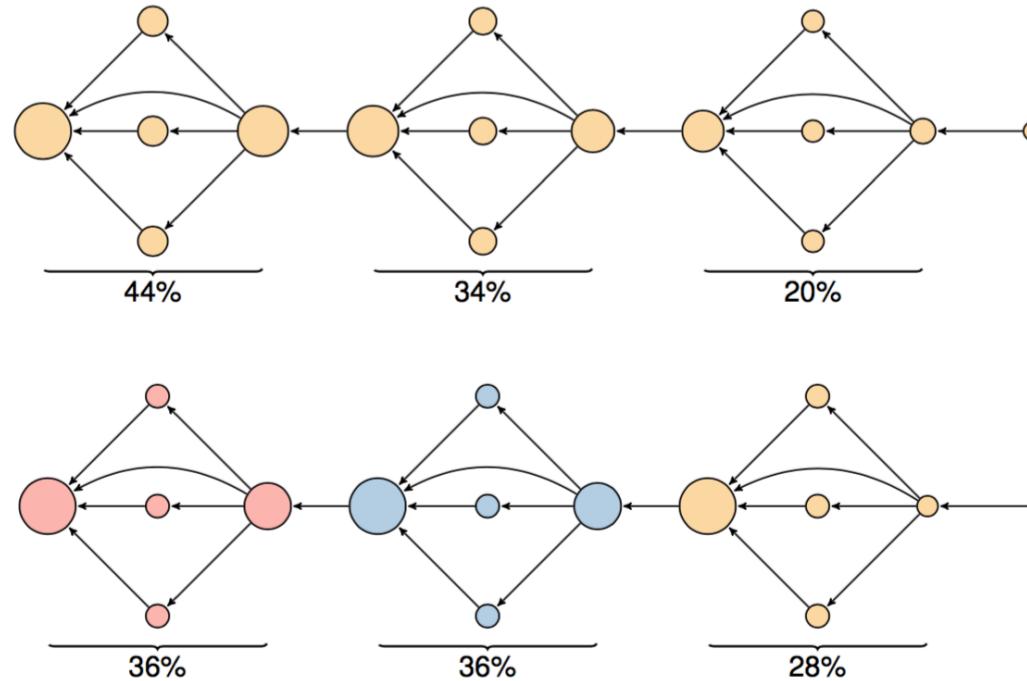


# Article-Level EigenFactor (ALEF)



Jevin West, Information School, University of Washington

Ian Wesley-Smith, Information School, University of Washington

Carl T. Bergstrom, Department of Biology, University of Washington

# WSDM CUP CHALLENGE

*SIGN-UPS FOR THE WSDM CUP CHALLENGE ARE NOW CLOSED*

## The Graph

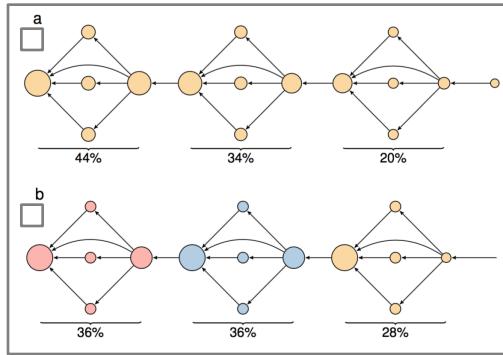
The Microsoft Academic Graph is a heterogeneous graph containing scientific publication records, citation relationships between publications, as well as authors, institutions, journal and conference "venues," and fields of study.

## The Data

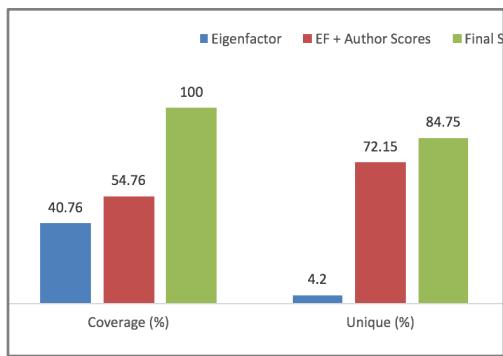
This data is available as a set of zipped text files stored in Microsoft Azure blob storage and available via HTTP. The file size (zipped) is ~30GB and may be downloaded [here](#).

## The Challenge

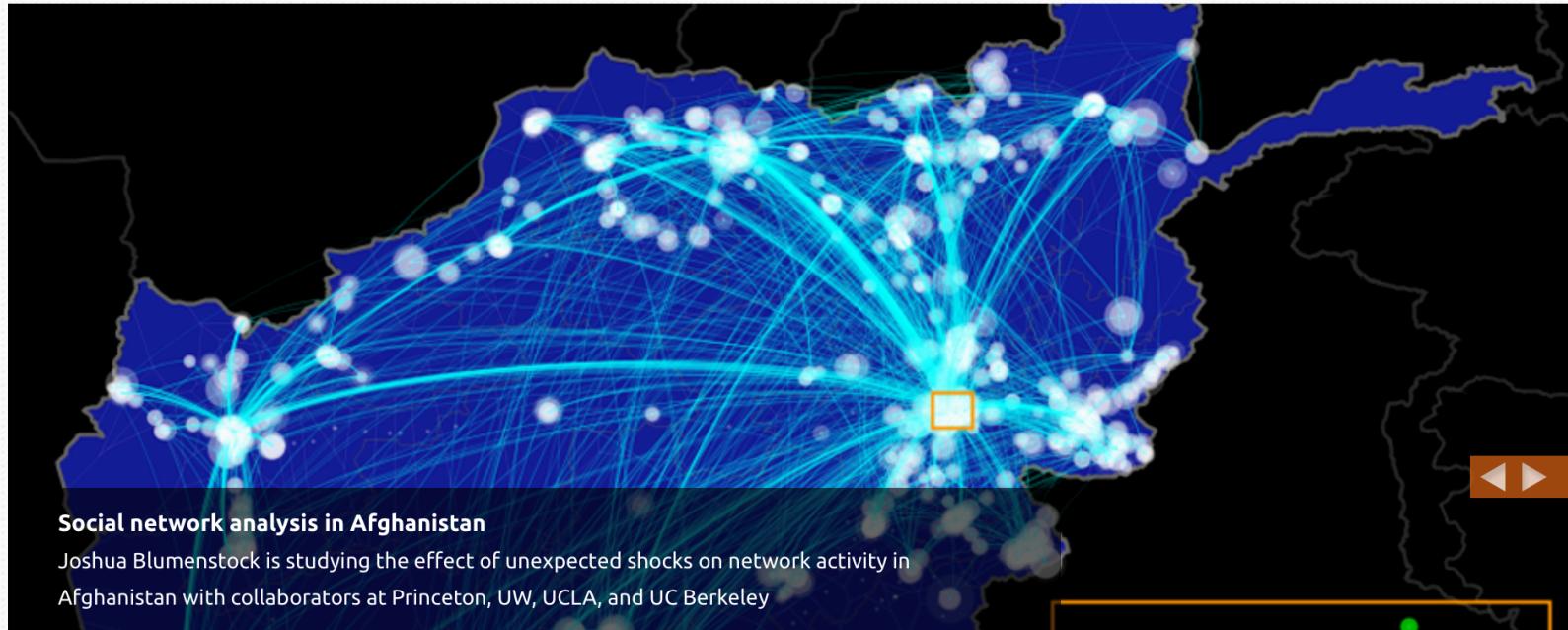
The goal of the Ranker Challenge is to assess the query-independent importance of scholarly articles, using data from the Microsoft Academic Graph.



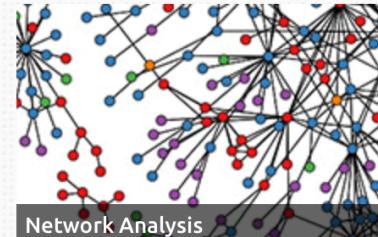
# Article-level Eigenfactor



# WSDM Cup Challenge



### Research Focus Areas



### News and Updates

28

Blumenstock at Population Association of America

### What we do

The DataLab is the nexus for research on Data Science and Analytics at the UW iSchool. We study **large-scale, heterogeneous human data** in an

# Journal Ranking

$$P = \alpha H + (1 - \alpha) a.e^T$$

Matrix representing the random walk over citations

Probability of not teleporting

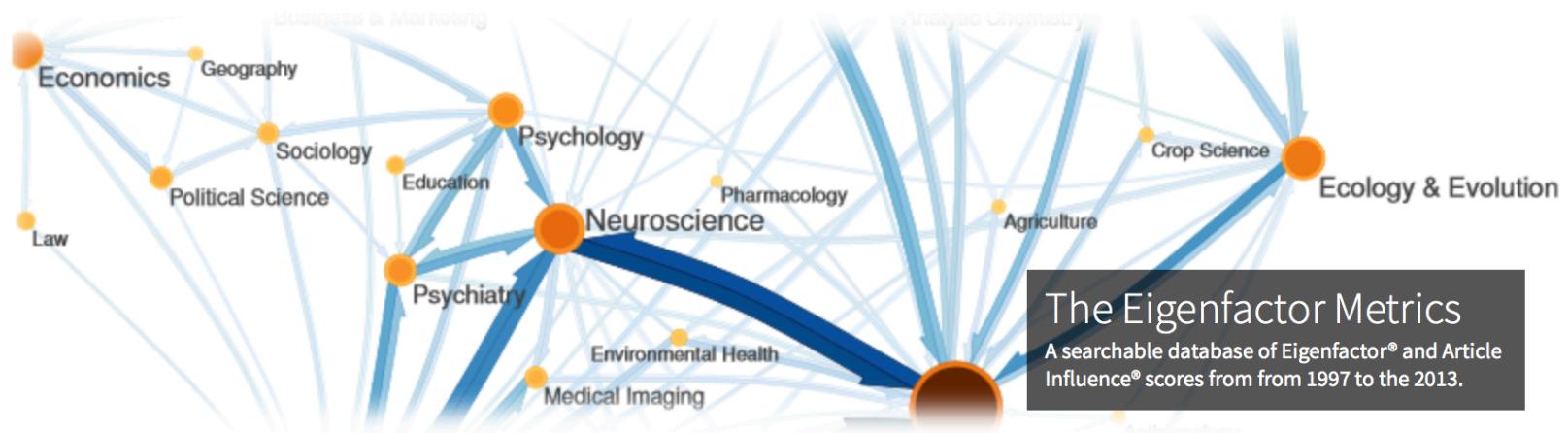
Cross-citation Matrix dictating the structure of the citation network

Probability of teleporting to completely new journal weighted by the number of articles in that journal

$$EF = 100 \frac{H\pi}{\sum_i [H\pi]_i}$$

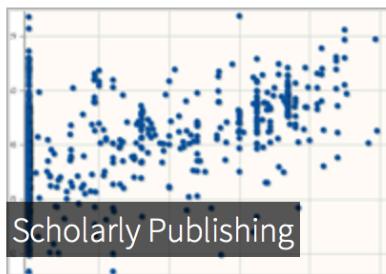
Leading eigenvector of the random walk matrix  $P$ .

Normalization

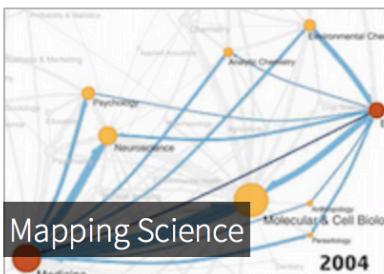


The Eigenfactor Metrics  
A searchable database of Eigenfactor® and Article Influence® scores from 1997 to 2013.

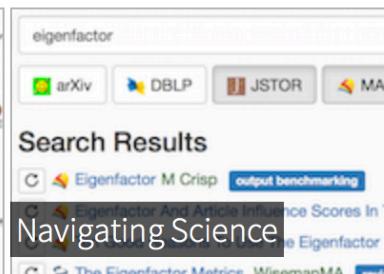
## RESEARCH AREAS



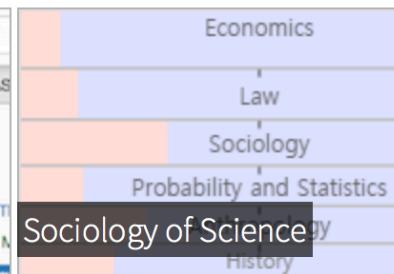
Scholarly Publishing



Mapping Science



Search Results  
Eigenfactor M Crisp output benchmarking  
Eigenfactor And Article Influence Scores In T...  
The Eigenfactor Metrics - WisemanMA



Economics  
Law  
Sociology  
Probability and Statistics  
History

## NEWS

23

Nov. JEVIN WEST ON MEGAJOURNALS IN THE *CHRONICLE OF HIGHER EDUCATION*  
Jevin West discusses the rise of the megajournal and our open access cost effectiveness tool in the *Chronicle of Higher Education*.

23

Nov. EIGENFACTOR TEAM PLACES SECOND IN MICROSOFT RESEARCH'S WSDM CUP  
The WSDM Cup Challenge asked teams to use 30GB of data from the Microsoft Academic Graph to rank the importance of individual articles. Using a mix of the article-level Eigenfactor algorithm and machine learning,

# Ranking and mapping article-level citation networks

Martin Rosvall\*  
IceLab, Umeå University\*

Jevin West  
Information School, University of Washington, Seattle, WA 98195-1800

Daril Vilhena and Carl T. Bergstrom  
Department of Biology, University of Washington, Seattle, WA 98195-1800  
(Dated: August 15, 2014)

Time-directed networks pose a challenge for flow-based methods of network analysis. Such networks are acyclic or nearly acyclic and thus very far from the nearly ergodic structures that flow-based methods are designed to handle. Without suitable modification, flow-directed ranking algorithms such as the Eigenfactor score put too much weight on older documents. Flow-based methods of cluster detection, such as the map equation approach, can fail to resolve important structures. Here we show how flow-directed methods can be modified to avoid these problems and thereby perform well on time-directed networks. To demonstrate the power of the new *article level Eigenfactor* metrics, we rank the 1.8 millions articles in JSTOR. To illustrate the power of our clustering approach, we create a hierarchical citation map of the JSTOR corpus.

Science is a massively parallel human endeavor to explain and predict the nature of the physical world. Thousands of individual scholars build cumulatively upon the prior work of a yet greater number of authors, and report upon their progress through their scholarly publications. Following the conventions of scholarly citation, each author references those predecessors most important in the development of her ideas. The most innovative research opens up avenues between new ideas—novel citation trails. Other researchers follow these tracks, guided by the citations that the pioneers laid down. What emerges is a latticework of citations from which we can in principle map the geography of scientific thought and retrace the pathways along which intellectual activity has proceeded. As Derek de Solla Price famously noted in 1965, this lattice is in fact a vast network of citations, growing dynamically and organically, doubling in size every ten to twenty years [1]. Our aim is to map out the way that ideas flow through scientific communities, so that we can comprehend large-scale patterns, identify important contributions, and better navigate the literature.

Network theory offers a rich set of tools for ranking and

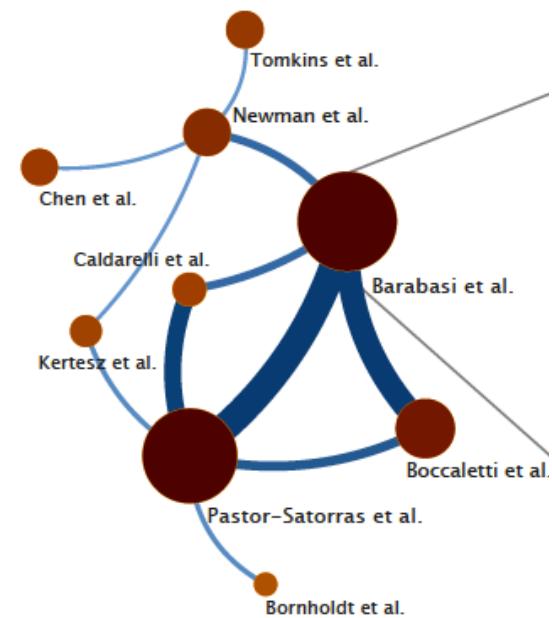
trails of citations move inexorably back in time.

In this paper, we show how flow-based methods of network analysis can be adapted for use with time-directed and acyclic networks. This will allow us to apply flow-based methods to map science at the scale of individual articles. To use network analysis to better navigate the literature, we also present a method to label the structures that we uncover at every scale, using the textual content of the articles themselves. We will need to do all of this in a way that is scalable to the full universe of scholarly publication, and that is updatable, so that researchers may always be navigating with maps that are current not to years, but to days. These methods will also be suitable for studying patent networks, court case networks, and other similarly time-directed structures.

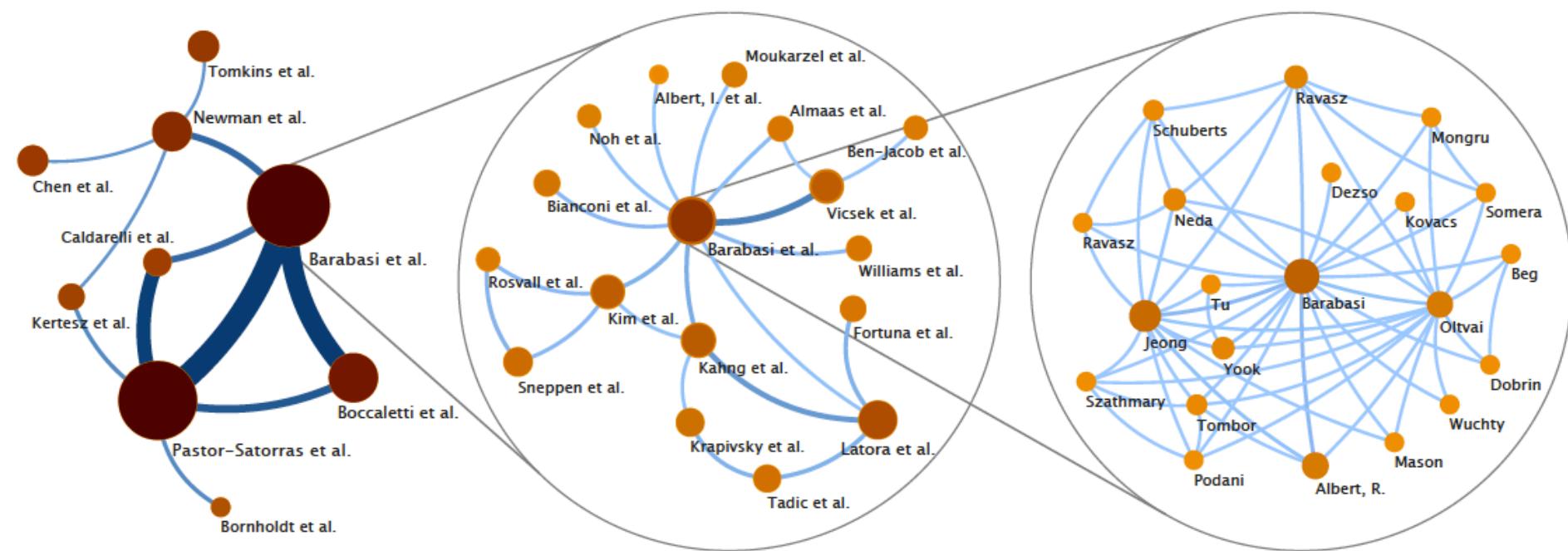
## I. RANKING

The standard PageRank algorithm can be viewed as tracing the path of a random walker on a directed, and possibly weighted, network. Most of the time, the random walker

# Hierarchical Mapping without ALEF

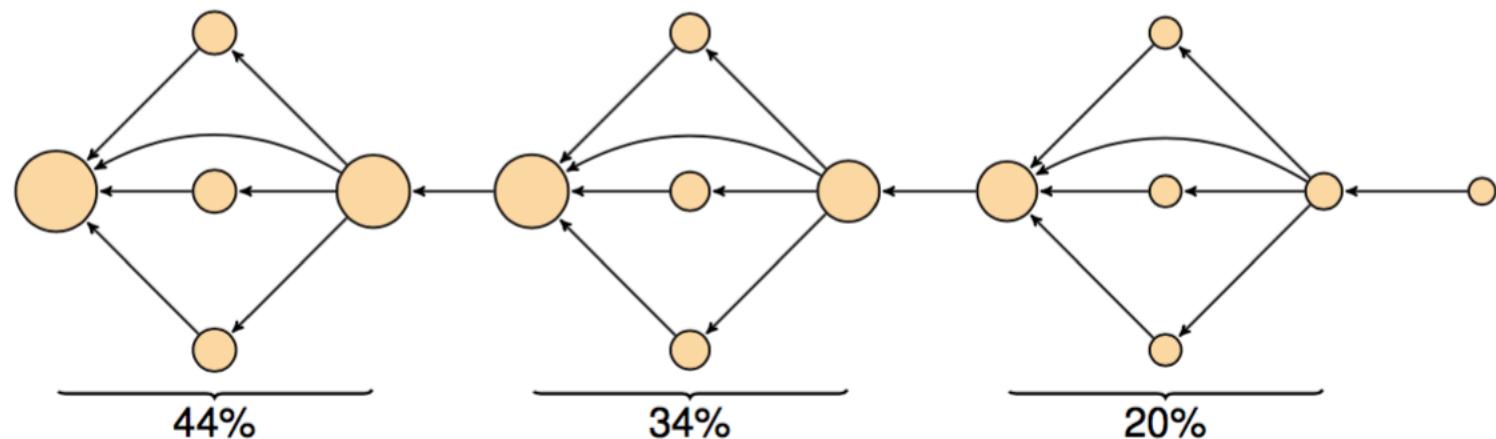


# Hierarchical Mapping with ALEF

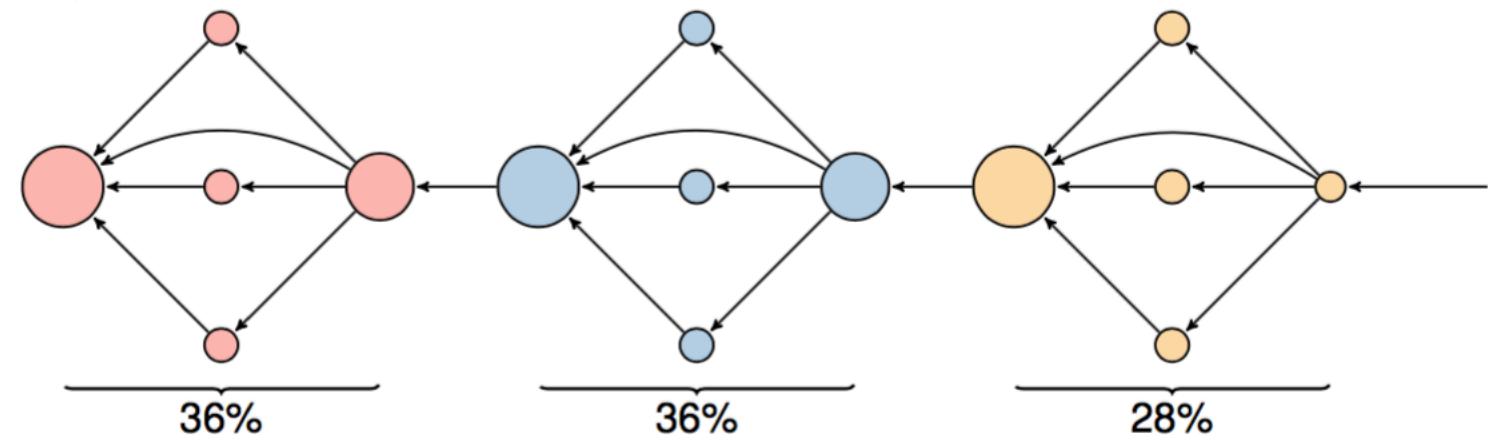


# Flow Distribution

PageRank

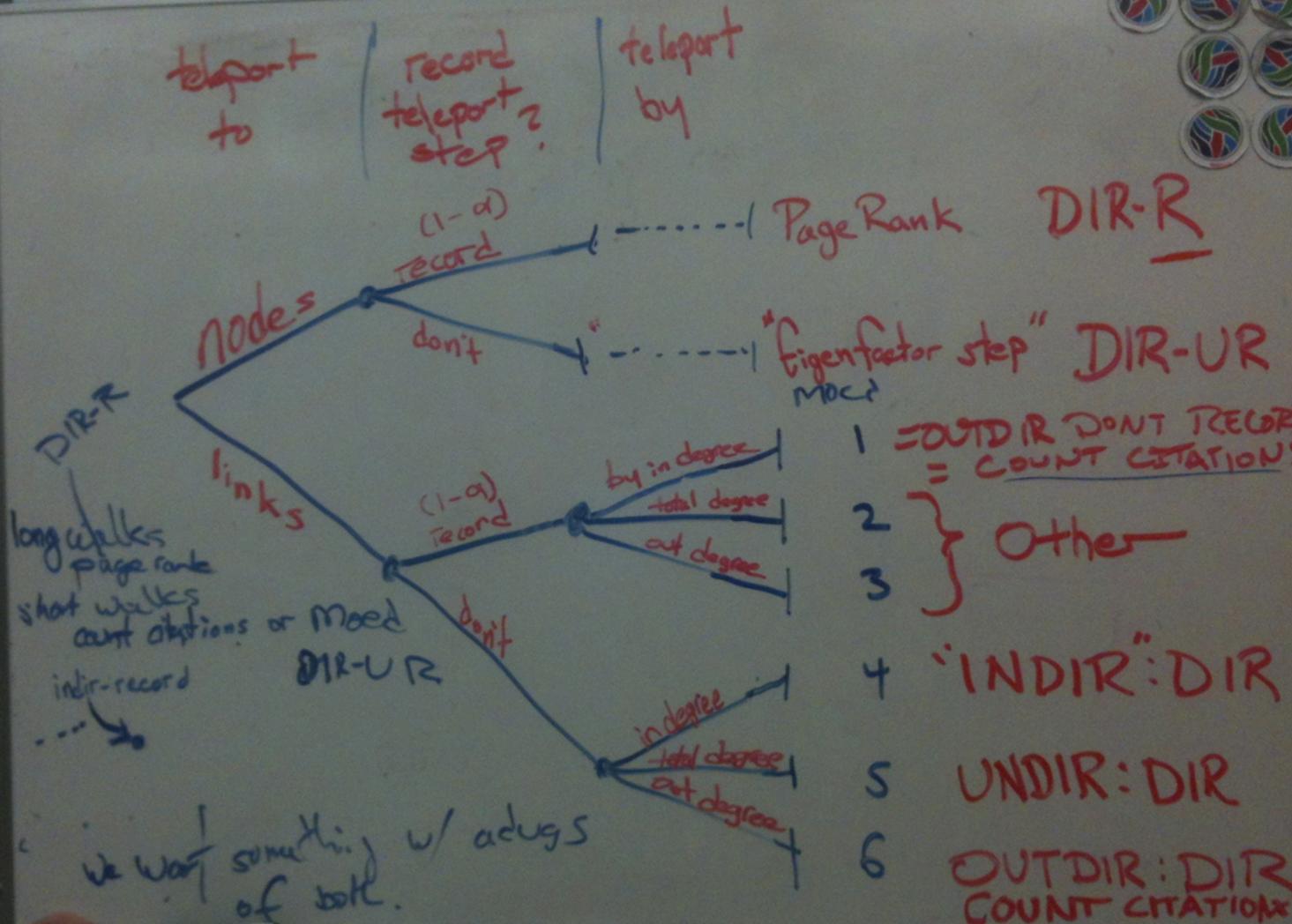


ALEF



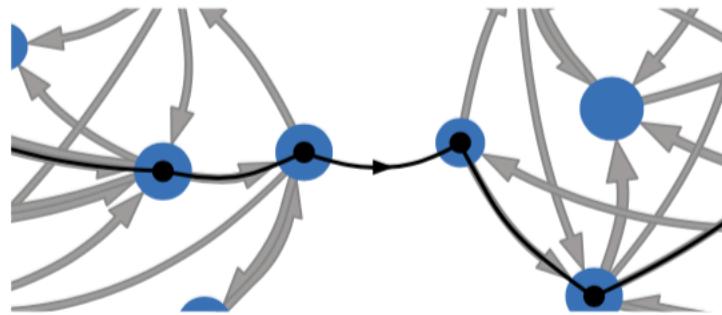
Time



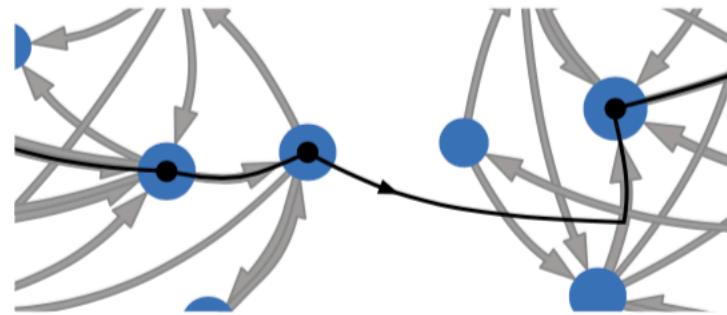


# Smart Teleportation

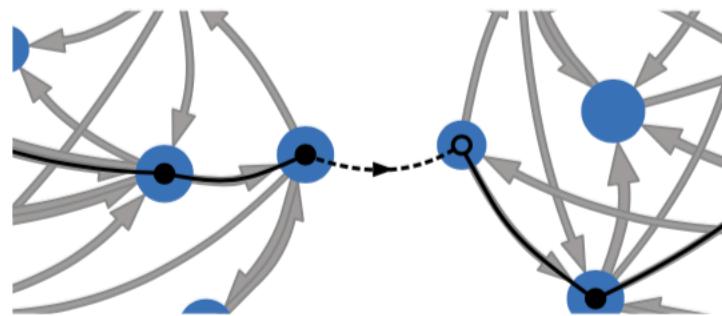
(a) Recorded node teleportation



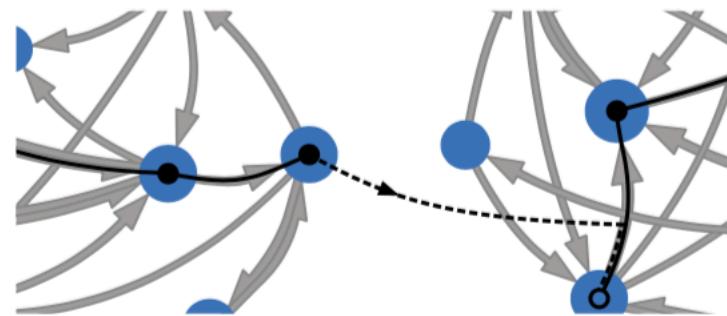
(b) Recorded link teleportation



(c) Unrecorded node teleportation



(d) Unrecorded link teleportation



# Mechanics

1. calculate step weight

$$w_i = \sum_j^n (Z_{ij} + Z_{ij}^T)$$

adjacency matrix



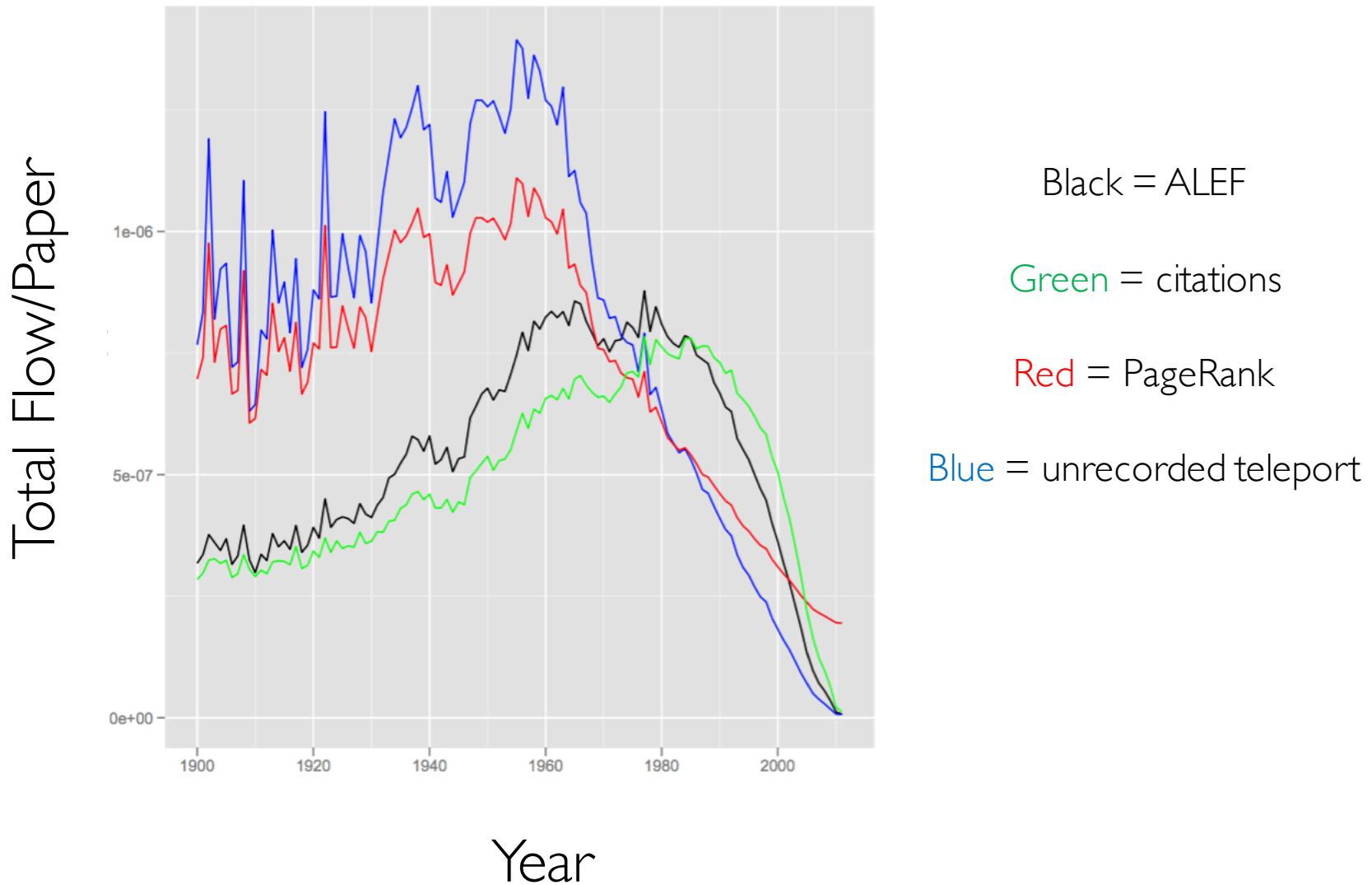
2. make row stochastic

$$\mathbf{H}_{ij} = \frac{\mathbf{Z}_{ij}}{Z_i}$$

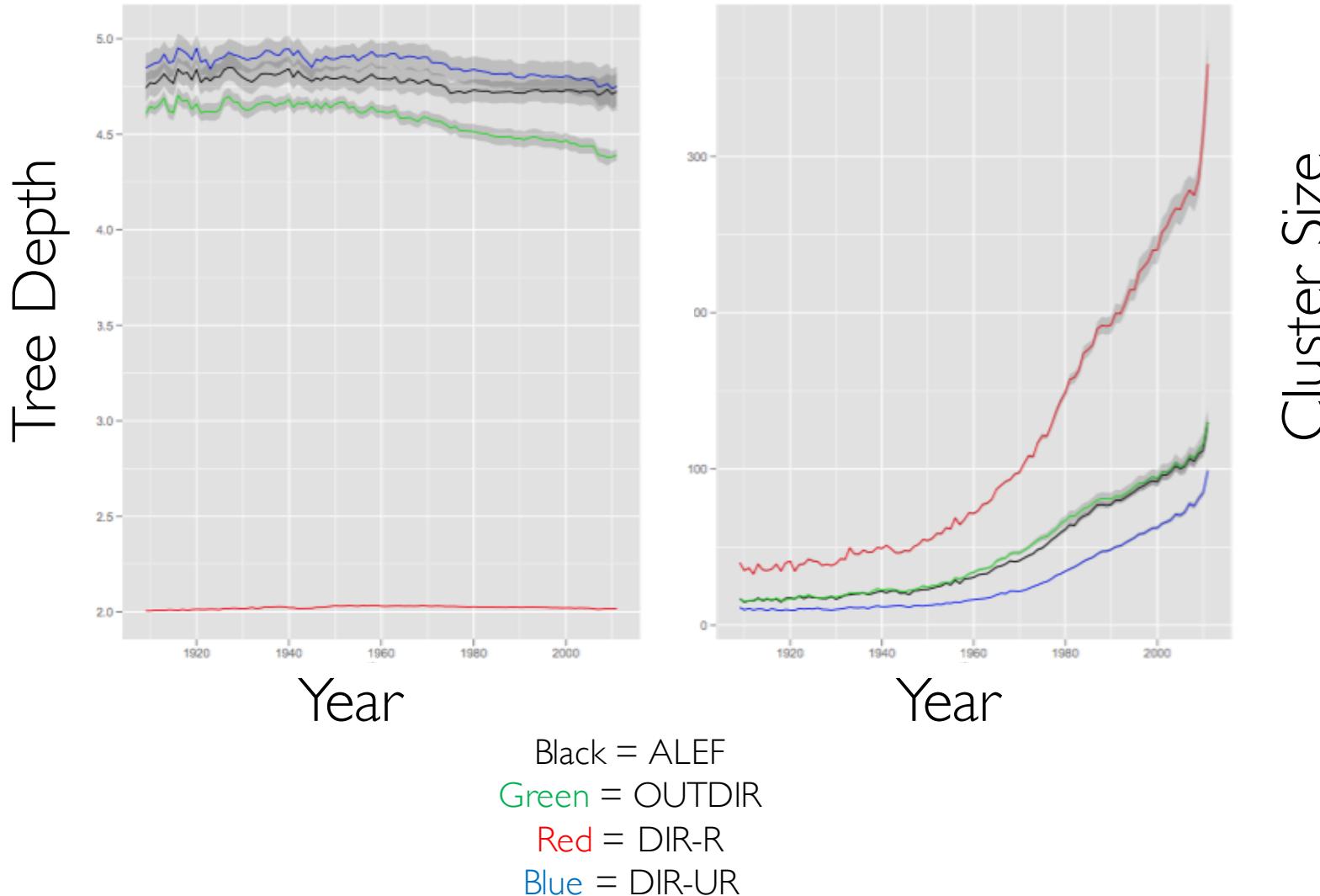
3. one-step on network

$$\text{ALEF} = n \frac{\mathbf{H}_{ij}^T \cdot w_i}{\sum_j [\mathbf{H}_{ij}^T \cdot w_i]_j}$$

# Flow Distribution (JSTOR)



# Tree Depth and Cluster Size



# ALEF Strengths

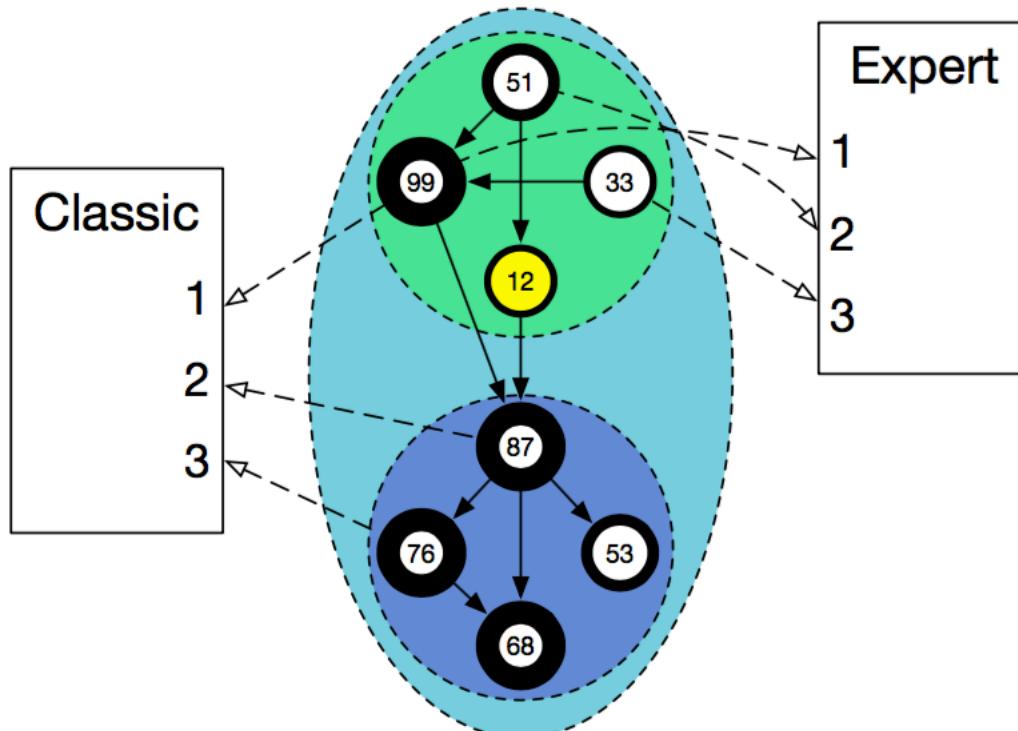
Performs well

Simple mechanics

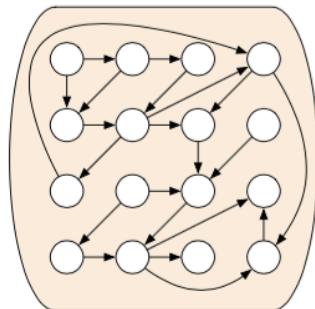
Fast calculation

High resolution partitions

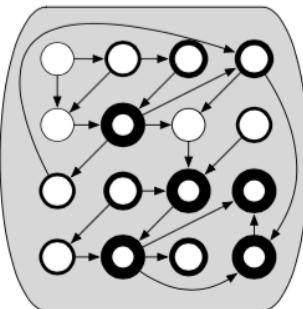
# Recommend



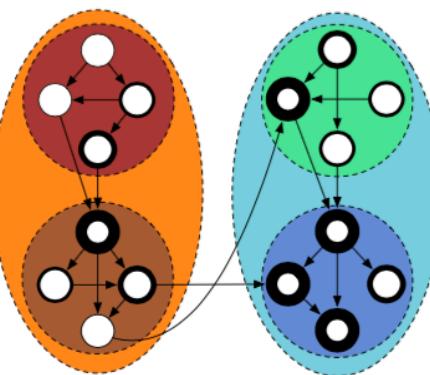
Assemble



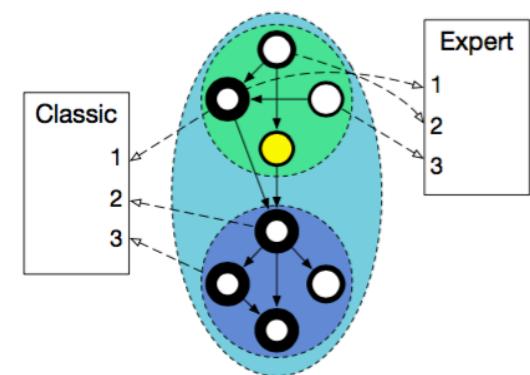
Rank



Cluster



Recommend



West, Wesley-Smith, Bergstrom (2016) A recommendation system based on hierarchical clustering of an article-level citation network. *IEEE Transactions on Big Data*

# Papers

- J.D. West, M. Rosvall, C.T. Bergstrom (2016) Ranking and mapping article-level citation networks, *in prep*
- J.D. West, I. Wesley-Smith, C.T. Bergstrom (2016) A recommendation system based on hierarchical clustering of an article-level citation network. IEEE Transactions on Big Data
- I. Wesley-Smith, C.T. Bergstrom, J.D. West (2016) Static Ranking of Scholarly Papers using Article-Level Eigenfactor (ALEF), WSDM Conference: Entity Ranking Challenge Workshop
- I. Wesley-Smith, J.D. West (2016) Babel: A platform for research in scholarly article recommendation. WWW Conference, Workshop on Big Scholarly Data

oren etzioni



DBLP

JSTOR

MAS

PLOS

PubMed

S

Statistical Methods For Analyzing Speedup Learning Experiments O Etzioni satisfaction programs

Face And Computer-Mediated Communities Amitai Etzioni, Oren Etzioni 1998 resources sustained

Document Clustering O Zamir document clustering

Communities: Virtual Vs. Real A Etzioni 1996 implications internet

Statistical Methods For Analyzing Speedup Learning Experiments. O Etzioni 1993 scheduling problems

Statistical Methods For Analyzing Speedup Learning Experiments O Etzioni 1993 generating abstractions

Get Related

Web Document Clustering: A Feasibility Demonstration O Zamir 1997 document clustering

Get Related

Web Document Clustering: A Feasibility Demonstration. O Zamir 1997 browsing large

Get Related

Sound And Efficient Closed- World Reasoning O Etzioni proving problem

Get Related

Appears In Comm. OfACM O Etzioni scalable comparison-shopping

« Previous

1

2

3

4

5

6

7

8

9

10

Next »

## Papers related to

Statistical Methods For Analyzing Speedup Learning Experiments O Etzioni satisfaction programs

Get Related

Automatically Configuring Constraint Satisfaction Programs: A Case Study S Minton 1995 satisfaction programs

Get Related

Abstraction Via Approximate Symmetry T Ellman 1992 satisfaction programs

Get Related

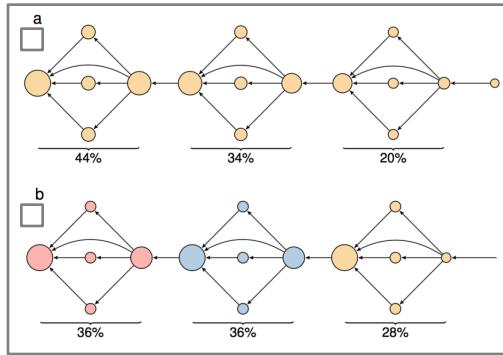
Integrating Heuristics For Constraint Satisfaction Problems: A Case Study S Minton 1992 satisfaction programs

Get Related

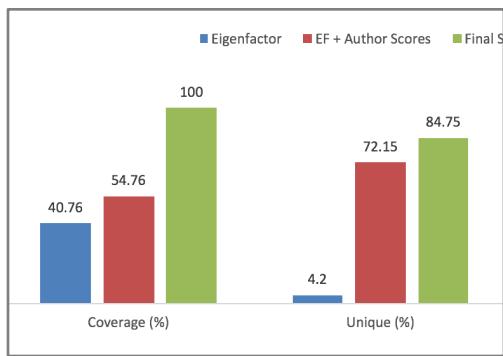
An Analytic Learning System For Specializing Heuristics S Minton 1992 satisfaction programs

Get Related

Automated Synthesis Of Constrained Generators W Braudaway 1988 satisfaction programs

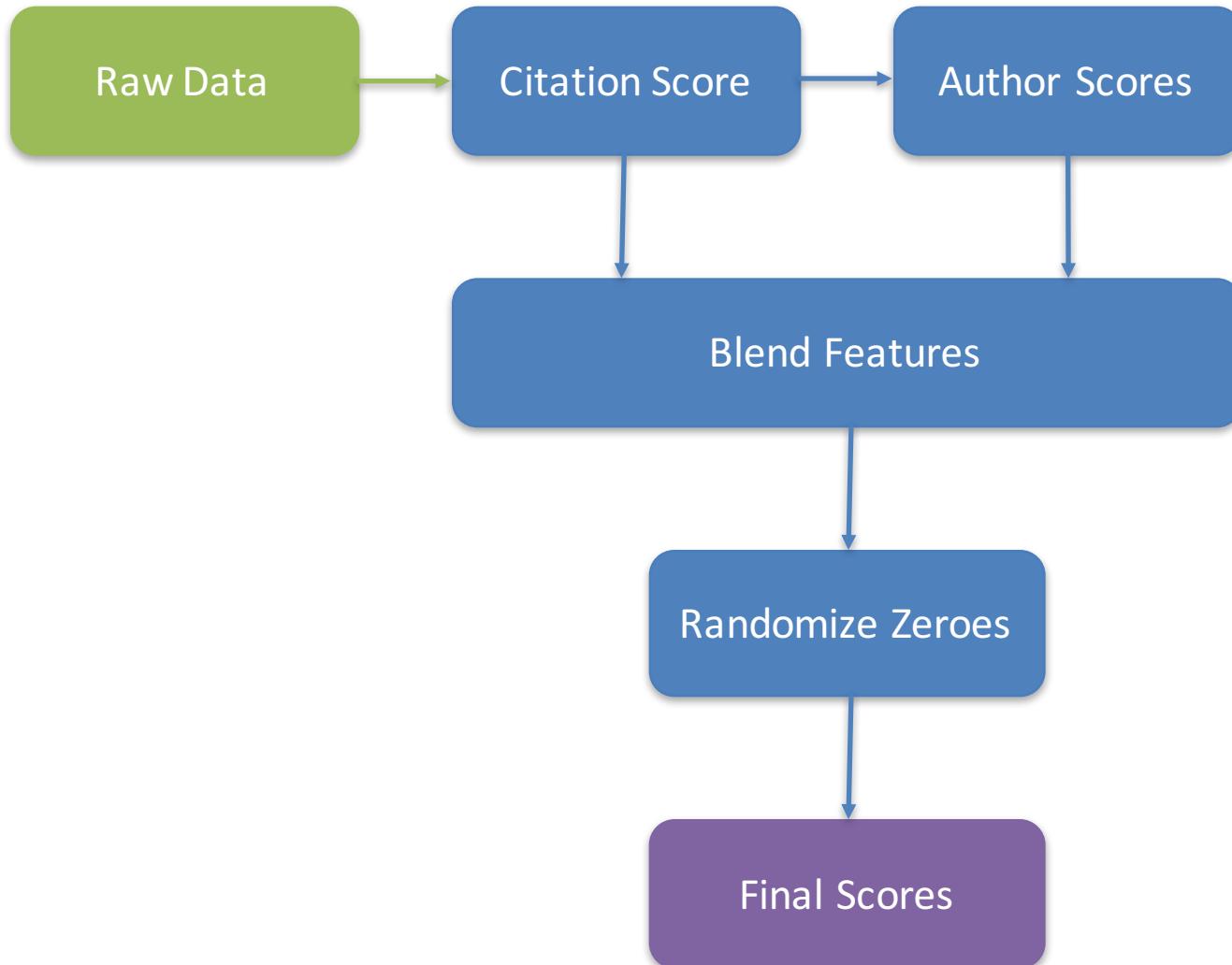


# Article-level Eigenfactor

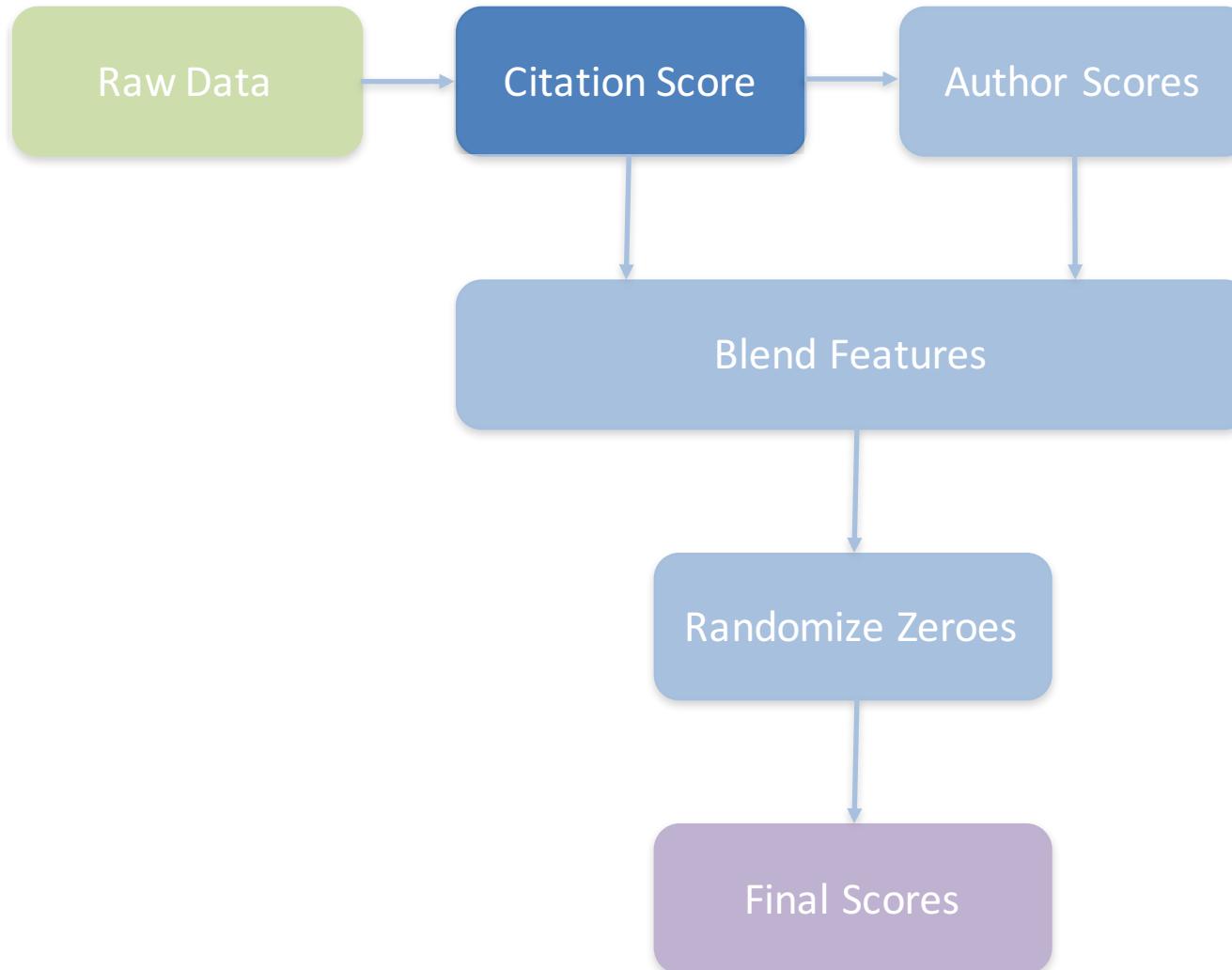


# WSDM Cup Challenge

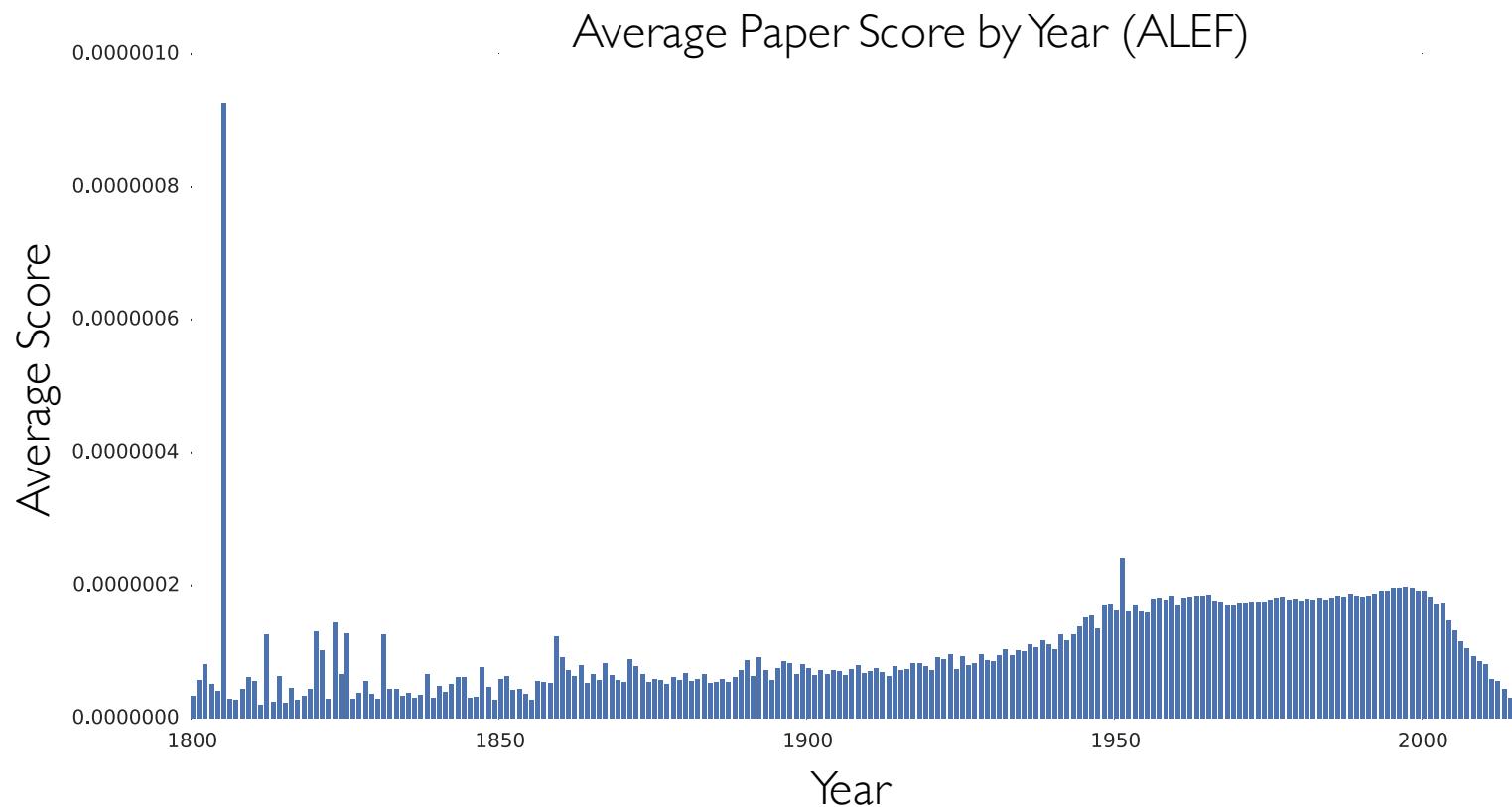
# Data Pipeline



# Citation Scores

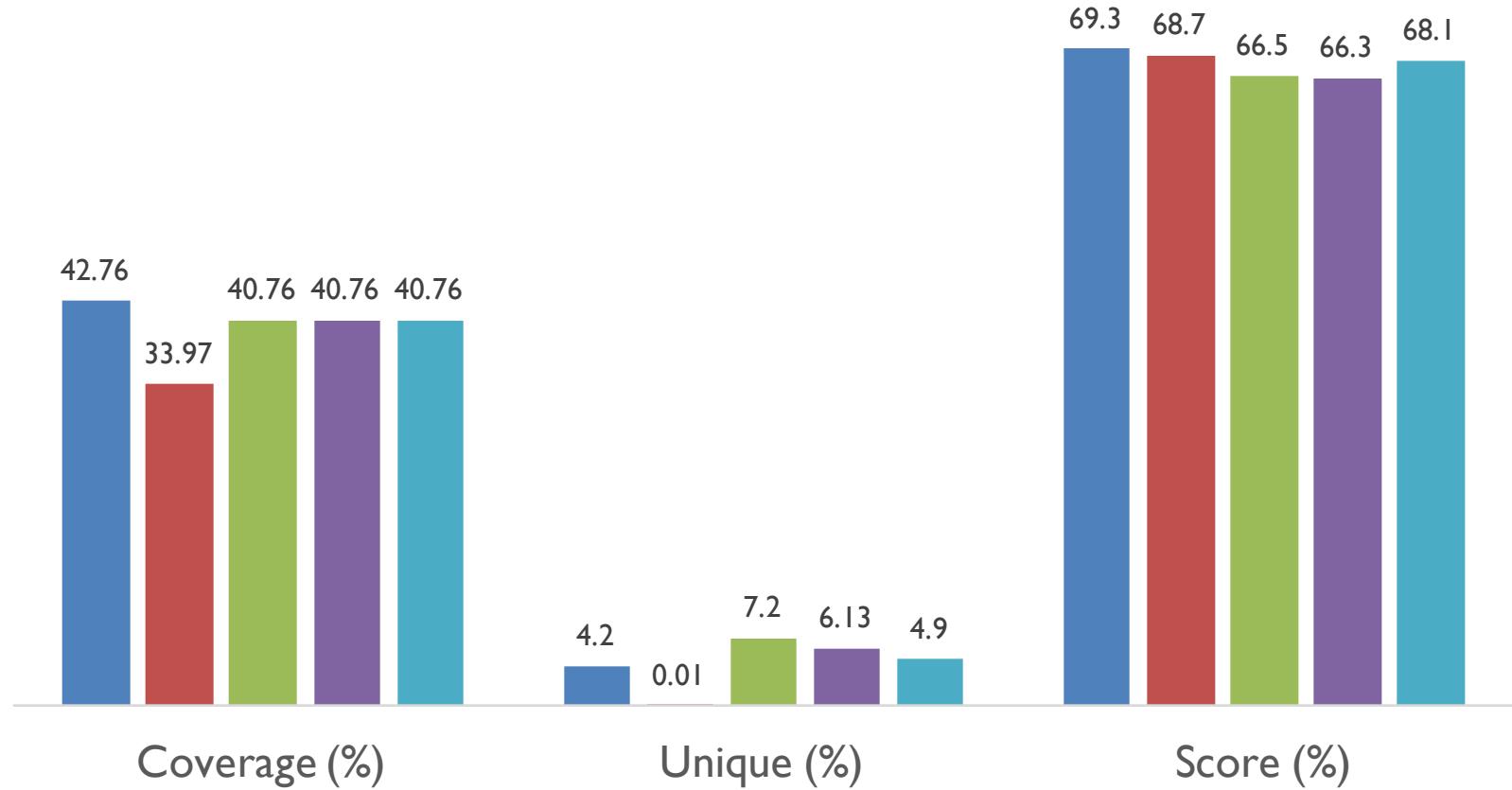


# Citation Scores

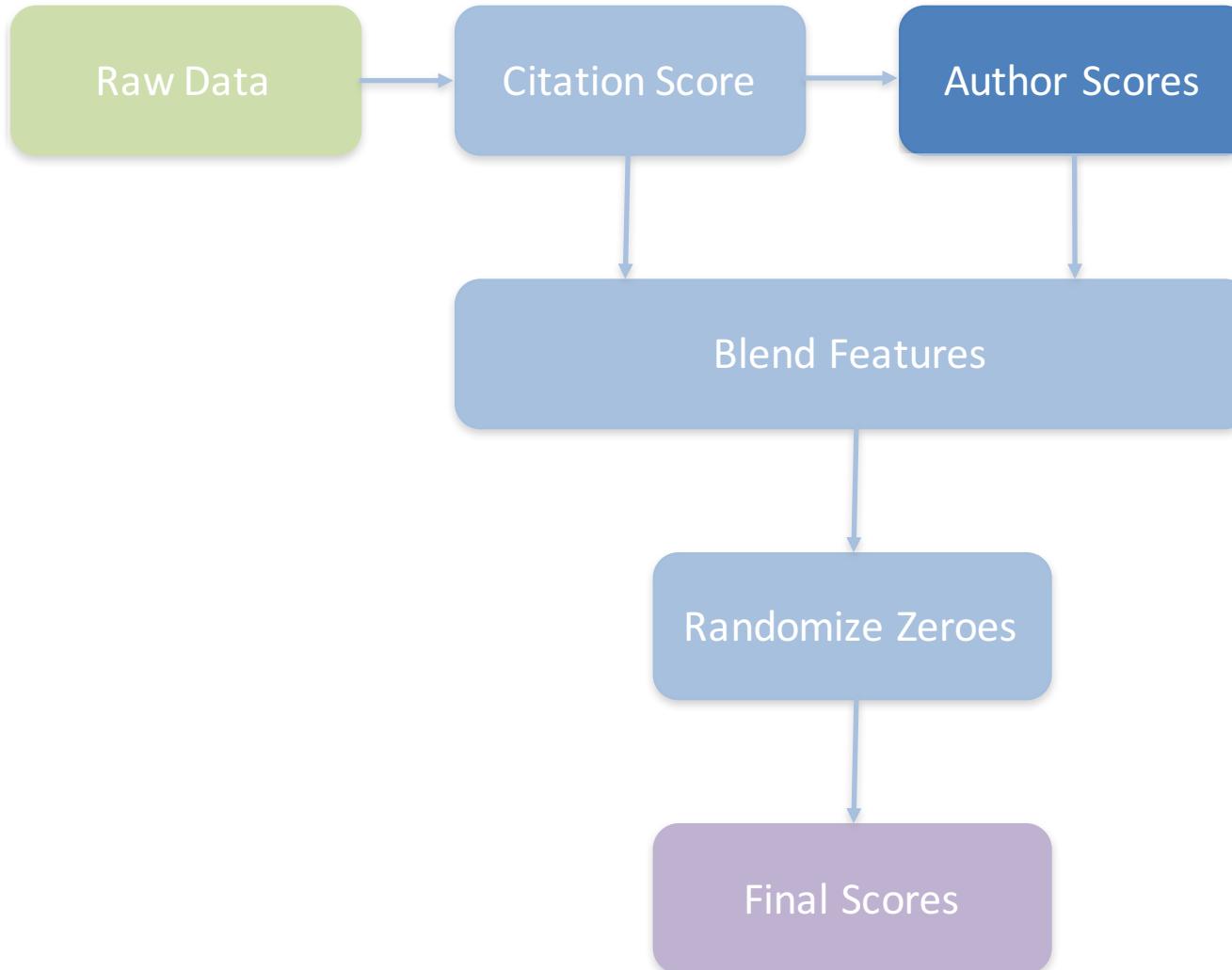


# Citation Variants

■ ALEF ■ Degree Centrality ■ 2-Step ■ In Citations ■ Uniform



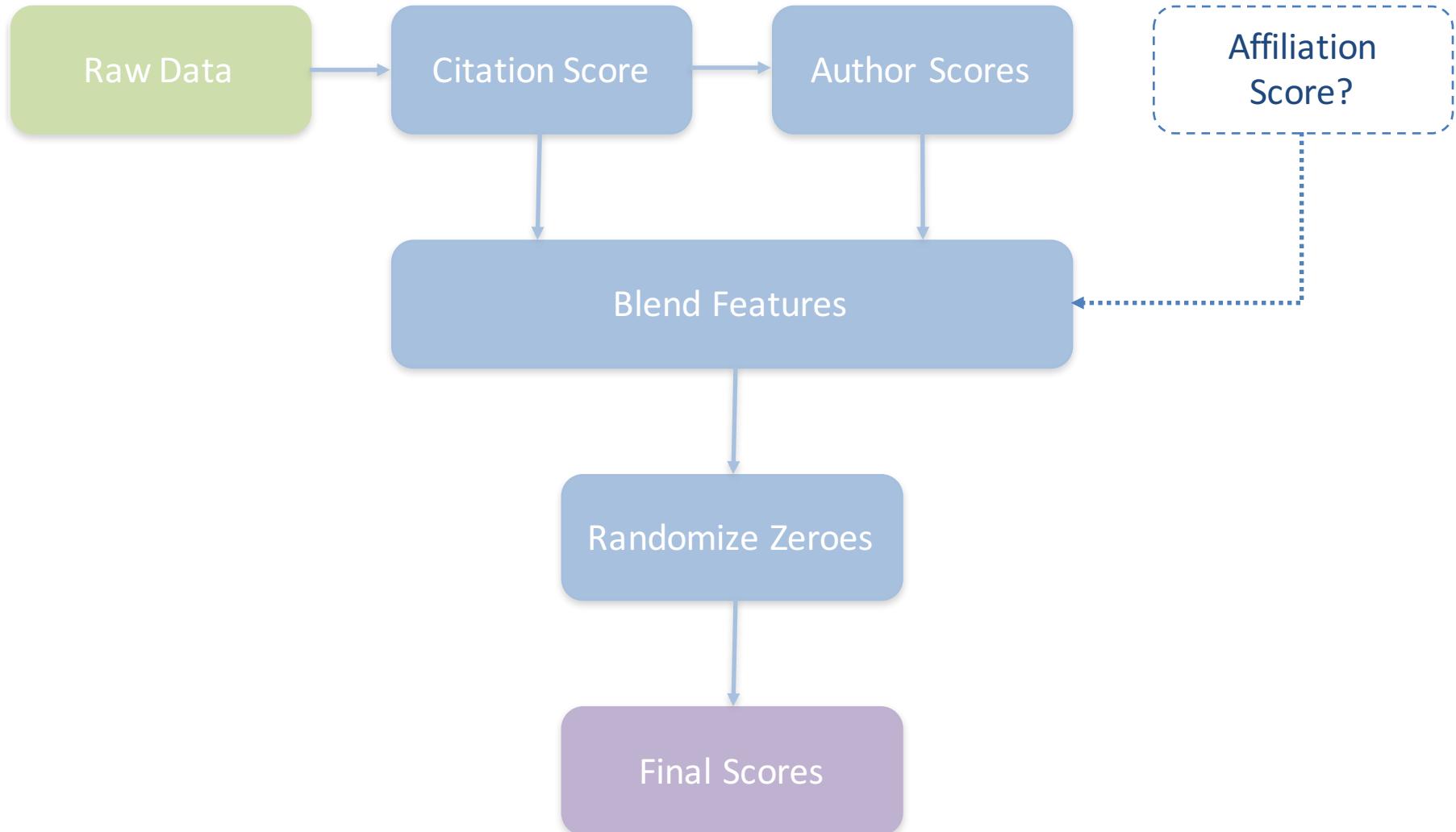
# Author Scores



# Author Scores

- Author Score = Average citation score of all papers
- How should paper credit be assigned?
  - Equally or Fractional?
- Why not sum?
  - Unique Scores: 72.15% vs 28.27%

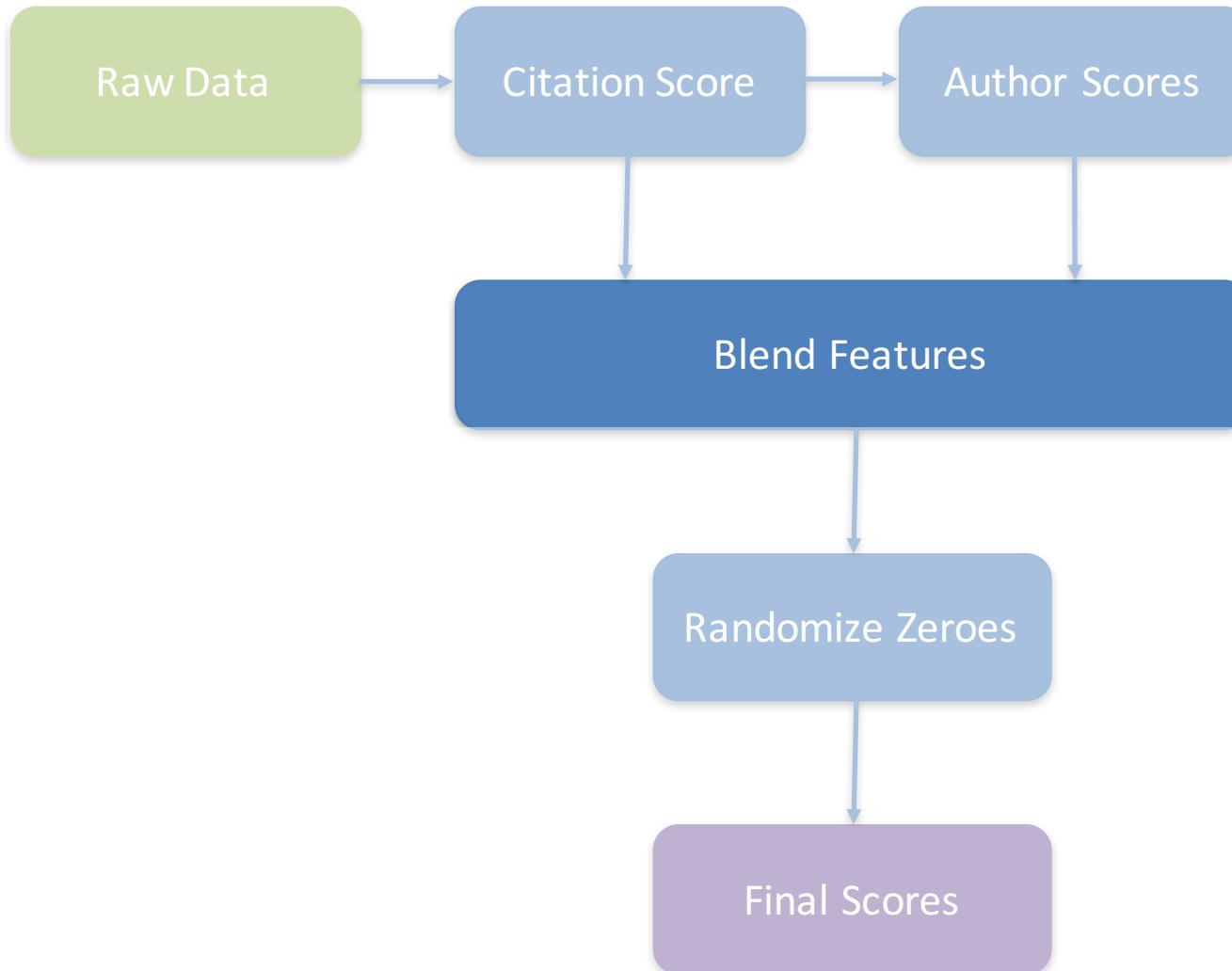
# Other Features?



# Other Features

- Matching datasets is hard
- Author Affiliation: University of Washington
  - george washington university
  - university of washington bioengineering
  - university of washington information school
  - university of washington school of law
  - university of washington tacoma
  - university of washington bothell
- Coverage is low: 25% of paper-author pairs have an affiliation

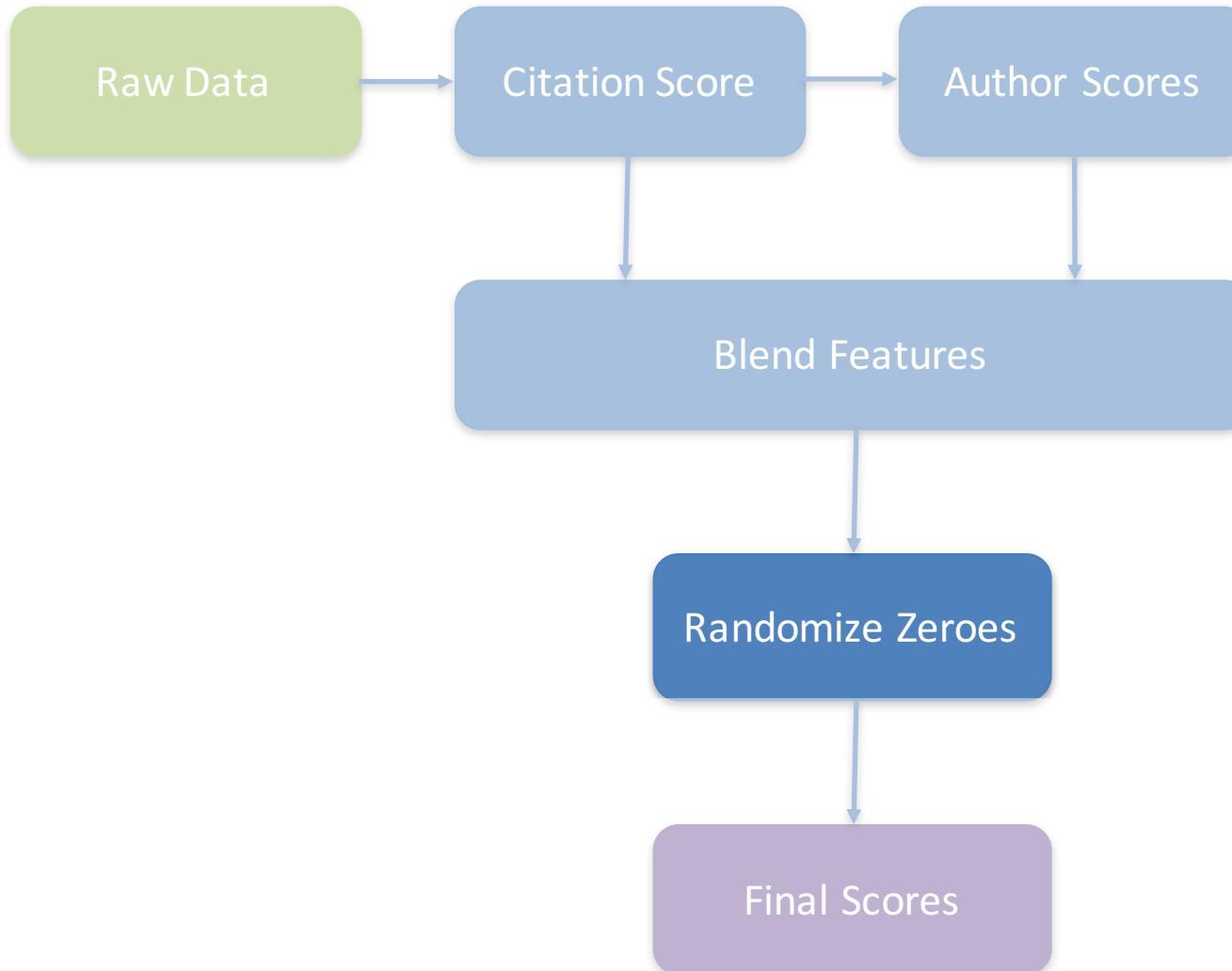
# Blend Features



# Blend Features

- Weighted Average
  - Weights found via manual parameter sweep
  - Citation Score: 70%
  - Author Score: 30%
- Axiom: Derived scores shouldn't outweigh the source

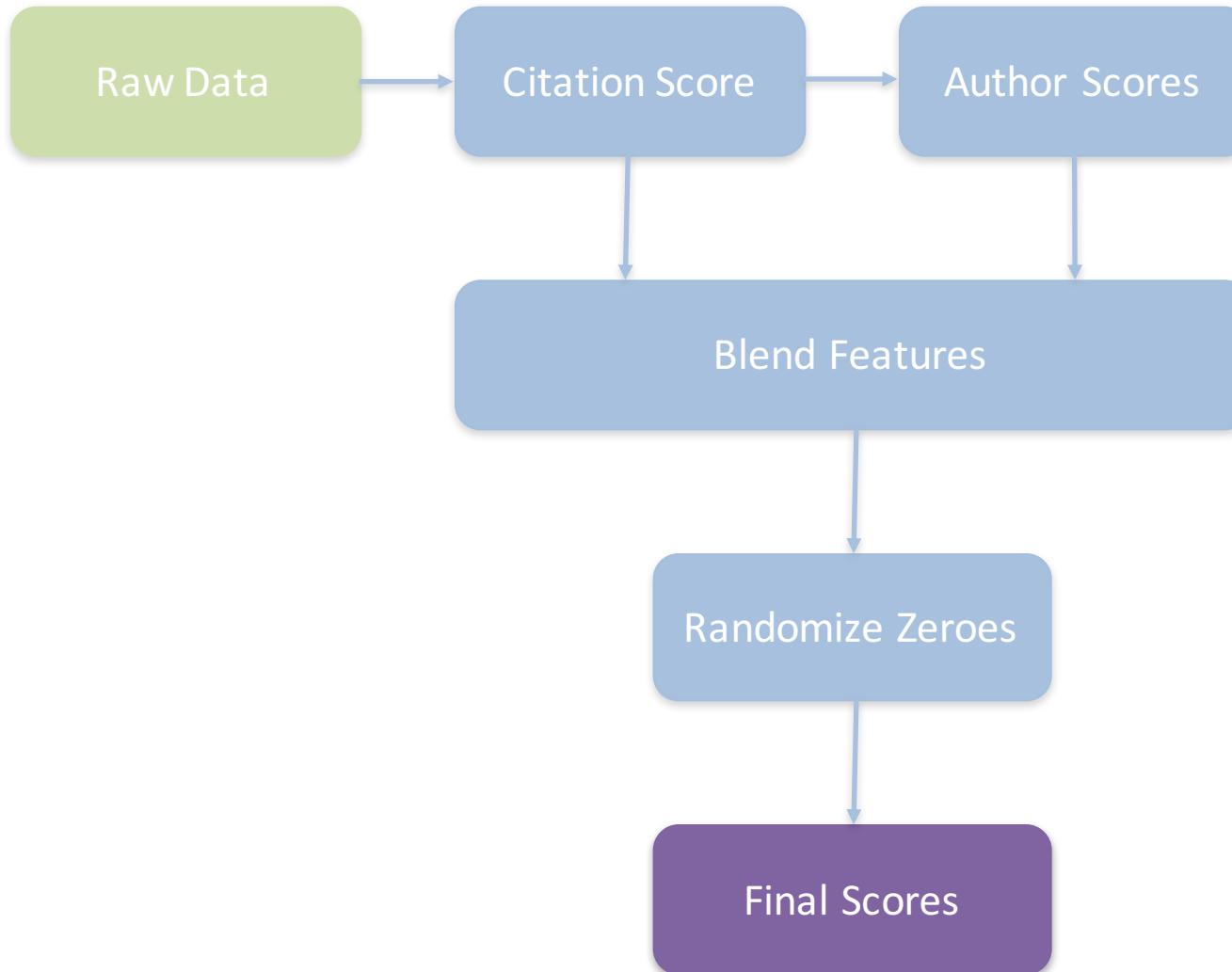
# Randomize Zeroes



# Random Chance?

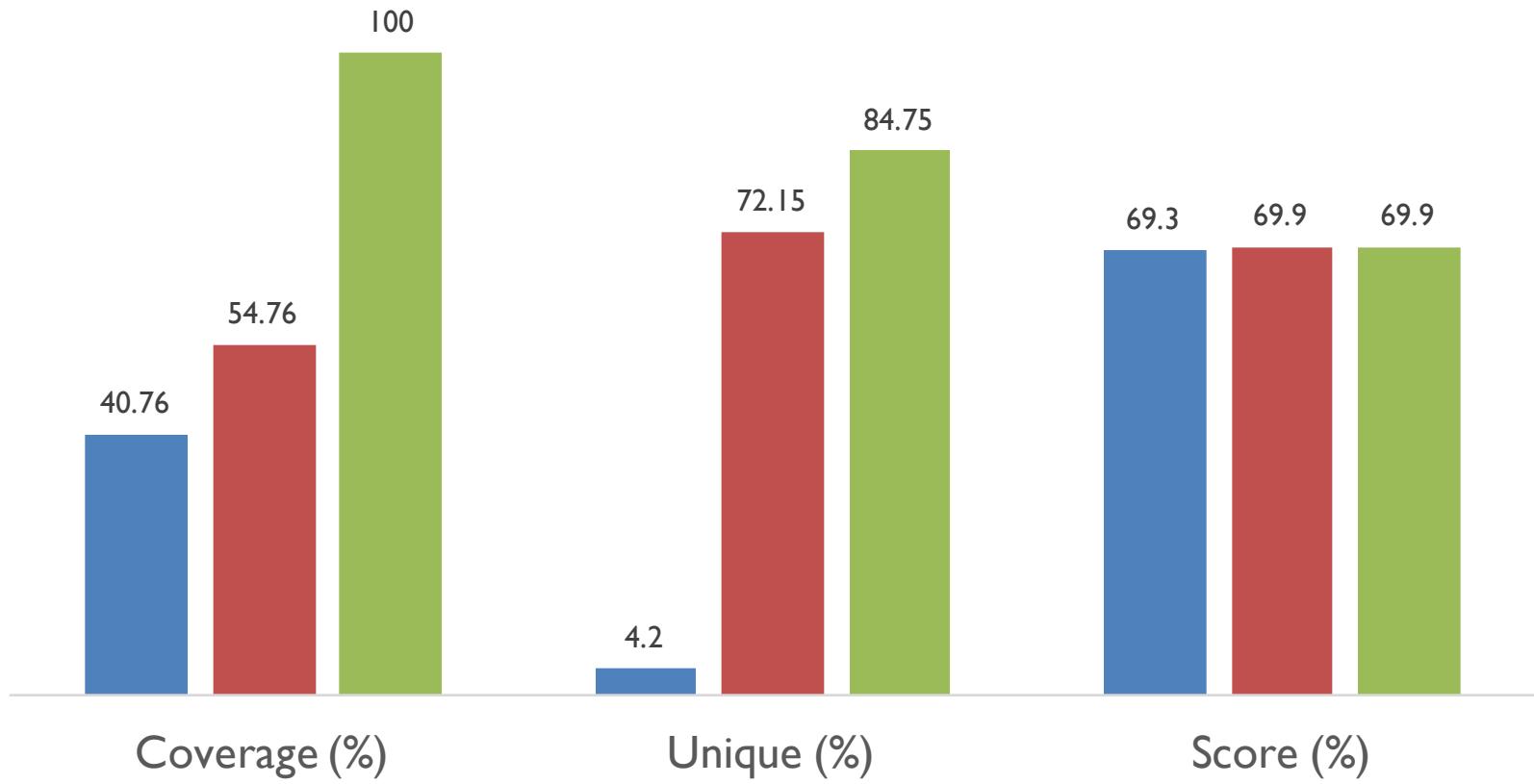
- Our best isn't much better than random
  - Random: 52.6%
  - 1<sup>st</sup>: 68.3% (+30%)
- This judging is favorable to random chance
- Unscored papers assigned [0, minval \* 0.999]

# Phase I Results



# Submissions

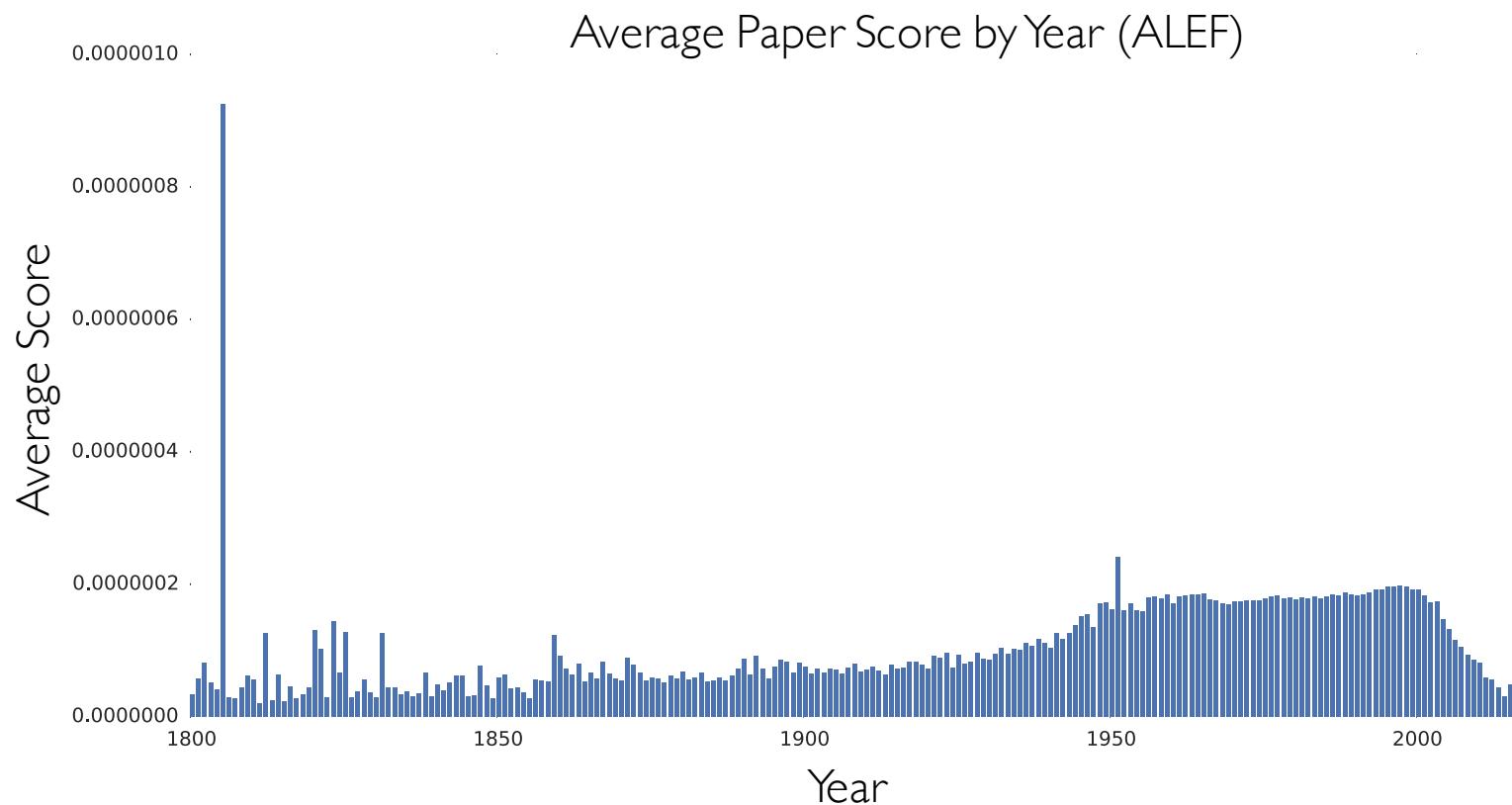
■ ALEF ■ ALEF + Author Scores ■ Final Submission



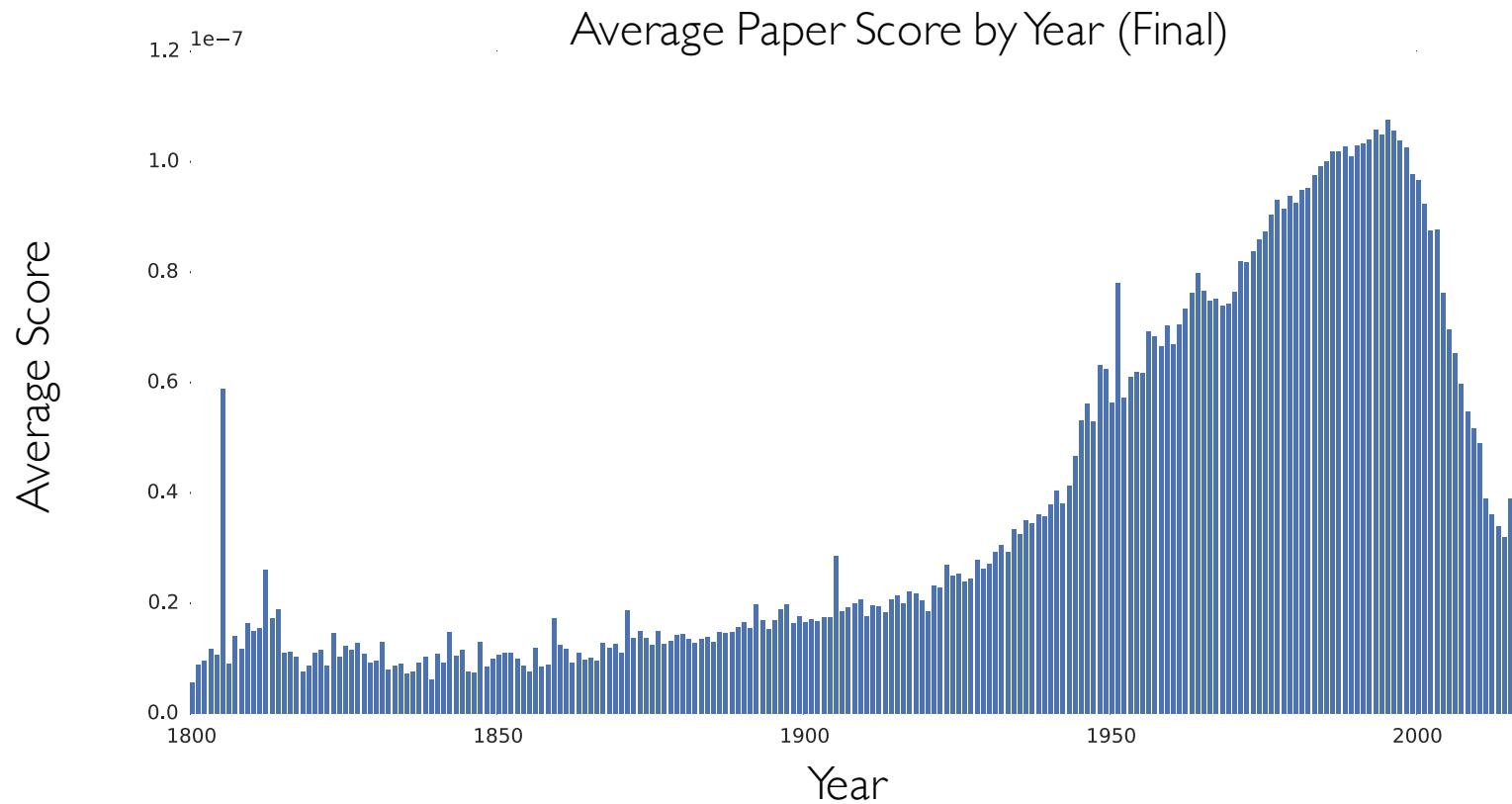
# Submissions

Run	Score	Density	Coverage
1	0.693	4.20%	40.76%
18	0.307	8.29%	15.10%
19	0.699	28.27%	54.76%
20	0.693	28.37%	54.76%
22	0.641	96.15%	100.00%
23	0.44	59.25%	9.44%
25	0.699	60.72%	100.00%
26	0	100.00%	45.24%
27	0.665	7.20%	40.76%
28	0.681	4.90%	40.76%
29	0.663	6.13%	40.76%
30	0.528	5.20%	31.76%
31	0.693	71.54%	54.76%
32	0.691	72.27%	54.76%
33	0.699	72.15%	54.76%
34	0.492	14.59%	50.93%
35	0.329	21.89%	33.90%
36	0.444	100.00%	100.00%
37	0.329	21.92%	33.90%
38	0.528	62.25%	66.19%
39	0.691	73.65%	54.76%
40	0.59	73.18%	54.76%
41	0.612	72.02%	54.76%
42	0.693	3.74%	54.76%
43	0.663	72.99%	57.85%
44	0.661	19.45%	42.12%
45	0.693	5.05%	33.90%

# ALEF Paper Scores



# Final Paper Scores



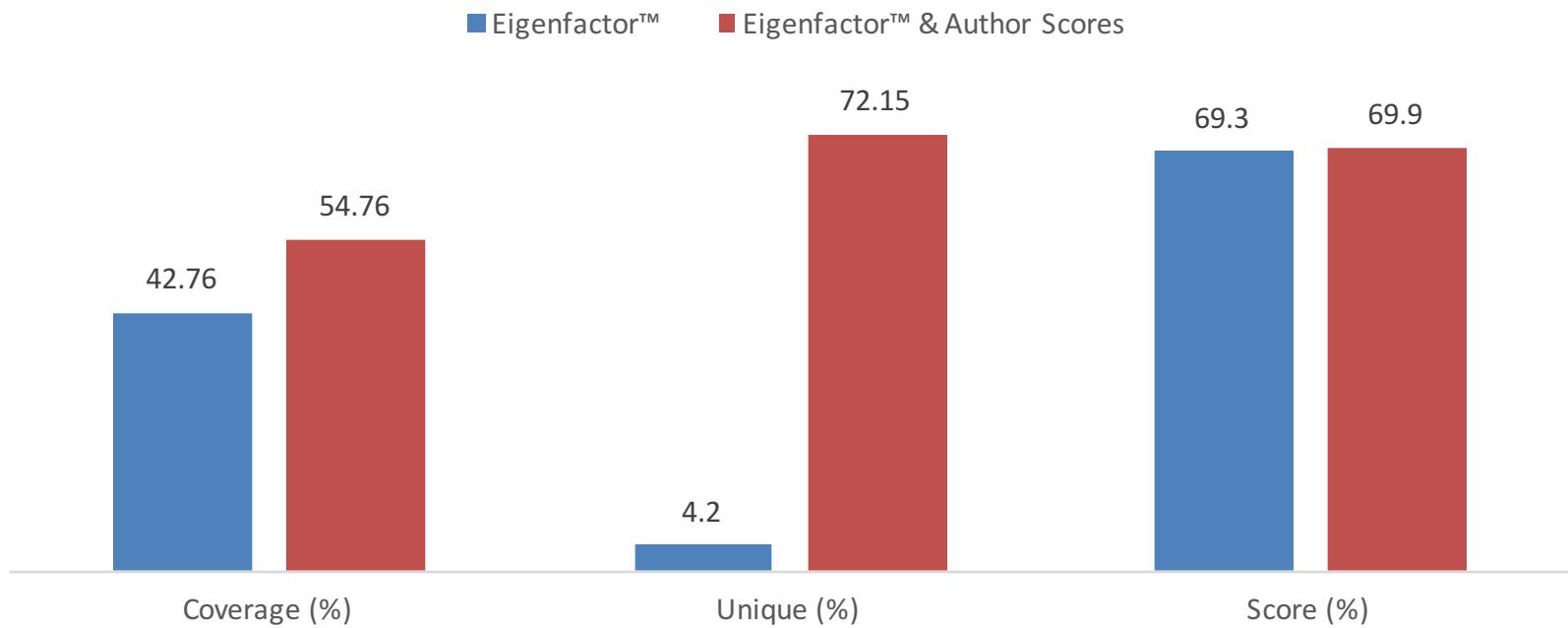
# Phase I – Evaluation Results

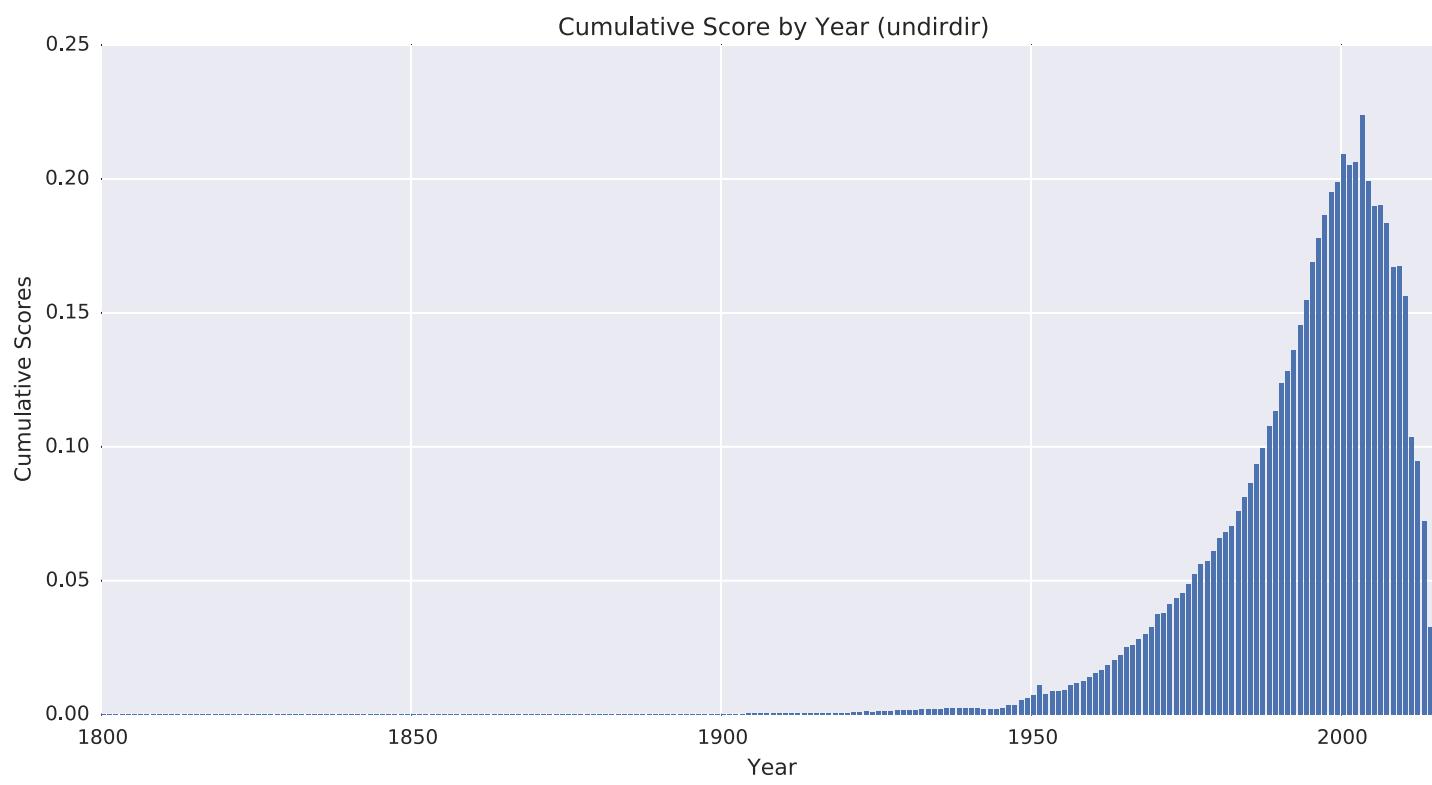
- 0.699
- 15<sup>th</sup>

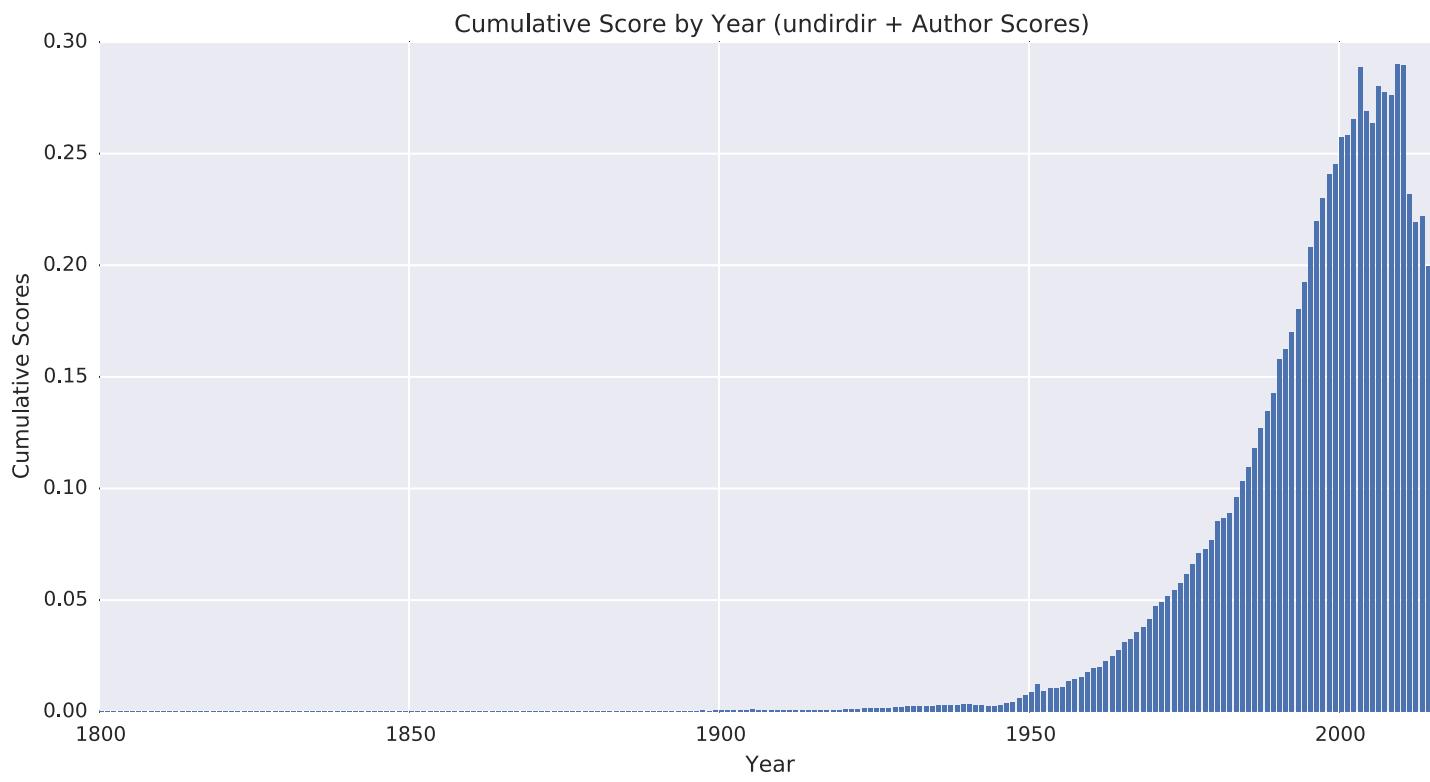
# Phase I – Test Results

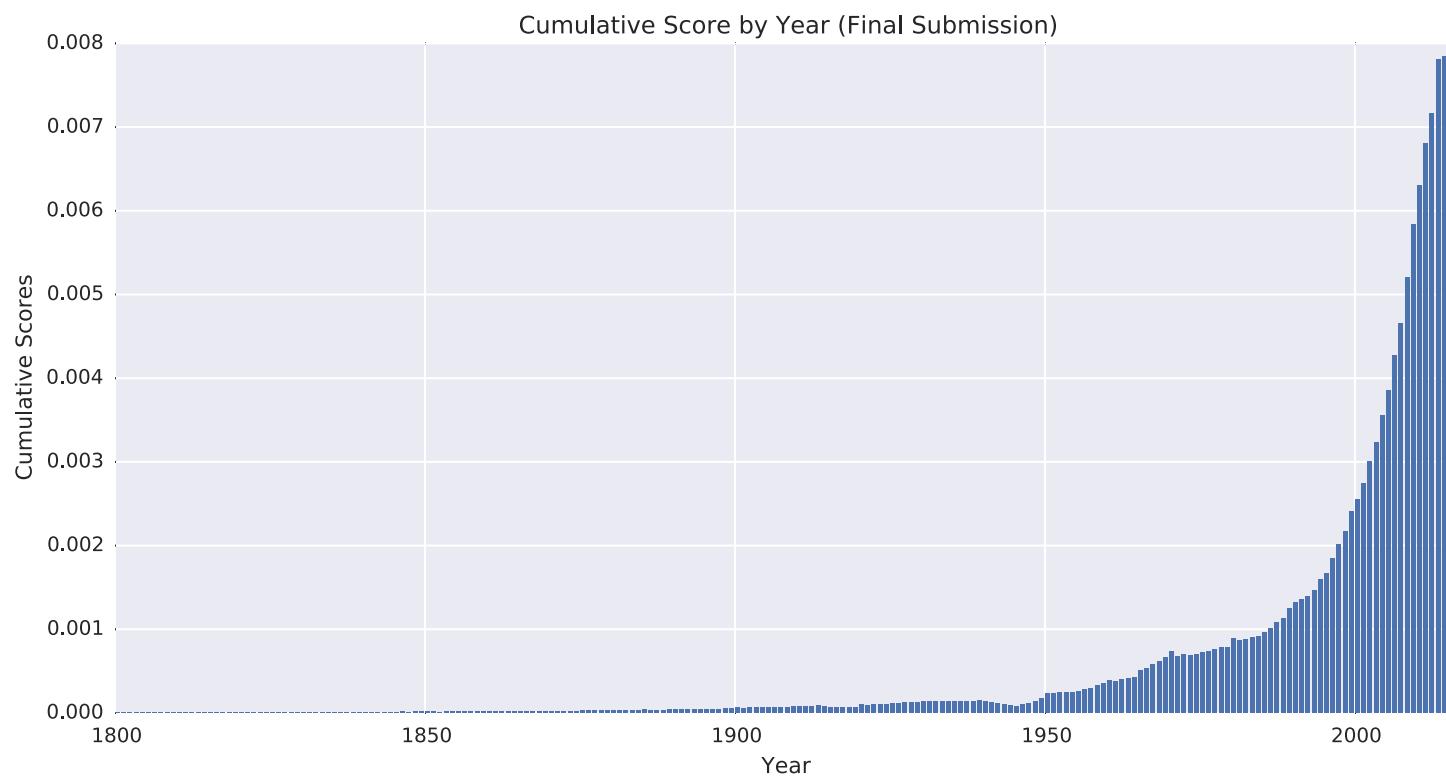
- 0.699 -> 0.676 (-3.3%)
- 15<sup>th</sup> -> 2<sup>nd</sup>

# Eigenfactor™ & Author Scores









# Logistics

- Phase II
  - Verticies 49,870,036
  - Edges 949,577,946
- Calculate Citation Scores: 34 minutes
- Build Paper-Author Matrix: ~2 hours
- Calculate Author Scores: 2 minutes
- Author Score Feature: 5 minutes
- Blending: 30 seconds

# ALEF Summary

- Simple, fast variant of PageRank for article-level citation networks
- Ranks and maps
- More experiments and modifications
- Data cleaning issues
- Thanks to Microsoft Academic Graph and WSDM Cup Challenge

# Acknowledgements

Carl Bergstrom, Department of Biology, University of Washington

Martin Rosvall, Department of Physics, Umea University

Daril Vilhena, Department of Biology, University of Washington

Aditya Gandhi, Information School, University of Washington



Metaknowledge Network,  
Templeton Foundation



# Resources

- Info, Data, Code - <http://www.eigenfactor.org/>
- Babel - <http://babel.eigenfactor.org/>
- J.D.West, M. Rosvall, C.T. Bergstrom (2016) Ranking and mapping article-level citation networks, *in prep*
- J.D.West, I.Wesley-Smith, C.T. Bergstrom (2016) [A recommendation system based on hierarchical clustering of an article-level citation network](#) *IEEE Transactions on Big Data*
- I.Wesley-Smith, C.T. Bergstrom, J.D.West (2016) [Static Ranking of Scholarly Papers using Article-Level Eigenfactor \(ALEF\)](#), WSDM Conference: Entity Ranking Challenge Workshop
- I.Wesley-Smith, J.D.West (2016) [Babel: A platform for research in scholarly article recommendation](#). WWW Conference, Workshop on Big Scholarly Data
- Jevin West - <http://www.jevinwest.org/>
- Ian Wesley-Smith – <http://iwsmitn.in/>