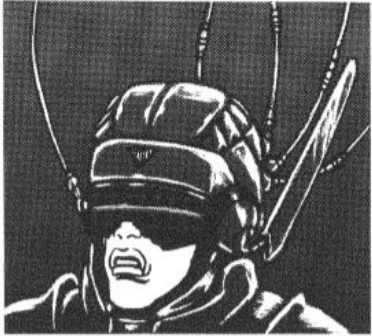


Portrait of Economic Academia: Evidence from Top 20 Economics Journals

Collaboration Network & Topic Modeling Analysis



Netrunners

Jerry Cheng
Hongzhang Xie
Yicheng Zhang

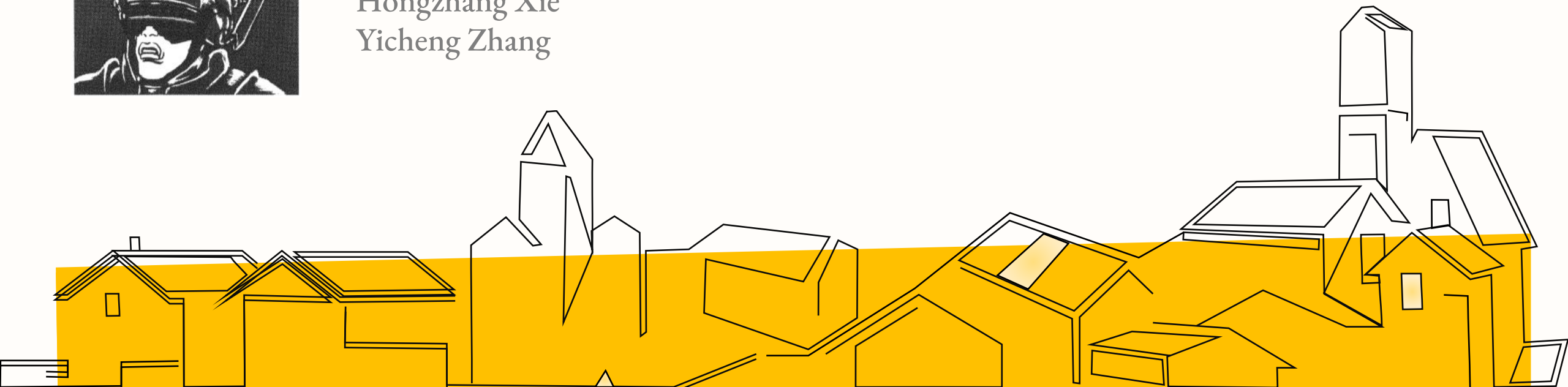
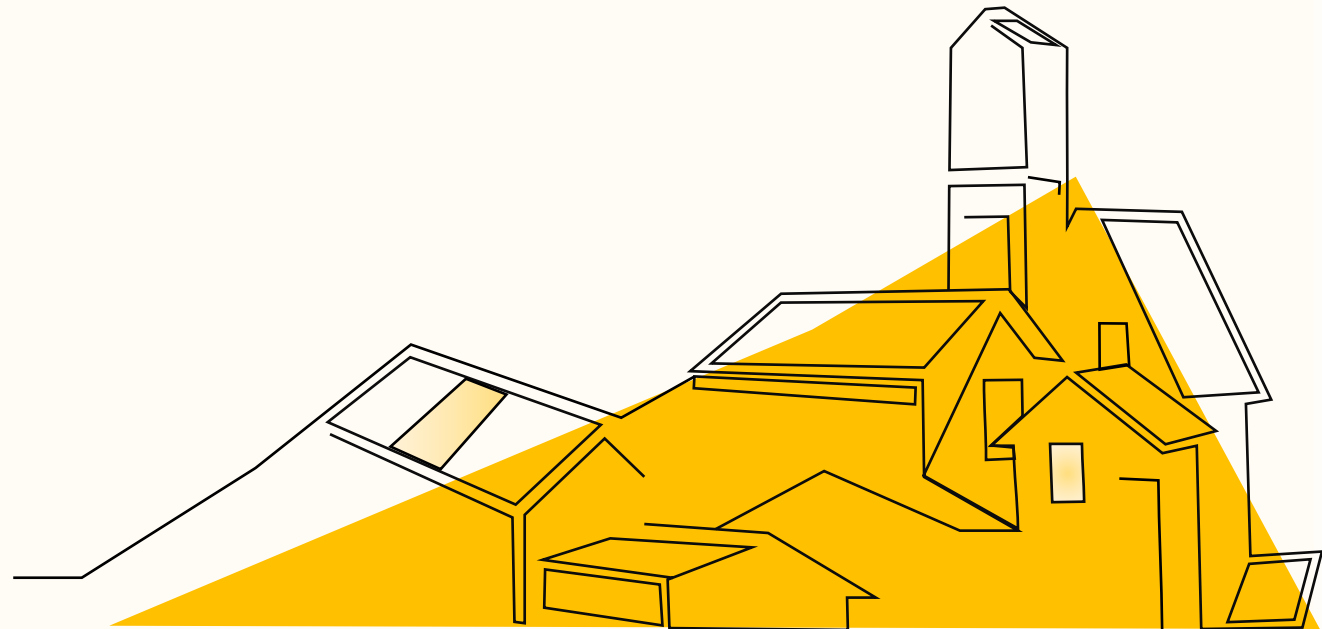


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2. Research Questions and Social Science Relevance
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4. Data Collection
5. Data Cleaning and Wrangling
6. Collaboration Network Analysis
7. Topic Modeling Analysis
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Research topic:

Examine the characters of the economic research collaboration structure

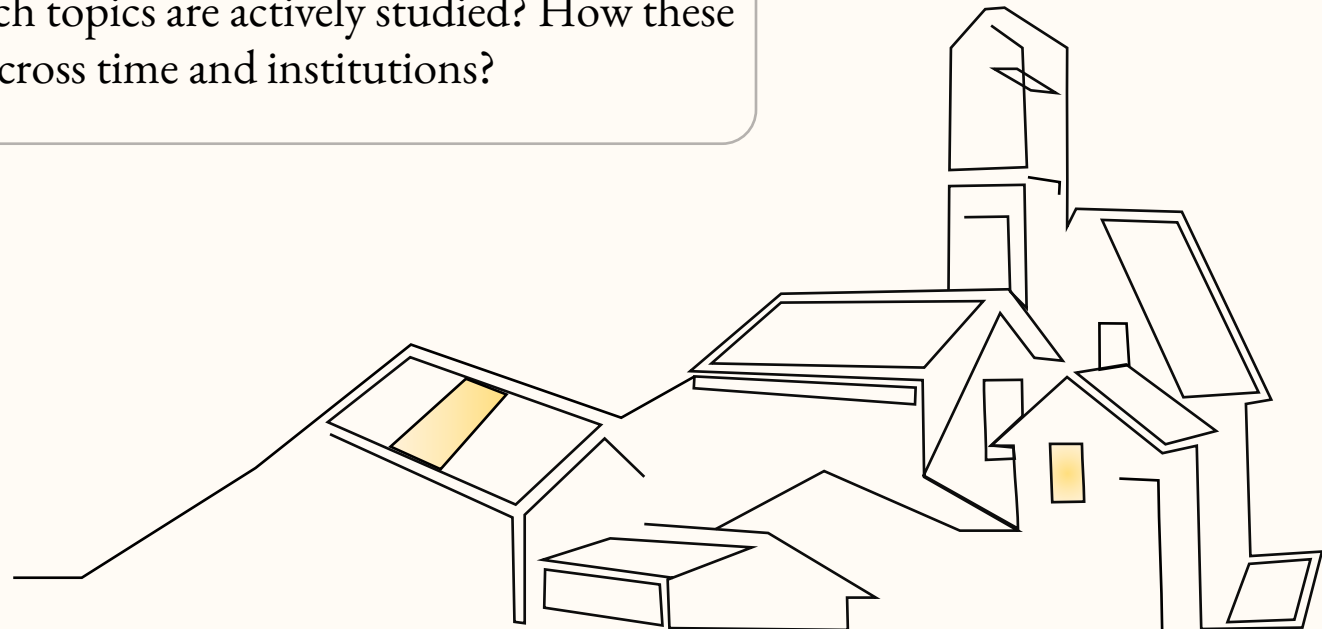
Question 1

What are the network features and collaboration patterns at the institutional level?

Question 2

What economics research topics are actively studied? How these topics vary and evolve across time and institutions?

Data Description

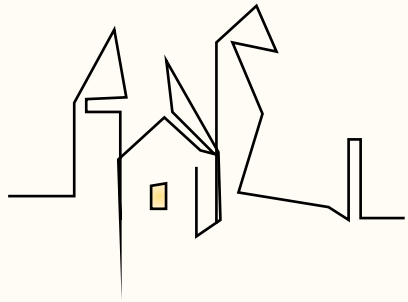


Data Description -Top 20 Economics Journals Data

- We collected the data of top 20 economics journals ranked by h5-index (2017 – 2021).
- Our raw data have 79,565 publications from 1886 till now.
- We select 18,086 publications from 2012 to 2022 as the research sample.
- We have three databases:
 - Top 20 Economics Journals Data
 - Google Scholar Author Data
 - Authors' Affiliations Data

Number of Collected Publications from 2012 to 2022 by Journal





Data Description

Top 20 Economics Journals Data

Data sources

Official Websites:

- American Economic Association
- Oxford Academic
- Science Direct

Third-party database :

- ideas.repec.org
- econpapers.repec.org

Title

Retail Pharmacies and Drug Diversion during the Opioid Epidemic

Authors

Aljoscha Janssen
Xuan Zhang

Journals

AMERICAN ECONOMIC REVIEW
VOL. 113, NO. 1, JANUARY 2023
(pp. 1-33)

VOL

[Download Full Text PDF](#)

Published time

Article Information

Abstract

Abstract

This study investigates the role of retail pharmacy ownership in the opioid epidemic. Using data of prescription opioid orders, we show that compared with chain pharmacies, independent pharmacies dispense 39.1 percent more opioids and 60.5 percent more OxyContin. After an independent pharmacy becomes a chain pharmacy, opioid dispensing decreases. Using the OxyContin reformulation, which reduced nonmedical demand but not the legitimate medical demand, we show that at least one-third of the difference in the amount of OxyContin dispensed can be attributed to nonmedical demand. We show that differences in competitive pressure and whether pharmacists own the pharmacy drive our estimates.

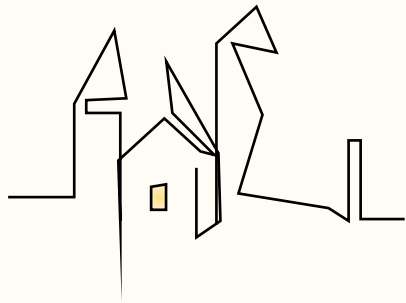
Citation

Citation

Janssen, Aljoscha, and Xuan Zhang. 2023. "Retail Pharmacies and Drug Diversion during the Opioid Epidemic." *American Economic Review*, 113 (1): 1-33.

DOI

DOI: 10.1257/aer.20210357



Data Description

Google Scholar Author Data

Links of two databases

- We input the **DOI** in **Top 20 Economics Journals Database** to search the **Authors** of the publications in Google Scholar.

Data sources

- Google Scholar Author Database
- Semantics Scholar Database (For incompleteness data)

DOI

Author id

Name

Affiliation*

Email

H-index

I10-index

Citations

Zhiguo He (何治国)

Fuji Bank and Heller Professor of Finance,
University of Chicago: NBER
Verified email at chicagobooth.edu - [Homepage](#)
[Financial Economics](#) [Chinese Economy and Fina...](#) [FinTech](#)

[FOLLOW](#) [GET MY OWN PROFILE](#)

Cited by	All	Since 2018
Citations	9695	6560
h-index	35	32
i10-index	45	42

TITLE	CITED BY	YEAR
Intermediary asset pricing Z He, A Krishnamurthy American Economic Review 103 (2), 732-770	1441	2013
Blockchain Disruption and Smart Contracts LW Cong, Z He Review of Financial Studies 32, 1754-1797	993	2017
Rollover risk and credit risk Z He, W Xiong The Journal of Finance 67 (2), 391-430	681	2012
Intermediary asset pricing: New evidence from many asset classes Z He, B Kelly, A Manela Journal of Financial Economics (126), 1-35	651	2017

Public access [VIEW ALL](#)

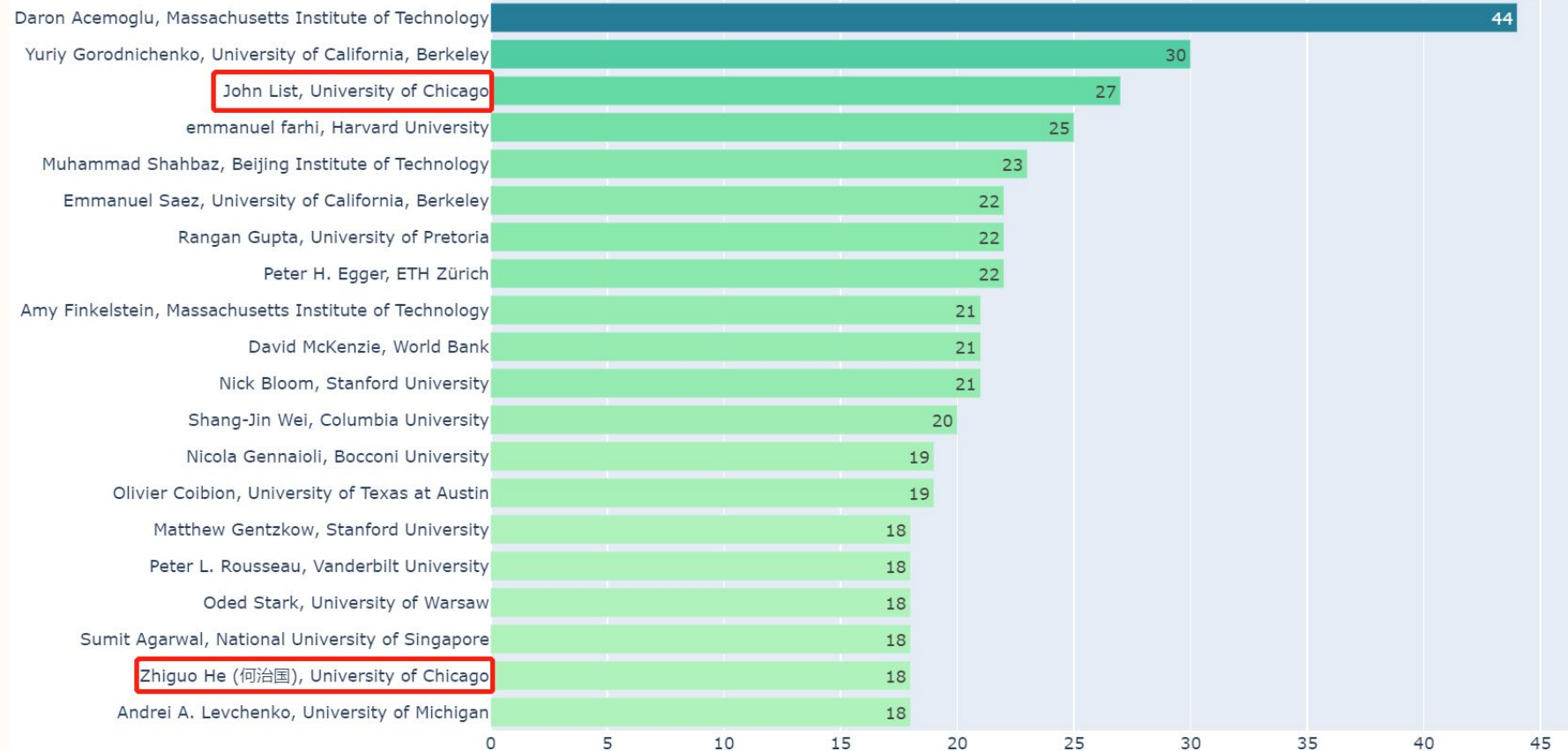
Bar chart showing annual citations from 2016 to 2023. The y-axis ranges from 0 to 1800. The x-axis shows years: 2016, 2017, 2018, 2019, 2020, 2021, 2022, 2023. The citation counts are approximately: 2016: 450, 2017: 500, 2018: 600, 2019: 700, 2020: 800, 2021: 900, 2022: 1000, 2023: 1100.

* If the author has **Multiple Affiliations**, we use the **Affiliation of the Email**.

Data Description - Google Scholar Author Data

- Our raw data have 19,585 records of authors
- We select 13,654 authors who have publications from 2012 to 2022 as the research sample.

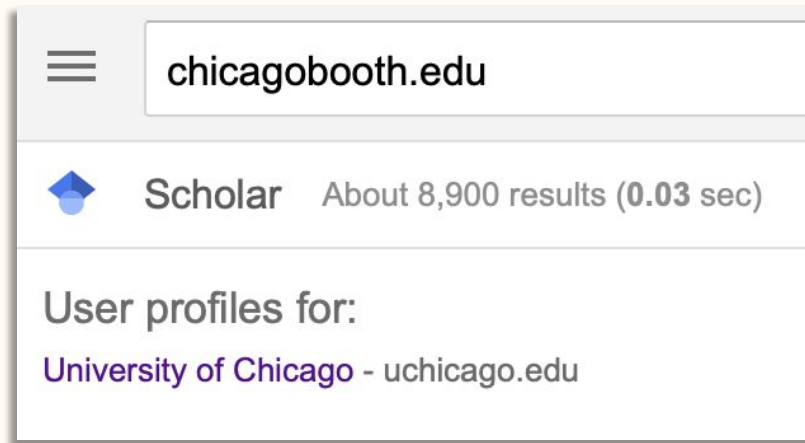
Number of Collected Publications from 2012 to 2022 by Authors



Data Sources and Description - Authors' Affiliations Data

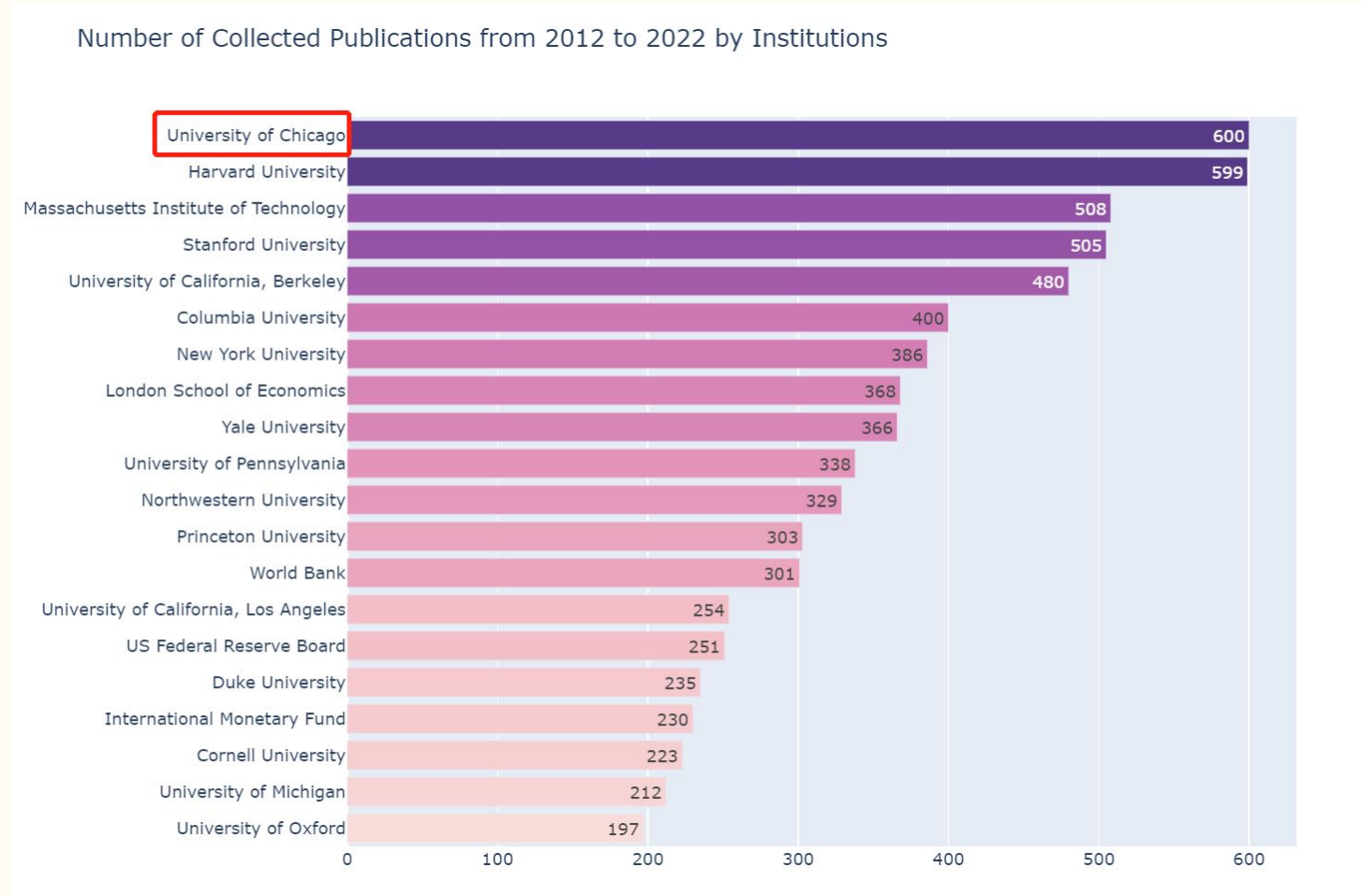
Data sources

- Google Scholar Institutions Information

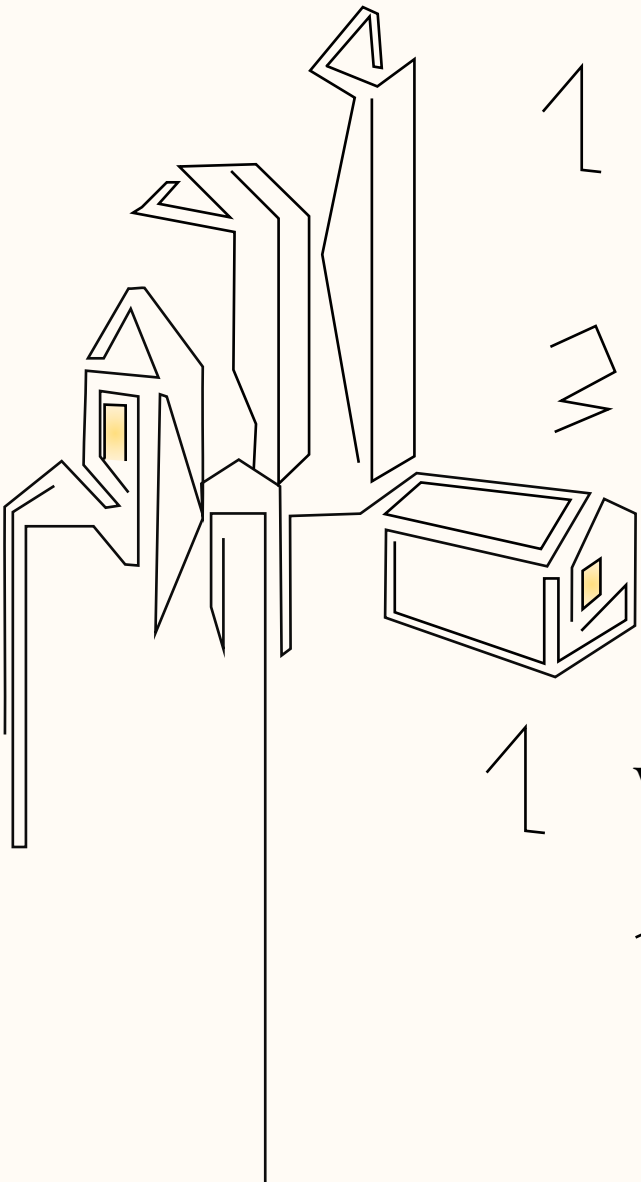


Data description:

- Our raw data have 1,392 records of institutions.
- We select 1,329 institutions that have authors who publish Top articles from 2012 to 2022 as the research sample.



Advantages



1

Large sample size with multiple attributes.

2

High data quality,
Crawl data from authoritative databases

3

Innovation, few studies have used similar data sources.

4

Importance, analyzing those data can help us better understanding the research collaboration mechanisms, and improve efficiency.

Limitations

1

We do not analyze the publication information outside the top 20 economic journals.

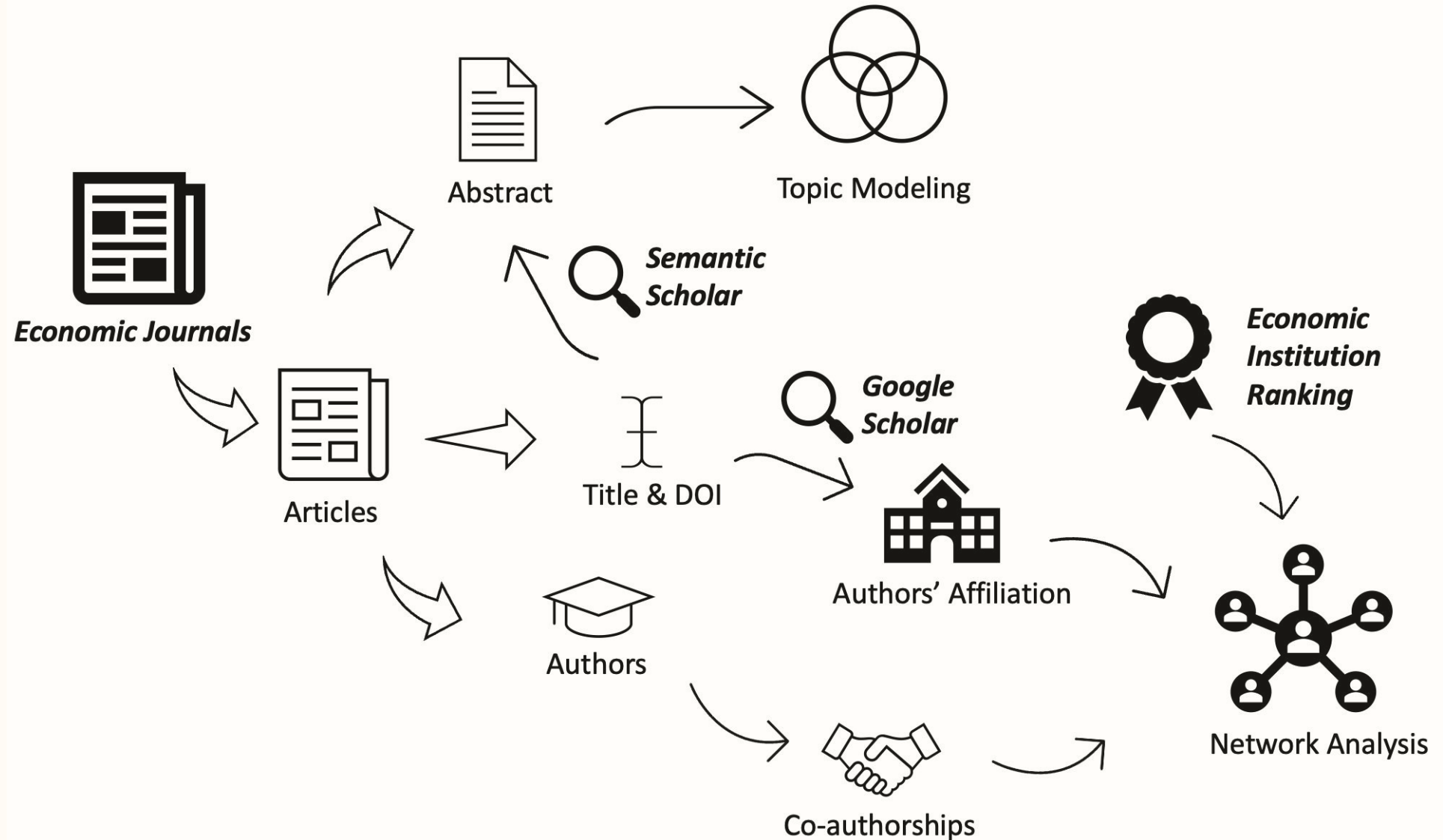
2

There are still some missing data in Publish time, DOI, and Authors in the publications before 2000.

3

We did not crawl the data for the authors who do not have Google Scholar profiles.


Data Collection Roadmap




Building Authors Database

(1) Retrieve Authors' Google Scholar Page

- Iterate through the article database. For every article, search the article's DOI (if available) in Google Scholar.
- Since the DOI is unique for every article, the researching result is either one or none.
- Scholars that has their own Google Scholar page will have the links embedded in their names.
- Collected *all the embedded links* to revisit later.


 DOI: 10.1257/000282802320188961

 Scholar

Bank bailouts and aggregate liquidity

DW Diamond, RG Rajan - American Economic Review, 2002 - pubs.aeaweb.org

Governments sometimes bail out banks by recapitalizing them, or by offering to insure their liabilities. The government's goal may be to rescue borrowers, bankers, or depositors, and economists have developed rationales why each of these constituencies may merit protection (eg, Diamond, 2001). These potential benefits have to be weighed against the costs of a bailout, which are typically thought to be the damage to long-run incentives that such intervention engenders. In this paper, we present a different effect of bank bailouts ...

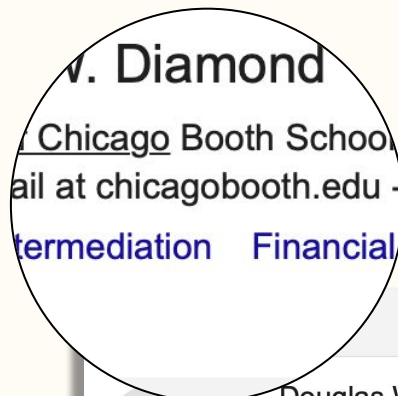
☆ Save  Cite Cited by 82 Related articles All 9 versions

Showing the best result for this search. [See all results](#)

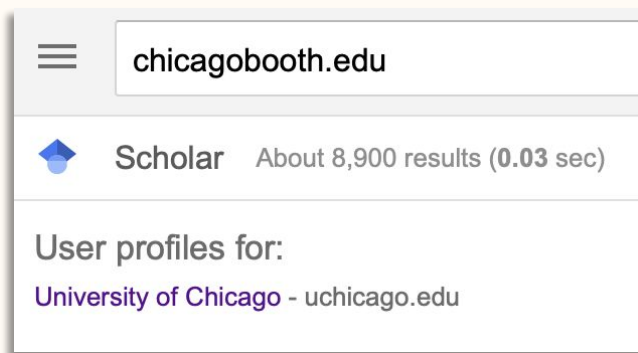
Building Authors Database

(2) Collect author's info from Google Scholar page

- Revisit the embedded URLs collected.
 - Collected information includes
 - the *email address* for this scholar.(Google Scholar affiliation section is messy)
 - Citations by year
 - h-index and i10-index listed
-
- Finally, search the email address collected directly in the Google Scholar.
 - The affiliation match with this address will appear in the top box.



A screenshot of the full Google Scholar profile page for Douglas W. Diamond. The page includes a profile picture, a 'FOLLOW' button, and a 'GET MY OWN PROFILE' button. The profile information shows his affiliation with the University of Chicago Booth School of Business and his verified email at chicagobooth.edu. Below this, there are links for 'Financial Intermediation' and 'Financial Crises'. The main section of the page lists his publications with columns for 'TITLE', 'CITED BY', and 'YEAR'. The first three publications are: 'Bank runs, deposit Insurance, and liquidity' (14470 citations, 1983), 'Financial intermediation and delegated monitoring' (12608 citations, 1984), and 'Disclosure, liquidity, and the cost of capital' (5095 citations, 1991). The fourth publication is 'Monitoring and reputation: The choice between bank loans and...' (4456 citations, 1991). On the right side of the page, there is a 'Cited by' section with a 'VIEW ALL' link. This section shows a table with columns for 'All' and 'Since 2018'. The table lists 'Citations' (60710), 'h-index' (35), and 'i10-index' (40). Below the table is a bar chart showing the number of citations per year from 2016 to 2023. The y-axis ranges from 0 to 3500. The x-axis shows the years 2016, 2017, 2018, 2019, 2020, 2021, 2022, and 2023. The bars show a general downward trend in citations over the years.



Building Institution Database

Top 10% Economic Institutions (Last 10 Years Publications), as of January 2023 (with details)

[Explanations](#)
[Ranking](#)
[More](#)

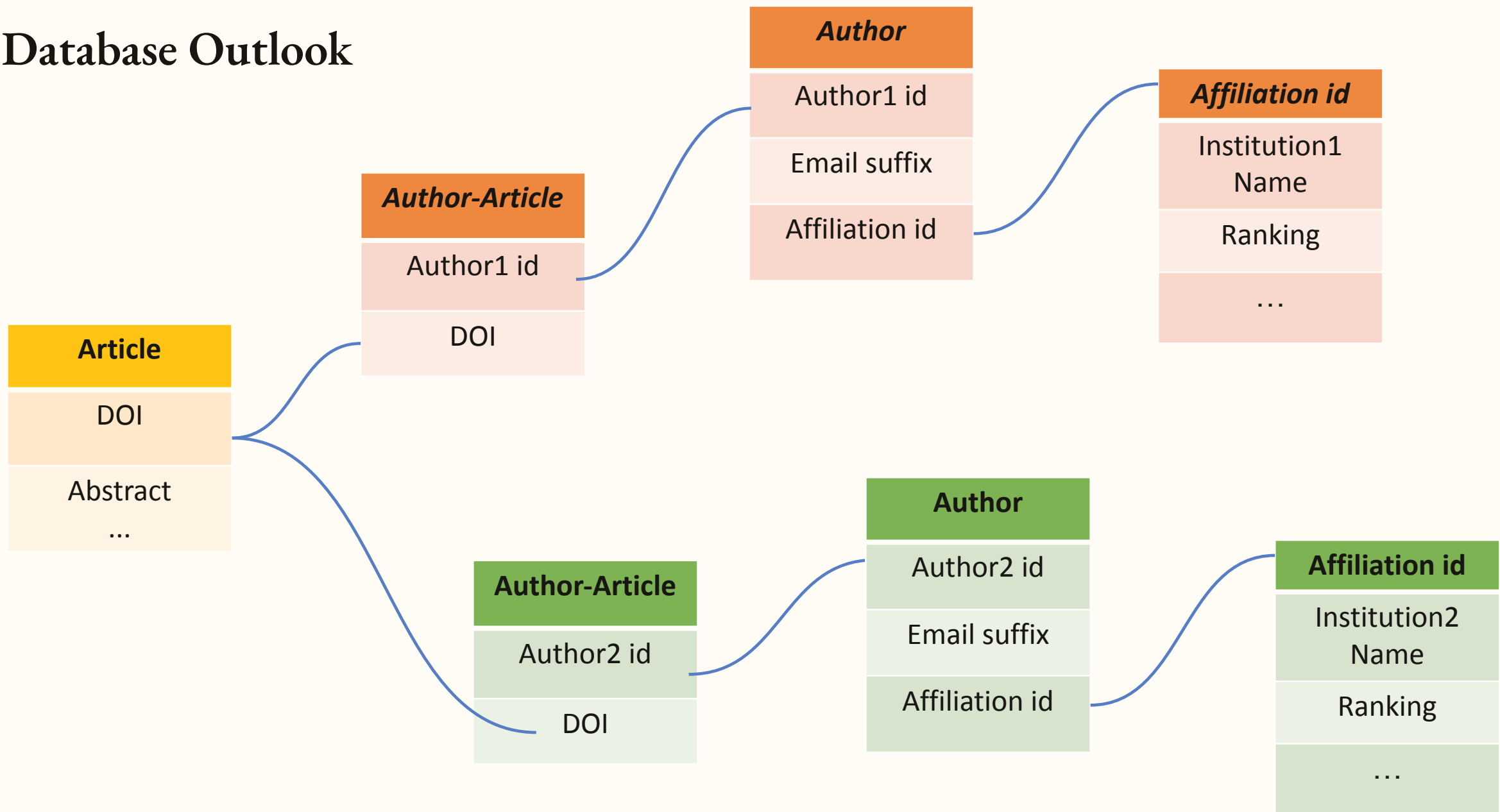
The rankings

Rank	Institution	Score	Nb Works	DNb Works	Sc Works	WSc Works	AI Works
1	London School of Economics (LSE)	2.04	3	5	1	1	
2	World Bank Group	3.25	2	1	2	4	
3	National Bureau of Economic Research (NBER)	3.77	20	26	4	2	
4	International Monetary Fund (IMF)	4.25	5	2	3	3	
5	Department of Economics, Harvard University	4.65	55	91	9	8	
6	European Central Bank	6.74	8	11	6	11	
7	Department of Economics, University of California-Berkeley	8.35	39	77	10	10	
8	Economics Department, Massachusetts Institute of Technology (MIT)	9.56	48	65	11	9	
9	Department of Economics, University of Chicago	9.81	37	61	8	6	
10	Federal Reserve Board (Board of Governors of the Federal Reserve System)	10.02	18	18	7	7	

affiliationid	name	email	match_...	univer_ins	Rank
au.aalto.fi	Aalto University	aalto.fi	0.7962962...	Carleton University	309
kriot.kth.se	KTH Royal Institu...	kth.se	0.7470084...	Beijing Institute of ...	78
uocla.ucla.edu	University of Calif...	ucla.edu	0.682900...	Boston College	100
aei.aei.org	American Enterpr...	aei.org	0.7599879...	Economic and Soci...	212
miot.mit.edu	Massachusetts In...	mit.edu	1	Massachusetts Inst...	8
ciot.caltech.edu	California Institut...	caltech.edu	0.8677729...	Beijing Institute of ...	78
hu.harvard.edu	Harvard University	harvard.edu	1	Harvard University	5
su.stanford.edu	Stanford University	stanford.edu	1	Stanford University	14
uop.upenn.edu	University of Pen...	upenn.edu	1	University of Penns...	55
ai.amazon.com	Amazon Inc.	amazon.com	0.6931818...	Chapman University	448
au.american.edu	American Univers...	american.e...	0.746666...	Academia Romana	129
ac.amherst.edu	Amherst College	amherst.edu	0.8002011...	Dartmouth College	83
cmu.cmu.edu	Carnegie Mellon ...	cmu.edu	1	Carnegie Mellon Un...	360
anu.anu.edu.au	Australian Nation...	anu.edu.au	1	Australian National ...	99
au.ariel.ac.il	Ariel University	ariel.ac.il	0.7460317...	Drexel University	196
uoa.arizona.edu	University of Ariz...	arizona.edu	0.7925925...	University of Alberta	416
asu.asu.edu	Arizona State Uni...	asu.edu	1	Arizona State Unive...	177
frboa.atl.frb.org	Federal Reserve ...	atl.frb.org	1	Federal Reserve Ba...	257
auob.aub.edu.lb	American Univers...	aub.edu.lb	0.666379...	Central Bank of Irel...	386

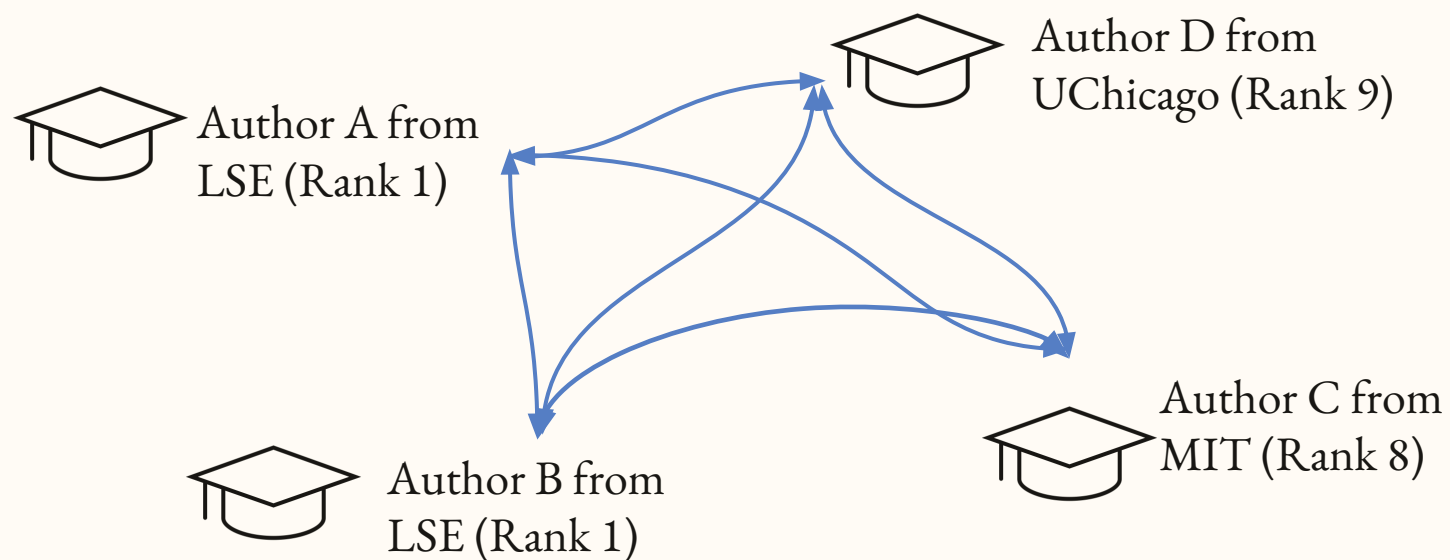
- IDEAS ranking
 - Academic Oriented
 - Comprehensive ranking that include non-university institutions
- Full record linkage
 - Common word is removed (“university of”)
 - Take the highest score
 - Exactly 200 institutions were matched

Database Outlook



Constructing Co-authorship links

Article – Author – Institution – Ranking

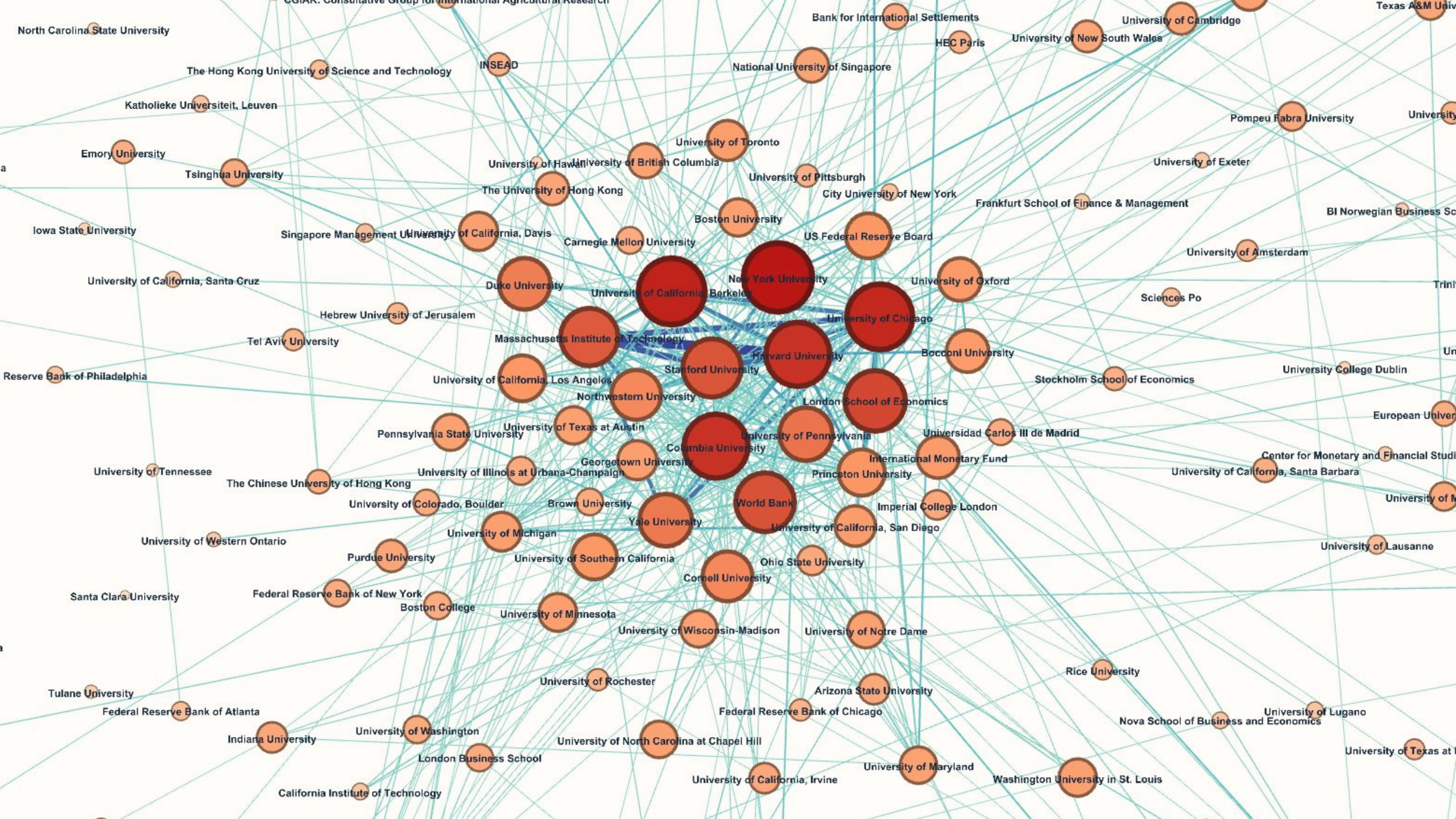


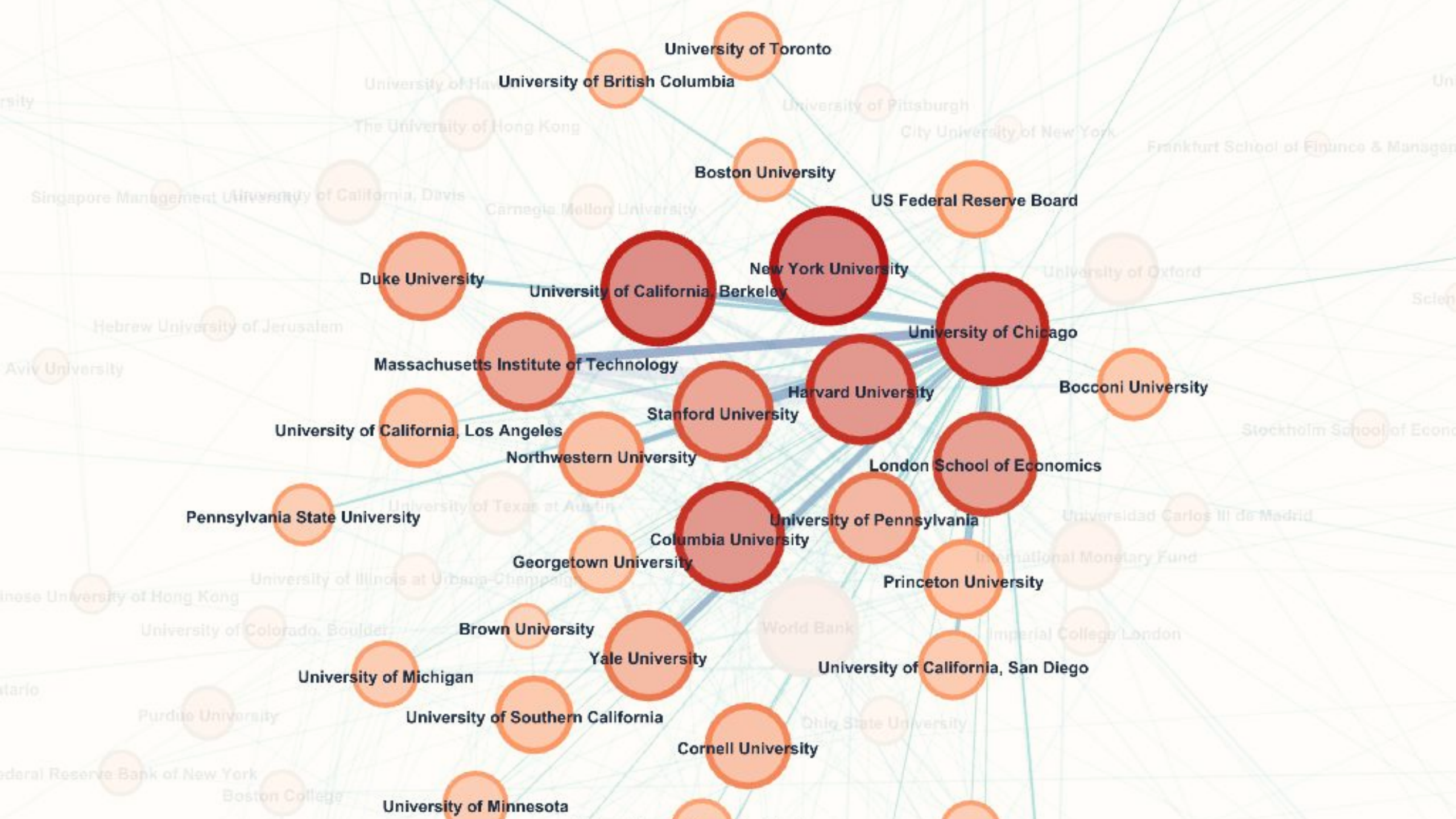
LSE (Whole Dataset)

Name	Number of links
LSE	10
UChicago	5
MIT	5

(example)

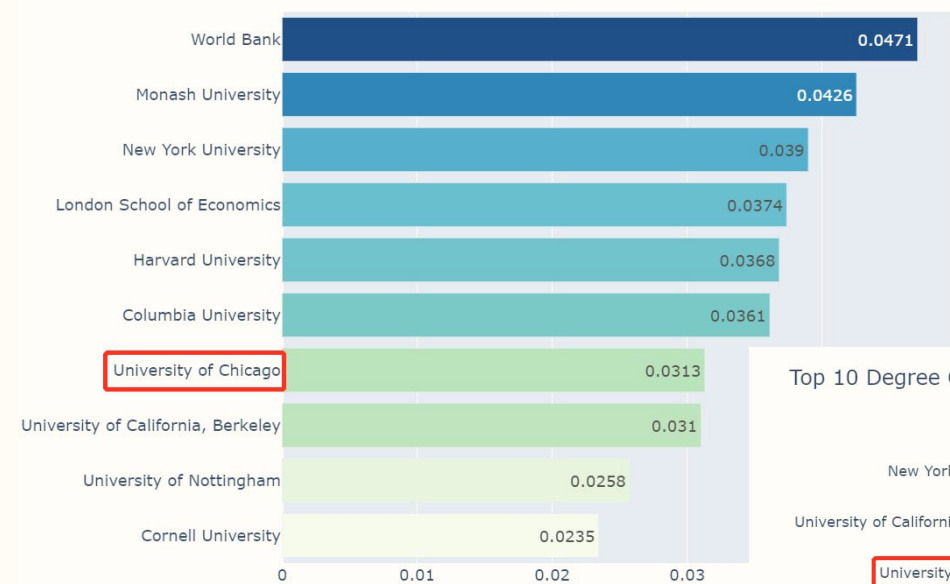
- Compile to Institution Level links



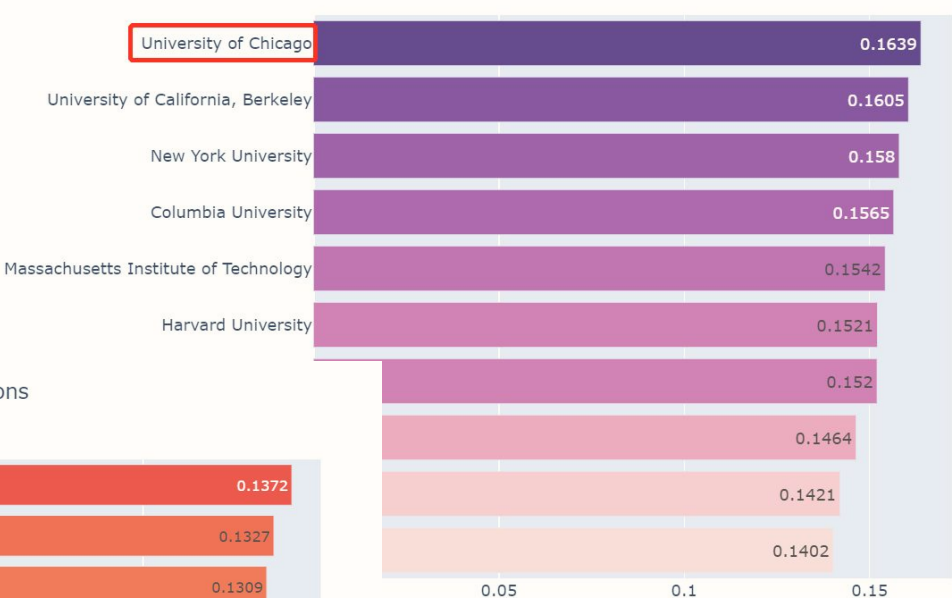


Centralities Analysis

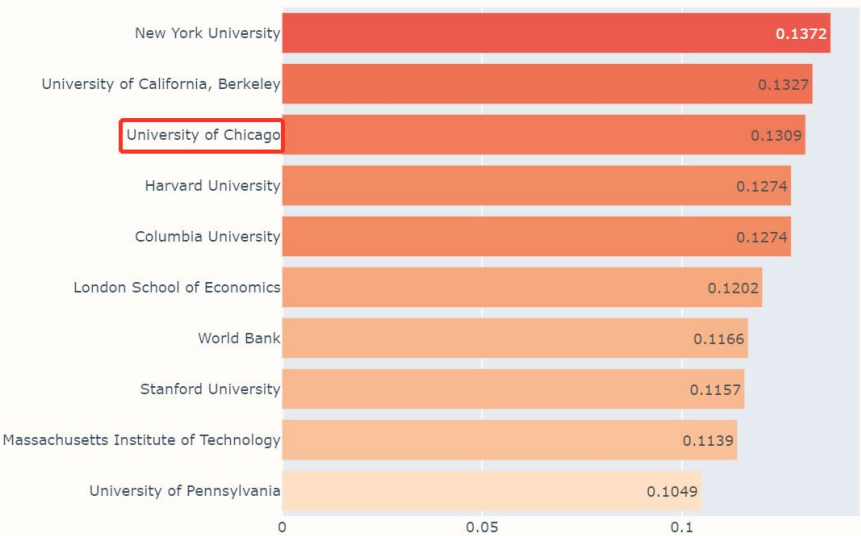
Top 10 Betweenness Centralities of Economic Institutions



Top 10 Eigenvector Centralities of Economic Institutions



Top 10 Degree Centralities of Economic Institutions



Quantifying Collaboration links

(1) Institution level analysis

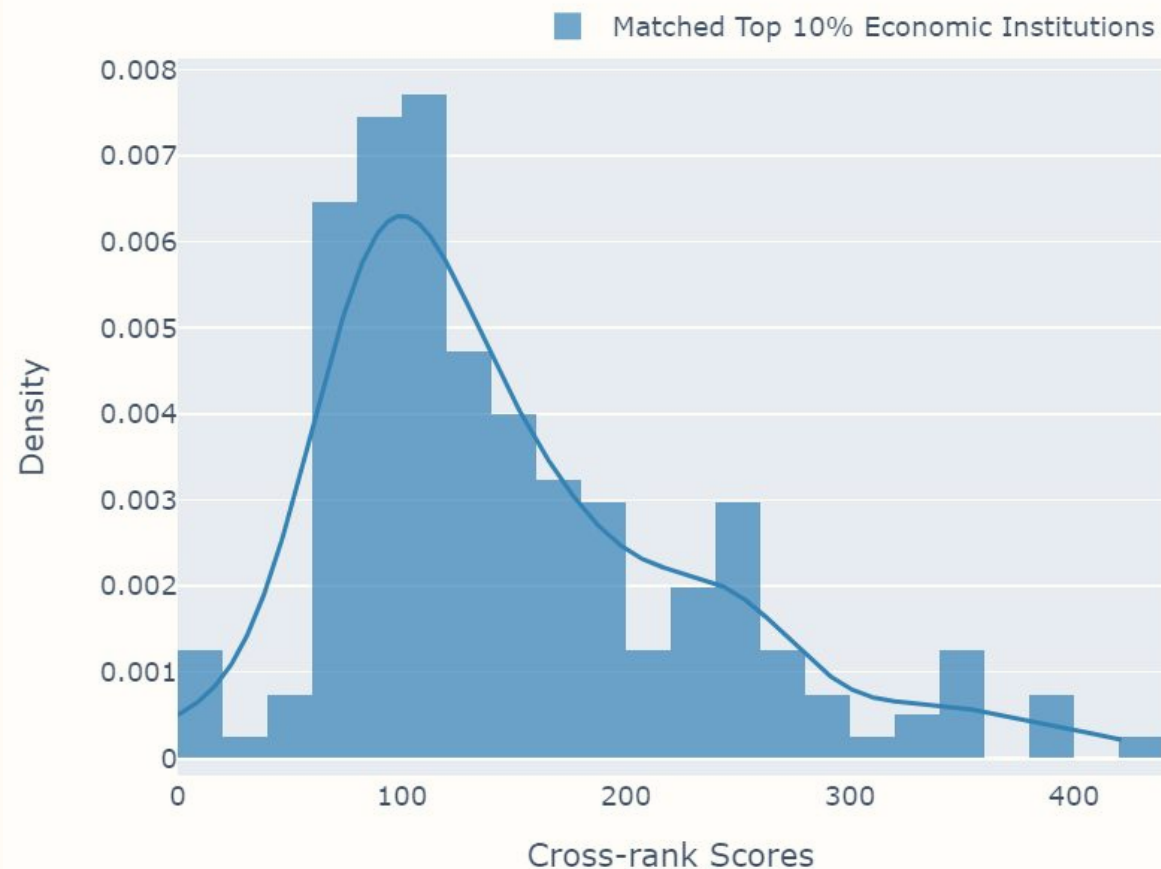
LSE (Whole Dataset)

Name	Number of links	Contribution To the score
LSE (Rank 1)	10	0
UChicago (Rank 9)	5	$(9-1) * 5$
MIT (Rank 8)	5	$(8-1) * 5$

(example)

Cross-rank score = $(0 + (9-1) * 5 + (8-1) * 5) / 20 = 3.75$

Cross-rank Scores Distributuion



Quantifying Collaboration links

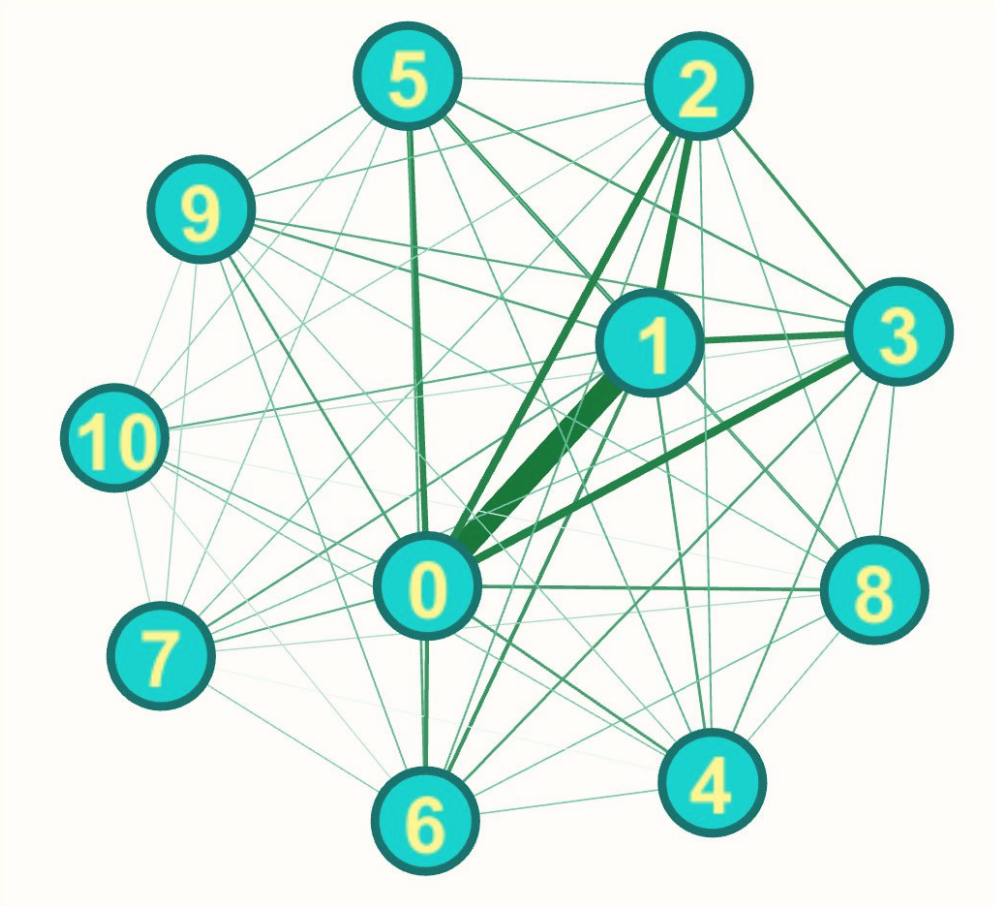
(1) Institution level analysis

- Institutions that rank higher tend to collaborate more with each other - a low cross-rank score.
- Cross-ranking collaboration is more common for institutions that rank later.

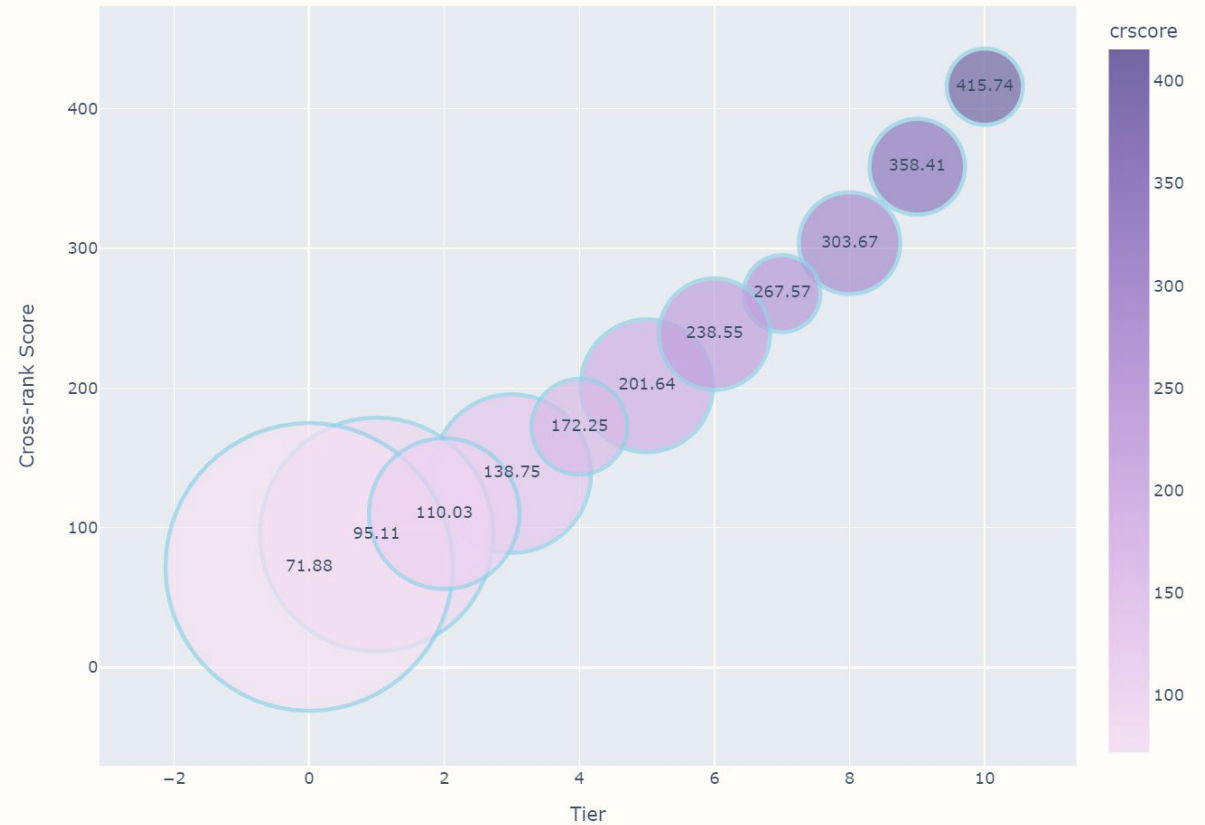


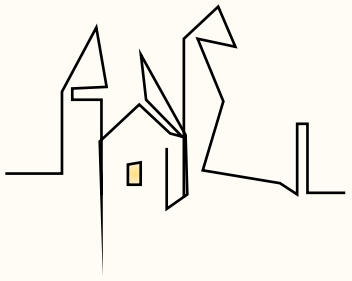
Quantifying Collaboration links

(2) Institution Tier level analysis



Cross-rank Score vs Tier by Economic Research





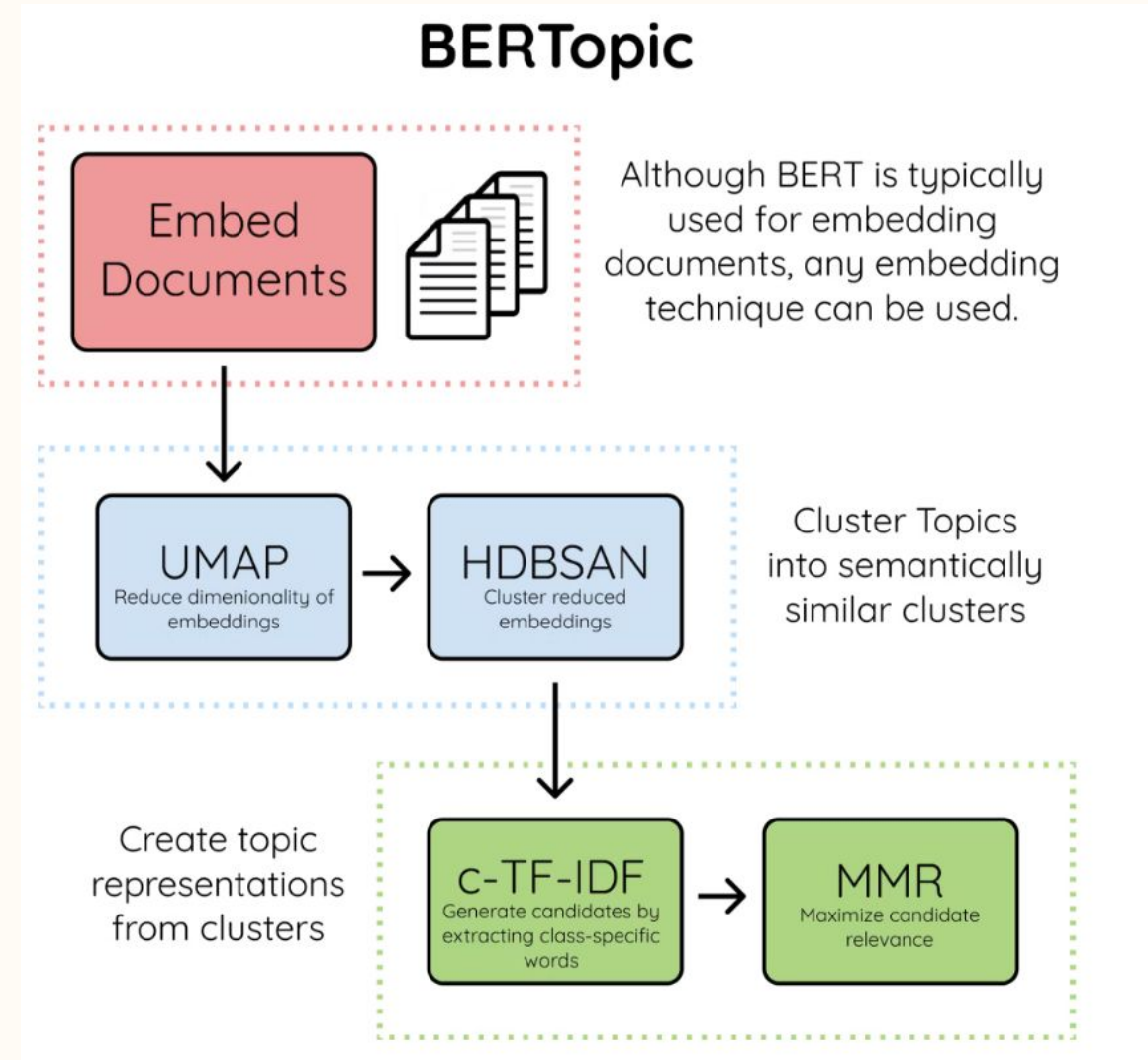
Topic Modeling Analysis

BERTopic Modeling

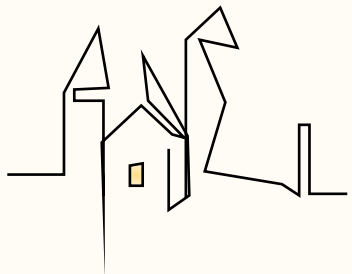
- An emerging TM technique that uses language embedding and class-based TF-IDF
- Original Paper (Grootendorst 2022):
<https://arxiv.org/abs/2203.05794>
- Documentation:
<https://maartengr.github.io/BERTopic/index.html>

Advantages over LDA TM

- Easy to fit in model: only need to clear stopwords and re-concatenate into paragraphs
- Much more stable: results may vary a little across runs but not as much as LDA
- Much less training time: 15,000+ abstracts take less than 1 minute to train



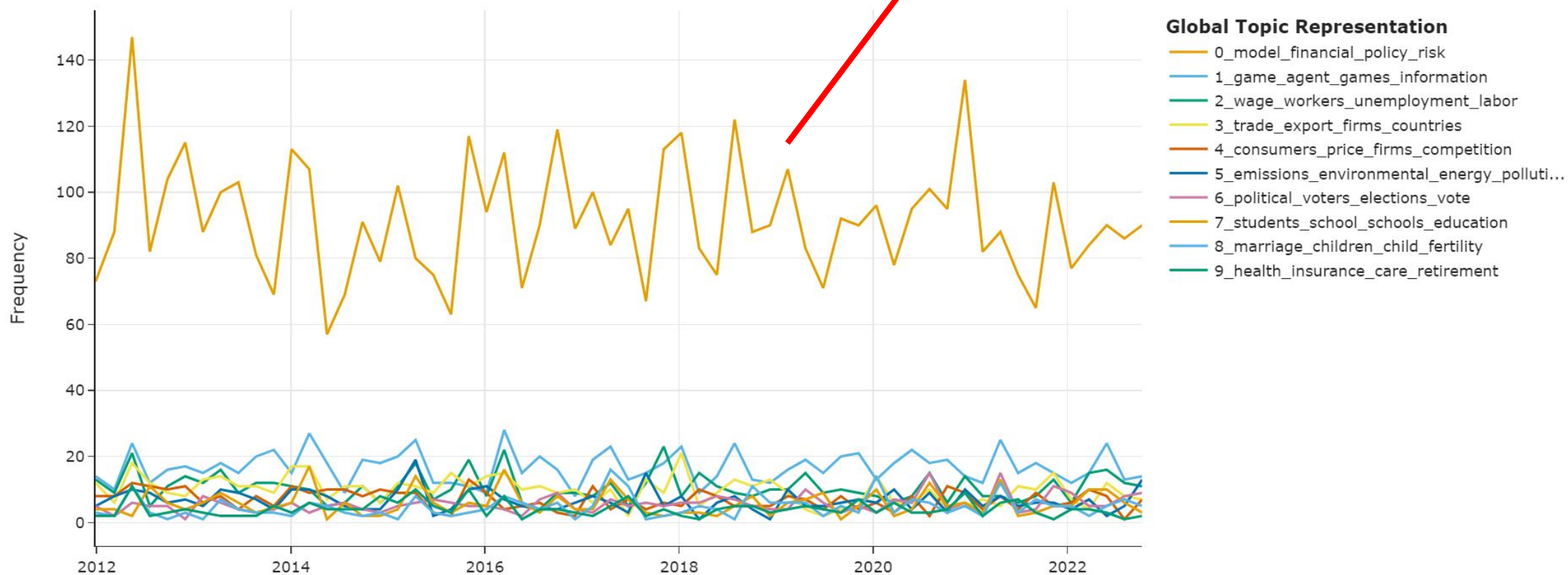
Source: <https://hackernoon.com/nlp-tutorial-topic-modeling-in-python-with-bertopic-372w3519>



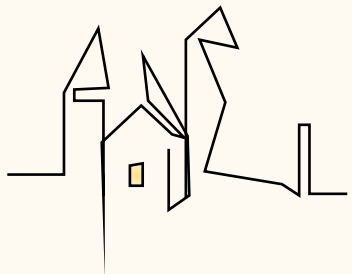
Topic Modeling Analysis

Dynamic Economic Research TM over Time

Topics over Time

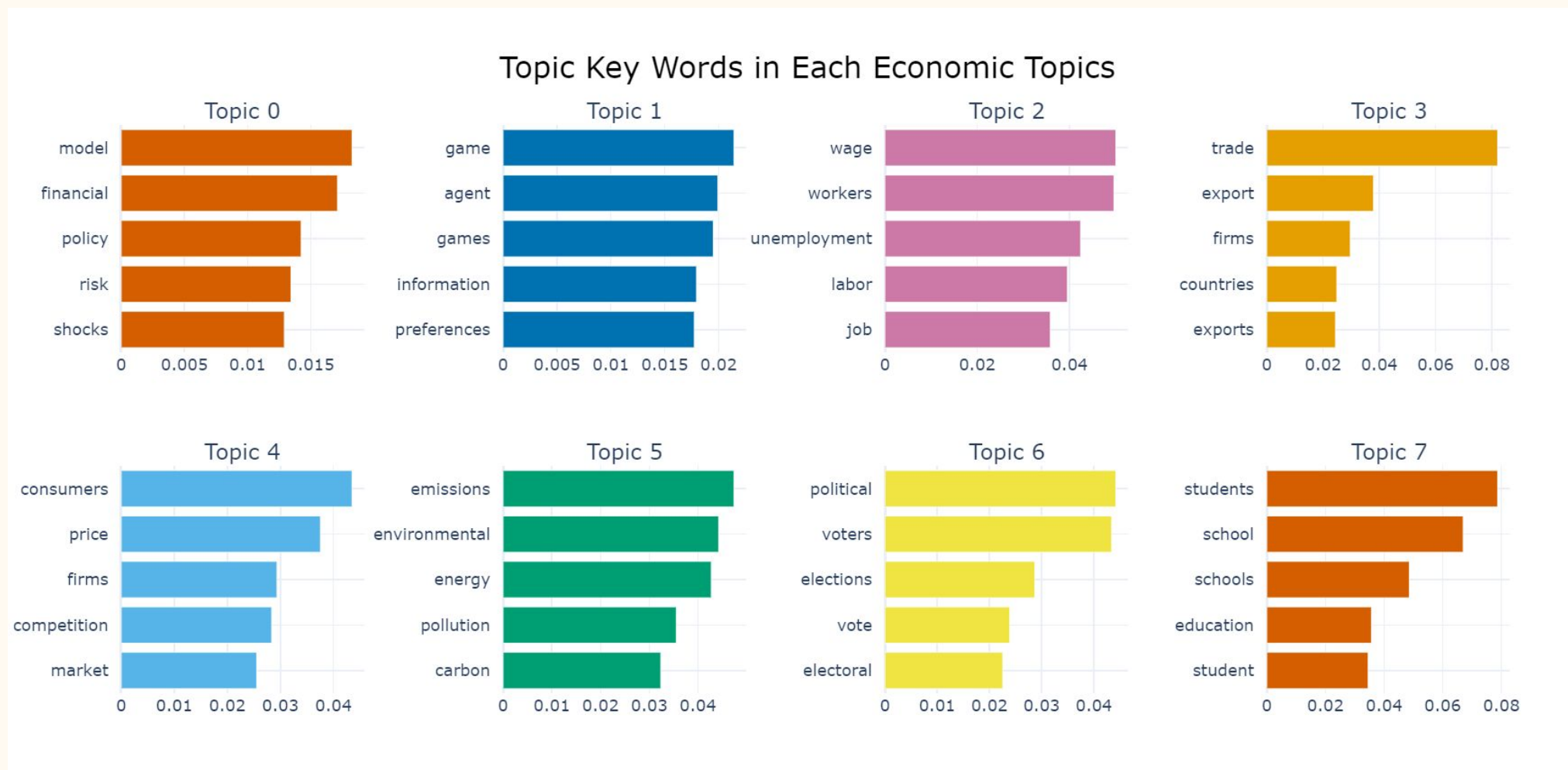


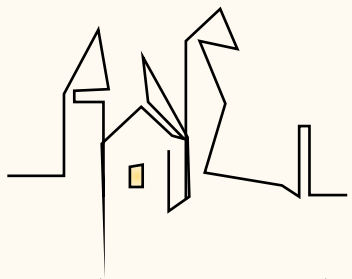
Dominating, but NON-trivial! Need further splitting among these documents for better results.



Topic Modeling Analysis

Top 5 Keywords for each Topic:



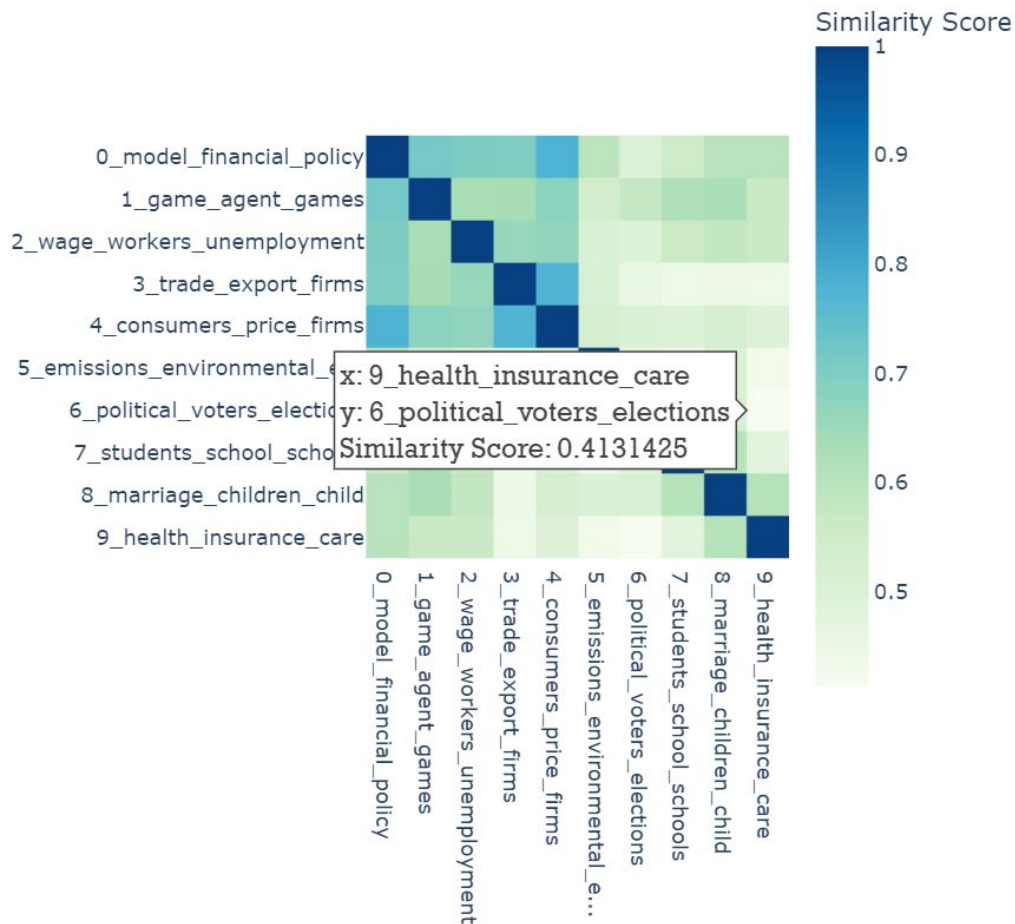


Topic Modeling Analysis

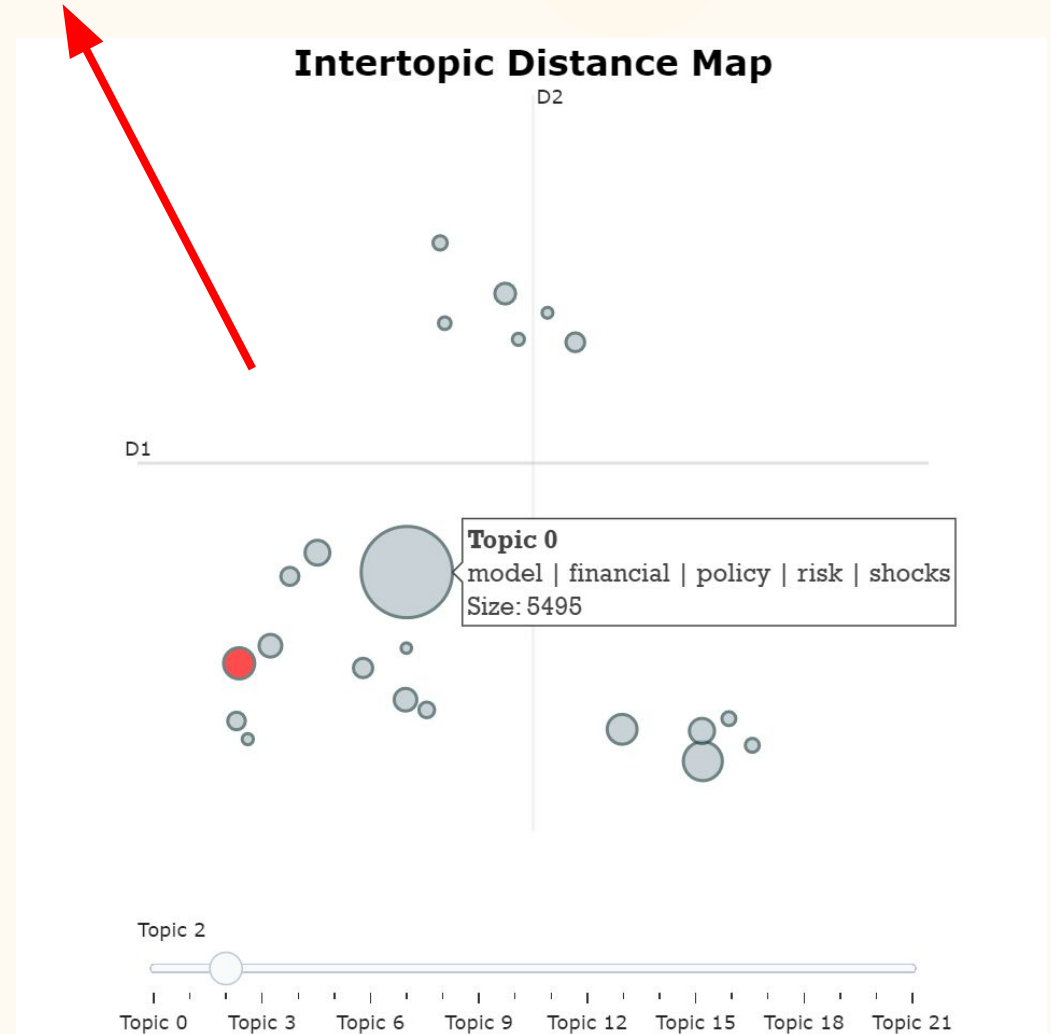
Similarity and Intertopic Distance

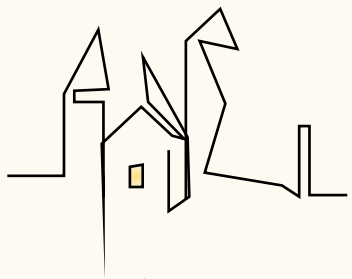
All topics do NOT overlap on Intertopic Distance Map, so the training result should be fine!

Similarity Matrix



Intertopic Distance Map





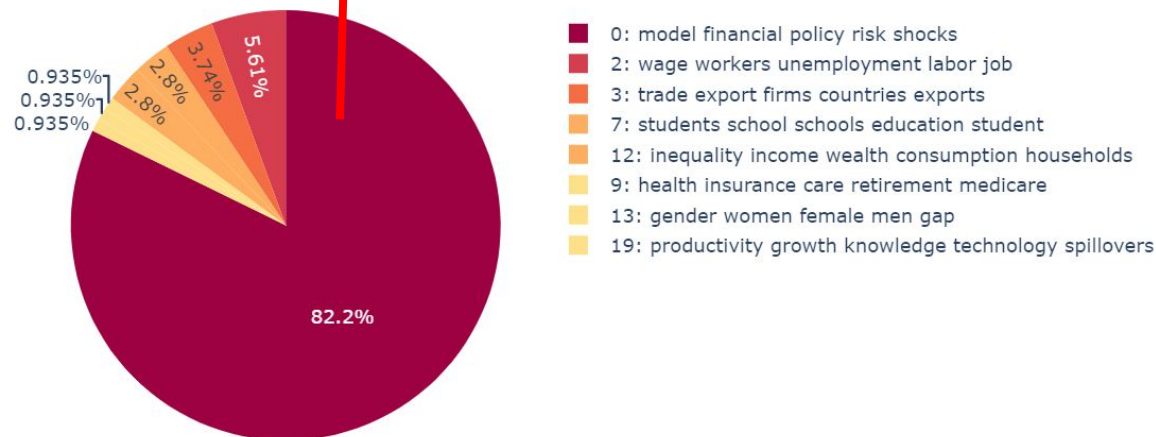
Topic Modeling Analysis

Specialization Areas of Different Institutions

- Every document is assigned with a single MOST RELEVANT topic.
- We can use this result to see what topics an institution is focusing on.

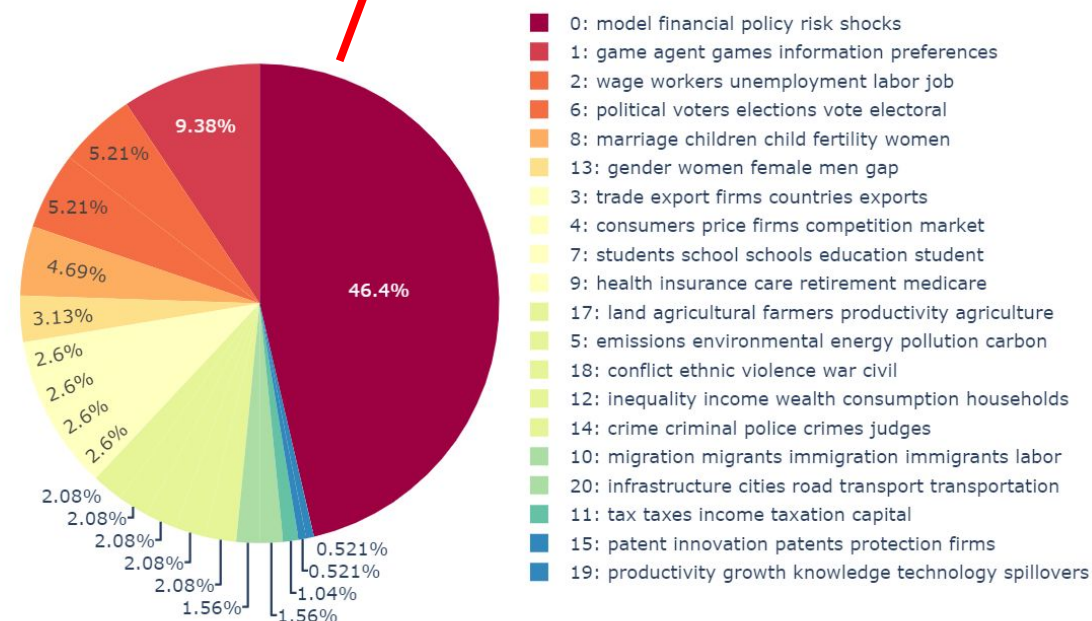
FED is mostly researching about “financial policy, risk, shocks”, macro related topics. Make sense!

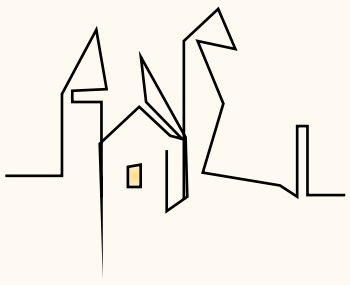
US Federal Reserve Board: Topics Distribution of Top Econ Publications



UChicago has a much broader interests of topics, almost covering all 21 topics we identify!

University of Chicago: Topics Distribution of Top Econ Publications





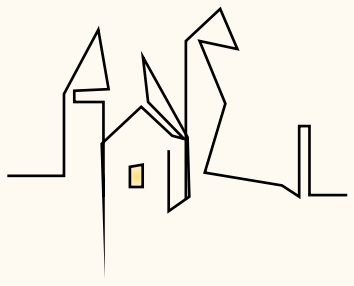
Take-home Points

From Technical Experience:

- **SQL** is really convenient to structure datasets and extract relevant information!
- Except for speed, **Selenium** is way better than Requests for web scraping!
- **Plotly** is fantastic for fancier visualization and interactive plots!
- When doing Topic Modeling, try out **BERTopic** package! Do NOT stick to Gensim LDA.

Social Science Significance:

- We form a whole picture of the current Economic academia by 1): identifying its collaboration patterns on Top Econ Journals through network analysis 2): research topics evolvement across time and institutions through dynamic topic modeling.
- *From network*, Top tier institutions are more likely to collaborate with institutions of similar levels; but for lower tier institutions, they often collaborate with those from very different tiers. In other words, an institution's cross-rank/tier collaboration likelihood decreases as its tier/rank gets higher.
- *From topic modeling*, we see topics are stable in trends across time with some fluctuations but can vary a lot across different institutions.
- *Future work*: 1) explore more parameters in network like centrality, shortest path; 2) further split the dominating topic into subtopics for better classification; 3) combining network and TM, see collaboration patterns under different topics.



This is the end of Presentation.
Thanks for Listening!

Q & A

