Understanding Hierarchies in Computer Science Conferences

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Motivation

- Different conferences are rated differently
- We attempt to understand the structure of the community associated with conferences at different tiers of rating
- We find various trends of
 - Similarity between venues/conferences
 - Measure of research carried out
 - Measure of people getting into research
- These metrics are evaluated year on year as well as tier wise

The data

	ACM	DBLP	CiteSeerX	Google Scholar	Web of Science
Free	Partly	Yes	Yes	Yes	No
Downloadable	No	Yes	Yes	No	Yes
Citation info	Yes	Few records	Yes	Yes	Yes
# records	1.59 mil	1.46 mil	32.23 mil	NA	45.68 mil

The data

<u>Initial hurdles</u> – 5 datasets but still couldn't get the data we were looking for

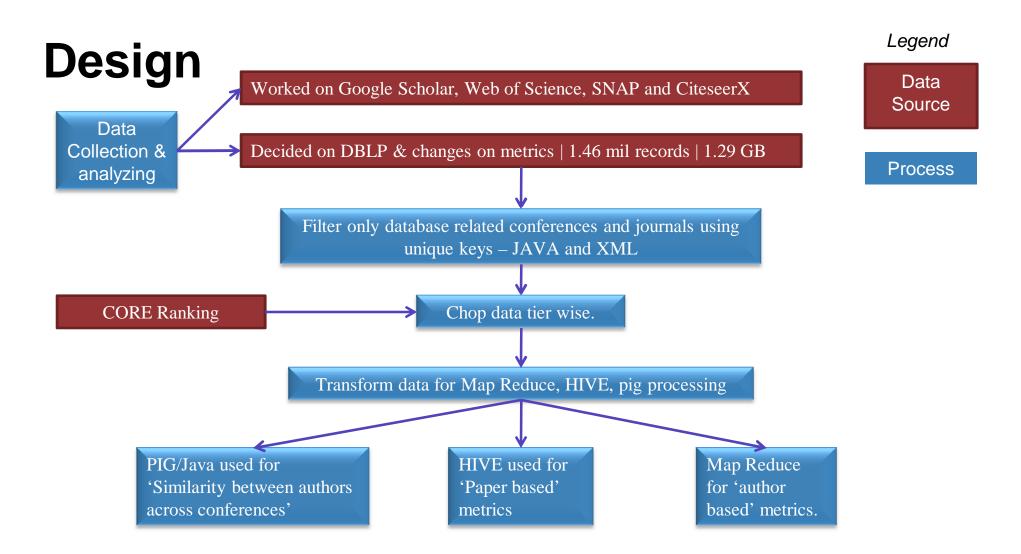
- •SNAP database: very small dataset, for Physic papers not in interest set.
- •Google Scholar: Used *Scrapy* to start building a web scraper. Involved sharp learning curve, limit on downloads. Required restarts, cleaning and manipulation.
- •DBLP: 1.3 Gb of clean annotated metadata for papers published in Computer Science. Missing citation and keyword information.
- •CiteseerX from the website : <xml> data, no citation information present.

The data

<u>Initial hurdles</u> – 5 datasets but still couldn't get the data we were looking for

<u>CiteseerX directly from Penn State:</u>

- •Over 40 Gb of data in XML and over 60 Gb MYSQL dump stored in Amazon S3.
- •Stored on HPC.
- •No mySQL on HPC.
- •Tried parsing using Perl, Python and Bash scripts one line at a time.
- •Data not indexed.
- •Cleaning and clustering required due to data being automatically scraped by a crawler.

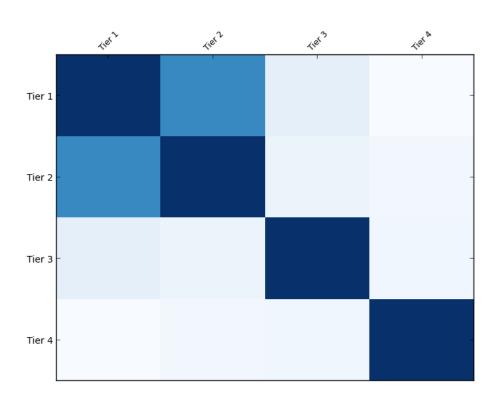


Design

Tiers

- *CORE* Computing Research & Education computer science rankings
- We decided to split the data into 4 different tiers as per ranking of the conference/journal.
- This gave us a better picture about the kind/measure of research at what level.
 - Tier 1 flagship conferences
 - Tier 2 excellent conferences
 - Tier 3 good conferences
 - Tier 4 other honorable conferences

Similarity between authors across conferences



Focused on the Database community in DBLP

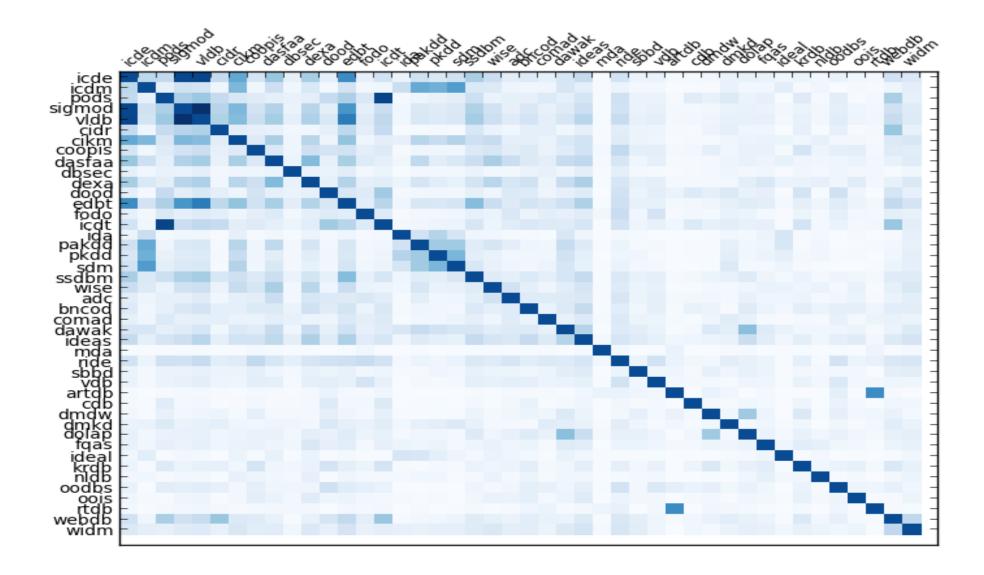
Similarity measured using Jaccard distance on the author sets

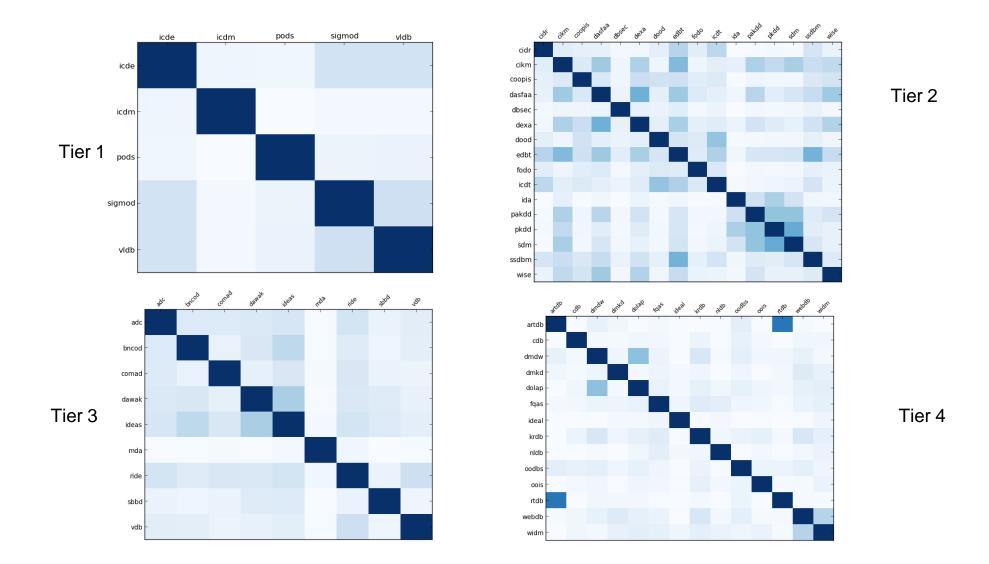
#1 Hive partitioning by conference.

2 Hive streaming with Python map reduce functions.

#3 Pig followed by Java

Jaccard distance = $A \cap B/A \cup B$





Hadoop MR

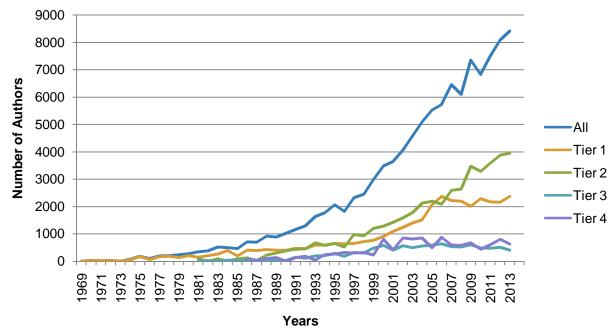
- Selected <xml> data was transformed and flattened to get each record into a single line for easier implementation of MR.
- E.g.

Inproceedings author:Roberto Brunelli|Ornella Mich title:Efficient Image Retrieval by Examples. year:2000 pages:145-162 crossref:conf/vdb/2000 booktitle:VDB

url:db/conf/vdb/vdb2000.html#BrunelliM00

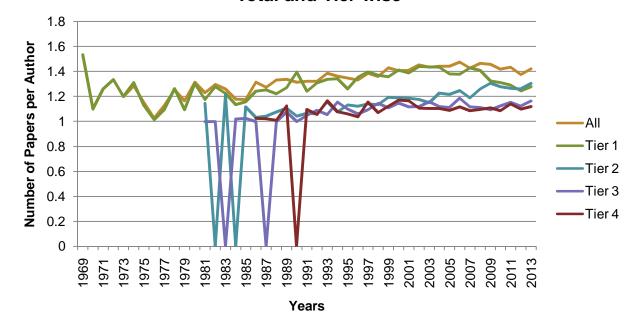
- •More people getting into research
- •Healthy increase of top tier conferences instead of lower tier

Number of Authors over the years - Total and Tier wise



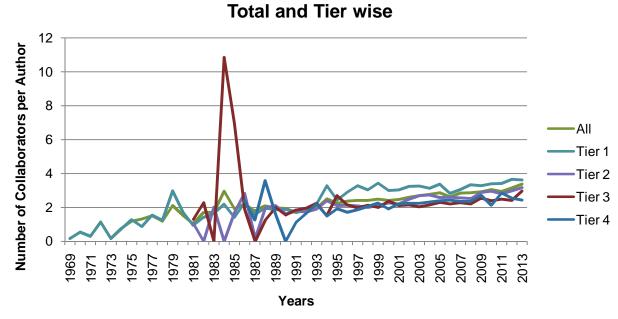
- •Steady amount of research per author; good as number of authors increasing
- •Top tier conferences performing better than lower tier
- •Lotka's Law of 60%

Average number of papers per author over the years Total and Tier wise



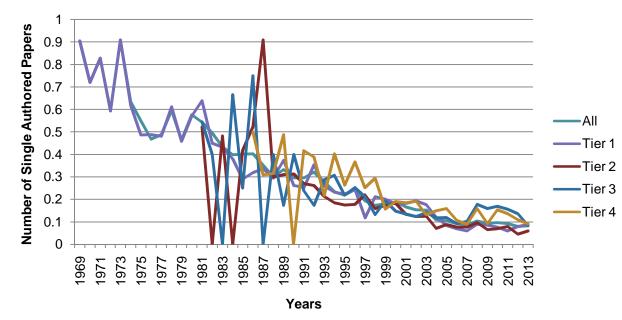
- •Steady number of collaborators
- •Many conferences also have limit on maximum number of collaborators

Average number of collaborators per author over the years



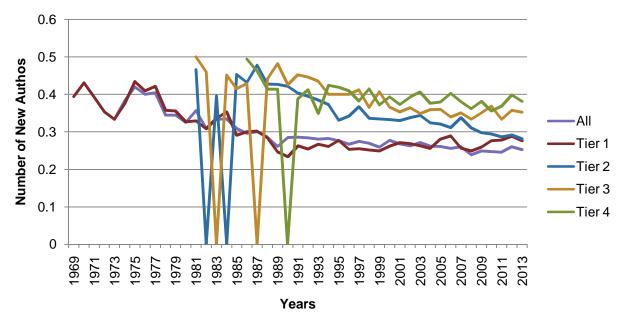
- •Number of single authored papers reducing over time. Better networking & collaboration opportunity.
- •Good as collaborators allow different perspective and usually allow deeper research

Number of single authored papers over the years Total and Tier wise - normalized



- •New people coming into research more or less stable
- •Higher in lower ranked conferences. Makes sense since usually people will enter with lower ranked conferences

Number of new authors over the years Total and Tier wise - normalized



Paper based statistics

Hive:

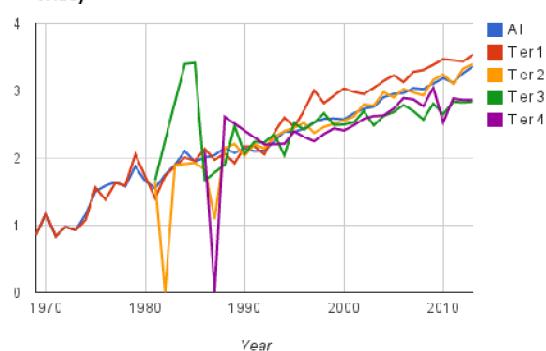
- As only certain fields were required for Hive analytics, the files were transformed into a tab delimited file having the following format key year conference authors-array
- Then these were put in the tables through Hive
- The configured file was also split according to the Tiers and the analytics were also found on the tier files

Paper based statistics

Avg Aurhors per Paper

- •Steady increase shows increasing trend of collaboration amongst authors.
- •Publish or Perish Theory

Average authors per Paper (Cumulative and Tier Wise)

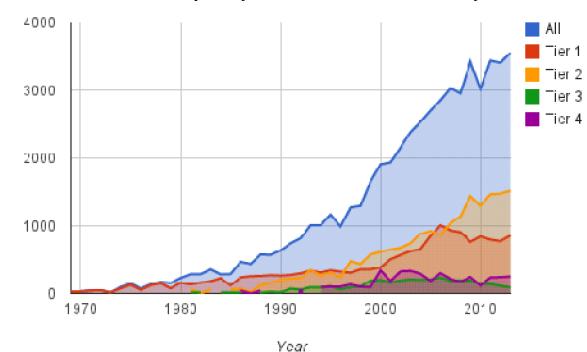


Paper based statistics

Number of Papers

- •Healthy and steady increase in number of papers published
- •More papers published in top tier conferences.

Number of Papers (Cumulative and Tier Wise)



Future Work

- Our data-set and associated techniques can also be used for further predictions like
 - What makes a paper acceptable
 - What gets a paper cited more often than others
 - Do people who get published, work in groups or alone?
 - What are the likely venues to publish given the authors one has worked with
 - Keyword analysis to identify what gets a paper cited more often
 - Structure of collaboration network/degrees of separation

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Special Thanks

- Professor McIntosh for her continuous support in providing guidance and data options.
- Penn State University