

## **Social Connectedness: Measurement, Determinants, and Effects**

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**S**ocial networks can shape many aspects of social and economic activity: migration and trade, job-seeking, innovation, consumer preferences and sentiment, public health, social mobility, and more. In turn, social networks themselves are associated with geographic proximity, historical ties, political boundaries, and other factors. Traditionally, the unavailability of large-scale and representative data on social connectedness between individuals or geographic regions has posed a challenge for empirical research on social networks. More recently, a body of such research has begun to emerge using data on social connectedness from online social networking services such as Facebook, LinkedIn, and Twitter. To date, most of these research projects have been built on anonymized administrative microdata from Facebook, typically by working with coauthor teams that include Facebook employees. However, there is an inherent limit to the number of researchers that will be able to work with social network data through such collaborations.

In this paper, we therefore introduce a new measure of social connectedness at the US county level. Our Social Connectedness Index is based on friendship links on Facebook, the global online social networking service. Specifically, the Social Connectedness Index corresponds to the relative frequency of Facebook friendship

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links between every county-pair in the United States, and between every US county and every foreign country. Given Facebook's scale, with 2.1 billion active users globally and 239 million active users in the United States and Canada (Facebook 2017), as well as the relative representativeness of Facebook's user body, these data provide the first comprehensive measure of friendship networks at a national level. Moreover, the Social Connectedness Index data can be made accessible to members of the broader research community. Interested researchers are invited to email [sci\\_data@fb.com](mailto:sci_data@fb.com) to learn about the current process for working with the Social Connectedness Index data.

We begin this article by describing the construction of the Social Connectedness Index (SCI). The bulk of the paper then explores various patterns related to social connectedness. We first use the SCI data to analyze patterns of social connectedness between US counties. We find that the intensity of friendship links is strongly declining in geographic distance, with the elasticity of the number of friendship links to geographic distance ranging from about  $-2.0$  over distances less than 200 miles, to about  $-1.2$  for distances larger than 200 miles. We also look at how social connectedness is shaped by political boundaries such as state lines, exposure to large within-US population movements, and other historical and contemporaneous factors.

We then explore heterogeneity across counties in the geographic concentration of their populations' social networks. For the average county, 62.8 percent of all friendship links are to individuals living within 100 miles, but this number ranges from 46.0 percent at the 5th percentile to 76.9 percent at the 95th percentile of the across-county distribution. We find that the populations of counties with a larger fraction of friends living more than 100 miles away are on average better off along a number of socioeconomic dimensions. For example, counties with more geographically dispersed social networks have higher incomes, higher education levels, and higher social mobility.

We then turn to the question of how the intensity of social connectedness between regions correlates with bilateral economic and social activity. We first document a strong correlation between social connectedness and trading activity, consistent with recent research that argues that social networks help overcome informational and cultural frictions that can inhibit trade. Social connectedness is also positively correlated with the spread of innovation and within-US migration. When we look at friendship links between US regions and foreign countries, we find further strong correlations with both past migration patterns and present-day trade flows.

Throughout this essay, our focus is on documenting and describing salient patterns of social connectedness across a variety of settings. We do not seek to provide causal analyses, nor do we want to imply causal relationships behind the correlations we document. Nevertheless, we do believe that our findings can guide future research on the causal effects of social networks. More generally, the patterns discussed here highlight significant opportunities for using data from online social networking services such as Facebook to help alleviate the measurement challenges faced by researchers across the social sciences trying to better understand the role of social connectedness.

## Measuring Social Connectedness

The Social Connectedness Index is constructed using aggregated and anonymized information from the universe of friendship links between all Facebook users as of April 2016. Duggan, Ellison, Lampe, Lenhart, and Madden (2015) report that as of September 2014, more than 58 percent of the US adult population and 71 percent of the US online population used Facebook. The same source reports that, among online US adults, Facebook usage rates are relatively constant across income groups, education groups, and racial groups. Usage rates among online US adults are declining in age, from 87 percent of 18-to-29 year-olds to 56 percent of above-65 year-olds.

In the United States, Facebook mainly serves as a platform for real-world friends and acquaintances to interact online, and people usually only add connections on Facebook to individuals whom they know in the real world (Jones et al. 2013; Gilbert and Karahalios 2009; Hampton, Goulet, Rainie, and Purcell 2011). Establishing a friendship link on Facebook requires the consent of both individuals, and the total number of friends for a person is limited to 5,000. As a result, Facebook data have a unique ability to provide a large-scale representation of US friendship networks.

To measure the social connectedness between geographies, we map Facebook users to their respective county and country locations, and thus obtain the total number of friendship links between these geographies. Locations are assigned to users based on the users' information and activity on Facebook, including the stated city on their Facebook profile, and device and connection information. We only consider friendship links among Facebook users who have interacted with Facebook over the 30 days prior to the April 2016 snapshot.<sup>1</sup> We treat each friendship link identically.

We then construct the Social Connectedness Index between all pairs of 3,136 US counties, and between every US county and every foreign country, as the normalized total number of friendship links for each geographic pair. In particular, the Social Connectedness Index is constructed to have a maximum value of 1,000,000, and relative differences in the index correspond to relative differences in the total number of friendship links. The highest Social Connectedness Index value of 1,000,000 is assigned to Los Angeles County–Los Angeles County connections (Los Angeles County is where people have the most friends with other people in their county).

## The Determinants of Social Connectedness

The Social Connectedness Index can be used to analyze the correlates of the intensity of social connectedness between US counties. We first analyze the role

<sup>1</sup>Facebook formally defines such “monthly active users” in its 10Q statements as follows: “We define a monthly active user as a registered Facebook user who logged in and visited Facebook through our website or a mobile device, or used our Messenger application (and is also a registered Facebook user), in the last 30 days as of the date of measurement.”

of geographic distance in shaping social connectedness in the United States. The effects of geographic proximity on friendship formation and social interactions have been studied in a number of papers, including Zipf (1949), Verbrugge (1983), and Marmaros and Sacerdote (2006).

As a motivating example, compare San Francisco County and Kern County in California. These two counties have roughly the same population of slightly under one million, but Kern County is 175 times larger in area. Moreover, San Francisco County, which is home to the city of San Francisco, is surrounded by the urbanized Bay Area economy including Oakland and San Jose. Kern County includes the Bakersfield metro area, but it is not surrounded by an urban area.

We construct a measure that we call the “relative probability of friendship” by taking the Social Connectedness Index between counties  $i$  and  $j$  and dividing it by the product of the number of Facebook users in the two counties. This allows us to take into account the fact that we will see more friendship links between counties with more Facebook users.<sup>2</sup> If this measure is twice as large, this means that a given Facebook user in county  $i$  is about twice as likely to be connected with a given Facebook user in county  $j$ . The heat maps in Figure 1 show the relative probability that a given Facebook user in San Francisco County (Figure 1A) or Kern County (Figure 1B) is connected to a given Facebook user in another county.

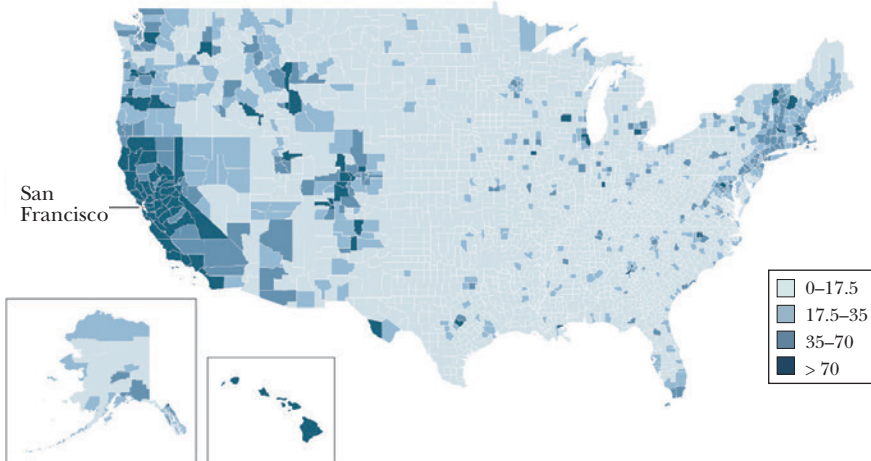
For both San Francisco County and Kern County, a significant proportion of friendship links (dark shading indicates more links) are to geographically close counties across the West Coast. However, there are also noticeable differences in the social connectedness of the two counties. The population of San Francisco County has significant social connections to counties located in the northeastern United States, while the population of Kern County has far fewer of these friendship links. Instead, Kern County’s friendship network is very concentrated in the West Coast and Mountain States, with the exception of a pocket of strong connections to individuals living in Oklahoma and Arkansas. These connections are likely related to past migration patterns, because Kern County was a major destination for migrants fleeing the Dust Bowl in the 1930s. Kern County also has substantial friendship links to the oil-producing regions of North Dakota, perhaps not surprising given that Kern County produces more oil than any other county in the United States.

Overall, the friendship networks of the Kern County population are much more geographically concentrated than those of the San Francisco County population: Kern County has 57 percent of friends living within 50 miles, relative to 27 percent for San Francisco County. In comparison with the summary statistics for the whole United States, displayed in Table 1, the geographic concentration of the friendship network of Kern County is similar to the US average while San Francisco County’s friendship network is extremely geographically dispersed. For the average (population-weighted) US county, 55.4 percent of friends live within 50 miles, with a

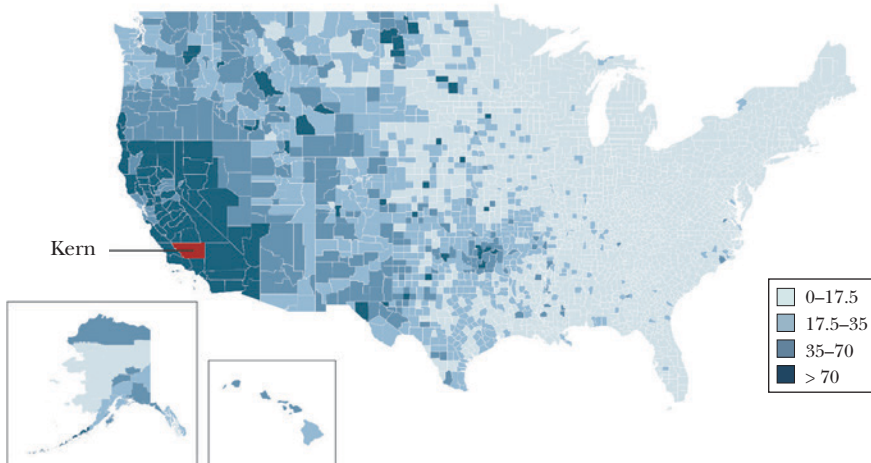
<sup>2</sup>While the number of Facebook users per county is not part of the public data release, very similar patterns for “relative probability of friendship” would be obtained if we instead divided the Social Connectedness Index by the product of county-level populations.

Figure 1  
**County-Level Friendship Maps**

A: Relative Probability of Friendship Link to San Francisco County, CA



B: Relative Probability of Friendship Link to Kern County, CA



Note: The heat maps show the relative probability that a Facebook user in each county  $j$  has a friendship link to San Francisco County, CA (Panel A) and Kern County, CA (Panel B). Darker colors correspond to counties in which there is a higher probability of a friendship link between a person in home county  $i$  (San Francisco or Kern) and county  $j$ . The “relative probability of friendship” is constructed by taking the Social Connectedness Index between counties  $i$  and  $j$  and dividing it by the product of the number of Facebook users in the two counties.

10–90 percentile range of 42.5 to 67.4 percent; and over 70 percent of friends live within 200 miles, with a 10–90 percentile range of 57.1 to 81.2 percent. This despite the fact that, for the average county, only 1.3 percent and 6.6 percent of the US population live within 50 miles and 200 miles, respectively.

Table 1

**Distance and Friendship Links: Across-County Summary Statistics for the United States**

	<i>Share of friends living within:</i>			<i>Share of US population living within:</i>		
	<i>50 Miles</i>	<i>100 Miles</i>	<i>200 Miles</i>	<i>50 Miles</i>	<i>100 Miles</i>	<i>200 Miles</i>
Mean	55.4%	62.8%	70.3%	1.3%	2.8%	6.6%
P5	38.1%	46.0%	54.2%	0.1%	0.3%	1.0%
P10	42.5%	49.6%	57.1%	0.1%	0.6%	2.1%
Median	55.4%	63.9%	71.6%	0.7%	2.1%	5.8%
P90	67.4%	74.8%	81.2%	3.2%	6.2%	15.0%
P95	70.3%	76.9%	83.2%	5.4%	9.2%	15.6%

*Note:* Table shows across-county summary statistics for the share of friends of a county's population living within a certain distance of that county as well as the share of the US population living within those distances. P5, P10, P90, and P95 are the 5th, 10th, 90th, and 95th percentiles, respectively. Counties are weighted by their populations.

The regressions in Table 2 offer a more systematic account of the relationship between geographic distance and social connectedness across county-pairs. The unit of observation is a county-pair. The dependent variable is the log of the Social Connectedness Index between the two counties. The log of the geographic distance between the counties is the explanatory variable in column 1. We include fixed effects for both counties, which controls for population levels and any other characteristics that vary at the county level. In this specification, geographic distance is able to explain a significant amount of the cross-county-pair variation in social connectedness. The estimated elasticity of social connectedness to geographic distance suggests that a 10 percent increase in the distance between two counties is associated with a 14.8 percent decline in the number of friendship links between those counties. Similar to gravity equations estimated in the trade literature, this estimates the equilibrium relationship between geographic distance and social connectedness, not necessarily the causal effect of one on the other.

In column 2, we include an additional control indicating whether both counties are within the same state. The social connectedness of a county is often strongest with other counties within the same state, even compared to nearby counties in other states. This finding is not the result of non-log linearities in the distance relationship, and it can be found for both border counties and nonborder counties (as we discuss further in the Appendix). Why social connectedness varies so strongly at state borders, and the extent to which this is driven by institutional, social, or economic factors, is an interesting avenue for future research. Possible explanations include the importance of common state-level identities or the role of state universities as meeting places for residents from the same state.

In columns 3 and 4, we restrict the sample to county-pairs that are more and less than 200 miles apart, respectively. In the sample of county-pairs that are less than

200 miles apart, the estimated elasticity between geographic distance and friendship links is  $-1.99$ . In the sample of county-pairs that are more than 200 miles apart, the magnitude of the elasticity falls by nearly half to  $-1.16$ . These findings suggest that while social connectedness is declining in geographic distance, the elasticity of this relationship is less negative as we include county-pairs that are progressively further apart. In turn, this pattern highlights that in the theoretical modeling of friendship links, the appropriate elasticity depends on the geographic distances studied. This finding may help to explain why previous estimates of the elasticity of friendship probability with respect to geographic distance vary so significantly across settings, including an estimate of  $-2$  in a study of cell-phone communication networks in the United Kingdom (Lambiotte et al. 2008); an estimate of  $-1$  among bloggers (Liben-Nowell, Novak, Kumar, Raghavan, and Tomkins 2005); and an estimate of  $-0.5$  in location-based online social networks such as Brightkite, Foursquare, and Gowalla (Scellato, Noulas, Lambiotte, and Mascolo 2011).

A substantial literature has documented that individuals are more likely to be associated with other individuals of similar characteristics. Following Lazarsfeld and Merton (1954), this empirical regularity is referred to as “homophily.” Homophily has been documented for a large number of individual characteristics, including racial identity, gender, age, religion, and education, as well as intangible aspects such as attitudes and beliefs (for a comprehensive review of the literature, see McPherson, Smith-Lovin, and Cook 2001). Thus, in column 5 of Table 2 we add a number of variables measuring the similarity of counties on measures such as per capita income, education levels, and religiosity. We find that county pairs that are more similar on these dimensions have more friendship links. However, while the magnitude of the effect of these socioeconomic differences on social connectedness is potentially meaningful, adding them barely affects the coefficients on other explanatory variables or the  $R^2$  relative to the specification in column 2.

Table 2 highlights that social connectedness drops off strongly at state borders. A related question is how closely the existing state borders resemble the borders that would form if we grouped together US counties to create communities with the aim of maximizing within-community social connectedness. There are a number of possible algorithms to facilitate such a grouping of counties. Here, we use a method called hierarchical agglomerative linkage clustering (which we describe further in the online Appendix).

Figure 2 shows the result when we use this algorithm to group the United States into 20 distinct communities. All resulting communities are spatially contiguous, which is a result of the strong dependence of social connectedness on geographic distance. In addition, and consistent with finding social connectedness to decline at state borders, many of the community borders line up with state borders. All of the West Coast States together with Nevada form one community. Similarly, all counties in states between New England and Pennsylvania are grouped into the same community. Another group of states is Florida, Georgia, and Alabama. However, some states are split into separate communities. The Texas panhandle is grouped with Oklahoma and Kansas, and Colorado’s Western Slope forms its own community.

Table 2

**Determinants of Social Connectedness across County Pairs**

	Dependent Variable: Log(SCI)				
	(1)	(2)	(3)	(4)	(5)
log(Distance in Miles)	-1.483*** (0.065)	-1.287*** (0.061)	-1.160*** (0.059)	-1.988*** (0.043)	-1.214*** (0.055)
Same State		1.496*** (0.087)	1.271*** (0.083)	1.216*** (0.044)	1.496*** (0.085)
$\Delta$ Income (\$1,000)					-0.006*** (0.001)
$\Delta$ Share Population White (%)					-0.012*** (0.001)
$\Delta$ Share Population No High School (%)					-0.012*** (0.002)
$\Delta$ 2008 Obama Vote Share (%)					-0.006*** (0.001)
$\Delta$ Share Population Religious (%)					-0.002*** (0.001)
County Fixed Effects	Y	Y	Y	Y	Y
Sample			>200 miles	<200 miles	
Number of observations	2,961,968	2,961,968	2,775,244	186,669	2,961,968
$R^2$	0.907	0.916	0.916	0.941	0.922

*Note:* Table shows results from a regression of the log of the Social Connectedness Index on a number of explanatory variables. The log of the geographic distance between the counties is the explanatory variable in column 1. In column 2, we include an additional control indicating whether both counties are within the same state. In columns 3 and 4, we restrict the sample to county-pairs that are more and less than 200 miles apart, respectively. The unit of observation is a county-pair. Standard errors are given in parentheses. The online Appendix (<http://e-jep.org>) provides more details on the data sources and exact specifications.

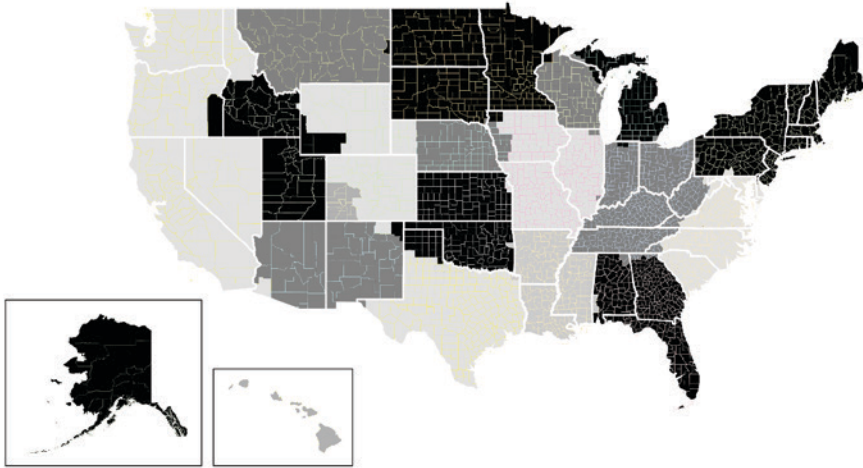
\*, \*\*, and \*\*\* indicate significance levels of  $p < 0.1$ ,  $p < 0.05$ , and  $p < 0.01$ , respectively.

These findings suggest that it might be interesting to study the economics and politics of US “regions” as defined by joint social connectedness, rather than alternative groupings such as Census regions or divisions.

We have explored a number of additional correlates of friendship links across counties. For example, we document that the strength of social connections can be affected by physical obstacles such as large rivers and mountain ranges. We highlight that counties with military bases exhibit strong connections across the entirety of the United States, as do counties in North Dakota that have seen a recent shale oil boom and an associated significant in-migration. Counties with Native American reservations are strongly connected to one another. Similarly, areas with ski resorts in the Rocky Mountains and New England have high social connectedness. Counties in Florida with significant retiree populations are strongly connected to the Rust Belt and the Northeast. In addition, large cities in the Midwestern



Figure 2

**Connected Communities within the United States—20 Units**

*Note:* Figure shows US counties grouped together when we use hierarchical agglomerative linkage clustering to create 20 distinct groups of counties.

United States with significant African American populations, such as Milwaukee and Chicago, have strong links to the South around Mississippi and Alabama, consistent with friendship links persisting following the Great Migration of southern African Americans to northern cities. For more details on these patterns, see the online Appendix (<http://e-jep.org>). In general, many of these patterns of friendship connections are unsurprising, but it is new that such patterns can now be measured and documented in systematic national data.

### **Concentration of Social Networks and County Characteristics**

The geographic concentrations of the friendship networks of different counties reveal a great deal of heterogeneity: for example, the earlier Table 1 shows that the 5th–95th percentile range across population-weighted counties in the share of friends living within 100 miles is 46.0 percent to 76.9 percent. Existing theoretical work suggests that the diversity of social networks is an important determinant of economic development; conversely, tightly clustered social ties can limit access to a broad range of social and economic opportunities (for example, Granovetter 1973). However, empirical studies of the relationship between the structure of social networks and economic outcomes of communities are rare. One exception is Eagle, Macy, and Claxton (2010), who use UK cellphone data to document that the diversity of individuals' social networks is correlated with regional economic well-being. In this section, we provide evidence that the geographic dispersion of friendship links across US counties is highly correlated with social and economic

outcomes at the county level, such as average income, educational attainment, and social mobility.

If we define the concentration of a friendship network as the share of friends who live within 100 miles, then friendship networks in the South, the Midwest, and Appalachia are the most geographically concentrated. Counties in the Rocky Mountains have the smallest share of friends living within 100 miles, in large part because these areas are often less-densely populated. Among the western United States, Utah and inland California have the most geographically concentrated friendship networks. The online Appendix shows heat maps of this and other measures of the geographic concentration of friendship networks.

What are the effects of differentially structured social networks on county-level outcomes? As a first step toward answering this question, we correlate our measure of the concentration of friendship links with county-level characteristics. Figure 3 presents county-level binned scatterplots using the share of friends living within 100 miles and a number of socioeconomic outcomes. The overall message is that counties where people have more concentrated social networks tend to have worse socioeconomic outcomes along a number of dimensions: on average, they have lower income, lower education, higher teenage birth rate, lower life expectancy, less social capital, and less social mobility.

These correlations cannot be interpreted as causal (although the online Appendix discusses a number of causal mechanisms proposed by the literature that are consistent with our findings). Our goal here, as in the rest of the paper, is to document patterns that can guide future research investigating the causal effects of social network structure on socioeconomic outcomes, and to describe the Social Connectedness Index data that can help with such analyses. More generally, the strong correlation between social connectedness and socioeconomic outcomes suggests that controlling for the geographic concentration of social networks is important to minimize omitted variables bias across a number of research agendas that study economic and social outcomes at the county level.

## **Social Connectedness and Cross-County Activity**

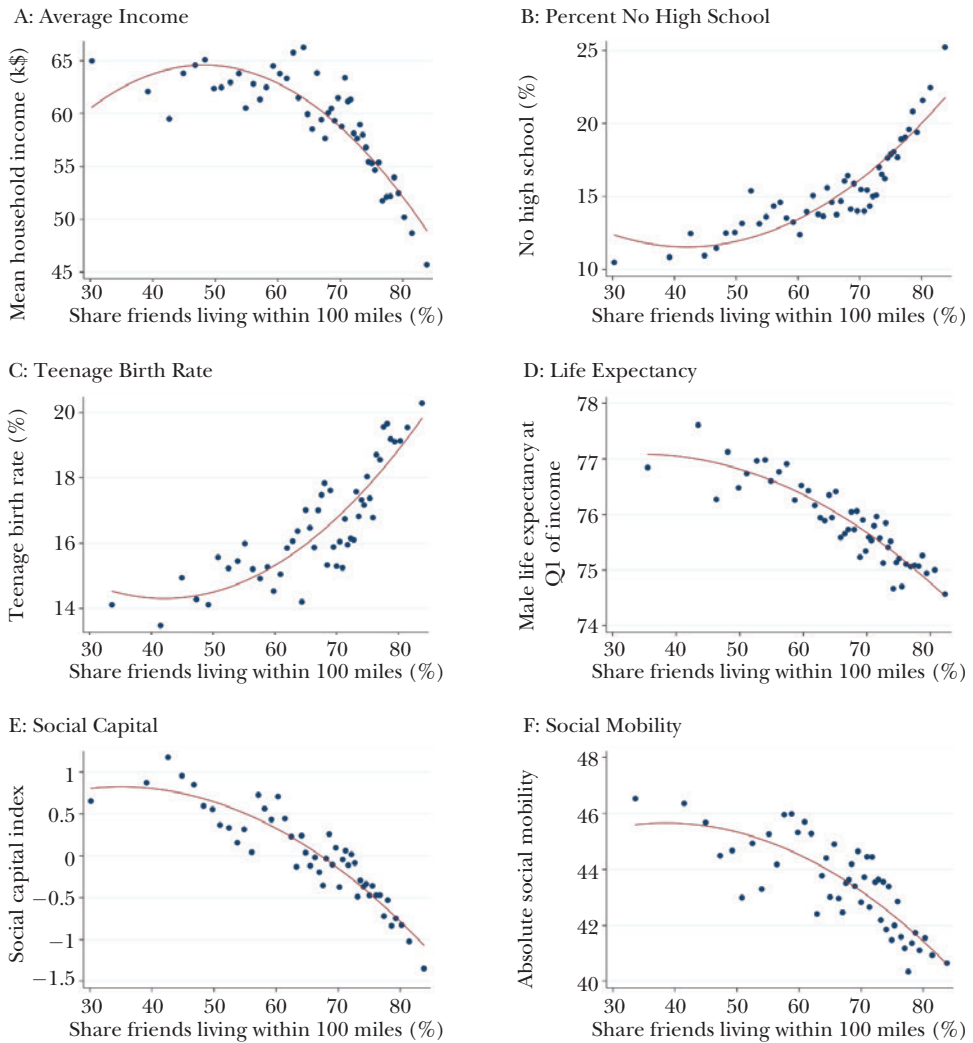
Social connectedness between two regions may be related to other economic and social interactions between these regions. Indeed, we next document correlations between the number of friendship links and trade flows, patent citations, and migration patterns. As before, we illustrate some salient patterns in the data rather than providing full-fledged causal analyses. For each of the patterns documented below, the online Appendix (<http://e-jep.org>) provides more details on the variables, data construction, specifications, and additional exploration.

### **Social Connectedness and Within-US Trade Flows**

A well-established empirical result in the trade literature is that bilateral trade between two regions decreases with geographic distance, although the explanations

Figure 3

Network Concentration and County-Level Characteristics



Notes: Panels show binned scatterplots with counties as the unit of observation. To generate each binned scatterplot, we group the x-axis variable into 50 equal-sized bins. We then compute the mean of the x-axis and y-axis variables within each bin and create a scatterplot of these 50 data points. The horizontal axes measure the share of friends of the county population that live within 100 miles. On the vertical axes are a number of county-level measures of socioeconomic outcomes: the mean county income in Panel A; the share of the population with no high school degree in Panel B; the teenage birth rate as provided by Chetty, Hendren, Kline, and Saez (2014) in Panel C; the life expectancy of males in the first quarter of the national income distribution from Chetty et al. (2016) in Panel D; the measure of social capital in 2009 as defined by Rupasingha, Goetz, and Freshwater (2006) in Panel E; and the absolute measure of social mobility from Chetty et al. (2014) in Panel F. The red line shows the fit of a quadratic regression. The online Appendix (<http://e-jep.org>) provides more details.

for this finding are still being debated (for a review, see Anderson and van Wincoop 2004). Many studies have highlighted that the distance effect is too large to be fully explained by trade costs alone, and that geographic distance might serve as a proxy for other trade frictions such as cultural differences, lack of familiarity, or information asymmetries. Social connections may alleviate the trade costs associated with these factors, and some empirical work has examined the causal effect of stronger social networks on trade (Rauch 1999; Combes, Lafourcade, and Mayer 2005; Cohen, Gurun, and Malloy 2012; Burchardi and Hassan 2013; Chaney 2014, 2016). However, much of this literature has struggled to measure the social connectedness between trading partners, and thus had to rely on indirect proxies, such as the ethnic composition of regions or past migration patterns.

The Social Connectedness Index data allow us to examine directly the empirical relationship between trade flows and social connectedness at the state level. Panel A of Table 3 shows some results. For the dependent variable, we measure interstate trading volumes using data from the Commodity Flow Survey. We focus on data from 2012, the latest year with comprehensively available data. Specifically, the dependent variable captures the log of the value of trade in 2012 between origin state  $i$  and destination state  $j$ .

For our main explanatory variables, we use the log of geographic distance between states  $i$  and  $j$ , as well as the log of the Social Connectedness Index between states  $i$  and  $j$  (constructed from a weighted average of county-level SCI measures). We also include fixed effects for each state, dummy variables for own-state flows, and dummy variables if the states are adjacent to each other.

We observe two main patterns. First, social connectedness is strongly correlated with state–state trade flows, even after controlling for geographic distance. The magnitude of the elasticity of trade with social connectedness is large and statistically significant.<sup>3</sup> In fact, when comparing them across columns 1 and 2, it appears as if social connectedness can explain marginally more of the variation in state–state trade flows than geographic distance.

Second, controlling for social connectedness significantly reduces the estimated distance elasticities of trade. A comparison of columns 1 and 3 shows that the distance elasticity of trade halves in magnitude after controlling for social connectedness. In column 4, we further control for differences across the states in GDP per capita, unemployment rates, sectoral composition, union share, and population density. The addition of these further controls has essentially no effect on the estimated elasticity between social connectedness and trade.

The observed reduction in the distance elasticities of trade, after controlling for social connectedness, is consistent with theories described above which suggest that geographic distance might be proxying for other factors affecting trade between

<sup>3</sup>In the online Appendix, we explore these patterns across industries. We find that the magnitude of the elasticity of trade flows with respect to friendship links rises with the share of high-skilled workers in the sector and is not affected by the share of labor compensation in total costs.

Table 3

**Social Connectedness and Across-Region Economic Interactions**

	(1)	(2)	(3)	(4)
<i>Panel A: Dependent Variable: log(State-Level Trade Flows)</i>				
log(Distance)	-1.057*** (0.071)		-0.531*** (0.084)	-0.533*** (0.085)
log(SCI)		0.999*** (0.051)	0.643*** (0.071)	0.637*** (0.060)
State Fixed Effects	Y	Y	Y	Y
Other State Differences	N	N	N	Y
Observations	2,219	2,220	2,219	2,219
R <sup>2</sup>	0.912	0.918	0.926	0.930
<i>Panel B: Dependent Variable: Indicator for Patent Citation</i>				
log(Distance)	-0.048*** (0.002)		-0.011** (0.005)	-0.021** (0.009)
log(SCI)		0.063*** (0.003)	0.049*** (0.006)	0.066*** (0.012)
Technological Category + County Fixed Effects	Y	Y	Y	Y
Cited + Issued Patent Fixed Effects, Other County Differences	N	N	N	Y
Observations	2,171,754	2,171,754	2,171,754	2,168,285
R <sup>2</sup>	0.056	0.059	0.059	0.101
<i>Panel C: Dependent Variable: log(County-Level Migration)</i>				
log(Distance)	-0.973*** (0.048)		0.023 (0.021)	0.031 (0.021)
log(SCI)		1.134*** (0.019)	1.148*** (0.024)	1.159*** (0.024)
County Fixed Effects	Y	Y	Y	Y
Other County Differences	N	N	N	Y
Observations	25,305	25,305	25,305	25,287
R <sup>2</sup>	0.610	0.893	0.893	0.893

*Note:* Table shows the relationship between bilateral economic activity across geographic units and the geographic distance and social connectedness between these units. “SCI” stands for Social Connectedness Index. In Panel A, the unit of observation is a state-pair, and the dependent variable is the log of the value of 2012 trade flows between the states. All specifications include state fixed effects, dummies for own state, and dummies for neighboring states; column 4 also controls for differences across states on important socioeconomic indicators. In Panel B, the unit of observation is a patent-pair. The dependent variable is an indicator of whether patent  $i$  cites patent  $j$ . All specifications control for the county and technology category fixed effects, and column 4 also controls for patent fixed effects and other differences across the counties of the patents on important socioeconomic indicators. In Panel C, the unit of observation is a county pair, and the dependent variable is the log of across-county migration between 2013 and 2014. All specifications control for county fixed effects, and column 4 also controls for other differences across counties on important socioeconomic indicators. Standard errors are given in parentheses. The online Appendix (<http://e-jep.org>) provides more details on the data sources and exact specifications.

\*, \*\*, and \*\*\* indicate significance levels of  $p < 0.1$ ,  $p < 0.05$ , and  $p < 0.01$ , respectively.

states. Further investigating the causal role of social connectedness in facilitating trade flows might therefore be a useful avenue for future research.

### **Social Connectedness and Patent Citations**

In many models of endogenous growth, knowledge spillovers among individuals or firms are an important driver of productivity and economic growth (Romer 1986; Lucas 1988; Aghion and Howitt 1992). Social connectedness might therefore have important effects on economic activity, by facilitating the diffusion of knowledge and ideas through society.<sup>4</sup> However, testing these theories is challenging, because both knowledge spillovers and the degree of social connectedness are hard to measure. To overcome these challenges, a large empirical literature has relied on patent citations as a measure of knowledge spillovers (Jaffe, Trajtenberg, and Henderson 1993; Thompson and Fox-Kean 2005). By studying the geographic distances between the locations where the issued patents and patent citations occur, these papers conclude that knowledge spillovers are highly localized. In turn, this finding is often interpreted as evidence for the importance of social interactions, which are more likely to happen at shorter distances. Other attempts to measure social connectedness have tried to proxy for an inventor's peer group based on characteristics such as common ethnicity (Agrawal, Kapur, and McHale 2008).

The Social Connectedness Index has the potential to provide more direct evidence for the role of social connectedness in facilitating knowledge spillovers. We obtain data containing information on all patents granted by the US Patent and Trademark Office in the years 2002–2014, and the location of the company or institution from which the patent originated. If the company or institution is not available, then the patent is assigned to the location of the first inventor with an available location (as in Berkes and Gaetani 2017). The patents cover 107 different technological classes, defined based on the International Patent Classification. For each granted patent, we observe all other patents that it cites.

We follow the approach in the existing literature to explore the relationship between social connectedness and patent citations (for example, Jaffe, Trajtenberg, and Henderson 1993). This approach matches each “citing patent” with a “non-citing patent” issued at the same time and in the same technological class to serve as a control, as we will explain below. Knowledge spillovers are then measured as the extent to which the citation probability increases with the social connectedness of the geographies associated with the patents, after controlling for the patent's technological class and the geographic distance between the geographies. The literature has argued that this approach can help to separate knowledge spillovers

<sup>4</sup>For examples, see Jovanovic and Rob (1989), Kortum (1997), Benhabib and Spiegel (2005), Alvarez, Buera, and Lucas (2008), Comin and Hobijn (2010), Comin, Dmitriev, and Rossi-Hansberg (2012), Fogli and Veldkamp (2012), and Buera and Oberfield (2016). Social networks can also affect the exposure of the region to new ideas and thus how quickly the region adopts a new idea (for instance, Glaeser 1999; Black and Henderson 1999; Moretti 2012).

from correlations that might be induced by patterns in the geographic location of technologically related activities across regions that are connected through social networks.

To implement this approach, for each US patent granted in 2014, we create an observation for every patent cited by the 2014 patent, so that the unit of observation is a patent–citation pair. For example, if a particular 2014 patent cites 10 other patents, this will generate 10 patent–citation pairs. We then construct a control observation for each of these patent–citation pairs. In particular, for each 2014 patent *A* that cites a previous patent *B*, we randomly select another 2014 patent *C* that is in the same technology class as patent *A*, but that does not cite patent *B*. We focus on patent classes with at least 1,000 patents issued in 2014, to ensure that there is a sufficient sample to select the control patents randomly.

Panel B of Table 3 shows results from our analysis. The dependent variable in the regressions equals one if an issued patent *i* cites patent *j*, and zero otherwise. The first two rows show the coefficients on the log of geographic distance and the log of the Social Connectedness Index between the counties of the issued and cited patents. We include fixed effects for the technology classes and for the counties of patents *i* and *j*.

Comparing columns 1 and 2, social connectedness explains marginally more of the variation in the probability of a patent citation than geographic distance, as the  $R^2$  in column 2 is higher. In terms of economic magnitudes, the probability of a patent citation is 6.3 percentage points higher when the social connectedness between the counties of the issued and cited patents doubles.

In column 3, we jointly estimate the relationship of geographic distance and social connectedness with the probability of a patent citation. The effect of doubling social connectedness on the probability of citation remains significant and large, at 4.9 percent, even after controlling for geographic distance. In comparison, the effect of doubling geographic distance on the probability of citations falls from  $-4.8$  to  $-1.1$  percent.

In column 4, we also control for a host of across-county differences on important socioeconomic indicators: 2008 vote share of Obama, mean income, share of population without a high school degree, share of population that is white, share of population that is religious, and share of workforce employed in manufacturing. We also add fixed effects for the cited and the issued patents. If anything, the estimated relationship between social connectedness and patent citation increases somewhat as a result of these further controls.

This finding suggests that the relationship between geographic distance and the probability of patent citation, viewed in isolation, may be partially capturing effects of information flows associated with social connectedness. More generally, our results suggest a significant correlation between social connectedness and knowledge spillovers, innovation, and, ultimately, economic growth. These findings highlight the potential of the Social Connectedness Index data to help uncover possible causal relationships behind these correlations.

### Social Connectedness and Migration

Understanding the factors driving migration patterns is important. For example, within-US migration is one mechanism for equilibrating the US labor market following regional shocks (Blanchard and Katz 1992). An existing literature has documented that social networks can play an important role in facilitating migration by providing information as well as social and economic support (for a review, see Munshi 2016). While a lot of the research has focused on international migration (for example, Moretti 1999), similar forces might be at work in explaining within-US migration.

We find that the Social Connectedness Index has significant explanatory power for migration between regions, beyond what is predicted by geographic distance. Panel C of Table 3 shows some results. The dependent variable captures the log of total migration between counties  $i$  and  $j$  between 2013 and 2014, as measured by the Statistics of Income (SOI) Tax Stats Migration Data provided by the IRS. The key explanatory variables are the log of geographic distance between those counties and the log of the Social Connectedness Index. We also include fixed effects for each county, which allows us to control for the size of its population and other county-level characteristics that might affect the degree of migration.

In column 1 of Table 3, Panel C, we do not include the social connectedness variable. The estimated elasticity of migration to geographic distance is close to  $-1$ . In column 2, we find that the elasticity of migration to social connectedness is slightly larger than 1, with a somewhat higher  $R^2$  than in column 1. In other words, the Social Connectedness Index can explain a larger part of the variation of the migration flows across county-pairs than geographic distance can. In column 3, we control for both the geographic distance and social connectedness between counties. We find that geographic distance adds no additional predictive power compared with column 2. This finding suggests that much of the estimated effect of distance on migration might be coming from the relationship between distance and social connectedness, and that distance by itself has no additional explanatory power for migration. Column 4 shows that these conclusions are robust to further controlling for other differences across counties on important socioeconomic indicators.

Overall, our results are consistent with stories in which individuals are more likely to move to counties where they already have friends. Such a mechanism could, for example, result in larger cities attracting even more new movers and thereby help explain the very right-tailed city size distribution (Gabaix 1999). Exploring the causal mechanisms behind the observed relationship between social connectedness and migration thus provides an exciting research agenda.

### International Dimension of Social Connectedness of US Counties

US counties vary considerably in the share of social connections to individuals living outside of the United States. For the median county, 4 percent of all



friendship links are to individuals living in foreign countries, but the 10–90 percentile range is 2.3 percent to 8.6 percent, and the 1–99 percentile range is 1.6 percent to 18.7 percent. Some of this variation is straightforward to explain. For example, areas close to the Mexican or the Canadian border have more international connections. Patterns of past immigration matter as well. For example, connections with Norway are particularly strong for those parts of the United States that saw major immigration from Norway in the late 19th and early 20th Centuries, like Wisconsin, Minnesota, and the Dakotas. Similarly, a number of counties in the northeastern United States have strong social connectedness to Italy. For heat maps of social connectedness to these and other countries, see the online Appendix available with this paper at <http://e-jep.org>.

The first three columns in Table 4 illustrate the extent to which past migration from a particular country is correlated with the strength of today's social connectedness of a US county with that country. In these columns, the dependent variable is the Social Connectedness Index between each county and foreign country. For the explanatory variables, geographic distance is measured between each county and the capital city of each foreign country. We use two measures of past migration: the number of residents who claim their primary ancestry as being from a given foreign country and the number of residents in each county who were born in a specific foreign country. The first measure is broader and can, for instance, include US-born individuals with immigrant parents or grandparents. All variables are measured in logs. We also include fixed effects for each county and foreign country.

The first column shows the correlation between geographic distance and international social connectedness: a 1 percent increase in the geographic distance is associated with a 1.2 percent decline in social connectedness. Interestingly, this elasticity is nearly identical to the elasticity of friendship links to geographic distance estimated for the United States for distances greater than 200 miles. The second column shows that a 1 percent increase in the number of residents with ancestry from a given foreign country correlates with an increase in social connections to that country by about one-third of a percent. In column 3, we obtain similar estimates for our second measure of past migration. Across columns 2 and 3, controlling for past migration reduces the estimated effect of geographic distance on social connectedness by between one-third and one-half.

In other regressions presented in the online Appendix, we find that the effect of past migration on today's social connections is stronger for countries from which immigration to the United States occurred more recently, such as Mexico or the Philippines, compared to countries from which immigration peaked earlier, such as Germany or Ireland. For example, the coefficient on a regression like that in column 2 is about 0.13 for counties with immigration waves that peaked pre-1900 or between 1900 and 1930, but more than twice as high for waves that peaked between 1930 and 1990 or for waves that have not yet peaked.

We also sought to estimate the relationship between social connectedness and international trade. Again, we used state-level data on social connectedness

Table 4

**Social Connectedness, Ancestry, and International Trade**

	log(SCI)			log (Exports + 1)	log (Imports + 1)
	(1)	(2)	(3)	(4)	(5)
log(Distance)	-1.159*** (0.258)	-0.690*** (0.162)	-0.493*** (0.174)	-2.092*** (0.391)	-1.627*** (0.378)
log(Ancestry in Foreign Country)		0.341*** (0.022)			
log(Born in Foreign Country)			0.367*** (0.033)		
log(SCI)				0.597*** (0.139)	0.470*** (0.103)
Fixed Effects	Y	Y	Y	Y	Y
Observations	33,146	33,146	16,527	11,015	11,014
R <sup>2</sup>	0.908	0.936	0.943	0.770	0.770
Number of Countries	105	105	52	216	216

*Note:* The table explores the international dimension of social connectedness. In columns 1 to 3, we explore how past migration patterns and geographic distance are correlated with international social connectedness. The unit of observation is a US county–foreign country pair. Each specification also includes fixed effects for the US state and the foreign country, and the dependent variable is the log of the Social Connectedness Index between those units. In columns 4 and 5, we explore how today’s international trading activity is correlated with social connectedness. The unit of observation is a US state–foreign country pair. Standard errors are given in parentheses. The online Appendix (<http://e-jep.org>) provides more details on the data sources and exact specifications.

\*, \*\*, and \*\*\* indicate significance levels of  $p < 0.1$ ,  $p < 0.05$ , and  $p < 0.01$ , respectively.

(by combining the counties of a given state into a population-weighted average), because data on international trade is only available at the state level. Adjusting for geographic distance, (in a specification similar to Table 3, Panel B, column 3), we find that a state with 10 percent higher social connectedness to a given foreign country on average imports 4.7 percent more from this country and exports 6.0 percent more to this country. These findings are highly consistent with our earlier estimates on within-US trade. In the online Appendix for this paper, we provide additional details on these variables and alternative specifications.

## Conclusion

We use data from the global online social networking site Facebook to construct the Social Connectedness Index (SCI). These data provide a new and comprehensive measure of social connectedness between US county pairs, as well as between US counties and foreign countries. The SCI should allow researchers to overcome some of the measurement challenges that have held back empirical research on the role of social interactions in finance, economics, and the broader social sciences. To illustrate

this point, we show how the SCI data can be used to better understand the geographic dimensions of real-world social networks, as well as to document that social connectedness correlates strongly with social and economic activity across regions. While these correlations should not be seen as identifying causal relationships, they provide starting points for investigating a variety of important questions.

A number of recent studies have used data from online social networks, in most cases by including coauthors from Facebook or other social networking services. For example, Gee, Jones, and Burke (2017) and Gee, Jones, Fariss, Burke, and Fowler (2017) use de-identified microdata from Facebook to analyze the role of social networks in the job-finding process. These researchers were able to assess the relative importance of strong and weak ties in helping job seekers find new employment. Social network data from Facebook have also been used to study a range of other topics: the relationship between the size of friendship networks and mortality (Hobbs, Burke, Christakis, and Fowler 2016); the structure of social networks in immigrant communities in the United States (Herdağdelen, State, Adamic, and Mason 2016); the evolution of information cascades (Cheng, Adamic, Kleinberg, and Leskovec 2016); and the effects of social influence and social advertising (Bakshy, Eckles, Yan, and Rosenn 2012). Other researchers have studied the effects of online social networks themselves. For example, Bakshy, Messing, and Adamic (2015) study how online networks influence exposure to perspectives that cut across ideological lines. In our own work, we have used social network data from Facebook to document that social interactions influence people's perceptions of local housing markets as well as their real estate investment decisions and mortgage leverage choices (Bailey, Cao, Kuchler, and Stroebel forthcoming; Bailey, Dávila, Kuchler, and Stroebel 2017). We have also explored the role of peer effects in product adoption decisions (Bailey, Kuchler, Stroebel, and Wong 2018), and are working with other coauthors to better understand the role of social connectedness in facilitating social mobility.

For many researchers, it should prove a considerable advantage that the Social Connectedness Index is now more broadly available. In addition to the topics that we have explored in this paper, here are five other examples of policy and research questions that we hope will be pursued with the SCI data.

First, many contagious illnesses and diseases, such as the flu or tuberculosis, spread through human contact. Combined with localized data on the prevalence of the flu, data on social connectedness might allow researchers and public health officials to better predict where to expect future outbreaks of the flu (Cauchemez et al. 2011; Christakis and Fowler 2010).

Second, the Social Connectedness Index data could also be used to track whether measures of sentiment—for example, those tracked by the Michigan Survey of Consumers or through geo-coded Twitter feeds—spread along social networks.

Third, sociolinguistic research has argued that social networks are an important force determining how languages evolve over time (for example, Milroy 1987). The Social Connectedness Index data would allow researchers to study the extent to which linguistic development in the United States is associated with patterns of social connectedness.

Fourth, the relationships between transportation networks and social connectedness may prove interesting. For example, significant social connectedness between two regions might be a strong indicator that providing transportation infrastructure between these regions, such as direct airline routes, is profitable. Using the Social Connectedness Index as a measure of the potential demand for various routes could address some of the identification issues in the literature analyzing airline scheduling in operations research and industrial organization. Moreover, increased transportation links might also have a causal effect on social connectedness. One approach using the SCI data is to compare the social connectedness of two counties that happen to lie on the straight line between two major cities, and which are therefore connected by a highway, to the connectedness of two similar counties that do not lie on the straight line between major cities (see Bailey et al. 2018).

Finally, the SCI might prove useful in testing theoretical models of network formation (Jackson 2014). Specifically, in models of geographic strategic network formation models, the costs of network formation are directly related to distance (for example, Johnson and Gilles 2000). Using data from the National Longitudinal Survey of Adolescent Health on close friends of individuals, Patacchini, Picard, and Zenou (2015) show that students living in central locations have higher levels of social interactions. Our estimates of the elasticities of friendship links with respect to distance often map directly into the parameters of these models and can be used to parameterize them.

While we hope that the county-level Social Connectedness Index will prove useful to researchers, it is of course only one aspect of the vast wealth of data on networks being created by online social networking services. As these data become available in various forms, the modeling and analysis of social networks will advance substantially.

## References

- Aghion, Philippe, and Peter Howitt.** 1992. "A Model of Growth through Creative Destruction." *Econometrica* 60(2): 323–51.
- Agrawal, Ajay, Devesh Kapur, and John McHale.** 2008. "How Do Spatial and Social Proximity Influence Knowledge Flows? Evidence from Patent Data." *Journal of Urban Economics* 64(2): 258–69.
- Alvarez, Fernando E., Francisco J. Buera, and Robert E. Lucas, Jr.** 2008. "Models of Idea Flows." NBER Working Paper 14135.
- Anderson, James E. and Eric van Wincoop.** 2004. "Trade Costs." *Journal of Economic Literature* 42(3): 691–751.
- Bailey, Michael, Ruiqing Cao, Theresa Kuchler, and Johannes Stroebel.** Forthcoming. "The Economic Effects of Social Networks: Evidence from the Housing Market." *Journal of Political Economy*.
- Bailey, Michael, Eduardo Dávila, Theresa Kuchler, and Johannes Stroebel.** 2017. "House Price Beliefs and Mortgage Leverage Choice." NBER Working Paper 24091.
- Bailey, Michael, Theresa Kuchler, Johannes Stroebel, and Arlene Wong.** 2018. "Peer Effects in

Product Adoption." Unpublished paper.

**Bakshy, Eytan, Dean Eckles, Rong Yan, and Itamar Rosenn.** 2012. "Social Influence in Social Advertising: Evidence from Field Experiments." In *Proceedings of the 13th ACM Conference on Electronic Commerce (EC '12)*, pp. 146–61. New York, NY: ACM.

**Bakshy, Eytan, Solomon Messing, and Lada A. Adamic.** 2015. "Exposure to Ideologically Diverse News and Opinion on Facebook." *Science* 348(6239): 1130–32.

**Benhabib, Jess, and Mark M. Spiegel.** 2005. "Human Capital and Technology Diffusion." Chap. 13 in *Handbook of Economic Growth*, vol. 1A, edited by Philippe Aghion and Steven N. Durlauf, 935–966. Elsevier.

**Berkes, Enrico, and Ruben Gaetani.** 2017. "The Geography of Unconventional Innovation." [https://cpb-us-east-1-juc1ugur1qwqqo4.stackpathdns.com/sites.northwestern.edu/dist/4/638/files/2017/06/Berkes\\_Gaetani\\_Submission\\_June\\_2017-2ao3fck.pdf](https://cpb-us-east-1-juc1ugur1qwqqo4.stackpathdns.com/sites.northwestern.edu/dist/4/638/files/2017/06/Berkes_Gaetani_Submission_June_2017-2ao3fck.pdf).

**Black, Duncan, and Vernon Henderson.** 1999. "A Theory of Urban Growth." *Journal of Political Economy* 107(2): 252–84.

**Blanchard, Olivier Jean, and Lawrence F. Katz.** 1992. "Regional Evolutions." *Brookings Papers on Economic Activity* no. 1, pp. 1–75.

**Buera, Francisco J., and Ezra Oberfield.** 2016. "The Global Diffusion of Ideas." NBER Working Paper 21844.

**Burchardi, Konrad B., and Tarek A. Hassan.** 2013. "The Economic Impact of Social Ties: Evidence from German Reunification." *Quarterly Journal of Economics* 128(3): 1219–71.

**Cauchemez, Simon, Achuyt Bhattarai, Tiffany L. Marchbanks, Ryan P. Fagan, Stephen Ostroff, Neil M. Ferguson, David Swerdlow, and the Pennsylvania H1N1 Working Group.** 2011. "Role of Social Networks in Shaping Disease Transmission during a Community Outbreak of 2009 H1N1 Pandemic Influenza." *PNAS* 108(7): 2825–30.

**Chaney, Thomas.** 2014. "The Network Structure of International Trade." *American Economic Review* 104(11): 3600–34.

**Chaney, Thomas.** 2016. "Networks in International Trade." Chap. 28 in *Oxford Handbook of the Economics of Networks*, edited by Yann Bramoullé, Andrea Galeotti, and Brian Rogers. Oxford University Press.

**Cheng, Justin, Lada A. Adamic, Jon M. Kleinberg, and Jure Leskovec.** 2016. "Do Cascades Recur?" 671–681, International World Wide Web Conferences Steering Committee. arXiv:1602.01107 [cs.SI].

**Chetty, Raj, Nathaniel Hendren, Patrick Kline, and Emmanuel Saez.** 2014. "Where is the Land of Opportunity? The Geography of Intergenerational

Mobility in the United States." *Quarterly Journal of Economics* 129(4): 1553–1623.

**Chetty, Raj, Michael Stepler, Sarah Abraham, Shelby Lin, Benjamin Scuderi, Nicholas Turner, Augustin Bergeron, and David Cutler.** 2016. "The Association between Income and Life Expectancy in the United States, 2001–2014." *JAMA* 315(16): 1750–66.

**Christakis, Nicholas A., and James H. Fowler.** 2010. "Social Network Sensors for Early Detection of Contagious Outbreaks." *PLoS ONE* 5(9): e12948.

**Cohen, Lauren, Umit G. Gurun, and Christopher J. Malloy.** 2012. "Resident Networks and Firm Trade." NBER Working Paper 18312.

**Combes, Pierre-Philippe, Miren Lafourcade, and Thierry Mayer.** 2005. "The Trade-Creating Effects of Business and Social Networks: Evidence from France." *Journal of International Economics* 66(1): 1–29.

**Comin, Diego A., Mikhail Dmitriev, and Esteban Rossi-Hansberg.** 2012. "The Spatial Diffusion of Technology." NBER Working Paper 18534.

**Comin, Diego, and Bart Hobijn.** 2010. "An Exploration of Technology Diffusion." *American Economic Review* 100(5): 2031–59.

**Duggan, Maeve, Nicole B. Ellison, Cliff Lampe, Amanda Lenhart, and Mary Madden.** 2015. "Social Media Update 2014." Pew Research Center, January 9.

**Eagle, Nathan, Michael Macy, and Rob Claxton.** 2010. "Network Diversity and Economic Development." *Science*, May 21, 328(5981): 1029–31.

**Facebook.** 2017. "Facebook Form 10-Q, Quarter 4, 2017." [https://s21.q4cdn.com/399680738/files/doc\\_financials/2017/Q4/Q4-2017-Earnings-Presentation.pdf](https://s21.q4cdn.com/399680738/files/doc_financials/2017/Q4/Q4-2017-Earnings-Presentation.pdf).

**Fogli, Alessandra, and Laura Veldkamp.** 2012. "Germs, Social Networks and Growth." NBER Working Paper 18470.

**Gabaix, Xavier.** 1999. "Zipf's Law for Cities: An Explanation." *Quarterly Journal of Economics* 114(3): 739–67.

**Gee, Laura K., Jason Jones, and Moira Burke.** 2017. "Social Networks and Labor Markets: How Strong Ties Relate to Job Finding on Facebook's Social Network." *Journal of Labor Economics* 35(2): 485–518.

**Gee, Laura K., Jason J. Jones, Christopher J. Fariss, Moira Burke, and James H. Fowler.** 2017. "The Paradox of Weak Ties in 55 Countries." *Journal of Economic Behavior & Organization* 133: 362–72.

**Gilbert, Eric, and Karrie Karahalios.** 2009. "Predicting Tie Strength with Social Media." In *Proceedings of the SIGCHI Conference on Human Factors in Computing Systems*, 211–20. ACM.

**Glaeser, Edward.** 1999. "Learning in Cities."

*Journal of Urban Economics* 46(2): 254–77.

**Granovetter, Mark S.** 1973. “The Strength of Weak Ties.” *American Journal of Sociology* 78(6): 1360–80.

**Hampton, Keith, Lauren Sessions Goulet, Lee Rainie, and Kristen Purcell.** 2011. “Social Networking Sites and Our Lives.” Pew Internet and American Life Project, Washington, DC.

**Herdagdelen, Amaç, Bogdan State, Lada Adamic, and Winter Mason.** 2016. “The Social Ties of Immigrant Communities in the United States.” *Proceedings of the 8th ACM Conference on Web Science*, 78–84.

**Hobbs, William R., Moira Burke, Nicholas A. Christakis, and James H. Fowler.** 2016. “Online Social Integration is Associated with Reduced Mortality Risk.” *PNAS* 113(46): 12980–984.

**Jackson, Matthew O.** 2014. “Networks in the Understanding of Economic Behaviors.” *Journal of Economic Perspectives* 28(4): 3–22.

**Jaffe, Adam B., Manuel Trajtenberg, and Rebecca Henderson.** 1993. “Geographic Localization of Knowledge Spillovers as Evidenced by Patent Citations.” *Quarterly Journal of Economics* 108(3): 577–98.

**Johnson, Cathleen, and Robert Gilles.** 2000. “Spatial Social Networks.” *Review of Economic Design* 5(3): 273–99.

**Jones, Jason J., Jaime E. Settle, Robert M. Bond, Christopher J. Fariss, Cameron Marlow, and James H. Fowler.** 2013. “Inferring Tie Strength from Online Directed Behavior.” *PLoS ONE* 8(1): e52168.

**Jovanovic, Boyan, and Rafael Rob.** 1989. “The Growth and Diffusion of Knowledge.” *Review of Economic Studies* 56(4): 569–82.

**Kortum, Samuel S.** 1997. “Research, Patenting, and Technological Change.” *Econometrica* 65(6): 1389–1419.

**Lambiotte, Renaud, Vincent D. Blondel, Cristobal de Kerchove, Etienne Huens, Christophe Prieur, Zbigniew Smoreda, and Paul Van Dooren.** 2008. “Geographical Dispersal of Mobile Communication Networks.” *Physica A: Statistical Mechanics and Its Applications* 387(21): 5317–25.

**Lazarsfeld, P., and R. K. Merton.** 1954. “Friendship as a Social Process: A Substantive and Methodological Analysis.” In *Freedom and Control in Modern Society*, edited by Morroe Berger, Theodore Abel, and Charles H. Page. New York: Van Nostrand.

**Liben-Nowell, David, Jasmine Novak, Ravi Kumar, Prabhakar Raghavan, and Andrew Tomkins.** 2005. “Geographic Routing in Social Networks.” *PNAS* 102(33): 11623–28.

**Lucas, Robert E., Jr.** 1988. “On the Mechanics of Economic Development.” *Journal of Monetary Economics* 22(1): 3–42.

**Marmaros, David, and Bruce Sacerdote.** 2006. “How Do Friendships Form?” *Quarterly Journal of Economics* 121(1): 79–119.

**McPherson, Miller, Lynn Smith-Lovin, and James M. Cook.** 2001. “Birds of a Feather: Homophily in Social Networks.” *Annual Review of Sociology* 27: 415–44.

**Milroy, Lesley.** 1987. *Language and Social Networks*. 2nd edition. Wiley-Blackwell.

**Moretti, Enrico.** 1999. “Social Networks and Migrations: Italy 1876–1913.” *International Migration Review* 33(3): 640–57.

**Moretti, Enrico.** 2012. *The New Geography of Jobs*. Houghton Mifflin Harcourt.

**Munshi, Kaivan.** 2016. “Community Networks and Migration.” Chap. 23 in *The Oxford Handbook of the Economics of Networks*, edited by Yann Bramoullé, Andrea Galeotti, and Brian Rogers. Oxford University Press.

**Patacchini, Eleonora, Pierre M. Picard, and Yves Zenou.** 2015. “Urban Social Structure, Social Capital and Spatial Proximity.” CEPR Discussion Papers no. DP10501, Center for Economic Policy Research, March.

**Rauch, James E.** 1999. “Networks versus Markets in International Trade.” *Journal of International Economics* 48(1): 7–35.

**Romer, Paul M.** 1986. “Increasing Returns and Long-Run Growth.” *Journal of Political Economy* 94(5): 1002–37.

**Rupasingha, Anil, Stephan J. Goetz, and David Freshwater.** 2006. “The Production of Social Capital in US Counties.” *Journal of Socio-Economics* 35(1: Essays on Behavioral Economics): 83–101.

**Scellato, Salvatore, Anastasios Noulas, Renaud Lambiotte, and Cecilia Mascolo.** 2011. “Socio-Spatial Properties of Online Location-Based Social Networks.” *Proceedings of the Fifth International Conference on Weblogs and Social Media*, held in Barcelona, Catalonia, Spain, July 17–21. AAAI Digital Library.

**Thompson, Peter, and Melanie Fox-Kean.** 2005. “Patent Citations and the Geography of Knowledge Spillovers: A Reassessment.” *American Economic Review* 95(1): 450–60.

**Verbrugge, Lois M.** 1983. “A Research Note on Adult Friendship Contact: A Dyadic Perspective.” *Social Forces* 62(1): 78–83.

**Zipf, George Kingsley.** 1949. *Human Behavior and the Principle of Least Effort*. Cambridge, MA: Addison-Wesley Press.

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3. Grit Hein, Lynn Huestegge, Anne Böckler-Raettig, Lorenz Deserno, Andreas B. Eder, Johannes Hewig, Andreas Hotho, Sarah Kittel-Schneider, Anna Linda Leutritz, Andrea M.F. Reiter, Johannes Rodrigues, Matthias Gamer. 2024. A social information processing perspective on social connectedness. *Neuroscience & Biobehavioral Reviews* **167**, 105945. [[Crossref](#)]
4. Susana Herrero Olarte, Joan Torrent, Kamila Aguirre. 2024. Internet use at work and income inequality in Ecuador. *Technology in Society* **79**, 102738. [[Crossref](#)]
5. Kushagra Tiwari, M. Amin Rahimian, Mark S. Roberts, Praveen Kumar, Jeanine M. Buchanich. 2024. Measuring network dynamics of opioid overdose deaths in the United States. *Scientific Reports* **14**:1. . [[Crossref](#)]
6. Aparna Ananthasubramaniam, David Jurgens, Daniel M. Romero. 2024. Networks and identity drive the spatial diffusion of linguistic innovation in urban and rural areas. *npj Complexity* **1**:1. . [[Crossref](#)]
7. Nasim Sabah. 2024. The impact of social distancing on trading activity during the COVID-19 pandemic. *Financial Management* **53**:4, 833-865. [[Crossref](#)]
8. Junhua Hu, Yingling Zhou, Huiyu Li, Pei Liang. 2024. An interval forecast model for infectious diseases using fuzzy information granulation and spatial-temporal graph neural network. *Journal of Intelligent & Fuzzy Systems* **47**:1-2, 83-97. [[Crossref](#)]
9. Xiaoqing Zhou. 2024. Financial Technology and the Transmission of Monetary Policy: The Role of Social Networks. *Journal of Political Economy Macroeconomics* **127**. . [[Crossref](#)]
10. Sean Higgins. 2024. Financial Technology Adoption: Network Externalities of Cashless Payments in Mexico. *American Economic Review* **114**:11, 3469-3512. [[Abstract](#)] [[View PDF article](#)] [[PDF with links](#)]
11. Miguel González-Leonardo, Ruth Neville, Sofia Gil-Clavel, Francisco Rowe. 2024. Where have Ukrainian refugees gone? Identifying potential settlement areas across European regions integrating digital and traditional geographic data. *Population, Space and Place* **30**:8. . [[Crossref](#)]
12. Lena Abou El-Komboz, Moritz Goldbeck. 2024. Virtually borderless? Cultural proximity and international collaboration of developers. *Economics Letters* **244**, 111951. [[Crossref](#)]
13. Edem M. Azila-Gbettor, Jewel Dela Novixoxo, Ben Q. Honyenuga. 2024. Navigating the nexus of citizenship fatigue and knowledge sharing: the influence of employee vitality and tenure among university lecturers. *Journal of Workplace Learning* **50**. . [[Crossref](#)]
14. Abdul Aziz Abdul Rahman, Poh Ling Chong, Tze San Ong, Boon Heng Teh, Tze Chin Ong. 2024. Business network and balanced scorecard: an analysis of small and medium enterprises in Malaysia. *Arab Gulf Journal of Scientific Research* **42**:3, 621-635. [[Crossref](#)]
15. Sukrati Agrawal, Neha Agrawal, Rohit Bansal, Anjali Rawat. Online Social Networking 523-549. [[Crossref](#)]
16. Daniel J. Braconnier, John Liu. Comments & Replies: Discussion Forum Interaction Networks Stimulated by Peer-Review Assignments 1-6. [[Crossref](#)]
17. Franklin Allen, Meijun Qian. 2024. Alternative finance in the international business context: a review and future research. *Journal of International Business Studies* **36**. . [[Crossref](#)]

18. Andreas Mastrosavvas. 2024. The geography of partisan homophily in the 2020 US presidential election. *Applied Geography* **171**, 103371. [[Crossref](#)]
19. Chenxi Liu, Zhenghong Peng, Lingbo Liu, Hao Wu, Jan Kinne, Meng Cai, Shixuan Li. 2024. XAI in geographic analysis of innovation: Evaluating proximity factors in the innovation networks of Chinese technology companies through web-based data. *Applied Geography* **171**, 103373. [[Crossref](#)]
20. Christos A. Makridis, Tao Wang. 2024. Learning from Friends in a Pandemic: Social networks and the macroeconomic response of consumption. *European Economic Review* **169**, 104836. [[Crossref](#)]
21. Grant L. Ferguson, James G. Gimpel, Mark E. Owens, Daron R. Shaw. 2024. The surge of the small donate in U.S. elections: A view from Texas statewide campaigns. *Political Geography* **114**, 103191. [[Crossref](#)]
22. Casey F. Breen, Dennis M. Feehan. 2024. New Data Sources for Demographic Research. *Population and Development Review* **59**. . [[Crossref](#)]
23. Lin Peng, Linyi Zhang. 2024. Unleashing the Crowd: The Effect of Social Networks in Crowdfunding Markets. *Management Science* . [[Crossref](#)]
24. Erdinc Akyildirim, Ahmet Faruk Aysan, Oguzhan Cepni, Özge Serbest. 2024. Sentiment matters: the effect of news-media on spillovers among cryptocurrency returns. *The European Journal of Finance* **30**:14, 1577-1613. [[Crossref](#)]
25. Nuttavuth Nundhapana, Chiraphol N. Chiyachantana, David K. Ding, Sirimon Treepongkaruna. 2024. Social network centrality and the corporate environment: The case of sexual diversity policies. *Corporate Social Responsibility and Environmental Management* **31**:5, 4085-4100. [[Crossref](#)]
26. Zekai He, Xinyu Liu, Xiuzhen Shi, Xiaoyan Sun. 2024. Housing demolition and entrepreneurship: Evidence from China. *Cities* **152**, 105201. [[Crossref](#)]
27. Andreas Mastrosavvas. 2024. Social Networks and Brexit: Evidence from a Trade Shock. *Regional Science and Urban Economics* **108**, 104024. [[Crossref](#)]
28. Till Koebe, Theophilus Aidoo, Ridhi Kashyap, Douglas R. Leasure, Valentina Rotondi, Ingmar Weber. 2024. Social capital mediates knowledge gaps in informing sexual and reproductive health behaviours across Africa. *Social Science & Medicine* **357**, 117159. [[Crossref](#)]
29. Cillian Berragan, Alex Singleton, Alessia Calafiore, Jeremy Morley. 2024. Mapping cognitive place associations within the United Kingdom through online discussion on Reddit. *Transactions of the Institute of British Geographers* **49**:3. . [[Crossref](#)]
30. Patrick Clasen. 2024. Solidarity on a divided continent: Perceptions of 'centre' and 'periphery' determine European citizens' willingness to help other EU countries. *European Union Politics* **25**:3, 569-592. [[Crossref](#)]
31. Riley Wilson. 2024. The Isolated States of America: Home State Bias and the Impact of State Borders on Mobility. *Journal of Labor Economics* **84**. . [[Crossref](#)]
32. Shruthy Nair, Clio Andris. 2024. The Relative Probability of Facebook Friendship in the United States. *Environment and Planning B: Urban Analytics and City Science* . [[Crossref](#)]
33. J. Anthony Cookson, Runjing Lu, William Mullins, Marina Niessner. 2024. The social signal. *Journal of Financial Economics* **158**, 103870. [[Crossref](#)]
34. Rongfei Su, Xiao Huang, Ruishan Chen, Xiaona Guo. 2024. Spatial and social inequality of hierarchical healthcare accessibility in urban system: A case study in Shanghai, China. *Sustainable Cities and Society* **109**, 105540. [[Crossref](#)]
35. Michael Bailey, Drew Johnston, Martin Koenen, Theresa Kuchler, Dominic Russel, Johannes Stroebel. 2024. Social Networks Shape Beliefs and Behavior: Evidence from Social Distancing during the COVID-19 Pandemic. *Journal of Political Economy Microeconomics* **2**:3, 463-494. [[Crossref](#)]



36. MARK J. GARMAISE, YARON LEVI, HANNO LUSTIG. 2024. Spending Less after (Seemingly) Bad News. *The Journal of Finance* **79**:4, 2429-2471. [[Crossref](#)]
37. Panagiotis Adamopoulos, Vilma Todri. 2024. Consumer Social Connectedness and Persuasiveness of Collaborative-Filtering Recommender Systems: Evidence From an Online-to-Offline Recommendation App. *Production and Operations Management* **3**. . [[Crossref](#)]
38. Beining Chen, Li Li, Yichen Xiong, Jianjun Wang. Who will adapt faster: insights from studying the underlying impact of climate change perception on people's willingness to adopt household PV - BES system 368-372. [[Crossref](#)]
39. Jacob Kruse, Song Gao, Yuhan Ji, Daniel P. Szabo, Kenneth R. Mayer. 2024. Bringing spatial interaction measures into multi-criteria assessment of redistricting plans using interactive web mapping. *Cartography and Geographic Information Science* **51**:4, 513-532. [[Crossref](#)]
40. Wanxiang Cai, Friedemann Polzin, Erik Stam. 2024. Mitigating local bias in equity crowdfunding: a financial ecology perspective. *Journal of Economic Geography* **24**:4, 549-565. [[Crossref](#)]
41. Yilan Xu, Sébastien Box-Couillard. 2024. Social learning about climate risks. *Economic Inquiry* **62**:3, 1172-1191. [[Crossref](#)]
42. David Hirshleifer, Lin Peng, Qiguang Wang. 2024. News Diffusion in Social Networks and Stock Market Reactions. *The Review of Financial Studies* **125**. . [[Crossref](#)]
43. Evelyne Brie, Felix Mathieu. 2024. Strained ties in plurinational states: Analysing the social network divide between Canada's two solitudes. *Nations and Nationalism* **30**. . [[Crossref](#)]
44. Benjamin Enke, Raymond Fisman, Luis Mota Freitas, Steven Sun. 2024. Universalism and Political Representation: Evidence from the Field. *American Economic Review: Insights* **6**:2, 214-229. [[Abstract](#)] [[View PDF article](#)] [[PDF with links](#)]
45. Cillian Berragan, Alex Singleton, Alessia Calafiore, Jeremy Morley. 2024. Mapping Great Britain's semantic footprints through a large language model analysis of Reddit comments. *Computers, Environment and Urban Systems* **110**, 102121. [[Crossref](#)]
46. Alex Annan Abakah, Gunchang Kim, Hyacinthe Yirlier Somé. 2024. Social networks and start-up funding. *Finance Research Letters* **64**, 105480. [[Crossref](#)]
47. Xiangpeng Li, Yuqin Jiang, Ali Mostafavi. 2024. Hazard exposure heterophily in socio-spatial networks contributes to post-disaster recovery in low-income populations. *International Journal of Disaster Risk Reduction* **108**, 104537. [[Crossref](#)]
48. Ankitkumar Kariya, Chhavi Shekhawat. 2024. Distance lending & social connectedness. *Journal of Financial Stability* **72**, 101249. [[Crossref](#)]
49. Xuesong You, Carolyn Kousky. 2024. Improving household and community disaster recovery: Evidence on the role of insurance. *Journal of Risk and Insurance* **91**:2, 299-338. [[Crossref](#)]
50. Nadia Ferreira, Isaac T. Kwao, Ingrid L. Potgieter. 2024. Psychological career pre-occupation and social connectedness in Ghanaian education staff's career management. *South African Journal of Business Management* **55**:1. . [[Crossref](#)]
51. Markus Knell, Helmut Stix. 2024. The role of social effects in shaping beliefs about the economy. *Oxford Economic Papers* **90**. . [[Crossref](#)]
52. Oliver Gingrich, Dominik Havsteen-Franklin, Claire Grant, Alain Renaud, Daniel Hignell-Tully. 2024. Participatory presence – social connectedness through collaborative art practices. *International Journal of Performance Arts and Digital Media* **20**:2, 296-320. [[Crossref](#)]
53. Samira S. Abraham, Gianandrea Lanzara, Sara Lazzaroni, Paolo Masella, Mara P. Squicciarini. 2024. Spatial and historical drivers of fake news diffusion: Evidence from anti-Muslim discrimination in India. *Journal of Urban Economics* **141**, 103613. [[Crossref](#)]

54. John Levi Martin, Jan Overgoor, Bogdan State. 2024. Persistence and change in structural signatures of tie formation over time. *Social Networks* **77**, 5-16. [[Crossref](#)]
55. Joseph L. Breeden. 2024. An Age–Period–Cohort Framework for Profit and Profit Volatility Modeling. *Mathematics* **12**:10, 1427. [[Crossref](#)]
56. Debasis Rooj, Anurag Banerjee, Reshmi Sengupta, Prasad Pathak. 2024. Shedding light on consumer sentiments: evidence from India. *Spatial Economic Analysis* **13253**, 1-19. [[Crossref](#)]
57. Xiaofan Liang, César A. Hidalgo, Pierre-Alexandre Balland, Siqi Zheng, Jianghao Wang. 2024. Intercity connectivity and urban innovation. *Computers, Environment and Urban Systems* **109**, 102092. [[Crossref](#)]
58. Linh Thi My Nguyen, Phong Thanh Nguyen. 2024. Determinants of cryptocurrency and decentralized finance adoption - A configurational exploration. *Technological Forecasting and Social Change* **201**, 123244. [[Crossref](#)]
59. Fernando Ferreira, Jeanna Kenney, Benjamin Smith. 2024. Household Mobility, Networks, and Gentrification of Minority Neighborhoods in the United States. *Journal of Labor Economics* **42**:S1, S61-S94. [[Crossref](#)]
60. Sabino Kornrich, Blaine Robbins. 2024. The rise of online dating and racial homogamy in marriage. *Social Science Research* **119**, 102976. [[Crossref](#)]
61. Kiran Tomlinson, Austin R. Benson. 2024. Graph-based methods for discrete choice. *Network Science* **12**:1, 21-40. [[Crossref](#)]
62. Suin Lee, Christos Pantzalis, Jung Chul Park. 2024. Interstate migration networks and stock return comovement. *Journal of Financial Research* **47**:1, 89-121. [[Crossref](#)]
63. Jingbo Hou, Chen Liang, Pei-yu Chen. 2024. How Socially Perceived Threat Shapes Preventive Behavior in the Context of COVID-19. *Production and Operations Management* . [[Crossref](#)]
64. Lawrence McKay, Will Jennings, Gerry Stoker. 2024. Social ties, trust and the geography of discontent. *Cambridge Journal of Regions, Economy and Society* **17**:1, 17-36. [[Crossref](#)]
65. Haiyang Lu, Xiaolin Guo, Chengzheng Li, Wenrong Qian. 2024. Social ties and urban settlement intention of rural-to-urban migrants in China: The mediating role of place attachment and the moderating role of spatial pattern. *Cities* **145**, 104725. [[Crossref](#)]
66. Marianna Frangeskou, Alice Erthal, Rweyemamu Ndibalema. 2024. Managing the tensions of standardized work processes in healthcare operations: The job crafting lens. *Journal of Business Research* **173**, 114459. [[Crossref](#)]
67. James R. Barth, Qinyou Hu, Robin Sickles, Yanfei Sun, Xiaoyu Yu. 2024. Direct and indirect impacts of natural disasters on banks: A spatial framework. *Journal of Financial Stability* **70**, 101194. [[Crossref](#)]
68. Milan de Vries, Jae Yeon Kim, Hahrie Han. 2024. The unequal landscape of civic opportunity in America. *Nature Human Behaviour* **8**:2, 256-263. [[Crossref](#)]
69. Lisa Gianmoena, Vicente Rios. 2024. The diffusion of COVID-19 across Italian provinces: a spatial dynamic panel data approach with common factors. *Regional Studies* **58**:2, 285-305. [[Crossref](#)]
70. Yi Wang, Yangyang Jiang, Baojiang Geng, Ziqi Yan, Xiaorong Wang. 2024. Social networks, network interaction and resilience of B&Bs in rural China. *International Journal of Contemporary Hospitality Management* **36**:2, 400-421. [[Crossref](#)]
71. Luciano Ayala Cantú. 2024. Conectividad social y patrones migratorios: un estudio de caso de México a Estados Unidos. *Entreciencias: Diálogos en la Sociedad del Conocimiento* **12**:26, 1-12. [[Crossref](#)]
72. Michael Down, Duncan Picknoll, Gerard Hoyne, Ben Piggott, Caroline Bulsara. 2024. “When the real stuff happens”: A qualitative descriptive study of the psychosocial outcomes of outdoor adventure education for adolescents. *Journal of Outdoor and Environmental Education* **89**. . [[Crossref](#)]

73. Joseph Gibbons, Joshua Chanin, Tse-Chuan Yang. 2024. Policing a Pandemic in New York City: How Do Community Features Matter in the Location of Social Distancing Violations?. *Social Problems* 71:1, 128-156. [[Crossref](#)]
74. Felipe Carozzi, Sandro Provenzano, Sefi Roth. 2024. Urban density and COVID-19: understanding the US experience. *The Annals of Regional Science* 72:1, 163-194. [[Crossref](#)]
75. Simona Šulíková, Peter Vanya, Lukáš Kováč, Daniel Buc, Rastislav Farkaš. 2024. Use of comprehensive datasets to estimate the Slovak National Transportation Model. *Transportation Research Procedia* 78, 546-553. [[Crossref](#)]
76. Bryce J. Dietrich, Hyein Ko, Payel Sen. 2024. Stand up and be counted: Using traffic cameras to assess voting behavior in real time. *Research & Politics* 11:1. . [[Crossref](#)]
77. Flavia Ioana Patrascu, Ali Mostafavi. 2024. Spatial model for predictive recovery monitoring based on hazard, built environment, and population features and their spillover effects. *Environment and Planning B: Urban Analytics and City Science* 51:1, 39-56. [[Crossref](#)]
78. Francesco Scalamonti. Oceania's Capitalistic Development-Path: Stylized Facts, Distinctive Features, and Evidence of the Cultural Distance from a Cluster of Advanced and Emerging Economies 102, . [[Crossref](#)]
79. Matilde Faralli. 2024. What Drives Beliefs about Climate Risks? Evidence from Financial Analysts. *SSRN Electronic Journal* 150. . [[Crossref](#)]
80. Oğuzhan Çepni. 2024. Fifty Shades of the US States: News Media Coverage and Predictability of House Prices. *SSRN Electronic Journal* 37. . [[Crossref](#)]
81. Wei Li, Boluo Liu, Yuan Zhang. 2024. Social Connectedness and Information Acquisition: Evidence from EDGAR Searches. *SSRN Electronic Journal* 13. . [[Crossref](#)]
82. Brad Cannon, David A. Hirshleifer, Joshua Thornton. 2024. Friends with Benefits: Social Capital and Household Financial Behavior. *SSRN Electronic Journal* 106. . [[Crossref](#)]
83. Ruben Enikolopov, Maria Petrova, Gianluca Russo, David Yanagizawa-Drott. 2024. Socializing Alone: How Online Homophily Has Undermined Social Cohesion in the US. *SSRN Electronic Journal* 110. . [[Crossref](#)]
84. Olga Balakina, Claes Bäckman, Anastasia Parakhonyak. 2024. Beyond Connectivity: Stock Market Participation in a Network. *SSRN Electronic Journal* 108. . [[Crossref](#)]
85. Christine Zhuowei Huang. 2024. Green Neighbors, Greener Neighborhoods. *SSRN Electronic Journal* 80. . [[Crossref](#)]
86. Francesco Scalamonti. 2024. A quantitative and qualitative macroeconomic and sociopolitical outlook of the MEDA transitional economies: development-path, governance climate, and sociocultural factors. *SSRN Electronic Journal* 29. . [[Crossref](#)]
87. In Gyun Baek, Ben Charoenwong, Yupeng Lin. 2024. Reducing Small Entities' Information Disadvantages and Patent Abandonment with Online Accessibility. *SSRN Electronic Journal* 138. . [[Crossref](#)]
88. J. Anthony Cookson, William Mullins, Marina Niessner. 2024. Social Media and Finance. *SSRN Electronic Journal* 118. . [[Crossref](#)]
89. Samuel Bazzi, Martin Fiszbein, Maximiliano Garcia. 2024. The Moral Values of "Rugged Individualism". *SSRN Electronic Journal* 32. . [[Crossref](#)]
90. Raffi E. García, Alberto Ortega. 2024. Racial Protests and Credit Access. *SSRN Electronic Journal* 93. . [[Crossref](#)]
91. Saumitra Jha, Peter Koudijs, Marcos Salgado. Markets under Siege: How Political Beliefs Move Financial Markets 29, . [[Crossref](#)]

92. Francesco Scalamonti. The Cultural Uniqueness of Antipodean Capitalism and Its Historical Development-paths Dependencies **102**, . [[Crossref](#)]
93. Isaiah Hull, Yingjie Qi. The Impact of Finfluencers on Retail Investment **67**, . [[Crossref](#)]
94. Tim Baule. A Structural Account of Protest Occurrence – Evidence from the US **31**, . [[Crossref](#)]
95. Muhammed Yönaç, Leonard Kostovetsky, Lin Peng, Christopher Rauh. Measuring Local Climate Change Attention: Does it Affect Investors and Firms? **5**, . [[Crossref](#)]
96. Sebastien Box-Couillard, Yilan Xu. Does Flood Insurance Help Neighborhoods Recover from Flooding? **104**, . [[Crossref](#)]
97. Francesco Scalamonti. 2024. A quantitative and qualitative macroeconomic and sociopolitical outlook of the MEDA transitional economies: development-paths, governance climate, and sociocultural factors. *National Accounting Review* **6**:3, 407-448. [[Crossref](#)]
98. Abreham Adera. 2023. Do migrant remittances have state de-legitimizing tendencies? A micro-survey based evidence from Africa. *Cogent Economics & Finance* **11**:1. . [[Crossref](#)]
99. Hamed Nilforoshan, Wenli Looi, Emma Pierson, Blanca Villanueva, Nic Fishman, Yiling Chen, John Sholar, Beth Redbird, David Grusky, Jure Leskovec. 2023. Human mobility networks reveal increased segregation in large cities. *Nature* **624**:7992, 586-592. [[Crossref](#)]
100. Yusuf Ransome, Alberto D Valido, Dorothy L Espelage, Graceson L Clements, Crystal Harrell, Caroline Eckel, Natalie Price, Rachel Nassau, Kate Nyhan, Tamara L Taggart. 2023. A systematic review of how social connectedness influences associations between racism and discrimination on health outcomes. *Epidemiologic Reviews* **45**:1, 44-62. [[Crossref](#)]
101. Chien-Chiang Lee, Chuan Zhang, Dan Ma. 2023. Can the Social Network Hinder the Impact of COVID-19 on Economic Uncertainty? New Evidence from China. *Emerging Markets Finance and Trade* **59**:15, 4088-4106. [[Crossref](#)]
102. Benjamin F. Jarvis, Robert D. Mare, Monica K. Nordvik. 2023. Assortative mating, residential choice, and ethnic segregation. *Research in Social Stratification and Mobility* **88**, 100809. [[Crossref](#)]
103. Chia-Fu Liu, Ali Mostafavi. 2023. Revealing hazard-exposure heterophily as a latent characteristic of community resilience in social-spatial networks. *Scientific Reports* **13**:1. . [[Crossref](#)]
104. Eszter Bokányi, Eelke M. Heemskerk, Frank W. Takes. 2023. The anatomy of a population-scale social network. *Scientific Reports* **13**:1. . [[Crossref](#)]
105. Vicente Rios, Mercedes Beltrán-Esteve, Lisa Gianmoena, Jesús Peiró-Palomino, Andrés J Picazo-Tadeo. 2023. Quality of government and women's political empowerment: Evidence from European regions. *Papers in Regional Science* **102**:6, 1067-1097. [[Crossref](#)]
106. Clio Andris, Caglar Koylu, Mason A. Porter. 2023. Human-network regions as effective geographic units for disease mitigation. *EPJ Data Science* **12**:1. . [[Crossref](#)]
107. Ketevan Gallagher, Taylor Anderson, Andrew Crooks, Andreas Züfle. Synthetic Geosocial Network Generation 15-24. [[Crossref](#)]
108. Carlos Llano, Juan Pardo, Santiago Pérez-Balsalobre, Julián Pérez. 2023. Estimating multicountry tourism flows by transport mode. *Annals of Tourism Research* **103**, 103672. [[Crossref](#)]
109. Petre Caraiani. 2023. Oil news shocks, inflation expectations and social connectedness. *Energy Economics* **127**, 107054. [[Crossref](#)]
110. Zhou Lu, Yajie Huang, Peiliang Du, Fang Li, Zhenhui Li. 2023. Pandemics uncertainty and informational globalization in CEE countries: The role of innovation diffusion. *Heliyon* **9**:11, e21489. [[Crossref](#)]
111. Jisung Yoon, Jinseo Park, Jinhyuk Yun, Woo-Sung Jung. 2023. Quantifying knowledge synchronization with the network-driven approach. *Journal of Informetrics* **17**:4, 101455. [[Crossref](#)]

112. Nicolas Crouzet, Apoorv Gupta, Filippo Mezzanotti. 2023. Shocks and Technology Adoption: Evidence from Electronic Payment Systems. *Journal of Political Economy* **131**:11, 3003-3065. [[Crossref](#)]
113. Viral V. Acharya, Richard Berner, Robert Engle, Hyeyoon Jung, Johannes Stroebel, Xuran Zeng, Yihao Zhao. 2023. Climate Stress Testing. *Annual Review of Financial Economics* **15**:1, 291-326. [[Crossref](#)]
114. Subbarao NV, Bindu Chhabra, Manit Mishra. 2023. Fun at Work: Need or Fad? Understanding the Effect of Social Connectedness in Job Search with Social Media. *Global Business Review* **96**. . [[Crossref](#)]
115. Milad Abbasiharofteh, Miriam Krüger, Jan Kinne, David Lenz, Bernd Resch. 2023. The digital layer: alternative data for regional and innovation studies. *Spatial Economic Analysis* **18**:4, 507-529. [[Crossref](#)]
116. Paul Elhorst, Ugo Fratesi, Maria Abreu, Pedro Amaral, Steven Bond-Smith, Coro Chasco, Luisa Corrado, Jan Ditzen, Daniel Felsenstein, Franz Fuerst, Vassilis Monastiriotis, Francesco Quattraro, Dimitrios Tsiotas, Jihai Yu. 2023. Raising the bar (final). *Spatial Economic Analysis* **18**:4, 431-436. [[Crossref](#)]
117. Giang Nguyen, My Nguyen, Anh Viet Pham, Man Duy (Marty) Pham. 2023. Navigating investment decisions with social connectedness: Implications for venture capital. *Journal of Banking & Finance* **155**, 106979. [[Crossref](#)]
118. Yihui Pan, Xiaoxia Peng. 2023. Ancestral connections and corporate alliances. *Journal of Corporate Finance* **82**, 102450. [[Crossref](#)]
119. Martin Obschonka, Sam Tavassoli, P. Jason Rentfrow, Jeff Potter, Samuel D. Gosling. 2023. Innovation and inter-city knowledge spillovers: Social, geographical, and technological connectedness and psychological openness. *Research Policy* **52**:8, 104849. [[Crossref](#)]
120. STIJN VAN NIEUWERBURGH. 2023. Johannes Stroebel: Winner of the 2023 Fischer Black Prize. *The Journal of Finance* **78**:5, 2417-2420. [[Crossref](#)]
121. Daniel W. Elfenbein, Raymond Fisman, Brian McManus. 2023. The Impact of Socioeconomic and Cultural Differences on Online Trade. *Management Science* **69**:10, 6181-6201. [[Crossref](#)]
122. Ezequiel Garcia-Lembergman, Ina Hajdini, John Leer, Mathieu O. Pedemonte, Raphael S. Schoenle. The Expectations of Others . [[Crossref](#)]
123. Wanglin Ma, Enoch Owusu-Sekyere, Hongyun Zheng, Victor Owusu. 2023. Factors influencing smartphone usage of rural farmers: Empirical analysis of five selected provinces in China. *Information Development* **14**. . [[Crossref](#)]
124. Allan Davids, Gideon du Rand, Co-Pierre Georg, Tina Koziol, Joeri Schasfoort. 2023. Social learning in a network model of Covid-19. *Journal of Economic Behavior & Organization* **213**, 271-304. [[Crossref](#)]
125. Bruce Carlin, Tarik Umar, Hanyi Yi. 2023. Deputizing financial institutions to fight elder abuse. *Journal of Financial Economics* **149**:3, 557-577. [[Crossref](#)]
126. Bosco Rowland, Neha Swami, Jennifer Prattley, Jacob Duffy, Jacqui A Macdonald, Francisco Perales, Kayla A Mansour, Brendan Quinn. 2023. Depressive symptoms and social support among Australian men: A 7-year longitudinal study. *Australian & New Zealand Journal of Psychiatry* **57**:9, 1243-1252. [[Crossref](#)]
127. Diletta Goglia, Laura Pollacci, Alina Sirbu. 2023. Dataset of Multi-Aspect Integrated Migration Indicators. *Data* **8**:9, 139. [[Crossref](#)]
128. Ce Fang, Te Bu, Fang Fang. Research on credit-risk models via machine-learning algorithms and logistic regression for predicting CBA consumer behaviour 344-350. [[Crossref](#)]
129. Federico Carril-Caccia, Aitor Garmendia-Lazcano, Asier Minondo. 2023. Social ties and home bias in mergers and acquisitions. *Review of World Economics* **159**:3, 563-593. [[Crossref](#)]

130. Mengya Xia, Caitlin M. Hudac. 2023. Social Connection Constellations and Individual Well-Being Typologies: Using the Loglinear Modeling Approach with Latent Variables. *Journal of Happiness Studies* 24:6, 1991-2012. [[Crossref](#)]
131. Govindapuram Suresh. 2023. Financial Inclusion and Its Impact on Fertility: An Empirical Investigation. *Indian Journal of Human Development* 17:2, 344-358. [[Crossref](#)]
132. Eaman Jahani, Samuel P. Fraiberger, Michael Bailey, Dean Eckles. 2023. Long ties, disruptive life events, and economic prosperity. *Proceedings of the National Academy of Sciences* 120:28. . [[Crossref](#)]
133. Alexandre Afonso, Samir Negash. 2023. Selective European Solidarity: How Recipient Country Characteristics Shape Support for International Redistribution in Europe. *JCMS: Journal of Common Market Studies* . [[Crossref](#)]
134. Jeff Chan. 2023. Forced displacement and migrants' location choices: Evidence from the Japanese-Canadian experience during World War II. *Journal of Economic Behavior & Organization* 211, 206-240. [[Crossref](#)]
135. Chen Lin, Jianghong Zhou, Jing Zhang, Carl Yang, Eugene Agichtein. Graph Neural Network Modeling of Web Search Activity for Real-time Pandemic Forecasting 128-137. [[Crossref](#)]
136. Dimuthu Ratnadiwakara, Buvaneshwaran Venugopal. 2023. Climate risk perceptions and demand for flood insurance. *Financial Management* 52:2, 297-331. [[Crossref](#)]
137. BING HAN, DAVID HIRSHLEIFER, JOHAN WALDEN. 2023. Visibility Bias in the Transmission of Consumption Beliefs and Undersaving. *The Journal of Finance* 78:3, 1647-1704. [[Crossref](#)]
138. Bilgeçağ Aydoğdu, Hanif Samad, Shiqi Bai, Sami Abboud, Ilias Gorantis, Albert Ali Salah. 2023. Analyzing international airtime top-up transfers for migration and mobility. *International Journal of Data Science and Analytics* 23. . [[Crossref](#)]
139. Panle Jia Barwick, Yanyan Liu, Eleonora Patacchini, Qi Wu. 2023. Information, Mobile Communication, and Referral Effects. *American Economic Review* 113:5, 1170-1207. [[Abstract](#)] [[View PDF article](#)] [[PDF with links](#)]
140. Yongwei Nian, Chunyang Wang. 2023. Go with the Politician. *American Economic Journal: Economic Policy* 15:2, 467-496. [[Abstract](#)] [[View PDF article](#)] [[PDF with links](#)]
141. Xiaofan Liang, Joshua Baker, Daniel DellaPosta, Clio Andris. 2023. Is your neighbor your friend? Scan methods for spatial social network hotspot detection. *Transactions in GIS* 27:3, 607-625. [[Crossref](#)]
142. Megan Meyer, Karen Hopkins, Jenny Lee, Nicole Mattocks, Jonalyn Denlinger. 2023. Looking for impact in all the wrong places: Setting realistic expectations and measurable outcomes for small-scale community building initiatives. *Journal of Community Practice* 31:2, 215-234. [[Crossref](#)]
143. Benjamin Lucas, Behzad Vahedi, Morteza Karimzadeh. 2023. A spatiotemporal machine learning approach to forecasting COVID-19 incidence at the county level in the USA. *International Journal of Data Science and Analytics* 15:3, 247-266. [[Crossref](#)]
144. Ruoxi Wang, Qi Wang, Nan Li. 2023. Percolation transitions in urban mobility networks in America's 50 largest cities. *Sustainable Cities and Society* 91, 104435. [[Crossref](#)]
145. Ying Wang, Richard Woodward. A Simple Model on Social Connections, Wages, and Welfare 45, . [[Crossref](#)]
146. Maria Greve, Michael Fritsch, Michael Wyrwich. 2023. Long-term decline of regions and the rise of populism: The case of Germany. *Journal of Regional Science* 63:2, 409-445. [[Crossref](#)]
147. David Parra Valcarce, Santiago Martínez Arias, Sergio Mena Muñoz. Towards a Real Comprehension of Social Networks Effects on Cybermedia Web Traffic: Quantitative Impact in the 27 European Union Countries Most Popular Online Journals 66, . [[Crossref](#)]

148. Jun Li, Melasutra Md. Dali, Nikmatul Adha Nordin. 2023. Connectedness among Urban Parks from the Users' Perspective: A Systematic Literature Review. *International Journal of Environmental Research and Public Health* **20**:4, 3652. [[Crossref](#)]
149. Jing Tang, Napatee Yaibuates, Theerat Tassanai, Natt Leelawat. 2023. Inter- and Intrastate Network Analysis of COVID-19 Spread Using the Social Connectedness Index. *Journal of Disaster Research* **18**:1, 40-47. [[Crossref](#)]
150. Sebastiano Manzan. Big Data and Computational Social Science for Economic Analysis and Policy 231-242. [[Crossref](#)]
151. Theresa Kuchler, Johannes Stroebel. Social Interactions, Resilience, and Access to Economic Opportunity: A Research Agenda for the Field of Computational Social Science 405-419. [[Crossref](#)]
152. Tobin South, Nick Lothian, Takahiro Yabe, Alex 'Sandy' Pentland. Building a Healthier Feed: Private Location Trace Intersection Driven Feed Recommendations 54-63. [[Crossref](#)]
153. Theresa Kuchler, Monika Piazzesi, Johannes Stroebel. Housing market expectations 163-191. [[Crossref](#)]
154. Christopher Carroll, Tao Wang. Epidemiological expectations 779-806. [[Crossref](#)]
155. Joshua Coven, Arpit Gupta, Iris Yao. 2023. JUE Insight: Urban flight seeded the COVID-19 pandemic across the United States. *Journal of Urban Economics* **133**, 103489. [[Crossref](#)]
156. Georgij Alekseev, Safaa Amer, Manasa Gopal, Theresa Kuchler, J. W. Schneider, Johannes Stroebel, Nils Wernerfelt. 2023. The Effects of COVID-19 on U.S. Small Businesses: Evidence from Owners, Managers, and Employees. *Management Science* **69**:1, 7-24. [[Crossref](#)]
157. Young Jae (Jay) Choi. 2023. When Losers Talk: Information Diffusion, Social Norms, and Conversations of Investors. *SSRN Electronic Journal* **62**. . [[Crossref](#)]
158. Viral V. Acharya, Richard Berner, Robert F. Engle, Hyeyoon Jung, Johannes Stroebel, Xuran Zeng, Yihao Zhao. 2023. Climate Stress Testing. *SSRN Electronic Journal* **65**. . [[Crossref](#)]
159. Andreas Mastrosavvas. Socio-Spatial Spillovers of All-Mail Voting: Evidence from North Carolina **15**, . [[Crossref](#)]
160. Samuli Knüpfer, Elias Henrikki Rantapuska, Theresa Spickers. 2023. The Banker in Your Social Network. *SSRN Electronic Journal* **99**. . [[Crossref](#)]
161. John M. Griffin, Samuel Kruger, Prateek Mahajan. 2023. Did Pandemic Relief Fraud Inflate House Prices?. *SSRN Electronic Journal* **59**. . [[Crossref](#)]
162. LINMEI HUANG, Shuting Ada Wang, Qiang Gao, Karl Reiner Lang. 2023. Take It or Leave It? Impact of Investments from Tech Giants on IT-Startups' Future Funding. *SSRN Electronic Journal* **23**. . [[Crossref](#)]
163. Ferdinando Monte, Charly Porcher, Esteban Rossi-Hansberg. 2023. Remote Work and City Structure. *SSRN Electronic Journal* **79**. . [[Crossref](#)]
164. Fernando V. Ferreira, Jeanna Kenney, Ben Smith. 2023. Household Mobility, Networks, and Gentrification of Minority Neighborhoods in the Us. *SSRN Electronic Journal* **32**. . [[Crossref](#)]
165. Ferdinando Monte, Charly Porcher, Esteban Rossi-Hansberg. 2023. Remote Work and City Structure. *SSRN Electronic Journal* **79**. . [[Crossref](#)]
166. J. Anthony Cookson, Marina Niessner. 2023. Investor Disagreement: Daily Measures from Social Media. *SSRN Electronic Journal* **59**. . [[Crossref](#)]
167. Huaixin Wang, Weichen Zhang. 2023. Economic Linkages from the Wisdom of Crowds. *SSRN Electronic Journal* **136**. . [[Crossref](#)]
168. Lin Cong, Pulak Ghosh, Jiasun Li, Qihong Ruan. 2023. Inflation Expectation and Cryptocurrency Investment. *SSRN Electronic Journal* **32**. . [[Crossref](#)]

169. Marit Hinno Saar, Elaine Liu, Eva Loeza-Albino. 2023. Externalities of Marijuana Legalization: Marijuana Use in Non-Legalizing States. *SSRN Electronic Journal* **4683257**. . [[Crossref](#)]
170. Jason (Pang-Li) Chen, Melissa Crumling, Torin McFarland. 2023. Externalities From Localized Labor Mobility Restrictions: The Role of Social Connections. *SSRN Electronic Journal* **218**. . [[Crossref](#)]
171. Kaixin Liu, Jiwei Zhou, Junda Wang. 2023. Can Black Lives Matter Movement Reduce Racial Disparity? Evidence from Medical Crowdfunding. *SSRN Electronic Journal* **68**. . [[Crossref](#)]
172. John M. Griffin, Samuel Kruger, Prateek Mahajan. 2023. Is Fraud Contagious? Social Connections and the Looting of COVID Relief Programs. *SSRN Electronic Journal* **80**. . [[Crossref](#)]
173. Ang LI. Social Connection and Local Bias based on Hometown Identity **56**, . [[Crossref](#)]
174. Filipe Correia, António Martins, Anthony Waikel. 2023. Consumer Credit Without Collateral, Regulation, or Intermediaries. *SSRN Electronic Journal* **45**. . [[Crossref](#)]
175. Travis Dyer, Gerrit Köchling, Peter Limbach. 2023. The Demand for Public Information by Social Connections: Evidence from Facebook Networks. *SSRN Electronic Journal* **125**. . [[Crossref](#)]
176. Alaina M Beauchamp, Christoph U Lehmann, Richard J Medford, Amy E Hughes. 2023. The Association of a Geographically Wide Social Media Network on Depression: County-Level Ecological Analysis. *Journal of Medical Internet Research* **25**, e43623. [[Crossref](#)]
177. Jun Sung Kim, Eleonora Patacchini, Pierre M. Picard, Yves Zenou. 2023. Spatial interactions. *Quantitative Economics* **14**:4, 1295-1335. [[Crossref](#)]
178. Pauline Grosjean, Federico Masera, Hasin Yousaf. 2022. Inflammatory Political Campaigns and Racial Bias in Policing. *The Quarterly Journal of Economics* **138**:1, 413-463. [[Crossref](#)]
179. Debasis Rooj, Anurag Banerjee, Reshmi Sengupta, Prasad Pathak. Shedding Light on Consumer Sentiments: Evidence from India **104**, . [[Crossref](#)]
180. Cornelius Fritz, Emilio Dorigatti, David Rügamer. 2022. Combining graph neural networks and spatio-temporal disease models to improve the prediction of weekly COVID-19 cases in Germany. *Scientific Reports* **12**:1. . [[Crossref](#)]
181. Viktor Stojkoski, Zoran Utkovski, Petar Jolakoski, Dragan Tevdovski, Ljupcho Kocarev. 2022. Correlates of the country differences in the infection and mortality rates during the first wave of the COVID-19 pandemic: evidence from Bayesian model averaging. *Scientific Reports* **12**:1. . [[Crossref](#)]
182. Ben Grunwald, Julian Nyarko, John Rappaport. 2022. Police agencies on Facebook overreport on Black suspects. *Proceedings of the National Academy of Sciences* **119**:45. . [[Crossref](#)]
183. Angela Andersen, Alexandre Garel, Aaron Gilbert, Alireza Tourani-Rad. 2022. Social capital, human capital, and board appointments. *Global Finance Journal* **54**, 100758. [[Crossref](#)]
184. Gizem Koşar, Tyler Ransom, Wilbert van der Klaauw. 2022. Understanding migration aversion using elicited counterfactual choice probabilities. *Journal of Econometrics* **231**:1, 123-147. [[Crossref](#)]
185. Chloé Duvivier, Claire Bussière. 2022. The contingent nature of broadband as an engine for business startups in rural areas. *Journal of Regional Science* **62**:5, 1329-1357. [[Crossref](#)]
186. Lisa Singh, Carole Roan Gresenz. 2022. Social Media Data for Firearms Research: Promise and Perils. *The ANNALS of the American Academy of Political and Social Science* **704**:1, 267-291. [[Crossref](#)]
187. Alaina M Beauchamp, Christoph U Lehmann, Richard J Medford, Amy E Hughes. The Association of a Geographically Wide Social Media Network on Depression: County-Level Ecological Analysis (Preprint) . [[Crossref](#)]
188. Lucila G. Alvarez-Zuzek, Casey M. Zipfel, Shweta Bansal. 2022. Spatial clustering in vaccination hesitancy: The role of social influence and social selection. *PLOS Computational Biology* **18**:10, e1010437. [[Crossref](#)]



189. K. F. Rafikova. 2022. Sociological research in the digital age: forming the Knowledge base of Computational sociology. *Humanities and Social Sciences. Bulletin of the Financial University* **12**:3, 36-40. [[Crossref](#)]
190. Riley Wilson. 2022. The Impact of Social Networks on EITC Claiming Behavior. *The Review of Economics and Statistics* **104**:5, 929-945. [[Crossref](#)]
191. Muhammad Farooq Ahmad, Eric De Bodt, Helen Bollaert. 2022. Mergers and Acquisitions Across Cultures. *Finance* **Vol. 43**:3, 37-117. [[Crossref](#)]
192. Christina Peterson, Ruben Ortiz, Louis Rocconi. 2022. Community Food Security: The Multi-Level Association Between Social Capital, Economic Capital, and Diet Quality. *International Journal of Community Well-Being* **5**:3, 571-585. [[Crossref](#)]
193. Christos A. Makridis. 2022. The social transmission of economic sentiment on consumption. *European Economic Review* **148**, 104232. [[Crossref](#)]
194. Christopher H. Seto, Corina Graif, Aria Khademi, Vasant G. Honavar, Claire E. Kelling. 2022. Connected in health: Place-to-place commuting networks and COVID-19 spillovers. *Health & Place* **77**, 102891. [[Crossref](#)]
195. Rupali Tamuly, Pranab Mukhopadhyay. 2022. Natural disasters and well-being in India: A household-level panel data analysis. *International Journal of Disaster Risk Reduction* **79**, 103158. [[Crossref](#)]
196. Raj Chetty, Matthew O. Jackson, Theresa Kuchler, Johannes Stroebel, Nathaniel Hendren, Robert B. Fluegge, Sara Gong, Federico Gonzalez, Armelle Grondin, Matthew Jacob, Drew Johnston, Martin Koenen, Eduardo Laguna-Muggenburg, Florian Mudekereza, Tom Rutter, Nicolaj Thor, Wilbur Townsend, Ruby Zhang, Mike Bailey, Pablo Barberá, Monica Bhole, Nils Wernerfelt. 2022. Social capital I: measurement and associations with economic mobility. *Nature* **608**:7921, 108-121. [[Crossref](#)]
197. Raj Chetty, Matthew O. Jackson, Theresa Kuchler, Johannes Stroebel, Nathaniel Hendren, Robert B. Fluegge, Sara Gong, Federico Gonzalez, Armelle Grondin, Matthew Jacob, Drew Johnston, Martin Koenen, Eduardo Laguna-Muggenburg, Florian Mudekereza, Tom Rutter, Nicolaj Thor, Wilbur Townsend, Ruby Zhang, Mike Bailey, Pablo Barberá, Monica Bhole, Nils Wernerfelt. 2022. Social capital II: determinants of economic connectedness. *Nature* **608**:7921, 122-134. [[Crossref](#)]
198. Angela Andersen, Alexandre Garel, Aaron Gilbert, Alireza Tourani-Rad. 2022. Disentangling Director Attributes: Human Capital versus Social Capital of Directors. *Journal of Risk and Financial Management* **15**:8, 336. [[Crossref](#)]
199. Muhammad Farooq Ahmad, Eric De Bodt, Helen Bollaert. 2022. Mergers and Acquisitions Across Cultures. *Finance* **Prépublication**:0, Ib-LXXXIb. [[Crossref](#)]
200. Michael Bailey, Drew Johnston, Theresa Kuchler, Johannes Stroebel, Arlene Wong. 2022. Peer Effects in Product Adoption. *American Economic Journal: Applied Economics* **14**:3, 488-526. [[Abstract](#)] [[View PDF article](#)] [[PDF with links](#)]
201. Nicholas Buttrick, Jessica Mazen. 2022. Historical prevalence of slavery predicts contemporary American gun ownership. *PNAS Nexus* **1**:3. . [[Crossref](#)]
202. Asad Ali Rind, Saeed Akbar, Sabri Boubaker, Souad Lajili-Jarjir, Sabur Mollah. 2022. The Role of Peer Effects in Corporate Employee Welfare Policies. *British Journal of Management* **33**:3, 1609-1631. [[Crossref](#)]
203. Xiaofan Liang, Seolha Lee, Hanzhou Chen, Benjamin De la Peña, Clio Andris. 2022. Characteristics of Jetties and Little Boxes: An Extensibility Study Using the Neighborhood Connectivity Survey. *Social Inclusion* **10**:3. . [[Crossref](#)]
204. Dan Ma, Chuan Zhang, Yarong Hui, Bing Xu. 2022. Economic uncertainty spillover and social networks. *Journal of Business Research* **145**, 454-467. [[Crossref](#)]

205. Muxin Zhai, Ruby P. Kishan, Dean Showalter. 2022. Social capital and suicidal behaviors: Evidence from the United States counties. *Journal of Behavioral and Experimental Economics* **98**, 101856. [[Crossref](#)]
206. Theresa Kuchler, Yan Li, Lin Peng, Johannes Stroebel, Dexin Zhou. 2022. Social Proximity to Capital: Implications for Investors and Firms. *The Review of Financial Studies* **35**:6, 2743-2789. [[Crossref](#)]
207. Zhongchen Hu. 2022. Social interactions and households' flood insurance decisions. *Journal of Financial Economics* **144**:2, 414-432. [[Crossref](#)]
208. Timothy Fraser. 2022. The Road More Traveled: Evacuation Networks in the US and Japan. *Environment and Behavior* **54**:4, 833-863. [[Crossref](#)]
209. Mark Igra. 2022. Donor Financial Capacity Drives Racial Inequality in Medical Crowdsourced Funding. *Social Forces* **100**:4, 1856-1883. [[Crossref](#)]
210. Sindhu Kiranmai Ernala, Moira Burke, Alex Leavitt, Nicole B. Ellison. Mindsets Matter: How Beliefs About Facebook Moderate the Association Between Time Spent and Well-Being 1-13. [[Crossref](#)]
211. Joseph Gibbons, Tse-Chuan Yang, Eyal Oren. 2022. Community Boosts Immunity? Exploring the Relationship Between Social Capital and COVID-19 Social Distancing. *Spatial Demography* **10**:1, 75-105. [[Crossref](#)]
212. Jana B. Berkessel, Tobias Ebert, Jochen E. Gebauer, Thorsteinn Jonsson, Shigehiro Oishi. 2022. Pandemics Initially Spread Among People of Higher (Not Lower) Social Status: Evidence From COVID-19 and the Spanish Flu. *Social Psychological and Personality Science* **13**:3, 722-733. [[Crossref](#)]
213. Matthias Flückiger, Erik Hornung, Mario Larch, Markus Ludwig, Allard Mees. 2022. Roman Transport Network Connectivity and Economic Integration. *The Review of Economic Studies* **89**:2, 774-810. [[Crossref](#)]
214. Martin Eichenbaum, Sergio Rebelo, Arlene Wong. 2022. State-Dependent Effects of Monetary Policy: The Refinancing Channel. *American Economic Review* **112**:3, 721-761. [[Abstract](#)] [[View PDF article](#)] [[PDF with links](#)]
215. Claes Bäckman, Tobin Hanspal. 2022. Participation and losses in multi-level marketing: Evidence from a Federal Trade Commission settlement. *FINANCIAL PLANNING REVIEW* **5**:1. . [[Crossref](#)]
216. Saloni Shah, Aos Muluhaish, Kayhan Zrar Ghafoor, Halgurd S. Maghdid. 2022. Prediction of global spread of COVID-19 pandemic: a review and research challenges. *Artificial Intelligence Review* **55**:3, 1607-1628. [[Crossref](#)]
217. Andreas Diemer, Tanner Regan. 2022. No inventor is an island: Social connectedness and the geography of knowledge flows in the US. *Research Policy* **51**:2, 104416. [[Crossref](#)]
218. Florian Haelg. 2022. Drivers of financial globalisation: The role of informational frictions. *The World Economy* **45**:3, 608-636. [[Crossref](#)]
219. Junteng Jia, Austin R. Benson. 2022. A Unifying Generative Model for Graph Learning Algorithms: Label Propagation, Graph Convolutions, and Combinations. *SIAM Journal on Mathematics of Data Science* **4**:1, 100-125. [[Crossref](#)]
220. Madeleine I. G. Daepf. 2022. Small-area moving ratios and the spatial connectivity of neighborhoods: Insights from consumer credit data. *Environment and Planning B: Urban Analytics and City Science* **49**:3, 1129-1146. [[Crossref](#)]
221. Marzena Czarnecka, Grzegorz Kinelski, Magdalena Stefańska, Mateusz Grzesiak, Borys Budka. 2022. Social Media Engagement in Shaping Green Energy Business Models. *Energies* **15**:5, 1727. [[Crossref](#)]
222. Shiwei Ye, Guoxiong Zhang. 2022. Measuring social connectedness using air travel data. *Applied Economics Letters* **29**:3, 188-194. [[Crossref](#)]

223. In Ji Jang, Namho Kang, Ari Yezegel. 2022. Common ownership, price informativeness, and corporate investment. *Journal of Banking & Finance* **135**, 106373. [[Crossref](#)]
224. VESA PURSIAINEN. 2022. Cultural Biases in Equity Analysis. *The Journal of Finance* **77**:1, 163-211. [[Crossref](#)]
225. A Rebecca Reuber, Eileen Fischer. 2022. Relying on the engagement of others: A review of the governance choices facing social media platform start-ups. *International Small Business Journal: Researching Entrepreneurship* **40**:1, 3-22. [[Crossref](#)]
226. C Matthew Leister, Yves Zenou, Junjie Zhou. 2022. Social Connectedness and Local Contagion. *The Review of Economic Studies* **89**:1, 372-410. [[Crossref](#)]
227. N. V. Subbarao, Bindu Chhabra, Manit Mishra. Antecedents and Consequences of Social Media Usage Behavior in Job Search: A Qualitative Study Toward a Conceptual Framework 215-241. [[Crossref](#)]
228. Theresa Kuchler, Dominic Russel, Johannes Stroebel. 2022. JUE Insight: The geographic spread of COVID-19 correlates with the structure of social networks as measured by Facebook. *Journal of Urban Economics* **127**, 103314. [[Crossref](#)]
229. Jeffrey Brinkman, Kyle Mangum. 2022. JUE Insight: The Geography of Travel Behavior in the Early Phase of the COVID-19 Pandemic. *Journal of Urban Economics* **127**, 103384. [[Crossref](#)]
230. Cornelius Fritz, Göran Kauermann. 2022. On the Interplay of Regional Mobility, Social Connectedness and The Spread of COVID-19 in Germany. *Journal of the Royal Statistical Society Series A: Statistics in Society* **185**:1, 400-424. [[Crossref](#)]
231. Vilma Todri. 2022. Frontiers: The Impact of Ad-Blockers on Online Consumer Behavior. *Marketing Science* **41**:1, 7-18. [[Crossref](#)]
232. Xiaoqing Zhou. 2022. Financial Technology and the Transmission of Monetary Policy: The Role of Social Networks. *SSRN Electronic Journal* **62**. . [[Crossref](#)]
233. Theresa Kuchler, Johannes Stroebel. 2022. Social Interactions, Resilience, and Access to Economic Opportunity: A Research Agenda for the Field of Computational Social Science. *SSRN Electronic Journal* **31**. . [[Crossref](#)]
234. Sean Flynn, Jing Wang. 2022. Social Connections and Bank Deposit Funding. *SSRN Electronic Journal* **126**. . [[Crossref](#)]
235. Theresa Kuchler, Monika Piazzesi, Johannes Stroebel. 2022. Housing Market Expectations. *SSRN Electronic Journal* **85**. . [[Crossref](#)]
236. Theresa Kuchler, Monika Piazzesi, Johannes Stroebel. 2022. Housing Market Expectations. *SSRN Electronic Journal* **85**. . [[Crossref](#)]
237. Pengfei Sui, Baolian Wang. 2022. Social Transmission Bias: Evidence from an Online Investor Platform. *SSRN Electronic Journal* **33**. . [[Crossref](#)]
238. Lin Peng, Qiguang Wang, Dexin Zhou. 2022. Social Networks, Trading, and Liquidity. *SSRN Electronic Journal* **125**. . [[Crossref](#)]
239. Kai Mäckle, Stefan Ruenzi. 2022. The Social Transmission of Non-Infectious Diseases: Evidence from the Opioid Epidemic. *SSRN Electronic Journal* **10**. . [[Crossref](#)]
240. Riley Wilson. 2022. The Isolated States of America: Home State Bias, State Identity, and the Impact of State Borders on Mobility. *SSRN Electronic Journal* **32**. . [[Crossref](#)]
241. James R. Barth, Qinyou Hu, Robin C. Sickles, Yanfei Sun, Fisher Yu. 2022. Spillover Effects of Natural Disasters on Banks: A Spatial Framework. *SSRN Electronic Journal* **80**. . [[Crossref](#)]
242. Hisham Farag, Biwesh Neupane, Andrew P. Marshall, Santosh Koirala. 2022. The Impact of Sovereign Wealth Funds Investment on Firm ESG Reputation Risk. *SSRN Electronic Journal* **30**. . [[Crossref](#)]

243. Jun Kim, Eleonora Patacchini, Pierre M. Picard, Yves Zenou. 2022. Spatial Interactions. *SSRN Electronic Journal* 37. . [[Crossref](#)]
244. Xuelin Li, Zihan Ye. 2022. Propagation of the Opioid Epidemic through the Banking Network. *SSRN Electronic Journal* 137. . [[Crossref](#)]
245. Olga Balakina, Claes Bäckman, Andreas Hackethal, Tobin Hanspal, Dominique Lammer. 2022. Good Peers, Good Apples? Peer Effects in Portfolio Quality. *SSRN Electronic Journal* 67. . [[Crossref](#)]
246. Jinoug Jeung, Jaemin Lee. 2022. Social Connection and Financing Cost of Municipal Governments. *SSRN Electronic Journal* 65. . [[Crossref](#)]
247. Jyothsna Harithsa. 2022. Disconnecting Financial Misconduct: Social Connectedness and Misconduct in Financial Advising. *SSRN Electronic Journal* 113. . [[Crossref](#)]
248. Lin Peng, Sheridan Titman, Muhammed Yönaç, Dexin Zhou. 2022. Social Ties, Comovements, and Predictable Returns. *SSRN Electronic Journal* 2. . [[Crossref](#)]
249. Raffi E. García, Alberto Ortega. 2022. Racial Protests and Credit Access. *SSRN Electronic Journal* 93. . [[Crossref](#)]
250. Felix Dornseifer, Oliver Rehbein. 2022. Invest in Friends or Foreigners? The Role of Social Connectedness in Foreign Direct Investment. *SSRN Electronic Journal* 121. . [[Crossref](#)]
251. Terence Kealey, Martin Ricketts. The Contribution Good as the Foundation of the Industrial Revolution 19-57. [[Crossref](#)]
252. Honglu Zhang, Yonghui Xu, Lei Liu, Xudong Lu, Xijie Lin, Zhongmin Yan, Lizhen Cui, Chunyan Miao. Multi-modal Information Fusion-powered Regional Covid-19 Epidemic Forecasting 779-784. [[Crossref](#)]
253. Serhii Puhach, Kostyantyn Mezentsev. 2021. The unevenly absorbed and induced intra-regional Facebook adoption in Western Ukraine. *AUC GEOGRAPHICA* 56:2, 157-167. [[Crossref](#)]
254. Inho Hong, Alex Rutherford, Manuel Cebrian. 2021. Social mobilization and polarization can create volatility in COVID-19 pandemic control. *Applied Network Science* 6:1. . [[Crossref](#)]
255. Jijian Fan, Daniel Friedman, Jonathan Gair, Sriya Iyer, Bartosz Redlicki, Chander Velu. 2021. A simulation study of how religious fundamentalism takes root. *Journal of Economic Behavior & Organization* 192, 465-481. [[Crossref](#)]
256. Gergő Tóth, Johannes Wachs, Riccardo Di Clemente, Ákos Jakobi, Bence Ságvári, János Kertész, Balázs Lengyel. 2021. Inequality is rising where social network segregation interacts with urban topology. *Nature Communications* 12:1. . [[Crossref](#)]
257. Behzad Vahedi, Morteza Karimzadeh, Hamidreza Zoraghein. 2021. Spatiotemporal prediction of COVID-19 cases using inter- and intra-county proxies of human interactions. *Nature Communications* 12:1. . [[Crossref](#)]
258. Zhenlong Li, Xiao Huang, Xinyue Ye, Yuqin Jiang, Yago Martin, Huan Ning, Michael E. Hodgson, Xiaoming Li. 2021. Measuring global multi-scale place connectivity using geotagged social media data. *Scientific Reports* 11:1. . [[Crossref](#)]
259. Dimuthu Ratnadiwakara. 2021. Flooded Social Connections. *The Quarterly Journal of Finance* 11:04. . [[Crossref](#)]
260. David Victorson, Gretchen Doninger, Scott Victorson, Gwen Victorson, Lars Hall, Carly Maletich, Bradley R. Corr, Kathy Scortino, Zachary Burns, Lori Allen, Ian Rosa, Kelley Quirk, Adekunle Adegbemi, Johanna Strokoff, Kile Zuidema, Kelle Sajdak, Todd Mckibben, Angie Roberts, Thomas W. McDade, Amanda Boes, Katie McAlinden, Karen Arredondo, Christina Sauer, Kristin Smith, John M. Salsman. 2021. Psychosocial and Biological Outcomes of Immersive, Mindfulness-Based Treks in Nature for Groups of Young Adults and Caregivers Affected by Cancer: Results from a Single

- Arm Program Evaluation from 2016–2021. *International Journal of Environmental Research and Public Health* **18**:23, 12622. [[Crossref](#)]
261. Michele Valsecchi, Ruben Durante. 2021. Internal migration networks and mortality in home communities: Evidence from Italy during the Covid-19 pandemic. *European Economic Review* **140**, 103890. [[Crossref](#)]
262. Theresa Kuchler, Johannes Stroebel. 2021. Social Finance. *Annual Review of Financial Economics* **13**:1, 37-55. [[Crossref](#)]
263. Beth Prusaczyk, Kathryn Pietka, Joshua M Landman, Douglas A Luke. Utility of Facebook's Social Connectedness Index in Modeling COVID-19 Spread: Exponential Random Graph Modeling Study (Preprint) . [[Crossref](#)]
264. Richard Bluhm, Maxim Pinkovskiy. 2021. The spread of COVID-19 and the BCG vaccine: A natural experiment in reunified Germany. *The Econometrics Journal* **24**:3, 353-376. [[Crossref](#)]
265. Emily Breza, Fatima Cody Stanford, Marcella Alsan, Burak Alsan, Abhijit Banerjee, Arun G. Chandrasekhar, Sarah Eichmeyer, Traci Glushko, Paul Goldsmith-Pinkham, Kelly Holland, Emily Hoppe, Mohit Karnani, Sarah Liegl, Tristan Loisel, Lucy Ogbu-Nwobodo, Benjamin A. Olken, Carlos Torres, Pierre-Luc Vautrey, Erica T. Warner, Susan Wootton, Esther Duflo. 2021. Effects of a large-scale social media advertising campaign on holiday travel and COVID-19 infections: a cluster randomized controlled trial. *Nature Medicine* **27**:9, 1622-1628. [[Crossref](#)]
266. Shenghao Yang, Priyabrata Senapati, Di Wang, Chris T. Bauch, Kimon Fountoulakis. 2021. Targeted pandemic containment through identifying local contact network bottlenecks. *PLOS Computational Biology* **17**:8, e1009351. [[Crossref](#)]
267. Rex N. Ali, Harvey Rubin, Saswati Sarkar. 2021. Countering the potential re-emergence of a deadly infectious disease—Information warfare, identifying strategic threats, launching countermeasures. *PLOS ONE* **16**:8, e0256014. [[Crossref](#)]
268. Karsten Müller, Carlo Schwarz. 2021. Fanning the Flames of Hate: Social Media and Hate Crime. *Journal of the European Economic Association* **19**:4, 2131-2167. [[Crossref](#)]
269. SURESH NAIDU, JAMES A. ROBINSON, LAUREN E. YOUNG. 2021. Social Origins of Dictatorships: Elite Networks and Political Transitions in Haiti. *American Political Science Review* **115**:3, 900-916. [[Crossref](#)]
270. Michael Zhao, David Holtz, Sinan Aral. 2021. Interdependent program evaluation: Geographic and social spillovers in COVID-19 closures and reopenings in the United States. *Science Advances* **7**:31. . [[Crossref](#)]
271. Philippe Lemey, Nick Ruktanonchai, Samuel L. Hong, Vittoria Colizza, Chiara Poletto, Frederik Van den Broeck, Mandev S. Gill, Xiang Ji, Anthony Lévasseur, Bas B. Oude Munnink, Marion Koopmans, Adam Sadilek, Shengjie Lai, Andrew J. Tatem, Guy Baele, Marc A. Suchard, Simon Dellicour. 2021. Untangling introductions and persistence in COVID-19 resurgence in Europe. *Nature* **595**:7869, 713-717. [[Crossref](#)]
272. Konstantinos Gavriilidis. 2021. Social connectedness and tourism demand. *Current Issues in Tourism* **24**:14, 1930-1934. [[Crossref](#)]
273. Maxim Ananyev, Michael Poyker, Yuan Tian. 2021. The safest time to fly: pandemic response in the era of Fox News. *Journal of Population Economics* **34**:3, 775-802. [[Crossref](#)]
274. Cuong Viet Nguyen. 2021. Can money buy friends? Evidence from a natural experiment. *European Economic Review* **136**, 103747. [[Crossref](#)]
275. Tamás Egedy, Bence Ságvári. 2021. Urban geographical patterns of the relationship between mobile communication, social networks and economic development – the case of Hungary. *Hungarian Geographical Bulletin* **70**:2, 129-148. [[Crossref](#)]

276. Erol Akçay, David Hirshleifer. 2021. Social finance as cultural evolution, transmission bias, and market dynamics. *Proceedings of the National Academy of Sciences* **118**:26. . [[Crossref](#)]
277. . Data as a resource for the private sector 91-116. [[Crossref](#)]
278. Andreas Filser, Richard Preetz. 2021. Do Local Sex Ratios Approximate Subjective Partner Markets?. *Human Nature* **32**:2, 406-433. [[Crossref](#)]
279. Pritha Guha, Avijit Bansal, Apratim Guha, Anindya S Chakrabarti. 2021. Gravity and depth of social media networks. *Journal of Complex Networks* **9**:2. . [[Crossref](#)]
280. Damian J. Ruck, R. Alexander Bentley, Joshua Borycz. 2021. Early warning of vulnerable counties in a pandemic using socio-economic variables. *Economics & Human Biology* **41**, 100988. [[Crossref](#)]
281. Abhilasha Sahay. The Silenced Women: Can Public Activism Stimulate Reporting of Violence against Women? **204**, . [[Crossref](#)]
282. Michael Bailey, Abhinav Gupta, Sebastian Hillenbrand, Theresa Kuchler, Robert Richmond, Johannes Stroebel. 2021. International trade and social connectedness. *Journal of International Economics* **129**, 103418. [[Crossref](#)]
283. Theresa Kuchler, Michaela Pagel. 2021. Sticking to your plan: The role of present bias for credit card paydown. *Journal of Financial Economics* **139**:2, 359-388. [[Crossref](#)]
284. Fabio Manfredini, Stefano Saloriani. Exploring New Workplaces with Social Network Analysis 33-49. [[Crossref](#)]
285. Xiaofan Liang, Yuhao Kang. A Review of Spatial Network Insights and Methods in the Context of Planning: Applications, Challenges, and Opportunities 71-91. [[Crossref](#)]
286. Fabio Milani. 2021. COVID-19 outbreak, social response, and early economic effects: a global VAR analysis of cross-country interdependencies. *Journal of Population Economics* **34**:1, 223-252. [[Crossref](#)]
287. Zhanbing Xiao, Yuxiang Zheng. 2021. Spillover Effects of Banks' Specialization in Corporate Lending on Mortgage Lending: The Industry Expertise Channel. *SSRN Electronic Journal* **79**. . [[Crossref](#)]
288. Matthew O. Jackson. 2021. Inequality's Economic and Social Roots: The Role of Social Networks and Homophily. *SSRN Electronic Journal* **25**. . [[Crossref](#)]
289. D Sadish. 2021. Trauma Propagation in Social Networks. *SSRN Electronic Journal* **111**. . [[Crossref](#)]
290. David A. Hirshleifer, Lin Peng, Qiguang Wang. 2021. Social Networks and Market Reactions to Earnings News. *SSRN Electronic Journal* **125**. . [[Crossref](#)]
291. Quentin Dupont. 2021. Do Cultural Institutions Affect Investor Trust? Evidence from the U.S. Catholic Clergy Abuse Scandal. *SSRN Electronic Journal* **1**. . [[Crossref](#)]
292. Kim Fe Cramer, Naz Koont. 2021. Peer Effects in Deposit Markets. *SSRN Electronic Journal* **337**. . [[Crossref](#)]
293. Christian Kubitzka. 2021. Investor-Driven Corporate Finance: Evidence from Insurance Markets. *SSRN Electronic Journal* **59**. . [[Crossref](#)]
294. Riley Wilson. 2021. Isolated States of America: The Impact of State Borders on Mobility and Regional Labor Market Adjustments. *SSRN Electronic Journal* **32**. . [[Crossref](#)]
295. Ashlynn R Daughton, Courtney D Shelley, Martha Barnard, Dax Gerts, Chrysm Watson Ross, Isabel Crooker, Gopal Nadiga, Nilesh Mukundan, Nidia Yadira Vaquera Chavez, Nidhi Parikh, Travis Pitts, Geoffrey Fairchild. 2021. Mining and Validating Social Media Data for COVID-19-Related Human Behaviors Between January and July 2020: Infodemiology Study. *Journal of Medical Internet Research* **23**:5, e27059. [[Crossref](#)]
296. Beth Prusaczyk, Kathryn Pietka, Joshua M Landman, Douglas A Luke. 2021. Utility of Facebook's Social Connectedness Index in Modeling COVID-19 Spread: Exponential Random Graph Modeling Study. *JMIR Public Health and Surveillance* **7**:12, e33617. [[Crossref](#)]

297. László Lőrincz, Guilherme Kenji Chihaya, Anikó Hannák, Dávid Takács, Balázs Lengyel, Rikard Eriksson. 2020. Global connections and the structure of skills in local co-worker networks. *Applied Network Science* 5:1. . [\[Crossref\]](#)
298. Bibo Liu, Huijun Wang, Jianfeng Yu, Shen Zhao. 2020. Time-varying demand for lottery: Speculation ahead of earnings announcements. *Journal of Financial Economics* 138:3, 789-817. [\[Crossref\]](#)
299. Mason Youngblood. 2020. Extremist ideology as a complex contagion: the spread of far-right radicalization in the United States between 2005 and 2017. *Humanities and Social Sciences Communications* 7:1. . [\[Crossref\]](#)
300. J Anthony Cookson, Joseph E Engelberg, William Mullins. 2020. Does Partisanship Shape Investor Beliefs? Evidence from the COVID-19 Pandemic. *The Review of Asset Pricing Studies* 10:4, 863-893. [\[Crossref\]](#)
301. Ben Charoenwong, Alan Kwan, Vesa Pursiainen. 2020. Social connections with COVID-19-affected areas increase compliance with mobility restrictions. *Science Advances* 6:47. . [\[Crossref\]](#)
302. Ivan Franch-Pardo, Brian M. Napoletano, Fernando Rosete-Verges, Lawal Billa. 2020. Spatial analysis and GIS in the study of COVID-19. A review. *Science of The Total Environment* 739, 140033. [\[Crossref\]](#)
303. Konstantin Büchel, Maximilian v. Ehrlich. 2020. Cities and the structure of social interactions: Evidence from mobile phone data. *Journal of Urban Economics* 119, 103276. [\[Crossref\]](#)
304. Zhang Zhao, Shuang Jing, Zhimin Yan, Lin Yu. 2020. Social change and birth cohort decrease in social support for older adults in China: A cross-temporal meta-analysis, 1994-2018. *Health & Social Care in the Community* 28:5, 1438-1447. [\[Crossref\]](#)
305. David Holtz, Michael Zhao, Seth G. Benzell, Cathy Y. Cao, Mohammad Amin Rahimian, Jeremy Yang, Jennifer Allen, Avinash Collis, Alex Moehring, Tara Sowrirajan, Dipayan Ghosh, Yunhao Zhang, Paramveer S. Dhillon, Christos Nicolaides, Dean Eckles, Sinan Aral. 2020. Interdependence and the cost of uncoordinated responses to COVID-19. *Proceedings of the National Academy of Sciences* 117:33, 19837-19843. [\[Crossref\]](#)
306. Ruqia Khan, Tahir Mumtaz Awan, Tayyba Fatima, Maria Javed. 2020. Driving forces of green consumption in sharing economy. *Management of Environmental Quality: An International Journal* 32:1, 41-63. [\[Crossref\]](#)
307. Fulya Y. Ersoy. 2020. The effects of the great recession on college majors. *Economics of Education Review* 77, 102018. [\[Crossref\]](#)
308. Allan Davids, Gideon Du Rand, Co-Pierre Georg, Tina Koziol, Joeri Anton Schasfoort. Social Learning in a Network Model of Covid-19 1, . [\[Crossref\]](#)
309. Michael Bailey, Patrick Farrell, Theresa Kuchler, Johannes Stroebel. 2020. Social connectedness in urban areas. *Journal of Urban Economics* 118, 103264. [\[Crossref\]](#)
310. Leila Hedayatifar, Alfredo J. Morales, Yaneer Bar-Yam. 2020. Geographical fragmentation of the global network of Twitter communications. *Chaos: An Interdisciplinary Journal of Nonlinear Science* 30:7. . [\[Crossref\]](#)
311. Ni Huang, Gordon Burtch, Yili Hong, Paul A. Pavlou. 2020. Unemployment and Worker Participation in the Gig Economy: Evidence from an Online Labor Market. *Information Systems Research* 31:2, 431-448. [\[Crossref\]](#)
312. Fabio Milani. COVID-19 Outbreak, Social Response, and Early Economic Effects: A Global VAR Analysis of Cross-Country Interdependencies 163, . [\[Crossref\]](#)
313. Viktor Stojkoski, Zoran Utkovski, Petar Jolakoski, Dragan Tevdovski, Ljupco Kocarev. The socio-economic determinants of the coronavirus disease (COVID-19) pandemic 395, . [\[Crossref\]](#)

314. Monika Piazzesi, Martin Schneider, Johannes Stroebel. 2020. Segmented Housing Search. *American Economic Review* **110**:3, 720-759. [[Abstract](#)] [[View PDF article](#)] [[PDF with links](#)]
315. Matthew O. Jackson. 2020. A typology of social capital and associated network measures. *Social Choice and Welfare* **54**:2-3, 311-336. [[Crossref](#)]
316. Guanghua Chi, Bogdan State, Joshua E. Blumenstock, Lada Adamic. Who Ties the World Together? Evidence from a Large Online Social Network 451-465. [[Crossref](#)]
317. Michael Bailey, Drew Johnston, Theresa Kuchler, Dominic Russel, Bogdan State, Johannes Stroebel. The Determinants of Social Connectedness in Europe 1-14. [[Crossref](#)]
318. Yanni Hu, Karl Lang. How Social Networks Dynamics can Affect Collaborative Decision Making on Crowdfunding Platforms 3-17. [[Crossref](#)]
319. Christine Laudenbach, Ulrike Malmendier, Alexandra Niessen-Ruenzi. 2020. The Long-Lasting Effects of Living under Communism on Attitudes towards Financial Markets. *SSRN Electronic Journal* **130**. . [[Crossref](#)]
320. Viktor Stojkoski, Zoran Utkovski, Petar Jolakoski, Dragan Tevdovski, Ljupco Kocarev. 2020. The Socio-Economic Determinants of the Coronavirus Disease (COVID-19) Pandemic. *SSRN Electronic Journal* **9**. . [[Crossref](#)]
321. Christos Makridis, Cary Wu. 2020. Ties That Bind (and Social Distance): How Social Capital Helps Communities Weather the COVID-19 Pandemic. *SSRN Electronic Journal* **6**. . [[Crossref](#)]
322. Christos Makridis, Tao Wang. 2020. Learning from Friends in a Pandemic: Social Networks and the Macroeconomic Response of Consumption. *SSRN Electronic Journal* **115**. . [[Crossref](#)]
323. Pauline A. Grosjean, Federico Masera, Hasin Yousaf. 2020. Whistle the Racist Dogs: Political Campaigns and Police Stops. *SSRN Electronic Journal* **70**. . [[Crossref](#)]
324. Allan Davids, Gideon du Rand, Co-Pierre Georg, Tina Koziol, Joeri Schasfoort. 2020. SABCoM: A Spatial Agent-Based COVID-19 Model. *SSRN Electronic Journal* **325**. . [[Crossref](#)]
325. Erol Akcay, David A. Hirshleifer. 2020. New Evolutionary Finance: Social Transmission Bias and Cultural Evolution in Financial Markets. *SSRN Electronic Journal* **192**. . [[Crossref](#)]
326. Dimuthu Ratnadiwakara. 2020. The Impact of Social Networks on Flood Insurance Take-up and Climate Risk Perceptions after Hurricane Harvey. *SSRN Electronic Journal* **126**. . [[Crossref](#)]
327. Maxim Ananyev, Mikhail Poyker, Yuan Tian. 2020. The Safest Time to Fly: Pandemic Response in the Era of Fox News. *SSRN Electronic Journal* **130**. . [[Crossref](#)]
328. In Ji Jang, Namho Kang, Ari Yezegel. 2020. Common Ownership, Price Informativeness, and Corporate Investment. *SSRN Electronic Journal* **95**. . [[Crossref](#)]
329. Lin Peng, Linyi Zhang. 2020. Social Ties and Peer Effects in Crowdfunding Markets. *SSRN Electronic Journal* **14**. . [[Crossref](#)]
330. Zhongchen Hu. 2020. Salience and Households' Flood Insurance Decisions. *SSRN Electronic Journal* **126**. . [[Crossref](#)]
331. Yihui Pan, Xiaoxia Peng. 2020. Ancestral Connection and Alliances. *SSRN Electronic Journal* **38**. . [[Crossref](#)]
332. Giorgio Topa. 2019. Social and spatial networks in labour markets. *Oxford Review of Economic Policy* **35**:4, 722-745. [[Crossref](#)]
333. Christa Brelsford, Gautam Thakur, Rudy Arthur, Hywel Williams. Using Digital Trace Data to Identify Regions and Cities 5-8. [[Crossref](#)]
334. Michael Bailey, Eduardo Dávila, Theresa Kuchler, Johannes Stroebel. 2019. House Price Beliefs And Mortgage Leverage Choice. *The Review of Economic Studies* **86**:6, 2403-2452. [[Crossref](#)]



335. Chuchu Liu, Xin Lu. 2019. Network Evolution of a Large Online MSM Dating Community: 2005–2018. *International Journal of Environmental Research and Public Health* **16**:22, 4322. [[Crossref](#)]
336. Xavier Giroud, Holger M. Mueller. 2019. Firms' Internal Networks and Local Economic Shocks. *American Economic Review* **109**:10, 3617–3649. [[Abstract](#)] [[View PDF article](#)] [[PDF with links](#)]
337. Alexandra M. Espinosa, Luís Horna, Rodrigo Mendieta Muñoz, Nicola Pontarollo. 2019. The statistical properties of the networks of emigrants: the Ecuadorian case. *International Migration* **57**:4, 40–57. [[Crossref](#)]
338. 2019. Report by the AEA Data Editor. *AEA Papers and Proceedings* **109**, 718–729. [[Citation](#)] [[View PDF article](#)] [[PDF with links](#)]
339. Johannes Wachs, Taha Yasserli, Balázs Lengyel, János Kertész. 2019. Social capital predicts corruption risk in towns. *Royal Society Open Science* **6**:4, 182103. [[Crossref](#)]
340. Swati Bhatt. Diminished Risk-Taking 117–151. [[Crossref](#)]
341. Elizabeth Bruch, M. Newman. 2019. Structure of Online Dating Markets in U.S. Cities. *Sociological Science* **6**, 219–234. [[Crossref](#)]
342. Hongbum Lee. 2019. Sentiment Spillover: Evidence from Two Environmental Disasters. *SSRN Electronic Journal* **78**. . [[Crossref](#)]
343. Turan G. Bali, David A. Hirshleifer, Lin Peng, Yi Tang. 2019. Attention, Social Interaction, and Investor Attraction to Lottery Stocks. *SSRN Electronic Journal* **5**. . [[Crossref](#)]
344. Olga Balakina, Anastasia Parakhonyak. 2019. The Value of Connections: Network Effects on Stock Market Participation. *SSRN Electronic Journal* **24**. . [[Crossref](#)]
345. Adrienne Wood, Adam M. Kleinbaum, Thalia Wheatley. 2019. Exposure to Cultural Diversity Predicts Connectedness in a Social Network. *SSRN Electronic Journal* **8**. . [[Crossref](#)]
346. Johannes Rode, Sven Müller. 2019. I Spot, I Adopt! A Discrete Choice Analysis on Peer Effects in Solar Photovoltaic System Adoption of Households. *SSRN Electronic Journal* **65**. . [[Crossref](#)]
347. Xueying Bian, Sergei Sarkissian, Jun Tu, Ran Zhang. 2019. Return Cross-Predictability in Firms with Similar Employee Satisfaction. *SSRN Electronic Journal* **97**. . [[Crossref](#)]
348. Mark J. Garmaise, Yaron Levi, Hanno N. Lustig. 2019. Spending Less After (Seemingly) Bad News. *SSRN Electronic Journal* **93**. . [[Crossref](#)]
349. Michael Bailey, Ruiqing Cao, Theresa Kuchler, Johannes Stroebel. 2018. The Economic Effects of Social Networks: Evidence from the Housing Market. *Journal of Political Economy* **126**:6, 2224–2276. [[Crossref](#)]
350. Shin Alexandre Koseki. 2018. The geographic evolution of political cleavages in Switzerland: A network approach to assessing levels and dynamics of polarization between local populations. *PLOS ONE* **13**:11, e0208227. [[Crossref](#)]
351. Fulya Ersoy. 2018. Reshaping Aspirations: The Effects of the Great Recession on College Major Choice. *SSRN Electronic Journal* **105**. . [[Crossref](#)]
352. Vesa Pursiainen. 2018. Cultural Biases in Equity Analysis. *SSRN Electronic Journal* **32**. . [[Crossref](#)]
353. Claes BBckman, Tobin Hanspal. 2018. The Geography of Alternative Work. *SSRN Electronic Journal* **11**. . [[Crossref](#)]
354. Vimal Balasubramaniam. 2018. The Effect of Lifespan Expectations on Financial Decisions: Evidence from Mass Shootings and Natural Disaster Experiences. *SSRN Electronic Journal* **85**. . [[Crossref](#)]
355. Roberto Mosquera, Mofioluwademi Odunowo, Trent McNamara, Xiongfei Guo, Ragan Petrie. 2018. The Economic Effects of Facebook. *SSRN Electronic Journal* **97**. . [[Crossref](#)]
356. Linda Allen, Lin Peng, Yu Shan. 2018. Social Networks and Supply and Demand on Online Lending Marketplaces. *SSRN Electronic Journal* **117**. . [[Crossref](#)]

357. Karsten MMller, Carlo Schwarz. 2017. Fanning the Flames of Hate: Social Media and Hate Crime. *SSRN Electronic Journal* 130. . [[Crossref](#)]
358. Christos Andreas Makridis. 2017. Sentimental Business Cycles and the Protracted Great Recession. *SSRN Electronic Journal* 107. . [[Crossref](#)]
359. Ni Huang, Gordon Burtch, Yili Hong, Paul A. Pavlou. 2017. Unemployment and Worker Participation in the Gig Economy: Evidence from an Online Labor Platform. *SSRN Electronic Journal* 33. . [[Crossref](#)]
360. Bing Han. 2016. Visibility Bias in the Transmission of Consumption Norms and Undersaving. *SSRN Electronic Journal* 80. . [[Crossref](#)]
361. Huijun Wang, Jianfeng Yu, Shen Zhao. 2016. Time-Varying Demand for Lottery: Speculation Ahead of Earnings Announcements. *SSRN Electronic Journal* 15. . [[Crossref](#)]