

Peer Effects in Product Adoption

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Introduction

- ▶ Peer interactions important driver of product adoption decisions
 - ▶ 401(k) participation; mortgage refinancing
- ▶ Specific nature of peer effects central to implications
 - ▶ Extra demand or retiming of future demand?
 - ▶ Characteristics of influential individuals? Correlation with price sensitivity?
 - ▶ Peer effects concentrated on product purchased by friends, or positive or negative spill-overs to competing products?
- ▶ **This project:** Explores these and other questions about peer effects in the market for phone purchases

Approach in this paper

- ▶ **Measurement Challenge:** Need to observe both peers and consumption or product adoption decisions in the same data set.
 - ▶ Anonymized data from Facebook to measure peers as well as product adoption from phone purchases of mobile users.
- ▶ **Identification Challenge:** Homophily \rightarrow common shocks & preferences \rightarrow Correlated Behavior \neq peer effects.
 - ▶ Exploiting quasi-random variation in peers purchasing phones induced by (i) breaking/losing phones, (ii) contract renewals.

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 - ▶ Exploiting quasi-random variation in peers purchasing phones induced by (i) breaking/losing phones, (ii) contract renewals.
- ▶ **Baseline Research Questions:**
 1. Are people more likely to buy **any** new phone if a friend recently bought a new phone?
 2. Are people more likely to buy a **specific** new phone if a friend recently bought that specific new phone?

Data Description

- ▶ Anonymized network data from Facebook
- ▶ Information on phones from mobile-active users
 - ▶ Phone model & carrier registered when logging into mobile app
 - ▶ Identify switches to new phones
- ▶ Unit of observation: Person-week
- ▶ Pool across weeks 2016-19, 2016-20, 2016-21, and 2016-22
 - ▶ Not close to major device release dates or shopping holidays

Research Design - Phone Purchase

$$\mathbb{1}(\text{BuysPhone})_{i,t} = \beta \text{FriendsBuyPhone}_{i,t-1} + \gamma X_{i,t} + \varepsilon_{i,t}$$

- ▶ **Identification challenges** (result of homophily):
 - ▶ Correlated preferences
 - ▶ Correlated shocks

 - ▶ **Our Approach:** Find instruments for *FriendsBuyPhone* that
 1. “Quasi-randomly” shifts probability of friends buying
 2. Does not affect own probability of buying, except through peer effects.
- Random phone loss
- Friends up for contract renewal

Instrument 1: “Random Phone Loss”

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- ▶ Use public posts on Facebook that signal “random” loss of phone.

13 November 2017 · 🌐

Phone stolen, contact me here if you need to reach me

👍👎👏 16 8 Comments

👍 Like 💬 Comment ➦ Share

31 December 2017 at 17:42 · Redding · 🌐

Well, my iphone took a tumble today and the screen shattered. So naturally, I am now the proud owner of an iphone x 😊

👍👎👏 7 7 Comments

👍 Like 💬 Comment ➦ Share

11 January at 16:01 · 🌐

Phone broke get my new 1 saturday!

👍 1

👍 Like 💬 Comment ➦ Share

16 January at 05:11 · 🌐

Phone broken...Ordered a new one but if anyone needs me urgently, call Joe. If not urgent, send me a message on FB.

👍 2

👍 Like 💬 Comment ➦ Share

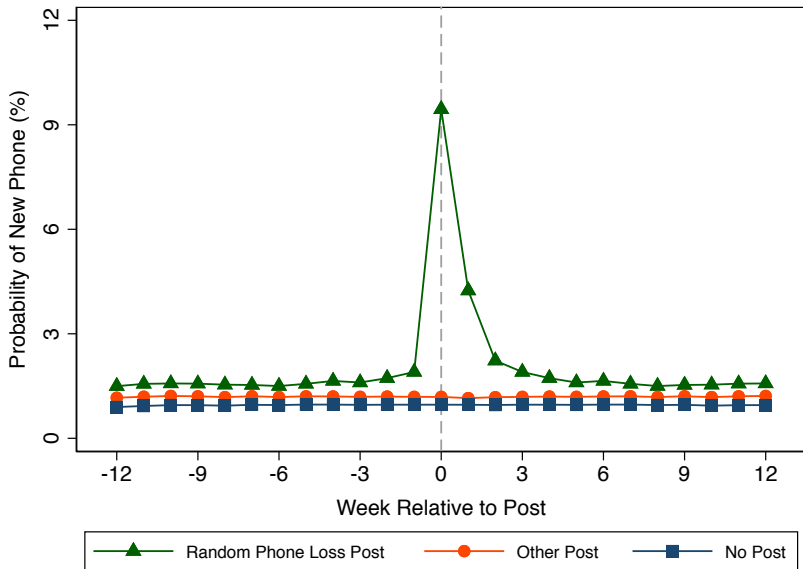
Instrument 1: “Random Phone Loss”

- ▶ Identify public posts on Facebook that signal “random” loss of phone.
- ▶ **Approach:** Word Embeddings & Convolutional Neural Networks
 - ▶ Neural network trained on about 15k hand-classified posts.

Instrument 1: “Random Phone Loss”

- ▶ Identify public posts on Facebook that signal “random” loss of phone.
- ▶ **Approach:** Word Embeddings & Convolutional Neural Networks
 - ▶ Neural network trained on about 15k hand-classified posts.
 - ▶ Advantages relative to regular expression search
 - ▶ Remove some **false positives**:
 - *“So...I dropped my phone in the toilet yesterday...!! Still works tho!!”*
 - ▶ Discover some **false negatives**:
 - *“R.I.P phone. You will be missed.”*
 - *“ughh... water + phone = new phone time.*
 - *“Long story short, my phone tried to light my house on fire last night and you’ll have to reach me on here for a while.”*
- ▶ Identify ~330,000 posts about “random phone loss”

Instrument 1: "Random Phone Loss" – First Stage



Results - Phone Purchase

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Controls in $X_{i,t}$ include:

- ▶ **User** characteristics FE:
age bucket \times gender \times education \times state \times week
- ▶ **Device** characteristics FE:
device \times carrier \times phone age bucket \times week
- ▶ **Friends** characteristics FE:
number of friends \times friends switching phones in last 6 months
 \times week
- ▶ Linear controls for
 - ▶ Individual probability of buying a new phone
 - ▶ Average purchase probability among friends
 - ▶ Individual and friend posting behavior (random phone loss instrument)

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	OLS	Second Stage	
	(1)	DV: Prob Buys New Phone (%)	
		(2)	(3)
		Broken Phone	Contract Threshold
# of Friends Buying (t-1 and t)	0.034*** (0.000)	0.041*** (0.005)	0.026** (0.013)
Controls + Fixed Effects	Y	Y	Y
Mean Dependent Variable	0.95	0.95	0.95
Number of Observations	335m	335m	335m
F-Statistic Instrument		339,156	55,592

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- ▶ ↑ 1 Friend Buys Phone → ↑ P(Buy Phone Next Week) by 0.04ppt

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- ▶ $\uparrow 1$ Friend Buys Phone $\rightarrow \uparrow$ P(Buy Phone Next Week) by 0.04ppt
- ▶ Effect not driven by family members
- ▶ Not caused by advertising responding to instrument

Results - Phone Purchase

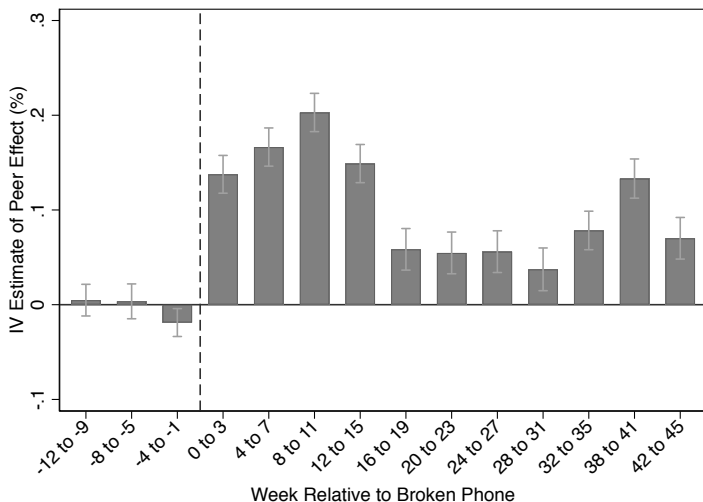
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- ▶ $\uparrow 1$ Friend Buys Phone $\rightarrow \uparrow P(\text{Buy Phone Next Week})$ by 0.04ppt
- ▶ OLS \approx IV: Common shocks/preferences less problematic at short horizon (conditional on controls)?
- ▶ Different instruments identified off of different individuals

Timing of Peer Effect: New Demand or Pulling Forward?

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- ▶ No evidence of a pre-trend, or reversal over 10 months.
- ▶ Implication: Value of customer > Direct effect on profit.

Specific Phone Purchase - Motivation

- ▶ So far: Effect of friends purchasing **any phone** on own probability of purchasing any phone.
- ▶ Next: Effect of friends purchasing **a specific brand of phone** (e.g., iPhone) on own probability of purchasing
 1. That specific brand of phone
 2. A different phone by a competing manufacturer (e.g., Samsung Galaxy)

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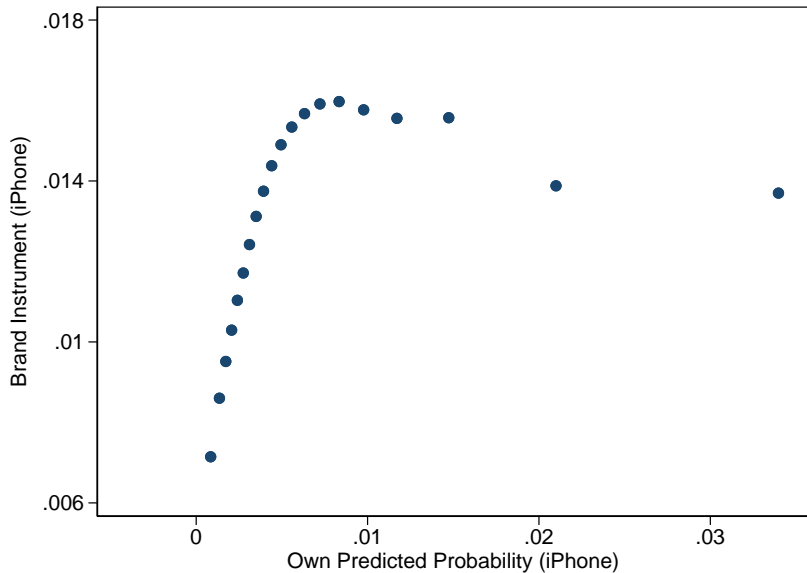
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 1. That specific brand of phone
 2. A different phone by a competing manufacturer (e.g., Samsung Galaxy)
- ▶ Conceptually two effects:
 1. Among those who are newly encouraged to buy, how many buy that specific phone vs. another phone (potential for positive demand spillover)
 2. Among those who would have bought anyways, what is the effect on the probability of buying that specific phone vs. another phone (potential for negative demand spillover)

Specific Phone Purchase - Research Design

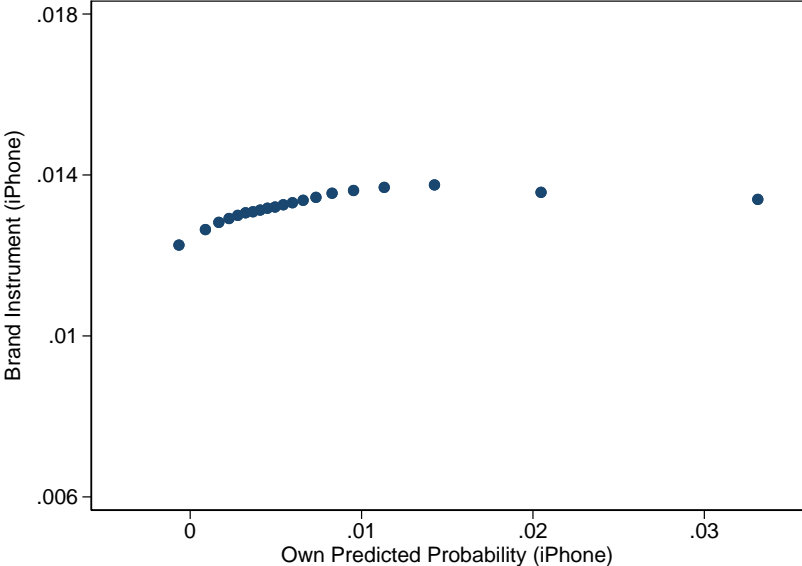
$$\mathbb{1}(BuysX)_{i,t} = \beta_1 FrBuysX_{i,t-1} + \beta_2 FrBuysY_{i,t-1} + \gamma X_{i,t} + \varepsilon_{i,t}$$

- ▶ **Common shocks + homophily:** You are more likely to buy the same phone as your friends, even in the absence of peer effects.
- ▶ **Observation:** Individuals differ in their (conditional) propensity to buy particular phones, $PropX$
 - ▶ Current iPhone users more likely to buy another iPhone
- ▶ **Identification Idea:**
 - ▶ IV: $PropX$ among all people who post about randomly losing their phone
 - ▶ Control for average of $PropX$ among all friends

Specific Phone Purchase - Research Design



Specific Phone Purchase - Research Design



Within and Across Brand Peer Effects

Cumulative Effects over 24 Weeks

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Cumulative Effects over 24 Weeks

	Dependent Variable: Buys between t and t+24 (%)			
	iPhone	Galaxy	Other	Any Phone
Friends buy iPhone	0.331*** (0.024)	-0.003 (0.018)	-0.121*** (0.017)	0.207*** (0.033)
Friends Buy Galaxy	-0.196*** (0.043)	0.670*** (0.037)	0.403*** (0.036)	0.877*** (0.063)
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- ▶ Largest positive peer effects for same brand
- ▶ Same brand effect smallest for iPhone (social learning?)

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► Negative across-OS spillover

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- ▶ Losing customers to a rival firm hurts me due to
 - ▶ Loss of future sales through positive peer effects
 - ▶ Loss of customers this person will bring to competitor who would have otherwise bought my product

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- ▶ Positive cross-brand spillovers for Android phones (social learning?)

Conclusion

- ▶ More likely to buy any new phone if friends recently bought new phone
 - ▶ Largest effect on specific device, some positive within-brand spillovers
 - ▶ Negative across-brand spillovers, but substantial new overall demand
- Value of customers; competitive implications; price setting
- Understanding precise nature of peer effects important for implications