auszubildender)	Bundesagentur für Arbeit (Federal Employment Agency)
Syrians Employed / in Train.	Number of Syrians employed or in training divided by Syrian population	Bundesagentur für Arbeit (Federal Employment Agency)
All Crimes 2014	Reported crimes (total) per population, 2014	Polizeiliche Kriminalstatistik (Police Crime Statistics)
Thefts 2014	Theft crimes per population, 2014	Polizeiliche Kriminalstatistik (Police Crime Statistics)
Violent crimes 2014	Violent crimes per population, 2014	Polizeiliche Kriminalstatistik (Police Crime Statistics)
% Christian	Number of Christians per population, 2011	Zensus Datenbank (Census Results)
% AfD 2014	Vote share Alternative für Deutschland (AfD), European elections, 2014, demeaned by state	Der Bundeswahlleiter (Federal Returning Officer)
% Voted 2014	Log turnout, European elections, 2014	Der Bundeswahlleiter (Federal Returning Officer)

Continued on next page

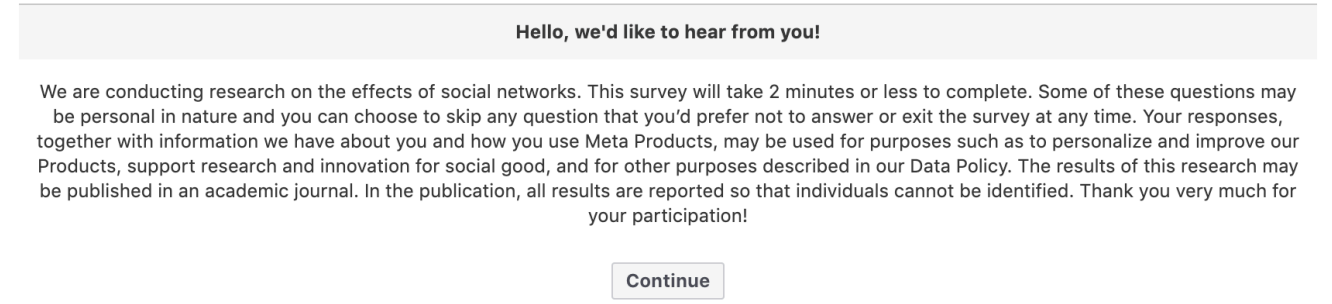
Table A16: Data Description of County-Level Covariates (Continued)

Variable	Description	Data Source
% Syrians 2010	Number of Syrians divided by population, 2010	German Statistical Office
% Syrians 2019	Number of Syrians divided by population, 2019	German Statistical Office
% Foreign 2010	Number of foreigners divided by population, 2010	German Statistical Office
% Foreign 2019	Number of foreigners divided by population, 2019	German Statistical Office
Integr. Courses per Syrian	Number of integration courses completed 2015-2019 per Syrian	Federal Office for Migration and Refugees
Pro-Immigr. Groups per Population	Number of groups affiliated with ProAsyl activist group per Population	ProAsyl (not publicly available, data received directly from organisation)
Integr. Sports Clubs per Syrian	Number of sports clubs that are part of Integration through Sport initiative	German Olympic Sports Confederation
Unemp. General Schools Teachers per Pop. 2014	Number of unemployed general school teachers divided by population, 2014	Bundesagentur für Arbeit (Federal Employment Agency) (not publicly available, data received directly from organisation)
Unemp. Higher Ed. School Teachers per Pop. 2014	Number of unemployed university and research institute teachers divided by population, 2014	Bundesagentur für Arbeit (Federal Employment Agency) (not publicly available, data received directly from organisation)
Unemp. Driving and Sports Teachers per Pop. 2014	Number of driving and sports teachers divided by population, 2014	Bundesagentur für Arbeit (Federal Employment Agency) (not publicly available, data received directly from organisation)
Unemp. Other School Teachers per Pop. 2014	Number of teachers in other education centers divided by population, 2014	Bundesagentur für Arbeit (Federal Employment Agency) (not publicly available, data received directly from organisation)

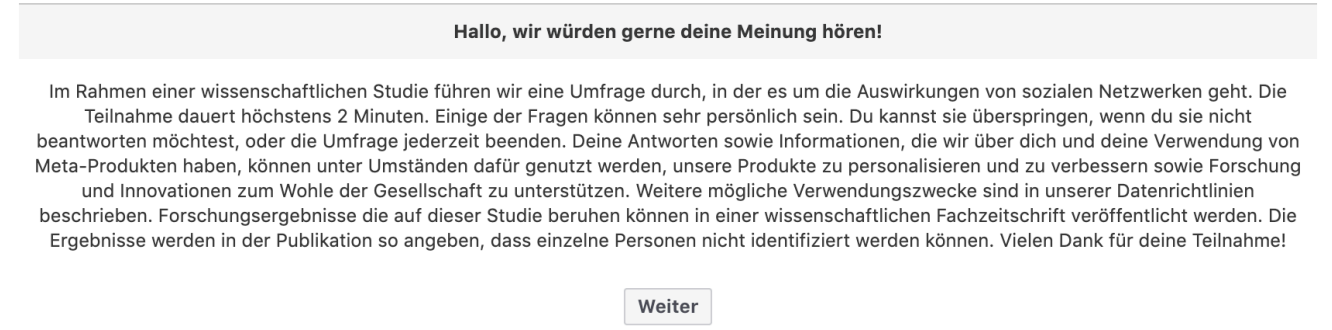
N Survey Screenshots

Figure A22: Survey Intro

(a) English



(b) German



(c) Arabic

Figure A23: Survey Question: Frequency of Social Interactions

(a) English

In the following we are going to ask you several questions about your interactions with the German population. By this, we mean individuals who have lived in Germany most of their lives.

In general, do you agree or disagree with the following statement: "I have many social interactions with Germans in the city I live in."

- ○ Strongly agree
- ○ Somewhat agree
- ○ Neither agree nor disagree
- ○ Somewhat disagree
- ○ Strongly disagree

(b) German

Im Folgenden stellen wir dir einige Fragen zu deinen Interaktionen mit der deutschen Bevölkerung. Damit meinen wir Personen, die den Großteil ihres Lebens in Deutschland verbracht haben.

Inwiefern stimmst du der folgenden Aussage zu: „Ich habe in der Stadt, in der ich wohne, viele soziale Interaktionen mit Deutschen."

- ○ Stimme völlig zu
- ○ Stimme eher zu
- ○ Keine Meinung
- ○ Stimme eher nicht zu
- ○ Stimme überhaupt nicht zu

(c) Arabic

سنطرح عليك فيما يلي عدة أسئلة حول تعاملاتك مع الشعب الألماني. ونعني بهذا الأفراد الذين عاشوا في ألمانيا معظم حياتهم.

بصفة عامة، هل توافق أم لا توافق على العبارة التالية: "لدي تعاملات اجتماعية كثيرة مع الألمان في المدينة التي أعيش فيها".

- أوافق بشدة ○
- أوافق نوعًا ما ○
- لست موافقًا ولا غير موافق ○
- لا أوافق نوعًا ما ○
- لا أوافق بشدة ○

Figure A24: Survey Question: Types of Interactions

(a) English

Which of the following interactions with Germans have you had in the past year? Please check all that apply.

- ☐ I have been invited to a German friend's home (for a dinner, a birthday party, etc.)
- ☐ I have invited a German friend to my home (for a dinner, a birthday party, etc.)
- ☐ I have gone to a restaurant, cafe, or bar with German friends
- ☐ I have been greeted on the street by German friends
- ☐ I have played sports with German friends

(b) German

Welche der folgenden Interaktionen mit Deutschen hattest du im letzten Jahr? Bitte wähle alle zutreffenden Antworten aus.

- ☐ Ein/e deutsche/r Freund/in hat mich zu sich nach Hause eingeladen (zum Abendessen, zu einer Geburtstagsfeier etc.)
- ☐ Ich habe eine/n deutschen Freund/in zu mir nach Hause eingeladen (zum Abendessen, zu einer Geburtstagsfeier etc.)
- ☐ Ich war mit deutschen Freunden in einem Restaurant, Café oder einer Bar
- ☐ Ich wurde auf der Straße von deutschen Freunden gegrüßt
- ☐ Ich habe mich mit deutschen Freunden zum Sport getroffen

(c) Arabic

أي من التعاملات التالية مع الألمان قمت بها خلال العام الماضي؟ يرجى تحديد كل الإجابات المناسبة.

- ☐ تمت دعوتي لمنزل صديق ألماني (لتناول وجبة طعام أو حضور حفلة عيد ميلاد، وما إلى ذلك)
- ☐ قمت بدعوة صديق ألماني إلى منزلي (لتناول وجبة طعام أو حضور حفلة عيد ميلاد، وما إلى ذلك)
- ☐ ذهبت إلى مطعم أو مقهى أو حانة مع أصدقاء ألمان
- ☐ تم الترحيب بي في الشارع من قبل أصدقاء ألمان
- ☐ مارست الرياضة مع أصدقاء ألمان

Figure A25: Survey Question: Effects of Social Integration

(a) English

Do you have German friends or acquaintances that have helped you or a member of your family? If so, please select all the ways in which they have helped.

- ☐ Finding a job
- ☐ Navigating the healthcare system (finding doctors, scheduling appointments, etc.)
- ☐ Finding an apartment or place to live
- ☐ Completing school work
- ☐ Navigating the bureaucracy (filling out official documents, identifying the right people to speak to, etc.)
- ☐ Finding language or integration courses

(b) German

Hast du deutsche Freunde oder Bekannte, die dir oder einem Mitglied deiner Familie bei etwas geholfen haben? Wenn ja, wähle bitte alle Dinge aus, bei denen dir geholfen wurde.

- ☐ Bei der Suche nach einem Job
- ☐ Beim Navigieren des Gesundheitssystems (Ärzte finden, Termine vereinbaren etc.)
- ☐ Bei der Suche nach einer Wohnung oder einem Ort zum Wohnen
- ☐ Bei Hausaufgaben (z. B. für die Schule oder Uni)
- ☐ Bei bürokratischen Angelegenheiten (offizielle Dokumente ausfüllen, richtige Ansprechpartner finden etc.)
- ☐ Bei der Suche nach Sprach- oder Integrationskursen

(c) Arabic

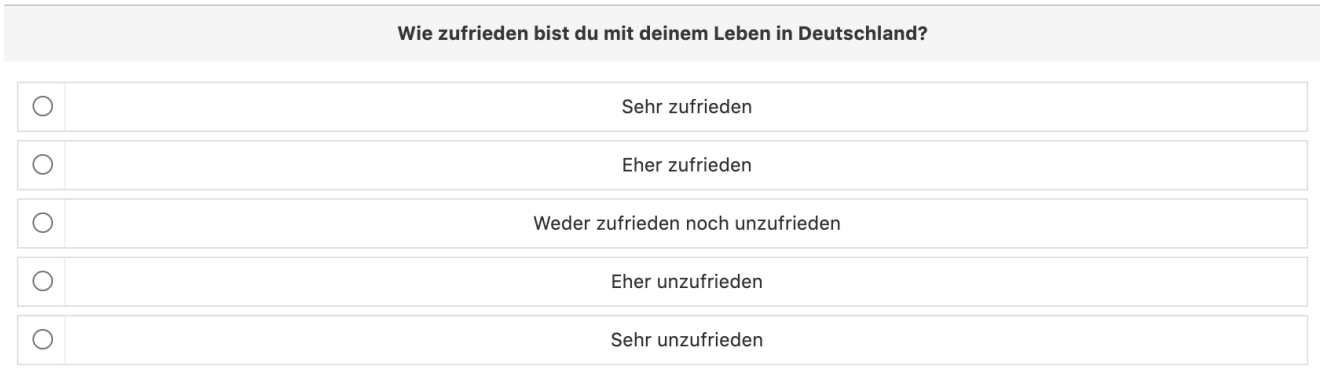
هل لديك أصدقاء أو معارف ألمان ساعدوك أو ساعدوا أحد أفراد عائلتك؟ إذا كان الأمر كذلك، يرجى تحديد جميع الطرق التي قدّموا بها المساعدة.

- ☐ البحث عن وظيفة
- ☐ التنقل ضمن نظام الرعاية الصحية (العثور على الأطباء وجدولة المواعيد، وما إلى ذلك)
- ☐ البحث عن شقة أو مكان للإقامة
- ☐ إكمال عمل مدرسي
- ☐ التعامل مع الإجراءات البيروقراطية (ملء المستندات الرسمية وتحديد الأشخاص المناسبين للتحدث معهم، وما إلى ذلك)
- ☐ البحث عن دورات تدريبية في اللغة أو الاندماج

Figure A26: Survey Question: Satisfaction in Germany

(a) English

(b) German



(c) Arabic

O Regional Measures of Integration and Friending

Table A17:

AGS	Name	NUTS3	Friending Integration	General Friendliness	Relative Friending	Language Integration
1001	Flensburg, Stadt	DEF01	5.53	84.1	0.0658	32.6
1002	Kiel, Landeshauptstadt	DEF02	5.86	119.3	0.0490	29.8
1003	Lübeck, Hansestadt	DEF03	5.05	106.2	0.0478	30.9
1004	Neumünster, Stadt	DEF04	3.46	84.8	0.0408	31.4
1051	Dithmarschen	DEF05	6.64	100.0	0.0665	33.5
1053	Herzogtum Lauenburg	DEF06	5.09	100.8	0.0504	29.1
1054	Nordfriesland	DEF07	5.28	92.9	0.0567	33.6
1055	Ostholstein	DEF08	5.44	88.0	0.0618	31.9
1056	Pinneberg	DEF09	4.24	106.4	0.0400	26.8
1057	Plön	DEF0A	4.62	102.0	0.0453	31.7
1058	Rendsburg-Eckernförde	DEF0B	4.21	109.6	0.0385	27.3
1059	Schleswig-Flensburg	DEF0C	4.87	101.4	0.0479	33.9
1060	Segeberg	DEF0D	4.49	105.4	0.0427	29.8
1061	Steinburg	DEF0E	3.90	96.4	0.0404	27.6
1062	Stormarn	DEF0F	5.39	110.3	0.0488	26.9
2000	Hamburg, Freie und Hansestadt	DE600	6.69	146.9	0.0456	30.3
3101	Braunschweig, Stadt	DE911	5.44	120.9	0.0451	33.2
3102	Salzgitter, Stadt	DE912	2.44	94.3	0.0259	22.2
3103	Wolfsburg, Stadt	DE913	5.51	87.6	0.0627	29.5
3151	Gifhorn	DE914	5.06	116.7	0.0432	29.2
3153	Goslar	DE916	4.31	88.3	0.0489	28.3
3154	Helmstedt	DE917	5.33	89.3	0.0594	26.7
3155	Northeim	DE918	6.63	113.0	0.0588	30.6
3157	Peine	DE91A	4.55	98.7	0.0460	30.7
3158	Wolfenbüttel	DE91B	7.78	101.9	0.0762	32.0
3159	Göttingen	DE91C	7.41	122.1	0.0607	38.7
3241	Region Hannover	DE929	5.78	126.3	0.0457	29.7
3251	Diepholz	DE922	7.27	121.0	0.0602	36.0
3252	Hameln-Pyrmont	DE923	6.16	104.9	0.0589	30.7
3254	Hildesheim	DE925	6.77	113.3	0.0596	36.1
3255	Holzminden	DE926	6.51	100.5	0.0648	28.3
3256	Nienburg (Weser)	DE927	11.65	124.9	0.0931	43.1
3257	Schaumburg	DE928	5.45	109.0	0.0502	31.2
3351	Celle	DE931	9.27	97.9	0.0949	43.7
3352	Cuxhaven	DE932	5.37	100.9	0.0533	33.6
3353	Harburg	DE933	6.03	114.1	0.0528	32.0

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Table A17: (Continued)

AGS	Name	NUTS3	Friending Integration	General Friendliness	Relative Friending	Language Integration
3354	Lüchow-Dannenberg	DE934	9.30	112.1	0.0834	32.0
3355	Lüneburg	DE935	5.79	110.4	0.0524	29.3
3356	Osterholz	DE936	4.48	118.4	0.0380	25.4
3357	Rotenburg (Wümme)	DE937	6.10	121.9	0.0500	32.6
3358	Heidekreis	DE938	7.48	101.5	0.0737	32.9
3359	Stade	DE939	4.68	129.3	0.0361	27.9
3360	Uelzen	DE93A	6.26	102.2	0.0612	28.0
3361	Verden	DE93B	6.95	118.6	0.0585	29.7
3401	Delmenhorst, Stadt	DE941	4.16	97.2	0.0428	26.3
3402	Emden, Stadt	DE942	8.94	117.9	0.0759	31.2
3403	Oldenburg (Oldenburg), Stadt	DE943	6.79	121.5	0.0559	31.3
3404	Osnabrück, Stadt	DE944	5.98	134.0	0.0444	27.0
3405	Wilhelmshaven, Stadt	DE945	3.54	79.4	0.0446	26.8
3451	Ammerland	DE946	8.54	134.1	0.0639	31.3
3452	Aurich	DE947	6.57	124.1	0.0533	31.9
3453	Cloppenburg	DE948	11.56	172.7	0.0669	44.3
3454	Emsland	DE949	7.24	153.6	0.0471	33.4
3455	Friesland	DE94A	5.57	91.9	0.0607	29.5
3456	Grafschaft Bentheim	DE94B	8.00	138.6	0.0576	34.7
3457	Leer	DE94C	4.61	125.6	0.0367	24.9
3458	Oldenburg	DE94D	6.74	110.5	0.0611	34.7
3459	Osnabrück	DE94E	6.67	141.0	0.0471	33.0
3460	Vechta	DE94F	6.03	156.2	0.0387	34.6
3461	Wesermarsch	DE94G	6.09	107.0	0.0569	21.4
3462	Wittmund	DE94H	6.76	116.7	0.0580	32.6
4011	Bremen, Stadt	DE501	5.23	122.2	0.0428	27.1
4012	Bremerhaven, Stadt	DE502	4.08	93.2	0.0439	25.0
5111	Düsseldorf, Stadt	DEA11	4.33	116.9	0.0370	30.5
5112	Duisburg, Stadt	DEA12	1.87	97.7	0.0192	23.3
5113	Essen, Stadt	DEA13	2.99	109.0	0.0275	24.6
5114	Krefeld, Stadt	DEA14	3.41	102.1	0.0334	29.1
5116	Mönchengladbach, Stadt	DEA15	2.92	103.9	0.0281	26.1
5117	Mülheim an der Ruhr, Stadt	DEA16	2.71	100.5	0.0270	24.9
5119	Oberhausen, Stadt	DEA17	1.75	91.7	0.0190	25.5
5120	Remscheid, Stadt	DEA18	2.91	105.1	0.0277	30.7
5122	Solingen, Klingenstadt	DEA19	3.11	96.8	0.0321	28.3
5124	Wuppertal, Stadt	DEA1A	2.84	112.2	0.0254	22.6
5154	Kleve	DEA1B	4.67	102.2	0.0456	28.5
5158	Mettmann	DEA1C	3.88	109.7	0.0353	28.0

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Table A17: (Continued)

AGS	Name	NUTS3	Friending Integration	General Friendliness	Relative Friending	Language Integration
5162	Rhein-Kreis Neuss	DEA1D	4.12	126.4	0.0326	31.0
5166	Viersen	DEA1E	5.93	103.5	0.0572	34.5
5170	Wesel	DEA1F	2.99	106.4	0.0281	25.7
5314	Bonn, Stadt	DEA22	7.24	134.3	0.0541	34.9
5315	Köln, Stadt	DEA23	6.68	152.9	0.0437	34.1
5316	Leverkusen, Stadt	DEA24	3.60	103.8	0.0347	31.0
5334	Städteregion Aachen	DEA2D	5.20	138.2	0.0378	33.7
5358	Düren	DEA26	4.11	132.9	0.0309	32.6
5362	Rhein-Erft-Kreis	DEA27	4.29	119.1	0.0359	33.2
5366	Euskirchen	DEA28	7.26	158.5	0.0457	33.6
5370	Heinsberg	DEA29	4.54	117.3	0.0388	32.2
5374	Oberbergischer Kreis	DEA2A	7.39	157.8	0.0469	35.5
5378	Rheinisch-Bergischer Kreis	DEA2B	5.83	131.9	0.0444	35.4
5382	Rhein-Sieg-Kreis	DEA2C	6.50	153.2	0.0424	34.9
5512	Bottrop, Stadt	DEA31	1.52	93.2	0.0163	25.7
5513	Gelsenkirchen, Stadt	DEA32	1.27	89.8	0.0141	18.5
5515	Münster, Stadt	DEA33	8.78	164.3	0.0536	32.7
5554	Borken	DEA34	6.66	149.8	0.0444	31.2
5558	Coesfeld	DEA35	5.13	142.6	0.0360	25.9
5562	Recklinghausen	DEA36	3.06	113.0	0.0271	24.5
5566	Steinfurt	DEA37	6.42	150.3	0.0427	28.1
5570	Warendorf	DEA38	5.40	135.0	0.0400	30.9
5711	Bielefeld, Stadt	DEA41	6.73	140.6	0.0482	33.8
5754	Gütersloh	DEA42	8.84	139.9	0.0630	36.5
5758	Herford	DEA43	6.46	125.6	0.0513	31.1
5762	Höxter	DEA44	6.70	135.8	0.0493	32.4
5766	Lippe	DEA45	5.95	129.0	0.0463	30.4
5770	Minden-Lübbecke	DEA46	6.87	134.6	0.0510	31.0
5774	Paderborn	DEA47	7.90	156.7	0.0505	32.7
5911	Bochum, Stadt	DEA51	3.47	118.5	0.0293	23.3
5913	Dortmund, Stadt	DEA52	3.52	116.8	0.0301	25.1
5914	Hagen, Stadt der FernUniversität	DEA53	2.50	103.5	0.0241	24.3
5915	Hamm, Stadt	DEA54	3.43	106.0	0.0324	25.7
5916	Herne, Stadt	DEA55	1.79	89.2	0.0200	22.0
5954	Ennepe-Ruhr-Kreis	DEA56	3.84	109.0	0.0352	29.3
5958	Hochsauerlandkreis	DEA57	8.18	170.0	0.0480	31.2
5962	Märkischer Kreis	DEA58	4.37	134.8	0.0325	26.3
5966	Olpe	DEA59	7.60	221.9	0.0344	31.8
5970	Siegen-Wittgenstein	DEA5A	6.23	187.3	0.0333	32.2

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Table A17: (Continued)

AGS	Name	NUTS3	Friending Integration	General Friendliness	Relative Friending	Language Integration
5974	Soest	DEA5B	6.69	140.8	0.0476	29.7
5978	Unna	DEA5C	3.26	109.6	0.0297	25.5
6411	Darmstadt, Wissenschaftsstadt	DE711	5.15	139.7	0.0369	35.8
6412	Frankfurt am Main, Stadt	DE712	7.53	147.7	0.0509	36.6
6413	Offenbach am Main, Stadt	DE713	4.96	138.9	0.0357	32.2
6414	Wiesbaden, Landeshauptstadt	DE714	7.91	141.2	0.0560	40.8
6431	Bergstraße	DE715	4.89	151.0	0.0324	35.7
6432	Darmstadt-Dieburg	DE716	3.73	143.3	0.0260	32.3
6433	Groß-Gerau	DE717	4.35	140.0	0.0311	31.4
6434	Hochtaunuskreis	DE718	5.98	140.6	0.0425	34.8
6435	Main-Kinzig-Kreis	DE719	5.11	162.2	0.0314	31.6
6436	Main-Taunus-Kreis	DE71A	5.42	124.6	0.0434	34.7
6437	Odenwaldkreis	DE71B	5.57	170.6	0.0326	45.7
6438	Offenbach	DE71C	4.48	138.8	0.0324	32.0
6439	Rheingau-Taunus-Kreis	DE71D	5.00	142.9	0.0351	36.0
6440	Wetteraukreis	DE71E	8.22	163.3	0.0505	36.3
6531	Gießen	DE721	7.62	177.2	0.0430	34.8
6532	Lahn-Dill-Kreis	DE722	4.62	195.6	0.0235	27.4
6533	Limburg-Weilburg	DE723	7.39	179.3	0.0411	32.6
6534	Marburg-Biedenkopf	DE724	7.17	193.7	0.0370	35.7
6535	Vogelsbergkreis	DE725	6.51	194.2	0.0335	36.2
6611	Kassel, documenta-Stadt	DE731	5.46	134.4	0.0406	27.9
6631	Fulda	DE732	7.66	203.5	0.0375	32.1
6632	Hersfeld-Rotenburg	DE733	8.04	164.2	0.0489	35.0
6633	Kassel	DE734	4.39	153.9	0.0286	32.2
6634	Schwalm-Eder-Kreis	DE735	7.86	157.9	0.0497	32.4
6635	Waldeck-Frankenberg	DE736	6.59	178.0	0.0370	30.2
6636	Werra-Meißner-Kreis	DE737	8.47	143.9	0.0587	33.0
7111	Koblenz, kreisfreie Stadt	DEB11	7.43	158.8	0.0468	36.3
7131	Ahrweiler	DEB12	7.98	163.9	0.0487	37.7
7132	Altenkirchen (Westerwald)	DEB13	6.03	190.0	0.0318	32.4
7133	Bad Kreuznach	DEB14	7.39	176.4	0.0417	40.0
7134	Birkenfeld	DEB15	7.99	185.8	0.0430	35.0
7135	Cochem-Zell	DEB1C	8.60	182.6	0.0472	41.7
7137	Mayen-Koblenz	DEB17	9.11	192.2	0.0473	39.5
7138	Neuwied	DEB18	6.80	162.8	0.0417	35.2
7140	Rhein-Hunsrück-Kreis	DEB1D	7.36	177.7	0.0415	31.4
7141	Rhein-Lahn-Kreis	DEB1A	6.93	174.9	0.0394	32.1
7143	Westerwaldkreis	DEB1B	7.50	198.1	0.0379	30.4

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Table A17: (Continued)

AGS	Name	NUTS3	Friending Integration	General Friendliness	Relative Friending	Language Integration
7211	Trier, kreisfreie Stadt	DEB21	7.78	132.5	0.0585	39.4
7231	Bernkastel-Wittlich	DEB22	7.85	168.8	0.0465	35.7
7232	Eifelkreis Bitburg-Prüm	DEB23	8.68	184.6	0.0472	35.5
7233	Vulkaneifel	DEB24	12.78	192.9	0.0661	36.8
7235	Trier-Saarburg	DEB25	9.33	173.1	0.0539	45.1
7311	Frankenthal (Pfalz), kreisfreie Stadt	DEB31	3.55	107.3	0.0331	31.8
7312	Kaiserslautern, kreisfreie Stadt	DEB32	4.13	121.3	0.0340	32.2
7313	Landau in der Pfalz, kreisfreie Stadt	DEB33	6.25	144.2	0.0434	34.9
7314	Ludwigshafen am Rhein, kreisfreie Stadt	DEB34	3.67	117.6	0.0312	30.9
7315	Mainz, kreisfreie Stadt	DEB35	6.79	148.5	0.0458	39.9
7316	Neustadt an der Weinstraße, kreisfreie Stadt	DEB36	5.96	137.3	0.0435	33.0
7317	Pirmasens, kreisfreie Stadt	DEB37	6.06	152.1	0.0397	34.1
7318	Speyer, kreisfreie Stadt	DEB38	6.72	124.9	0.0539	34.3
7319	Worms, kreisfreie Stadt	DEB39	4.66	136.4	0.0343	32.6
7320	Zweibrücken, kreisfreie Stadt	DEB3A	4.98	134.4	0.0371	31.8
7331	Alzey-Worms	DEB3B	5.57	155.4	0.0357	30.5
7332	Bad Dürkheim	DEB3C	4.65	139.1	0.0334	36.4
7333	Donnersbergkreis	DEB3D	7.58	145.2	0.0520	36.0
7334	Germersheim	DEB3E	4.48	154.0	0.0290	30.1
7335	Kaiserslautern	DEB3F	5.59	165.0	0.0338	35.6
7336	Kusel	DEB3G	7.52	172.3	0.0437	32.4
7337	Südliche Weinstraße	DEB3H	6.12	177.0	0.0345	40.0
7338	Rhein-Pfalz-Kreis	DEB3I	4.22	154.0	0.0274	34.1
7339	Mainz-Bingen	DEB3J	5.76	168.2	0.0343	33.9
7340	Südwestpfalz	DEB3K	8.29	192.1	0.0431	43.5
8111	Stuttgart, Stadtkreis	DE111	6.73	153.6	0.0438	33.3
8115	Böblingen	DE112	5.02	132.1	0.0379	32.8
8116	Esslingen	DE113	5.07	137.0	0.0369	32.0
8117	Göppingen	DE114	5.21	130.1	0.0400	35.1
8118	Ludwigsburg	DE115	5.10	128.4	0.0397	31.3
8119	Rems-Murr-Kreis	DE116	5.04	141.7	0.0355	29.5
8121	Heilbronn, Stadtkreis	DE117	5.55	127.5	0.0435	35.3
8125	Heilbronn	DE118	4.72	151.5	0.0312	31.6
8126	Hohenlohekreis	DE119	7.71	157.5	0.0493	35.5
8127	Schwäbisch Hall	DE11A	11.51	149.7	0.0771	36.7
8128	Main-Tauber-Kreis	DE11B	9.20	150.6	0.0611	35.7
8135	Heidenheim	DE11C	4.80	119.3	0.0404	35.9
8136	Ostalbkreis	DE11D	6.17	153.9	0.0401	29.1
8211	Baden-Baden, Stadtkreis	DE121	8.55	132.6	0.0644	34.6

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Table A17: (Continued)

AGS	Name	NUTS3	Friending Integration	General Friendliness	Relative Friending	Language Integration
8212	Karlsruhe, Stadtkreis	DE122	6.42	147.8	0.0433	40.6
8215	Karlsruhe	DE123	5.86	157.4	0.0371	33.1
8216	Rastatt	DE124	5.57	151.8	0.0366	36.1
8221	Heidelberg, Stadtkreis	DE125	7.67	145.1	0.0529	40.9
8222	Mannheim, Stadtkreis	DE126	6.12	139.8	0.0438	37.7
8225	Neckar-Odenwald-Kreis	DE127	6.45	167.2	0.0384	34.6
8226	Rhein-Neckar-Kreis	DE128	6.17	157.2	0.0391	36.2
8231	Pforzheim, Stadtkreis	DE129	4.31	126.8	0.0341	31.4
8235	Calw	DE12A	5.52	133.6	0.0413	34.1
8236	Enzkreis	DE12B	4.22	146.1	0.0288	32.7
8237	Freudenstadt	DE12C	5.38	133.8	0.0400	34.0
8311	Freiburg im Breisgau, Stadtkreis	DE131	10.30	151.4	0.0679	40.1
8315	Breisgau-Hochschwarzwald	DE132	6.22	150.9	0.0410	31.4
8316	Emmendingen	DE133	6.03	156.0	0.0386	32.2
8317	Ortenaukreis	DE134	5.95	159.5	0.0372	33.3
8325	Rottweil	DE135	6.04	147.8	0.0411	30.5
8326	Schwarzwald-Baar-Kreis	DE136	5.25	138.9	0.0378	29.6
8327	Tuttlingen	DE137	4.91	146.6	0.0336	30.8
8335	Konstanz	DE138	6.38	136.4	0.0466	34.7
8336	Lörrach	DE139	4.71	128.8	0.0365	28.2
8337	Waldshut	DE13A	5.44	128.6	0.0422	29.5
8415	Reutlingen	DE141	5.66	143.2	0.0396	31.1
8416	Tübingen	DE142	6.87	151.6	0.0455	32.9
8417	Zollernalbkreis	DE143	5.09	141.5	0.0360	41.6
8421	Ulm, Stadtkreis	DE144	5.78	117.5	0.0492	37.6
8425	Alb-Donau-Kreis	DE145	4.78	131.6	0.0364	37.1
8426	Biberach	DE146	6.91	157.5	0.0438	30.8
8435	Bodenseekreis	DE147	6.03	144.9	0.0418	29.3
8436	Ravensburg	DE148	7.49	161.4	0.0463	31.5
8437	Sigmaringen	DE149	8.18	158.7	0.0514	29.2
9161	Ingolstadt	DE211	6.06	116.9	0.0517	38.4
9162	München, Landeshauptstadt	DE212	7.44	141.7	0.0524	37.0
9163	Rosenheim	DE213	5.05	127.9	0.0395	45.5
9171	Altötting	DE214	8.16	143.7	0.0568	36.5
9172	Berchtesgadener Land	DE215	7.24	134.8	0.0537	54.4
9173	Bad Tölz-Wolfratshausen	DE216	7.11	145.7	0.0489	42.6
9174	Dachau	DE217	6.04	128.7	0.0469	32.8
9175	Ebersberg	DE218	3.93	103.7	0.0380	42.0
9176	Eichstätt	DE219	11.98	145.8	0.0821	44.9

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Table A17: (Continued)

AGS	Name	NUTS3	Friending Integration	General Friendliness	Relative Friending	Language Integration
9177	Erding	DE21A	6.74	122.0	0.0552	43.8
9178	Freising	DE21B	5.30	111.6	0.0476	44.4
9179	Fürstenfeldbruck	DE21C	5.63	127.2	0.0442	42.9
9180	Garmisch-Partenkirchen	DE21D	6.17	147.5	0.0419	48.9
9181	Landsberg am Lech	DE21E	8.70	124.8	0.0698	46.5
9182	Miesbach	DE21F	7.16	145.6	0.0492	40.2
9183	Mühldorf a.Inn	DE21G	6.46	156.2	0.0414	35.4
9184	München	DE21H	7.25	139.0	0.0521	38.8
9185	Neuburg-Schrobenhausen	DE21I	5.35	130.2	0.0411	32.8
9186	Pfaffenhofen a.d.Ilm	DE21J	6.09	121.5	0.0499	39.4
9187	Rosenheim	DE21K	6.99	157.5	0.0445	40.4
9188	Starnberg	DE21L	5.15	117.0	0.0440	41.4
9189	Traunstein	DE21M	7.66	163.4	0.0467	39.9
9190	Weilheim-Schongau	DE21N	6.40	136.2	0.0466	34.2
9261	Landshut	DE221	8.42	123.8	0.0684	41.8
9262	Passau	DE222	8.89	120.9	0.0733	41.0
9263	Straubing	DE223	8.48	80.6	0.1053	41.9
9271	Deggendorf	DE224	6.30	143.8	0.0439	39.2
9272	Freyung-Grafenau	DE225	11.09	175.8	0.0628	48.2
9273	Kelheim	DE226	4.88	126.9	0.0384	33.3
9274	Landshut	DE227	6.32	147.5	0.0428	38.8
9275	Passau	DE228	4.19	157.8	0.0265	27.2
9276	Regen	DE229	6.32	169.5	0.0374	28.8
9277	Rottal-Inn	DE22A	6.16	164.9	0.0374	30.2
9278	Straubing-Bogen	DE22B	5.32	148.3	0.0358	37.7
9279	Dingolfing-Landau	DE22C	4.96	142.4	0.0347	41.3
9361	Amberg	DE231	5.67	92.9	0.0612	33.7
9362	Regensburg	DE232	6.40	104.3	0.0616	34.2
9363	Weiden i.d.OPf.	DE233	3.19	84.4	0.0378	31.8
9371	Amberg-Sulzbach	DE234	4.09	115.8	0.0353	30.2
9372	Cham	DE235	8.31	170.8	0.0488	43.4
9373	Neumarkt i.d.OPf.	DE236	7.25	133.6	0.0544	37.1
9374	Neustadt a.d.Waldnaab	DE237	4.22	135.1	0.0311	22.5
9375	Regensburg	DE238	5.39	138.7	0.0390	30.7
9376	Schwandorf	DE239	7.96	125.7	0.0632	36.1
9377	Tirschenreuth	DE23A	3.00	136.0	0.0220	30.7
9461	Bamberg	DE241	6.58	103.4	0.0639	37.6
9462	Bayreuth	DE242	5.19	96.1	0.0541	38.6
9463	Coburg	DE243	5.76	90.6	0.0635	37.8

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Table A17: (Continued)

AGS	Name	NUTS3	Friending Integration	General Friendliness	Relative Friending	Language Integration
9464	Hof	DE244	1.40	82.9	0.0168	26.2
9471	Bamberg	DE245	2.97	126.2	0.0235	32.8
9472	Bayreuth	DE246	0.50	132.2	0.0037	23.7
9473	Coburg	DE247	6.25	121.0	0.0519	39.3
9474	Forchheim	DE248	3.59	111.5	0.0323	29.2
9475	Hof	DE249	2.75	116.8	0.0236	37.7
9476	Kronach	DE24A	9.77	138.8	0.0704	35.8
9477	Kulmbach	DE24B	8.21	107.1	0.0767	36.4
9478	Lichtenfels	DE24C	3.99	128.0	0.0311	34.9
9479	Wunsiedel i.Fichtelgebirge	DE24D	4.76	108.0	0.0440	35.8
9561	Ansbach	DE251	5.32	93.2	0.0571	33.4
9562	Erlangen	DE252	6.81	127.5	0.0535	35.1
9563	Fürth	DE253	4.33	96.8	0.0447	29.9
9564	Nürnberg	DE254	5.64	117.2	0.0481	35.4
9565	Schwabach	DE255	2.58	47.8	0.0540	51.1
9571	Ansbach	DE256	5.37	126.7	0.0424	39.4
9572	Erlangen-Höchstadt	DE257	5.10	110.6	0.0461	34.4
9573	Fürth	DE258	3.19	99.4	0.0322	37.2
9574	Nürnberger Land	DE259	4.90	105.1	0.0464	35.1
9575	Neustadt a.d.Aisch-Bad Windsheim	DE25A	8.15	109.6	0.0744	41.1
9576	Roth	DE25B	5.39	110.9	0.0487	36.0
9577	Weißenburg-Gunzenhausen	DE25C	4.48	128.3	0.0348	28.5
9661	Aschaffenburg	DE261	4.63	151.3	0.0307	27.5
9662	Schweinfurt	DE262	4.30	89.8	0.0480	31.0
9663	Würzburg	DE263	8.05	114.4	0.0706	41.1
9671	Aschaffenburg	DE264	4.54	176.2	0.0258	30.6
9672	Bad Kissingen	DE265	6.40	151.9	0.0421	30.4
9673	Rhön-Grabfeld	DE266	7.87	176.1	0.0446	34.0
9674	Haßberge	DE267	8.26	122.1	0.0677	43.3
9675	Kitzingen	DE268	5.68	120.4	0.0472	31.8
9676	Miltenberg	DE269	4.67	184.7	0.0254	25.5
9677	Main-Spessart	DE26A	6.93	169.8	0.0410	37.8
9678	Schweinfurt	DE26B	6.20	137.8	0.0450	31.5
9679	Würzburg	DE26C	5.60	147.8	0.0379	29.8
9761	Augsburg	DE271	6.42	133.5	0.0480	36.9
9762	Kaufbeuren	DE272	4.42	110.3	0.0401	35.5
9763	Kempten (Allgäu)	DE273	7.97	121.4	0.0656	34.5
9764	Memmingen	DE274	6.39	102.3	0.0625	30.9
9771	Aichach-Friedberg	DE275	6.84	145.0	0.0473	32.3

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Table A17: (Continued)

AGS	Name	NUTS3	Friending Integration	General Friendliness	Relative Friending	Language Integration
9772	Augsburg	DE276	5.07	135.7	0.0372	31.1
9773	Dillingen a.d.Donau	DE277	6.26	140.4	0.0444	37.0
9774	Günzburg	DE278	4.61	137.3	0.0336	36.3
9775	Neu-Ulm	DE279	4.37	112.5	0.0389	33.8
9776	Lindau (Bodensee)	DE27A	6.92	125.4	0.0552	34.2
9777	Ostallgäu	DE27B	8.06	142.5	0.0564	35.7
9778	Unterallgäu	DE27C	6.81	145.5	0.0468	38.4
9779	Donau-Ries	DE27D	6.05	147.9	0.0409	27.3
9780	Oberallgäu	DE27E	8.07	141.6	0.0569	36.0
10041	Regionalverband Saarbrücken	DEC01	6.15	184.0	0.0334	27.9
10042	Merzig-Wadern	DEC02	12.39	175.5	0.0707	39.3
10043	Neunkirchen	DEC03	6.84	196.5	0.0347	31.3
10044	Saarlouis	DEC04	8.69	196.8	0.0444	35.9
10045	Saarpfalz-Kreis	DEC05	6.59	187.7	0.0354	30.8
10046	St. Wendel	DEC06	9.61	197.4	0.0486	32.6
11000	Berlin, Stadt	DE300	7.84	156.4	0.0503	27.0
12051	Brandenburg an der Havel, Stadt	DE401	6.27	105.8	0.0592	27.6
12052	Cottbus, Stadt	DE402	3.58	97.7	0.0364	24.9
12053	Frankfurt (Oder), Stadt	DE403	5.74	89.6	0.0641	29.4
12054	Potsdam, Stadt	DE404	6.71	132.1	0.0509	29.3
12060	Barnim	DE405	5.79	115.1	0.0503	26.0
12061	Dahme-Spreewald	DE406	6.16	114.3	0.0537	24.8
12062	Elbe-Elster	DE407	8.29	104.4	0.0793	32.3
12063	Havelland	DE408	4.91	109.8	0.0449	26.9
12064	Märkisch-Oderland	DE409	5.12	109.7	0.0468	24.3
12065	Oberhavel	DE40A	5.41	107.5	0.0503	26.3
12066	Oberspreewald-Lausitz	DE40B	6.68	96.5	0.0692	32.4
12067	Oder-Spree	DE40C	6.29	116.5	0.0541	26.6
12068	Ostprignitz-Ruppin	DE40D	7.94	97.0	0.0819	34.9
12069	Potsdam-Mittelmark	DE40E	5.57	113.6	0.0491	27.6
12070	Prignitz	DE40F	5.96	91.9	0.0648	30.6
12071	Spree-Neiße	DE40G	5.36	98.7	0.0544	28.1
12072	Teltow-Fläming	DE40H	6.31	104.6	0.0603	28.7
12073	Uckermark	DE40I	3.94	95.8	0.0410	30.1
13003	Rostock	DE803	6.73	96.1	0.0701	29.0
13004	Schwerin	DE804	5.97	100.5	0.0595	25.7
13071	Mecklenburgische Seenplatte	DE80J	7.55	103.8	0.0727	33.4
13072	Landkreis Rostock	DE80K	6.67	97.3	0.0683	33.2
13073	Vorpommern-Rügen	DE80L	4.99	78.6	0.0636	32.0

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Table A17: (Continued)

AGS	Name	NUTS3	Friending Integration	General Friendliness	Relative Friending	Language Integration
13074	Nordwestmecklenburg	DE80M	4.91	90.3	0.0546	32.1
13075	Vorpommern-Greifswald	DE80N	7.46	97.7	0.0764	30.8
13076	Ludwigslust-Parchim	DE80O	6.26	99.8	0.0626	29.8
14511	Chemnitz, Stadt	DED41	5.39	113.1	0.0477	30.4
14521	Erzgebirgskreis	DED42	4.11	116.6	0.0353	24.0
14522	Mittelsachsen	DED43	5.18	108.0	0.0480	30.5
14523	Vogtlandkreis	DED44	5.75	113.9	0.0504	32.8
14524	Zwickau	DED45	6.36	106.4	0.0599	33.1
14612	Dresden, Stadt	DED21	7.84	125.8	0.0621	29.9
14625	Bautzen	DED2C	7.05	116.6	0.0605	31.7
14626	Görlitz	DED2D	7.06	94.2	0.0748	36.2
14627	Meißen	DED2E	6.92	112.3	0.0614	31.0
14628	Sächsische Schweiz-Osterzgebirge	DED2F	4.94	106.0	0.0468	30.9
14713	Leipzig, Stadt	DED51	6.21	110.6	0.0563	28.9
14729	Leipzig	DED52	4.54	104.8	0.0433	23.2
14730	Nordsachsen	DED53	4.66	103.9	0.0449	27.7
15001	Dessau-Roßlau, Stadt	DEE01	5.57	86.1	0.0648	30.2
15002	Halle (Saale), Stadt	DEE02	5.10	85.5	0.0598	28.1
15003	Magdeburg, Landeshauptstadt	DEE03	5.74	105.3	0.0545	27.1
15081	Altmarkkreis Salzwedel	DEE04	4.28	100.9	0.0424	34.5
15082	Anhalt-Bitterfeld	DEE05	5.33	84.4	0.0630	29.5
15083	Börde	DEE07	5.26	102.5	0.0512	26.3
15084	Burgenlandkreis	DEE08	7.87	101.3	0.0774	34.8
15085	Harz	DEE09	9.03	98.6	0.0914	33.6
15086	Jerichower Land	DEE06	3.72	101.8	0.0366	18.1
15087	Mansfeld-Südharz	DEE0A	7.15	98.5	0.0726	26.7
15088	Saalekreis	DEE0B	3.38	90.0	0.0375	24.5
15089	Salzlandkreis	DEE0C	7.80	98.7	0.0793	28.8
15090	Stendal	DEE0D	7.74	107.6	0.0721	29.8
15091	Wittenberg	DEE0E	6.02	96.7	0.0622	27.8
16051	Erfurt, Stadt	DEG01	6.44	98.3	0.0658	32.2
16052	Gera, Stadt	DEG02	4.74	78.5	0.0605	29.6
16053	Jena, Stadt	DEG03	7.15	108.6	0.0659	35.9
16054	Suhl, Stadt	DEG04	8.71	104.9	0.0827	46.0
16055	Weimar, Stadt	DEG05	7.27	89.2	0.0813	37.5
16056	Eisenach, Stadt	DEG0N	4.17	79.0	0.0528	25.5
16061	Eichsfeld	DEG06	6.78	141.9	0.0478	28.5
16062	Nordhausen	DEG07	6.50	108.5	0.0601	33.5
16063	Wartburgkreis	DEG0P	10.04	140.5	0.0717	39.5

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Table A17: (Continued)

AGS	Name	NUTS3	Friending Integration	General Friendliness	Relative Friending	Language Integration
16064	Unstrut-Hainich-Kreis	DEG09	5.42	100.4	0.0540	30.3
16065	Kyffhäuserkreis	DEG0A	3.57	108.1	0.0330	28.2
16066	Schmalkalden-Meiningen	DEG0B	7.23	134.4	0.0537	36.3
16067	Gotha	DEG0C	5.82	108.2	0.0537	34.3
16068	Sömmerda	DEG0D	3.98	107.9	0.0369	37.5
16069	Hildburghausen	DEG0E	9.14	114.5	0.0800	37.7
16070	Ilm-Kreis	DEG0F	6.39	117.2	0.0545	29.8
16071	Weimarer Land	DEG0G	4.71	106.4	0.0443	32.8
16072	Sonneberg	DEG0H	3.07	107.9	0.0285	27.8
16073	Saalfeld-Rudolstadt	DEG0I	8.19	107.0	0.0768	31.9
16074	Saale-Holzland-Kreis	DEG0J	4.20	108.7	0.0386	31.6
16075	Saale-Orla-Kreis	DEG0K	6.83	108.7	0.0631	30.3
16076	Greiz	DEG0L	8.61	110.6	0.0778	27.9
16077	Altenburger Land	DEG0M	5.93	83.0	0.0715	33.5

Note: Table shows county-level estimates. Friending integration is the measure mapped in Figure 2. General friendliness is the measure mapped in panel (a) of Figure 4. Relative friending is the measure mapped in panel (b) of Figure 4. Language integration is the share of Syrian migrants on Facebook who produce German content. Because of privacy restrictions, the estimates in this table may differ in small ways from those used to produce results in the paper.