Social Networks Shape Beliefs and Behavior: Evidence from Social Distancing During the COVID-19 Pandemic

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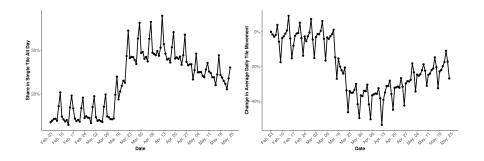
Introduction

- Global COVID-19 Pandemic
 - Persistent appeals for people to social distance to contain pandemic
 - Differences in compliance with and support for distancing policies
- What shapes beliefs about pandemic and social distancing behavior?
- → Focus on the role of social networks
- Use individual level de-identified data on social networks and mobility from Facebook to show that
 - 1 social network exposure to COVID-19 shapes social distancing
 - 2 social network exposure to COVID-19 shapes beliefs about pandemic

Data

- Anonymized network data from Facebook on U.S. users
 - Friendship links, location of friends
 - Individual level location and characteristics
- Proxy for Social Distancing: mobility behavior
 - Users with location history settings Summary Statistics
 - Two measures of mobility
 - Staying in home tile (600m square) on given day
 - Numbers of tiles visited on given day
- Public posting behavior and group memberships

Mobility Over Time



- Substantial reduction in mobility starting in mid-March
- Focus on staying in home tile but all results consistent for both

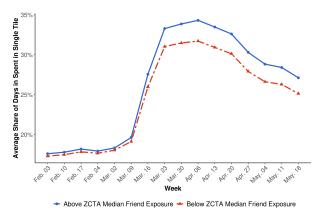
ullet Higher friend exposure to COVID cases o More social distancing?

- $\bullet \ \ \mbox{Higher friend exposure to COVID cases} \rightarrow \mbox{More social distancing?}$
- Friend Exposure to COVID-19
 - Measured as of March 15 when President Trump declared national emergency

$$\textit{FriendExposure}_{i}^{\textit{Mar}15} = \sum_{j=1}^{J} \textit{FracFriends}_{ij}^{\textit{Mar}15} \times \textit{COVID}19\textit{Cases}_{j}^{\textit{Mar}15}.$$

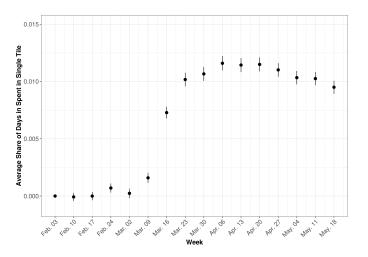
- Concern: Correlated with other characteristics
 - \rightarrow For now: Within ZCTA
 - \rightarrow Later:
 - Control for observables
 - Changes on changes specification

Two groups: Above and below median friend exposure within ZCTA

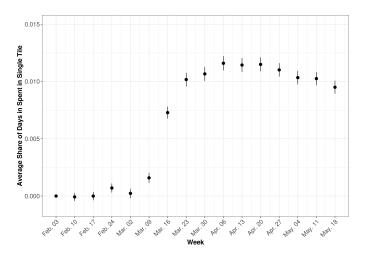


- Identical baseline of 18%
- ullet Above median friend exposure ightarrow 35% stay home in April
- Below median friend exposure \rightarrow less than 32%

- So far: Raw averages
- Concern: Differences between high and low exposure individuals
- Now: Include controls
 - Individual FE
 - Time varying effects of individual and location characteristics
 - College, age, gender, iPhone, tablet, ZCTA
 - Time varying effects of network characteristics
 - Friend weighted median HH income, population density, share urban



- Above median friend exposure \rightarrow Staying home \uparrow 1.2 ppt (3.8% increase relative to below median average of 32%)
- One SD friend exposure \uparrow Staying home $\uparrow \approx \frac{1}{2}$ college degree



- Differential ability to work from home?
 - Same patterns on weekdays and weekends Click here
 - Controlling for exact college (in addition to other demographics)

Changes in Friend Exposure

- So far: Friend exposure at onset of pandemic
- Concern: Friend exposure at onset correlated with unobservables
- Now: Changes in friend exposure and distancing
 - Early and subsequent hotspots very different
 - Correlation of observables with friend exposure varies over time
 - Onset of pandemic: Friend exposure positively correlated with college
 - Later: Friend exposure negatively correlated with college
 - ightarrow Unlikely that effect of unobservable characteristics changes over time in the same way as geographic spread of pandemic
- Maps
- Correlation Table

Changes in Friend Exposure and Social Distancing Behavior

$$\Delta Y_{it} = \sigma_0 + \sum_{j=1}^t \sigma_{1j} \textit{ChangeFriendExposure}_{ij} + \sigma_2 X_i + \sigma_3 N_i + \epsilon_{it}$$

	Monthly Change in ∆ Stay at Home								
	All months	All months	March	April	May	June	July		
Δ log(Friend Exposure + 1), All Months	0.206*** (0.029)	0.261*** (0.032)							
\log(Friend Exposure + 1), March			0.207***	0.006	-0.076** (0.048)	0.097 (0.054)	0.037		
\log(Friend Exposure + 1), April			,	0.035 (0.052)	0.096 (0.056)	0.329***	0.069**		
$\Delta \log(Friend Exposure + 1)$, May $\Delta \log(Friend Exposure + 1)$, June $\Delta \log(Friend Exposure + 1)$, July					0.379***	0.044	-0.057 (0.094)		
						0.854***	-0.329* (0.127)		
							0.323**		
Other Network Exposure FE	Y x Month	Y x Month	Υ	Υ	Υ	Υ	Υ		
ip Code x Age Group x Gender x Has College x Has Tablet x Has iPhone	Y x Month	Y x Month	Υ	Υ	Υ	Υ	Υ		
User FE		Υ							
R-Squared	0.211	0.287	0.174	0.141	0.150	0.146	0.145		
iample Mean	1.611	1.456	14.214	-0.923	-5.989	-1.068	0.679		
N	30,742,008	29,777,929	6,688,448	6,579,359	6,169,176	5,848,722	5,456,303		

ullet Increase in friend exposure o Increase in staying home

Changes in Friend Exposure and Social Distancing Behavior

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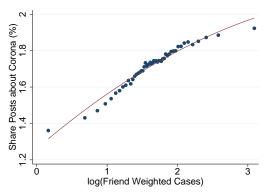
• Changes in friend exposure affect change in behavior in same month

- So far: Individuals with higher friend exposure engage in more social distancing. Why?
- Now: Effect of friend exposure on beliefs about COVID-19

$$Y_i = \delta_0 + \delta_1 \log(FriendExposure_i) + \delta_2 Z_i + \delta_3 C_i + \epsilon_i$$

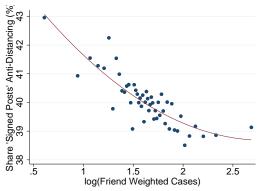
- Outcomes
 - Posting about COVID-19
 - Oppose or support social distancing
 - General sentiment of posts (positive vs. negative)
 - Membership in reopen groups
- Controls
 - Other network exposure FE
 - Interaction of location and individual characteristics (Zip Code * Age Group * Gender * Has College * Has Tablet * Has iPhone)

$$\textit{ShareOfPosts_COVID}_i = \delta_0 + \delta_1 \log(\textit{FriendExposure}_i) + \delta_2 Z_i + \delta_3 C_i + \epsilon_i$$



- ullet Higher friend exposure o More posts about COVID-19
- Doubling of friend exposure \rightarrow Share of posts about COVID \uparrow 17.2pp or 9.8% of baseline

$$\textit{ShareOfPosts_Oppose}_i = \delta_0 + \delta_1 \log(\textit{FriendExposure}_i) + \delta_2 Z_i + \delta_3 C_i + \epsilon_i$$



- Posts opposing restrictions among all opposing or supporting
- ullet Higher friend exposure o Fewer posts opposing social distancing
- Doubling of friend exposure \rightarrow Share opposing social distancing \downarrow 1.3pp or 3.7% of baseline

$$Y_i = \delta_0 + \delta_1 \log(FriendExposure_i) + \delta_2 Z_i + \delta_3 C_i + \epsilon_i$$

- Higher friend exposure (doubling) to COVID-19 cases:
 - More posts about COVID-19
 - Fewer posts opposing social distancing measures
 - General sentiment decrease
 - General sentiment ↓ 3.5% of baseline
 - Lower likelihood of being part of reopen group

$$Y_i = \delta_0 + \delta_1 \log(FriendExposure_i) + \delta_2 Z_i + \delta_3 C_i + \epsilon_i$$

- Higher friend exposure (doubling) to COVID-19 cases:
 - More posts about COVID-19
 - Fewer posts opposing social distancing measures
 - General sentiment decrease
 - Lower likelihood of being part of reopen group
 - Prob(Reopen Group) \downarrow 0.09 pp or 7.35% of baseline

ZCTA Friend Exposure: Establishments Visited and Spending

- ZCTA level analysis
 - Safegraph mobility and transaction data
 - Social Connectedness Index (SCI)
 - → More evidence on mechanism
 - → But: Higher potential for confounds
- ullet Behavioral response o reduce physical interactions
- Establishments visited
 - Less visits if non-essential (arts, entertainment, food & drinks)
 - No difference if essential or low contact (health care, social assistance, parks)
- Type of spending
 - Reduced spending at Starbucks
 - No differential spending at Amazon

Conclusion

- ullet Higher friend exposure to COVID-19 cases o More social distancing
 - Exposure at onset of pandemic affects social distancing for months
 - Not driven by ability to work from home
 - Changes in friend exposure affect changes in mobility
 - → Friends influence health behavior
- Higher friend exposure to COVID-19 cases
 - More posts about COVID-19
 - Post less likely to oppose social distancing measures
 - Less likely to be member of reopen group
 - → Friends influence beliefs about current events

Appendix

Summary Statistics - Mobility Sample

	Mean	SD	P10	P25	P50	P75	P90
Age	43.58	14.93	26	32	42	54	63
Female	0.53	0.50	0	0	1	1	1
Has College	0.53	0.50	0	0	1	1	1
Has iPhone	0.25	0.43	0	0	0	0	1
Has Tablet	0.53	0.50	0	0	1	1	1
Zip Code Income	\$58,792	\$21,961	\$36,160	\$43,648	\$54,000	\$69,203	\$88,096
Number of Friends	532.80	326.61	193	276	441	718	1047
Friend Exposure to Cases	10.35	19.34	0.74	1.77	4.49	11.12	26.31
Staying at home (Feb)							
- All	18.33	29.35	0	0	0	28.57	66.67
- Weekend	19.39	34.44	0	0	0	50.00	100.00
- Weekday	16.83	29.80	0	0	0	20.00	66.67
Bing tiles visited (Feb)							
- All	10.96	9.07	1.57	3.43	9.00	15.86	23.43
- Weekend	10.57	9.79	1.00	3.00	7.50	15.50	24.50
- Weekday	11.34	9.77	1.50	3.40	9.00	16.20	24.60

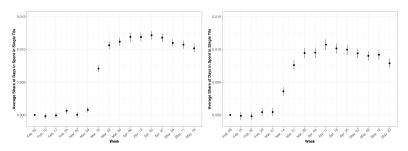


Friend Exposure and Social Distancing - Heterogeneity

	%Δ Stay at Home									
log(Friend Exposure) x I(Age < 35)	1.241***									
og(Friend Exposure) x I(Age 35-55)	0.960***									
og(Friend Exposure) x I(Age > 55)	(0.038)									
og(Friend Exposure) x Female		0.949***								
og(Friend Exposure) x Male		0.796*** (0.033)								
og(Friend Exposure) x College			1.321*** (0.034)							
og(Friend Exposure) x No College			0.443***							
og(Friend Exposure) x Zip Income First Tertile				0.386***						
og(Friend Exposure) x Zip Income Second Tertile				0.794*** (0.036)						
og(Friend Exposure) x Zip Income Third Tertile				1.608*** (0.045)						
og(Friend Exposure) x County Cases First Tertile					0.676***					
og(Friend Exposure) x County Cases Second Tertile					1.384***					
og(Friend Exposure) x County Cases Third Tertile					1.245*** (0.055)					
og(Friend Exposure - Rank 1 - 25)						0.204***				
og(Friend Exposure - Rank 26 - 50)						(0.017)				
og(Friend Exposure - Rank 51 - 75)						(0.017)				
og(Friend Exposure - Rank 76 - 100)						(0.017)				
Other Network Exposure FE	Υ	Υ	Y	Υ	Υ	Υ				
Zip Code x Age Group x Gender x Has College x Has Tablet x Has iPhone	Υ	Υ	Y	Υ	Υ	Υ				
R-Squared	0.175	0.175	0.175	0.175	0.175	0.177				
Sample Mean	13.800	13.800	13.800	13.800	13.800	14.488				
F Test (Rank 1-25 = Rank 76-100)						17.328**				
F Test (Kank 1-25 = Kank 76-100) N	6,400,738	6,400,738	6,400,738	6,400,738	6,400,738	5,6				

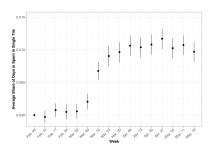


Friend Exposure and Social Distancing Behavior - Weekdays and Weekends



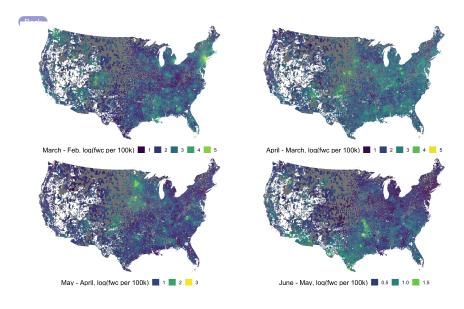
- Same pattern on weekdays and weekends
- $\bullet \to \text{evidence}$ against differences in ability to work from home being driving force behind observed patterns
- Back

Friend Exposure and Social Distancing Behavior - Controlling for Exact College



- Same pattern when controlling for exact college
- → evidence against differences in ability to work from home being driving force behind observed patterns
- Back

Changes in Friend Exposure



Determinants of Changes in Friend-Exposure to COVID-19

	Monthly Change in log(Friend Exposure + 1)									
	March	April	May	June	July	March	April	May	June	July
Age Group										
35-54	0.040***	0.014***	-0.013***	-0.008***	-0.001**	0.015***	0.005***	-0.004***	-0.003***	0.001***
	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)
55+	0.076***	0.015***	-0.026***	-0.018***	-0.004***	0.024***	0.007***	-0.006***	-0.006***	0.001
	(0.002)	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)	(0.000)	(0.000)	(0.000)	(0.000)
Female	-0.021***	0.006***	0.003***	0.006***	0.004***	-0.004***	0.002***	0.001***	0.000**	-0.000***
	(0.001)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)
Has College	0.039***	-0.031***	-0.007***	-0.008***	-0.004***	0.003***	-0.013***	-0.003***	0.002***	0.004***
	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)
Has iPhone	0.011***	0.005***	-0.007***	0.008***	0.013***	0.002***	-0.002***	-0.000	0.002***	0.003***
rias irrione	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)
Has Tablet	0.005***	-0.009***	-0.008***	-0.001***	0.002***	0.003***	-0.002***	-0.002***	-0.000***	0.001***
Tids Tablet	(0.001)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)
Network-Exposure Median HH Income (\$k)	0.015***	-0.004***	0.001***	-0.009***	-0.013***					
	(0.001)	(0.000)	(0.000)	(0.000)	(0.000)					
Network-Exposure Population Density	349.495***	-34.302***	-65.142***	-71.383***	-88.601***					
(residents/meter2)	(5.622)	(1.527)	(1.280)	(1.582)	(1.764)					
Network-Exposure Fraction of Pop. Urban	1.112***	-0.076**	-0.263***	0.319***	0.456***					
	(0.035)	(0.019)	(0.016)	(0.014)	(0.016)					
Zip Code Income										
Middle Tertile	-0.034***	-0.017**	0.007	-0.011**	-0.004					
	(0.011)	(0.008)	(0.006)	(0.005)	(0.006)					
Top Tertile	0.002	-0.026***	-0.008	-0.006	0.005					
	(0.011)	(0.008)	(0.005)	(0.005)	(0.006)					
Zip Code FE						Υ	Υ	Υ	Υ	Υ
Other Network Exposure FE						Υ	Υ	Υ	Υ	Υ
R-Squared	0.560	0.044	0.117	0.215	0.281	0.877	0.680	0.728	0.781	0.822
Sample Mean	2.800	2.303	0.810	0.476	0.615	2.800	2.303	0.810	0.476	0.615
N	7,090,255	6,981,142	6,571,618	6,251,614	5,859,728	7,090,254	6,981,141	6,571,617	6,251,614	5,859,728

