## The Network of Firms Implied by the News

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### Firm networks matter, but data are scarce

- Acemoglu et al. (2012): "This paper argues that, in the presence of intersectoral input—output linkages, microeconomic idiosyncratic shocks may lead to aggregate fluctuations."
- Herskovic (2018): "Changes in the [production] network are sources of systematic risk reflected in equilibrium asset prices."

However, existing databases are incomplete, sparse, or lagged

- Compustat segments, BEA input-output, variance decomposition, interbank links, alternative data (10-K similarity, internet co-searches)
- → Lack of network data hinders accurate measurement of risks

## Our solution: mine news reporting

News often reports about relationships between firms

## Our methodology

- GH-UAW contract seen hard to match fully by rivals
   By David Bailey
   Tue Oct 2, 2007 3:47pm EDT
   http://www.reuters.com/article/2007/10/02/us-gm-uaw-idUSN0242907020071002

  DETROIT (Bouters) Sourcel aspects of the tentative contract between General Motors Corp. GM.N.) and the United Auto Workers union will be hard for Ford Motor Co. (F.N.) and Chrysler LLC to match in labor talks expected to heat up in coming days, people familiar with the negotiations said.

  The adoption of second-tier wages for new hires at GM represents an attractive concession for Ford and Chrysler but the structure of a retiree health-care trust could prove difficult to transfer, sources familiar with the match said on Tuesday. The establishment of a Voluntary Employees Beneficiary Association trust, or VEBA, was a centerpiece of the LMW; agreement with GM; allowing the automaker to take some SSB billion of liabilities off its books. Privately held Chrysler has been focused on cash flow since Cerberus acquired the automaker over the summer, to the point that it has been taking daily cash flow reports. The GM-UAW health-care trust would on crowide savings until 2016, when the new trust is expected to take over some S3 billion in annuar betiree health care payments from the top U.S. automaker. Ford and Chrysler would be hard-pressed to match the bump-up in pension payments to their retirees that GM pas agreed to give to its UAW retirees up the tentative contract. The people amiliar with the talks said. UAW President Ron Gettelfinger said, privaly he expected to assess the state of talks who but fordy and privately held Chrysler a law to the people amiliar with the talks said. UAW President Ron Gettelfinger said, privaly held Chrysler in keeping with a long-held tradition that has kept at three Detroit-based automakers on a similar labor-cost footing IJ union's deal with GM service as a basic pattern for talks who held forty-sler have been poring over the details in the UAW ard traditions with Ford or Chrysler and the provice have been po
  - Identify mentions of firms in text data and establish link when two firms are mentioned in same sentence

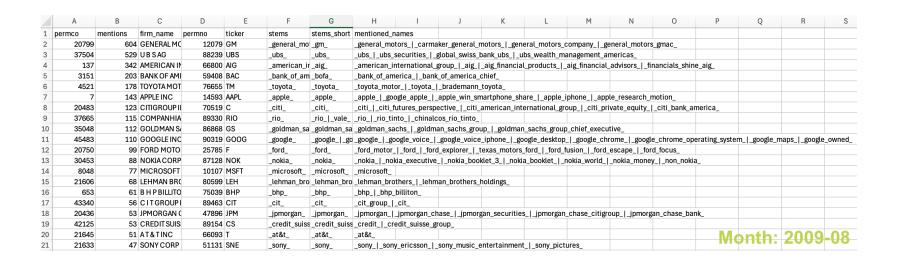
## Firm & link identification approach

- 1. Named Entity Recognition (spaCy in R) to identify mentions of *entities*: date, duration, location, money, number, ordinal, organization, percent, person, time, and other.
- 2. Keep organizations, filter out non-firms, and cluster same firms with unique name stem
- 3. Match with known firms from CRSP database
- 4. Establish link when firms are mentioned in same sentence
  - Can be relaxed to article co-mentions, with similar results

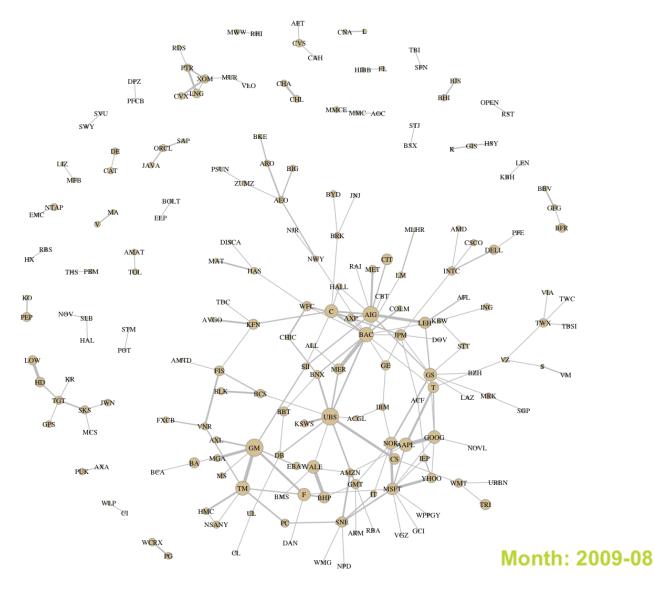
All codes available at www.news-networks.net

## **Data and output**

- We apply methodology to two news sample:
  - 106,521 articles from Reuters (2006-10 through 2013-11)
  - 430,770 articles from NYT (1981-01 through 2023-12)
- We identify 40,000 links among 13,500 firms in NYT sample
  - All data available at www.news-networks.net



## Monthly networks going back to 1981



## **Accuracy of methodology**

RIC mentions:	77,326
Unique firms associated with RICs:	1,861
Sentences with RICs:	60,615
RICs correctly matched by our algorithm (true positives or TP):	36,965
RICs incorrectly matched by our algorithm (false positives or FP):	1,347
RICs not matched by our algorithm (false negatives or FN):	38,571
Precision (TP $/$ (TP $+$ FP)):	96.48%
Recall (TP $/$ (TP $+$ FN)):	48.94%
F1 Score (2 $\times$ Precision $\times$ Recall / (Precision + Recall)):	64.94%
Total firm mentions identified by our algorithm:	514,015
Unique firms identified by our algorithm:	2,930
Sentences in which our algorithms identifies a firm:	410,120

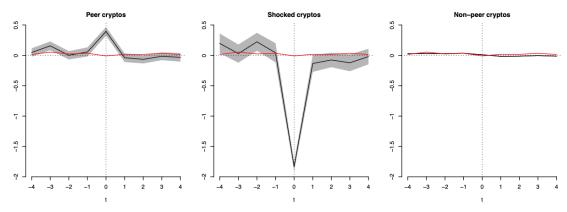
## **Application 1: predict aggregate outcomes**

		Forecast horizon T											
		1	2	3	4	5	6	7	8	9	10	11	12
Log consumption <sup>†</sup>													
	Density	-0.061		-0.135	* -0.377		* -0.716	* -0.685	* -0.817	* -0.893	** -0.811	** -0.851	** -0.967
		(-1.510)	(-1.725)	(-0.963)	(-2.050)	(-1.879)	(-1.995)	(-2.265)	(-2.397)	(-2.449)	(-2.781)	(-2.777)	(-2.678)
	Centrality	0.042	0.084	-0.025	0.140	0.283	0.307	0.245	0.288	0.207	0.033	0.131	0.210
	Interconnectivity	(0.963) -0.099	(1.087) -0.203	(-0.376) -0.157	(0.830) -0.357	(0.923) $-0.400$	(0.894) -0.277	$(0.745) \\ -0.132$	$(0.762) \\ -0.076$	(0.592) -0.033	$(0.144) \\ 0.117$	(0.503) $0.072$	$(0.678) \\ -0.021$
	Interconnectivity	(-1.362)	(-1.401)	(-0.960)	(-1.174)	(-1.101)	(-1.010)	(-0.725)	(-0.413)	(-0.207)	(0.603)	(0.366)	(-0.106)
	Adj. R <sup>2</sup>	0.009	0.065	0.091	0.122	0.158	0.153	0.117	0.133	0.158	0.133	0.122	0.159
	Auj. It	0.003	0.000	0.031	0.122	0.100	0.100	0.117	0.133	0.100	0.133	0.122	0.103
	Density	-0.113	-0.119	-0.150	-0.140	-0.191	-0.146	-0.151	-0.145	-0.166	-0.110	-0.143	-0.017
t <sub>s</sub>		(-1.561)	(-0.867)	(-0.674)	(-0.542)	(-0.588)	(-0.425)	(-0.416)	(-0.343)	(-0.383)	(-0.249)	(-0.277)	(-0.036)
Log defaults <sup>†</sup>	Centrality	* 0.170	0.201	0.284	0.341	0.432	0.534	0.626	0.761	0.984	° 1.076	° 1.219	° 1.231
		(2.039)	(1.376)	(1.278)	(1.269)	(1.363)	(1.584)	(1.581)	(1.462)	(1.523)	(1.675)	(1.825)	(1.760)
	Interconnectivity	-0.141	-0.148	-0.247	-0.341	-0.417	-0.521	-0.607	-0.728	-0.972	-1.096	^ -1.257	^ -1.306
	Adj. R <sup>2</sup>	(-1.864) $0.583$	(-1.035)	(-0.890) $0.700$	(-1.043) $0.726$	(-0.936) $0.736$	(-1.104)	(-1.189) $0.751$	(-1.223) $0.756$	(-1.307) $0.761$	(-1.530) $0.760$	(-1.664) $0.763$	(-1.909) $0.769$
	Adj. K	0.583	0.660	0.700	0.726	0.736	0.747	0.751	0.756	0.761	0.760	0.763	0.769
$\vdash$													
prod.†	Density	-0.026	-0.134	-0.164	-0.316	-0.559	-0.782	-0.789				* -1.085	* -1.209
		(-0.486)	(-1.366)	(-0.982)	(-1.349)	(-1.415)	(-1.521)	(-1.577)	(-1.701)	(-1.701)	(-1.862)	(-1.974)	(-2.031)
Į Į	Centrality	-0.003	0.072	0.013	0.001	0.136	0.182	0.052	0.004	-0.121	-0.387	-0.377	-0.254
ا ج		(-0.046)	(0.573)	(0.076)	(0.002)	(0.327)	(0.341)	(0.085)	(0.005)	(-0.152)	(-0.522)	(-0.471)	(-0.300)
ind.	Interconnectivity	-0.005	-0.128	-0.108	-0.250	-0.363	-0.399	-0.243	-0.143	-0.062	0.097	0.101	0.026
Log	A.1: D2	(-0.081)	(-0.767)	(-0.556)	(-0.803)	(-0.820)	(-0.952)	(-0.750)	(-0.484)	(-0.198)	(0.326)	(0.305)	(0.071)
ļ	Adj. R <sup>2</sup>	0.091	0.073	0.098	0.130	0.145	0.165	0.160	0.168	0.187	0.196	0.186	0.194

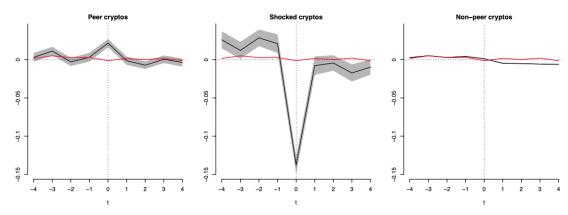
Table 11: Predictive regression estimates (Part 1). This table presents the estimates of the coefficient  $\beta_T$  of Regression (1) for the monthly density, centrality, and interconnectivity of the news-implied network extracted from the New York Times sample. The dependent variables are cumulative differences over T months of log consumption, log defaults, and log industrial production; see Section 5.1 for a description of these variables. We standardize all regressors and apply a Newey-West standard error adjustment in which the lag is equal to the sample size to the power one-over-four. The values in parentheses give t-statistics. \*\*\*, \*\*, and of denote significance on the 99.9%, 99%, 95%, and 90% confidence levels, respectively. indicates that the coefficients have been scaled by a factor of 100.

Also other variables: credit spreads, S&P 500 returns, VIX

## **Application 2: crypto peer trading**



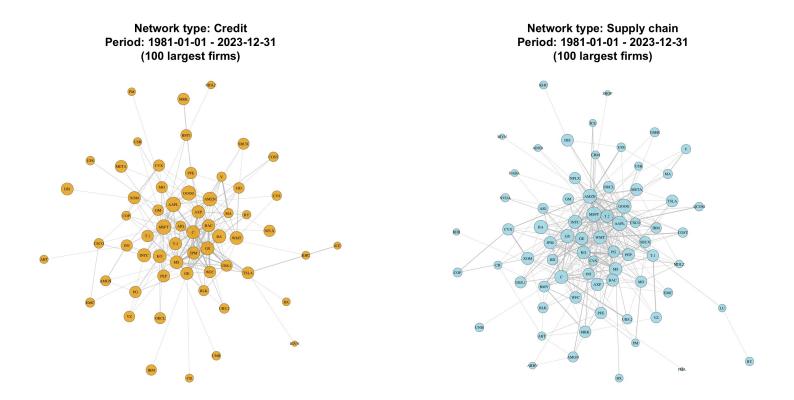
(a) Standardized abnormal return in week e + t, where e is the event week.



(b) Raw abnormal return in week e+t, where e is the event week.

6.5% annualized alpha, SR = 1.4

### **Application 3: network classification**



Also: peer & equity links (other links also possible)

### **Conclusion**

- We propose a methodology to extract firm links from text data
- We apply it to Reuters & NYT samples to identify 40,000 links among 13,500 firms
- Results enable several applications:
  - Predict aggregate outcomes
  - Predict returns among peers
  - Construct credit, supply chain, and other types of networks
- This is only the first step
  - Other asset classes, other text databases, ...
  - Data repository: www.news-networks.net

# Thank you!

### References

Acemoglu, Daron, Vasco M. Carvalho, Asuman Ozdaglar and Alireza Tahbaz-Salehi (2012), 'The network origins of aggregate fluctuations', *Econometrica* **80**(5), 1977–2016.

Herskovic, Bernard (2018), 'Networks in production: Asset pricing implications', *The Journal of Finance* **73**(4), 1785–1818.