Metadata Analytics: Workflows and Applications

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- The (very brief) history
- The motivation
- The perspective

- Theories
- Methodologies
- Applications

What is (big) metadata?

An earlier definition

"...the structured, semi-structured, or unstructured descriptions of scientific data stored in repositories" (Bratt et al., 2017)

An updated brief version

The structured or semistructured descriptions of information and/or data objects.

An updated long version

The structured or semi-structured descriptions of information and/or data objects in the forms of library catalogs, indexing databases, and metadata repositories.

The (very brief) history



Price, Derek J. de Solla, 1922-1983.

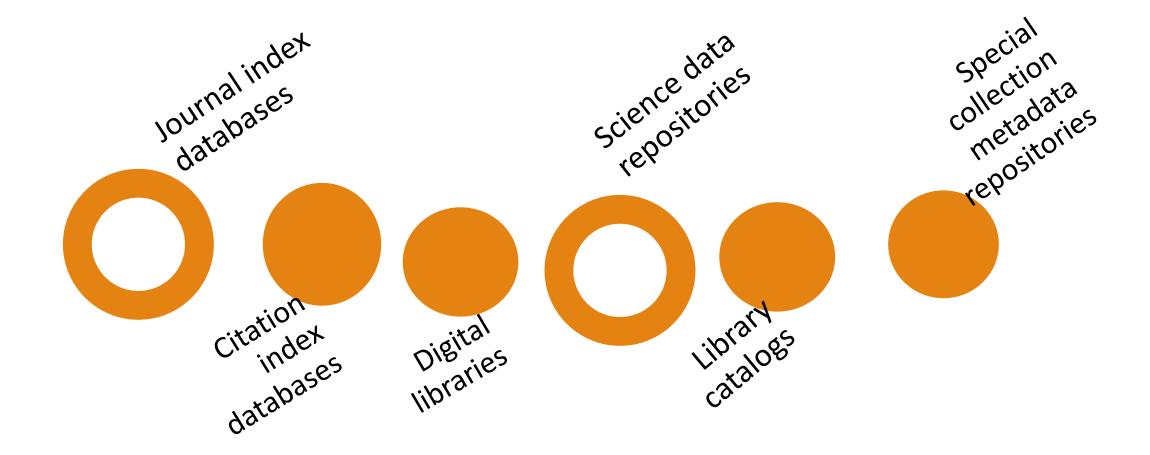


Eugene Eli Garfield, 1925-2017

- Price's model: Preferential attachment process
- Power law distribution of citation network, first example of scale-free network
- Price's law: square root law for relationships between authors and publications
- Exponential growth of science and half-life of science literature
- Science citation index (which inspired the PageRank algorithm by Google co-founders)
- Journal impact factor

Bibliometrics, Scientometrics:
Theories (laws) built on math
Quantitative methods
Macro- and micro-scale
Authors, publications, citations

The changing landscape of metadata...



Old issues, great challenges

- The perpetual ambiguous author names
- ∇ Difficult to reuse code and workflows
 →
 ∇ Reinventing the wheel

Name disambiguation

Constant data cleaning and processing

Very large volume of data and very complex structures require careful planning for metadata analytics to avoid reinventing the wheel and/or waste of time and efforts.

Workflows are a method for ensuring effectiveness and quality of metadata analytics.

What is a workflow?

- What is a workflow?
- "the activity of defining the sequence of tasks needed to manage a business or computational science or engineering process" with four broad aspects:
 - composition,
 - mapping,
 - execution, and
 - Provenance." (Deelman et al., 2009, p. 529).

An example of conceptual workflow in metadata analytics: Name disambiguation solutions (1)

Goal

- 1) improve accuracy of name-centric retrieval of information from GenBank
- 2) improve accuracy of data integration between GenBank and other sources

Documentation from Evernote

Task

1) Resolve each Author referenced in Genbank to a unique identifier (resolution).

In Genbank, authors are referenced by first initial and last name, giving rise to the following forms of ambiguity:

- A) Multiple authors with the same last name and first initial (polysemy). Example: multiple authors named 'Smith, J.'
- B) A single author with multiple name variants (synonymy). This can occur due to a name change, spelling variation (Anglicization of foreign names), or misspelling. Example: a single author referred to as 'Adams, E.' in one record and 'Adams-Hoffert, E.' in another.

An example of conceptual workflow in metadata analytics: Name disambiguation solutions (2)

2) Enhance metadata associated with each author referenced in GenBank (attribution).

To improve the ability to resolve a given GenBank author to an author referenced in another source, additional metadata will be associated with each uniquely identified author. This metadata could include:

- A) Full name
- B) Name variants
- C) Organizational affiliations
- D) Co-author affiliations
- E) Subject matter expertise
- F) etc.

Scope

The name disambiguation application will:

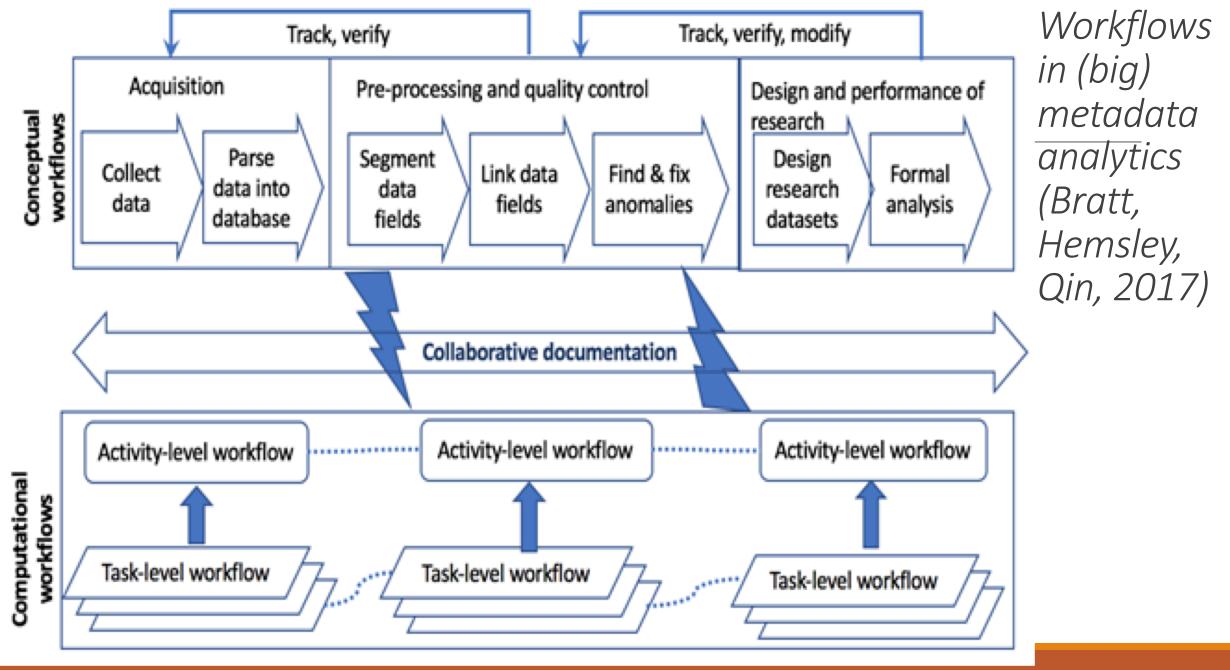
- 1) need to occasionally re-analyze the GenBank database when a
- significant amount of new information has been incorporated
- 2) need to handle updates to external resources such as author information in Pubmed
- 3) run as an offline (non-realtime) process

An example of conceptual workflow in metadata analytics: Name disambiguation solutions (3)

Steps

Resolution:

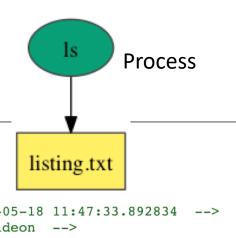
- 1) Get access to data and set up dev / test environment
- 2) Develop ground-truth / test data for assessing accuracy
- Develop algorithm for assigning similarity score between each pair of author refs in GenBank database based on available metadata
- 5) Determine similarity threshold for considering two refs as same individual
- 4) Apply clustering method to group all refs
- Measure accuracy and modify similarity / clustering algorithms as necessary



Why do we need workflows in metadata analytics?

- PAlign your data collection, processing, and analysis with your research goals
 - make sure the data you collected are you needed for answering your research question
- Sestablish provenance for your research project to assure the reproducibility and replicability of your research

Tools for (computational) workflow management





https://pegasus.isi.edu/

```
Gene Accession Number
                                           XML Entry Display
     AA045112
                                                            Sequence Getter Using XPath Sequence Display
                                                 Entry of Gene
                                                            HTML Generator Using XSLT
                                                                                      HTML Display
                                   rrors Sink
K .06-UsingWebServicesAndD...
File Tools Help
                                                          K .06-UsingWebServicesAndD...
(<SEQUENCE>cacctggagaaacttctgcactggcad
                                                           File Tools Help
              K .06-UsingWebServicesAndD...
              File Tools Help
                               K .06-UsingWebServicesAndD...
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<br/>
h>Program: </b><input
               CBASE COUNT A=
                                               <br/>
<br/>b>Database: </b><inpu
               SEQUENCE>cacc
                                               <b>Query: </b><input r
              </DDBJXML>
                                               <INPUT TYPE=\"submit\"
```

</form> </body></html>

Kepler Workflow Management System

https://kepler-project.org/

Now we have conceptual and computational workflows, what comes next?

Linked data

- Available on the web
- Available as structured data readable by machine
- Available in a non-proprietary format
- Expressed using open World Wide Web Consortium (W3C) standards
- PLinked to other data on the web

What implications are there to the fields of bibliometrics and scientometrics?

Name disambiguation solutions

- Traditional solution: use algorithms to automatically disambiguating author names
- However, if database producers keep current practice in abbreviating names, the problem will remain unresolved.
- New solution: creating globally unique ID for researchers and authors







LC Linked Data Service
Authorities and Vocabularies

Virtual International Authority File

Union List of Artist Names®
Getty Research Institute

Identifying things with globally unique identifiers

- **Subject terms** in controlled vocabularies
- **Events** (political, cultural, public health, social, ...)
- Publications (papers, versions of a paper, journals, ...)
- Patasets (research data, census data, observation data, sensor data...)
- Cultural objects (archives, museum objects, digital surrogates of physical objects...)

Metadata is changing to broaden the research horizon

Bibliometrics and scientometrics Knowledge discovery for humanities and social sciences Data services to support interdisciplinary large-scale research

Uniquely identified authors, organizations, taxonomic classes, subject terms, datasets, publications, etc. in structured, linkable formats

Index databases; library catalogs; metadata repositories for datasets; digital libraries for scholarly pubs, special collections, and cultural objects

Big metadata analytics

Semantic infrastructure

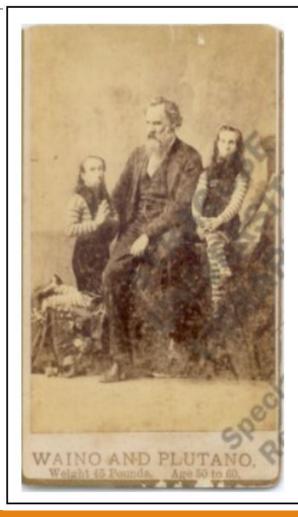
Data infrastructure

Case: Mining large (meta)datasets for the humanities (1)

date_issued	description	coverage	series	
	<p>"Perestroika, glasnost: they don't even talk about them in Castro's Cuba. Violent</p>	t		
	insurgencies? The Cubans support the Soviets say they prefer political solutions. What			
	are Gorbachev and Castro really talking Toy to the control of the	Text mining in metadata to		
	hopeful? Joining us live from Havana	lauala li	J	
	spokesman for the Soviet foreign mir			
1989-04-03	discover trends, p	atterns	and	
	<p>"Imagine a space con</p>	accerris,	ullu	
	Pollard gave the Israelis. What Pollard there to haunt US-Israeli relations phenomena for h	umanitie	25	
	Jonathan Pollard's father and the aut			
1989-04-04	Israel." Includes commercials.8 and social science	sc cchola	rc	
	allu social science	es scilola	13	
	<p>"It's been a horror-st</p>			
	Massachusetts where the criminally insane and those committed for civil reasons are held			
1989-04-05	together." Includes commercials.	Massachusetts	Nightline	
	<p>" that's the cost [\$100-200 million] of cleaning-up [the Exxon Valdez oil spill],</p>			
	but are Americans willing to pay the much higher price of not having it happen again?"			
1989-04-06	Includes commercials.	Alaska	Nightline	

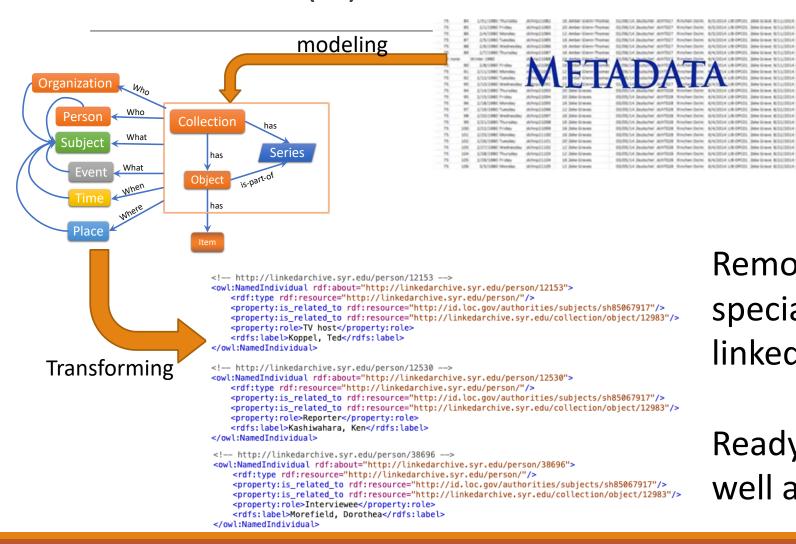
Case: Mining large (meta)datasets for the humanities (2)

Text analysis for metadata description led to new faceted approach in representing historical visual materials



Individual depicted (LCNAF)	Wano, 1825-1905	
Individual depicted (LCNAF)	Plutano, 1827-1912	
Groups depicted (LCSH)	Circus performers	
Groups depicted (LCSH)	Men	
Groups depicted (21st century		
terminology)	Little people	
Groups depicted (21st century		
terminology)	Dwarfs	
Groups depicted (19th century		
terminology)	Midgets	
Conditions depicted (MeSH)	Dwarfism	
Conditions depicted		
(SNOMED CT)	Short stature disorder	
Relations depicted (LCSH)	Brothers	
Nationalities depicted (LCSH)	Americans	
Places referenced (19th century		
terminology)	Borneo	
Time period depicted (LCSH)	Nineteenth century	

Case: Mining large (meta)datasets for the humanities (3)



Remodeling and transforming special collection metadata to linked archives

Ready for publishing and sharing as well as consuming by machines

Archival

collections

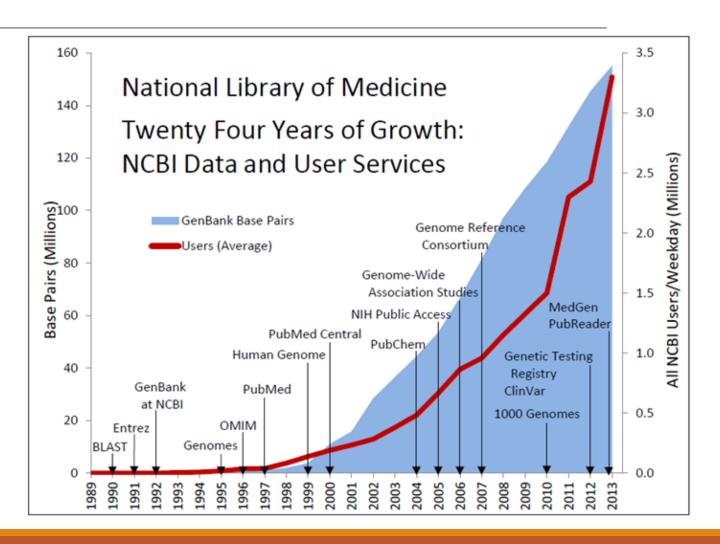
Case: GenBank Metadata mining (1)

"From 1982 to the present, the number of bases in GenBank has doubled approximately every 18 months."

-- NCBI. (2017). Growth of GenBank and WGS, http://www.ncbi.nlm.nih.gov/genbank/statistics.

Image credit:

https://www.nlm.nih.gov/about/2015CJ.html



Case: GenBank Metadata mining (2)

GenBank's big metadata as a source for quantitative studies of team science

```
LOCUS
                         5028 bp
                                                               21-JUN-1999
            SCU49845
                                    DNA
                                                    PLN
            Saccharomyces cerevisiae TCP1-beta gene, partial cds, and Ax12p
DEFINITION
            (AALZ) and Revip (REVI) genes, complete cds.
ACCESSION
            U49845
VERSION
            U49845.1 GI:1293613
KEYWORDS
SOURCE
            Saccharomyces cerevisiae (baker's yeast)
            Saccharomyces cerevisiae
  ORGANISM
            Eukaryota; Fungi; Ascomycota; Saccharomycotina; Saccharomycetes;
            Saccharomycetales; Saccharomycetaceae; Saccharomyces.
REFERENCE
            1 (bases 1 to 5028)
            Torpey, L.E., Gibbs, P.E., Nelson, J. and Lawrence, C.W.
  AUTHORS
            Cloning and sequence of REV7, a gene whose function is required for
  TITLE
            DNA damage-induced mutagenesis in Saccharomyces cerevisiae
            Yeast 10 (11), 1503-1509 (1994)
  JOURNAL
 PUBMED
            7871890
REFERENCE
            z (pases 1 to 5028)
            Roemer, T., Madden, K., Chang, J. and Snyder, M.
 AUTHORS
            Selection of axial growth sites in yeast requires Axl2p, a novel
  TITLE
            plasma membrane glycoprotein
            Genes Dev. 10 (7), 777-793 (1996)
  JOURNAL
  PUBMED
            8846915
REFERENCE
            3 (bases 1 to 5028)
            Roemer, T.
 AUTHORS
 ጥፐጥፒው
            Direct Submission
            Submitted (22-FEB-1996) Terry Roemer, Biology, Yale University, New
 JOURNAL
            Haven, CT, USA
FEATURES
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                     /db xref="taxon:4932"
                     /chromosome="IX"
```

/map="9"

Case: GenBank Metadata mining (3)

Collaboration across countries, labs, and fields

- P Big problems, big data (and big metadata), and big teams
- Relations between data production and paper publication
- ∇ Large scale studies of collaboration networks to find patterns, structures, and empirical evidence for in-depth exploration

```
SOURCE
            Bacillus subtilis subsp. subtilis str. 168
            Bacillus subtilis subsp. subtilis str. 168
  ORGANISM
            Bacteria; Firmicutes; Bacilli; Bacillales; Bacillaceae; Bacillus.
REFERENCE
            Kunst, F., Ogasawara, N., Moszer, I., Albertini, A.M., Alloni, G.,
  AUTHORS
            Azevedo, V., Bertero, M.G., Bessieres, P., Bolotin, A., Borchert, S.,
            Borriss, R., Boursier, L., Brans, A., Braun, M., Brignell, S.C.,
            Bron, S., Brouillet, S., Bruschi, C.V., Caldwell, B., Capuano, V.,
            Carter, N.M., Choi, S.K., Codani, J.J., Connerton, I.F., Cummings, N.J.,
            Daniel, R.A., Denizot, F., Devine, K.M., Dusterhoft, A., Ehrlich, S.D.,
            Emmerson, P.T., Entian, K.D., Errington, J., Fabret, C., Ferrari, E.,
            Foulger, D., Fritz, C., Fujita, M., Fujita, Y., Fuma, S., Galizzi, A.,
            Galleron, N., Ghim, S.Y., Glaser, P., Goffeau, A., Golightly, E.J.,
            Grandi, G., Guiseppi, G., Guy, B.J., Haga, K., Haiech, J., Harwood, C.R.,
```

The complete genome sequence of the Gram-positive bacterium Bacillus subtilis in Harring and Complete genome sequence of the Gram-positive bacterium Bacillus subtilis

```
F. Kunst → N. Ogasawara ← [...] A. Danchin

**No., M., M., C., P.,

**Nature 390, 249–256 (20 November 1997)

**doi:10.1038/36786

**Download Citation**

**Download Citation**

**No., M., M., C., P., C., P
```

TITLE

JOURNAL

Direct Submission

France

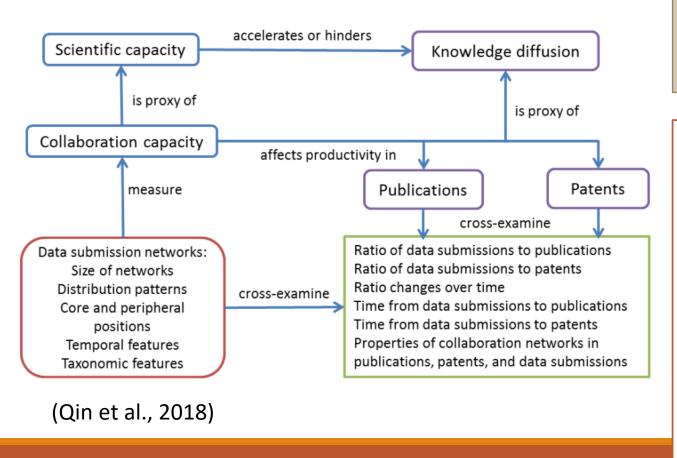
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Tosato, V., Uchiyama, S., Vandenbol, M., Vannier, F., Vassarotti, A.,
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            Winters, P., Wipat, A., Yamamoto, H., Yamane, K., Yasumoto, K., Yata, K.,
            Yoshida, K., Yoshikawa, H.F., Zumstein, E., Yoshikawa, H. and
            Danchin, A.
            The complete genome sequence of the gram-positive bacterium
  TITLE
            Bacillus subtilis
  JOURNAL
            Nature 390 (6657), 249-256 (1997)
   PUBMED
            9384377
            2 (bases 1 to 14210)
REFERENCE
  AUTHORS
            Glaser, P.
```

Submitted (25-JUN-1997) Philippe Glaser, Regulation de l'Expression

Genetique, Institut Pasteur, 28 Rue du Dr Roux, Paris, 75724,

Case: GenBank Metadata mining (4)

The collaboration capacity framework



Collaboration capacity: the ability of an individual researcher or a team of researchers to collaborate throughout the data production and publication lifecycle and sustain a network of collaborators over time.

Assumptions:

- Collaboration capacity is a proxy for studying scientific capacity
- Data, publication, and patent together can be used as a proxy for studying knowledge diffusion
- Collaboration capacity significantly affects the level of research productivity and extent of knowledge diffusion

Case: GenBank Metadata mining (5): Methods

Source: metadata describing molecular sequences in GenBank

- ♥ Exploratory data analysis (EDA)



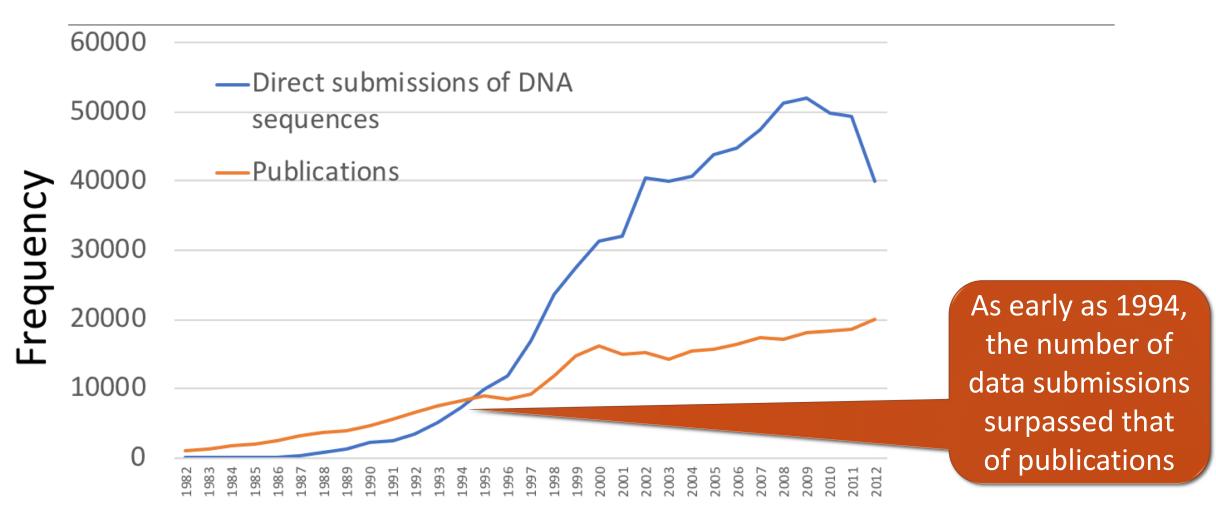
Purpose: using descriptive stats and visualization techniques to look for patterns, structures, and problems

- - Size of collaboration networks for data submission
 - Extent of knowledge diffusion
 - Rate of knowledge diffusion

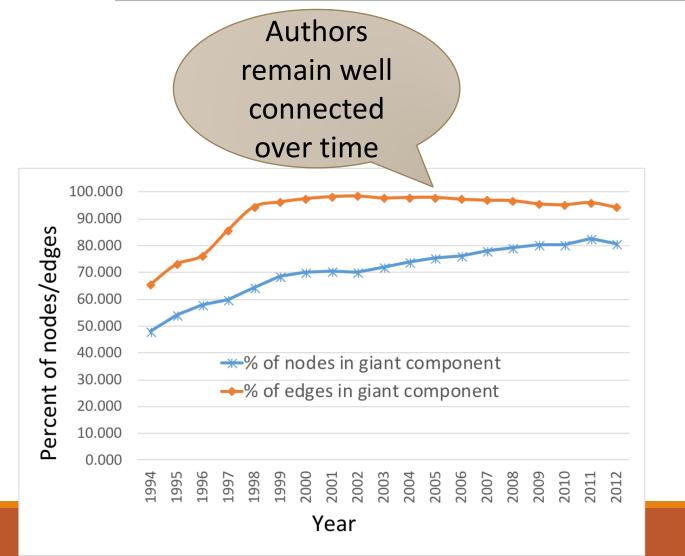
Case: GenBank Metadata mining (6): Findings

- **?** Connectedness of collaboration networks
- Ratio of data submissions to publications

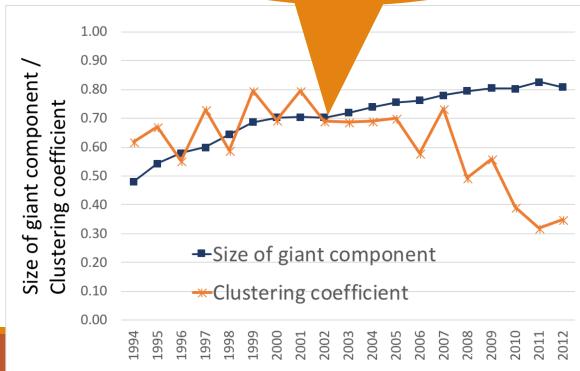
Case: GenBank Metadata mining (7): Data submissions vs. publications



Case: GenBank Metadata mining (8): Connectedness vs. distributedness



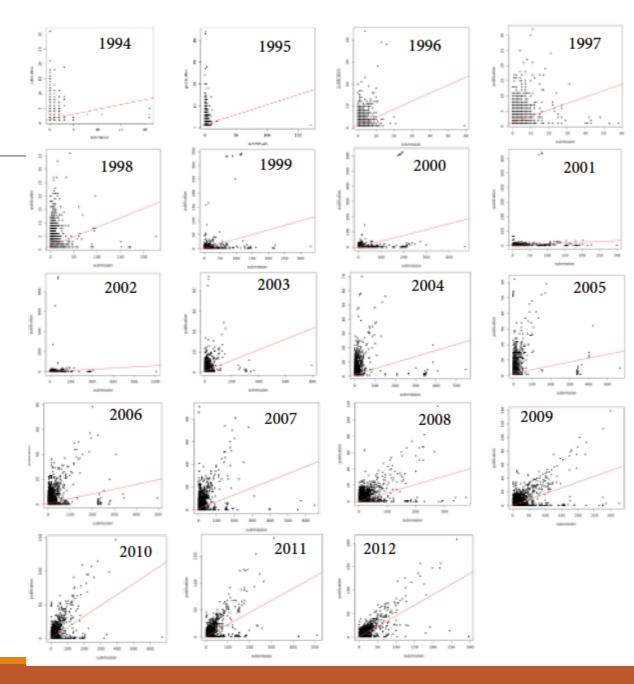
While more clusters of smaller communities emerged



Case: GenBank Metadata mining (9)

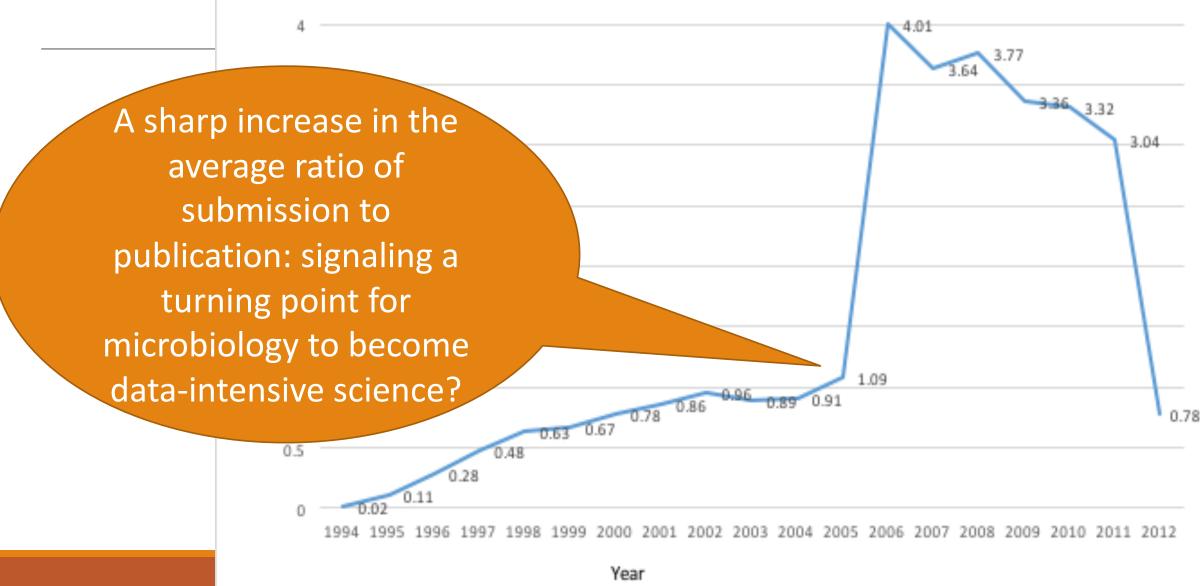
Ratio of submissions to publications

- x axis: # of authors who submitted sequence data
- y axis: # of authors who published a paper associated with the data submissions
- After 1998, more authors were involved in data production than those in paper publications
- Significant increment in productivity:
 - Before 1998, majority had a range between 20 publications and 50 data submissions
 - Since 2008, a sizable # of authors had a high productivity in the range of 50~100 publications and 100~300 data submissions



Average ratio of submission to publication

Case: GenBank Metadata mining (10)



Conclusion

- ⟨Big⟩ metadata analytics uses metadata as the data source to:
 - Study phenomena, trends, behaviors, and relations
 - Produce semantically precise, linked data for better discovery, access, and management of information resources and datasets
- PAs an emerging research field, it faces great challenges in
 - Methodologies: workflows, tools, and practices that reduce reinventing the wheel and enhance research reproducibility
 - Data: scattered, in different formats, messy, and over 80% of time spent in getting data ready for analysis

Thank you!

Questions?



References

- Paratt, S., Hemsley, J., Qin, J. & Costa, M. (2017), Big data, big metadata and quantitative study of science: A workflow model for big scientometrics. *Proc. Assoc. Info. Sci. Tech.*, 54: 36−45. doi:10.1002/pra2.2017.14505401005
- Qin, J., J. Hemsley, & S. Bratt. (2018). Collaboration capacity: Measuring the impact of cyberinfrastructure-enabled collaboration networks. Science of Team Science (SCITS) 2018 Conference, Galveston, Texas, May 21-24, 2018.