

# Next Generation Machine Learning: The Evolution of the Library as Research Partner, Project Catalyst and Digital Integrator



Xuemao Wang  
Vice Provost for Digital Scholarship  
Dean and University Librarian

James Lee  
Associate Vice Provost for Digital Scholarship  
Associate Dean of Libraries  
Director, Digital Scholarship Center

Kristen Burgess  
Operational Manager  
Digital Scholarship Center

# University of Cincinnati

Founded in  
**1819**

Enrollment

**46,798**

Degree Programs

**414**

Student:Teacher Ratio

**16:1**

## University of Cincinnati Quick Facts

**Location:** Cincinnati, Ohio

**Number of Buildings:** 118 facilities on 476 acres

**Majors & Programs:** 414 degree programs, 262 minors and certificates

**Athletics:** NCAA Division I; American Athletic Conference

**Mascot:** Bearcat

**Colors:** Red and Black

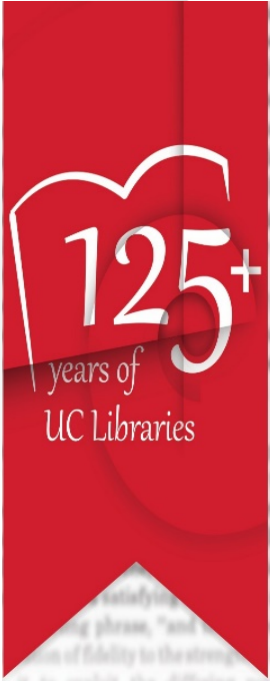
**Famous Alumni & Faculty:** Astronaut Neil Armstrong; President and later Chief Justice William Howard Taft; Eula Bingham, environmental scientist and one-time head of OSHA; Albert Sabin, developer of the oral polio vaccine; and prima ballerina Suzanne Farrell

**Students from:** 50 states and 114 countries

**Living Alumni:** over 300,000 with approximately half (more than 140,000 residing in the greater Cincinnati region).

Additional information available on the [UC Fact Sheet](#).

# University of Cincinnati Libraries



*Archives & Rare Books Library*



*Chem-Bio Library*



*Classics Library*



*CCM Library*



*DAAP Library*



*CECH Library*



*CEAS Library*



*Geo-Math-Phys Library*



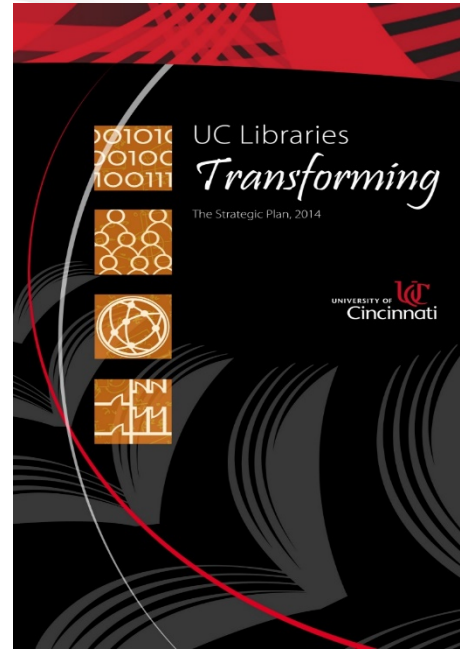
*Health Sciences Library*

\* Regional Libraries: Blue Ash, Clermont & Law

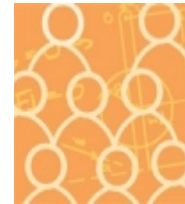
[www.libraries.uc.edu](http://www.libraries.uc.edu)

# University of Cincinnati Libraries' Strategic Direction

## Journey to Transformational Change



**DIGITAL  
TECHNOLOGIES &  
INNOVATION**



**PEOPLE**



**SPACE**



**DATA to INFORMATION  
to KNOWLEDGE**

The University of Cincinnati Libraries will become the  
***globally engaged, intellectual commons  
of the university***  
—positioning ourselves as the hub of **collaboration,  
digital innovation and scholarly endeavor** on campus.

# Our Journey Towards a Great Public Research University – Boldly *Bearcat*

**next  
lives  
here**

**UNLEASHING OUR VISION**

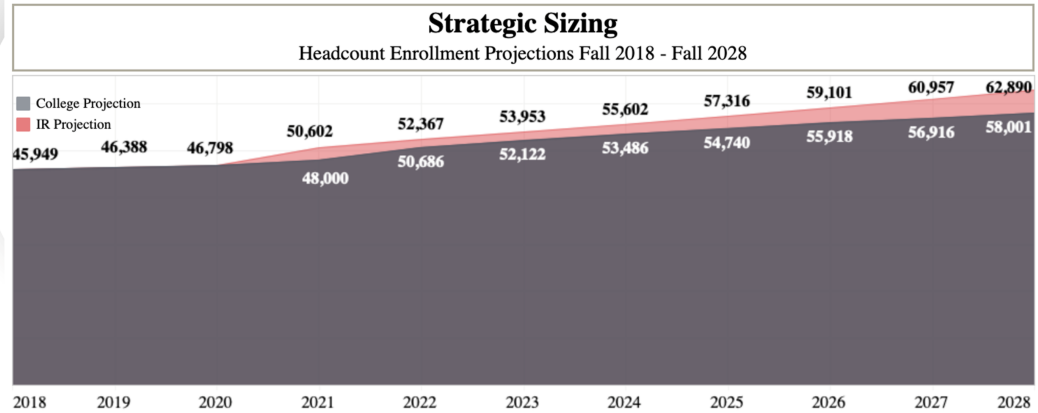
LEADING URBAN PUBLIC  
UNIVERSITIES INTO A  
NEW ERA OF INNOVATION  
AND IMPACT

# University of Cincinnati's Strategic Direction: Next Lives Here



# UC Strategic Sizing Plan

## ENROLLMENT GROWTH 2018-2028



## GROWTH IN ACCESS & QUALITY

TRENDS IN COLLEGE-LEVEL PROGRAM EFFORTS



COMPUTING  
& BIG DATA



HEALTH &  
WELLNESS



DIGITAL



INTERDISCIPLINARY



RETENTION

## GROWTH IN EDUCATION

INSTITUTION-WIDE EFFORTS TO INCREASE ACCESS & QUALITY



BEARCAT PROMISE



CO-OP 2.0



CPS STRONG



LAUNCH UC



DIRECT ADMIT



NEXT/NOW  
SCHOLARSHIPS



VETERANS



ADULT LEARNING  
INSTITUTE



UC ONLINE



PROFESSIONAL  
MASTERS



PHD



URM PROGRAMS

RECRUITMENT, RETENTION, & COMPLETION

## GROWTH IN URBAN/COMMUNITY IMPACT

TRENDS IN COLLEGE-LEVEL EFFORTS



CINCINNATI  
INNOVATION  
DISTRICT



CPS  
READINESS



SKILLING &  
CREDENTIALING



CLINICAL  
PARTNERSHIPS

# THEMATIC AREAS

## FOR STRATEGIC INVESTMENT



**DATA  
SCIENCES**



**HEALTH &  
WELLNESS**



**CREATIVITY,  
EXPLORATION &  
JUSTICE IN A DIGITAL  
WORLD**



**HUMAN-ENVIRONMENT  
INTERACTIONS**



# UC Research 2030 Plan



## RESEARCH 2030

### UC'S 10-YEAR STRATEGIC PLAN FOR RESEARCH

The University of Cincinnati is the leading R1 urban university in our region with unrivaled talent solving problems that matter. We are rigorously pursuing diversity, equity and inclusion in research and actively transforming society through the creation of game-changing new knowledge and application of disruptive discoveries.

[SEE THE PLAN](#)

OBJECTIVE

## NATIONAL PROMINENCE

GOAL

TOP 25 PUBLIC RESEARCH UNIVERSITY

GUIDING PRINCIPLE

Galvanizing our mission to serve the public good


## INVEST IN SUCCESS

 RECRUIT AND RETAIN TOP TALENT

 INNOVATE THE RESEARCH INFRASTRUCTURE

 CULTIVATE & GROW PROGRAMS OF EXCELLENCE

## INVEST TO ADVANCE

 URBAN FUTURES PATHWAY

 RETHINKING THE WHERE

 COALITION FOR CHANGE

OBJECTIVE

## IMPACTFUL RESEARCH

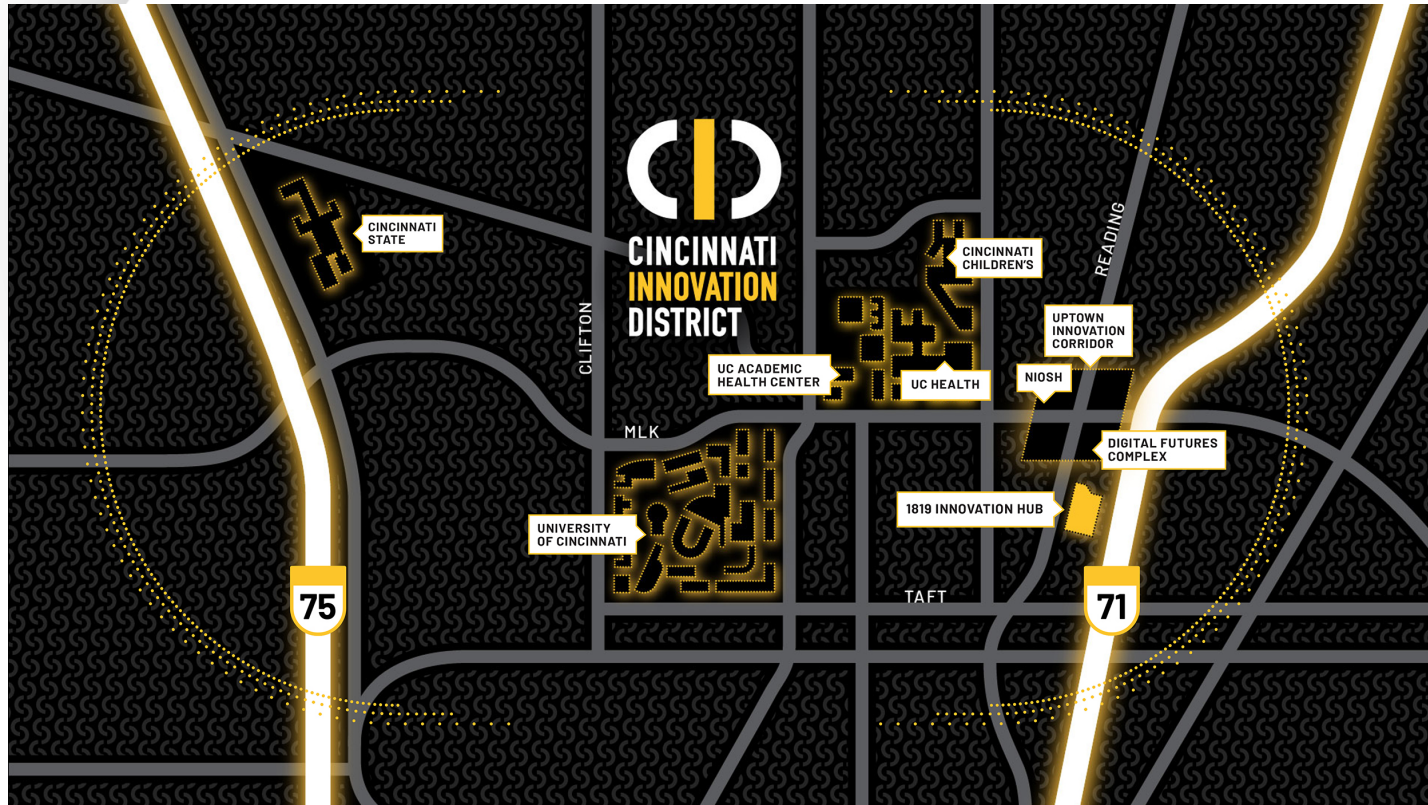
GOAL

IMPROVING PEOPLE'S LIVES

GUIDING PRINCIPLE

Foundational Partnerships to Solve Real World Problems

# UC Innovation District



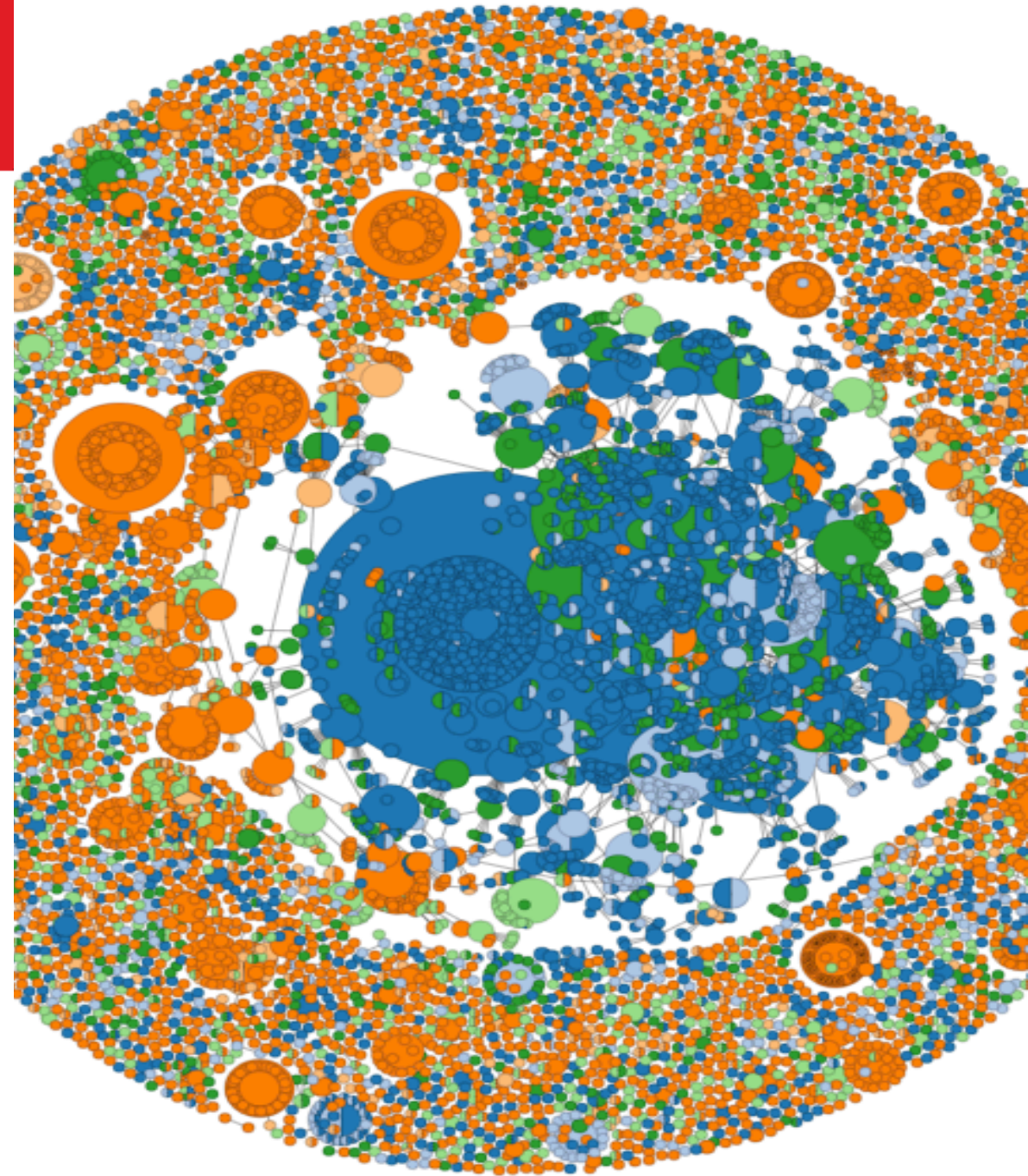
# DSC: Mission

- Core mission - To break silos and cross wires across the university. We work at the intersection of data science, the arts and humanities, and the libraries.
- Academic Center (Libraries + Arts & Sciences)
- Mellon Digital Integrator (One of six, projects with eight colleges).
- UC Digital Futures Anchor Team

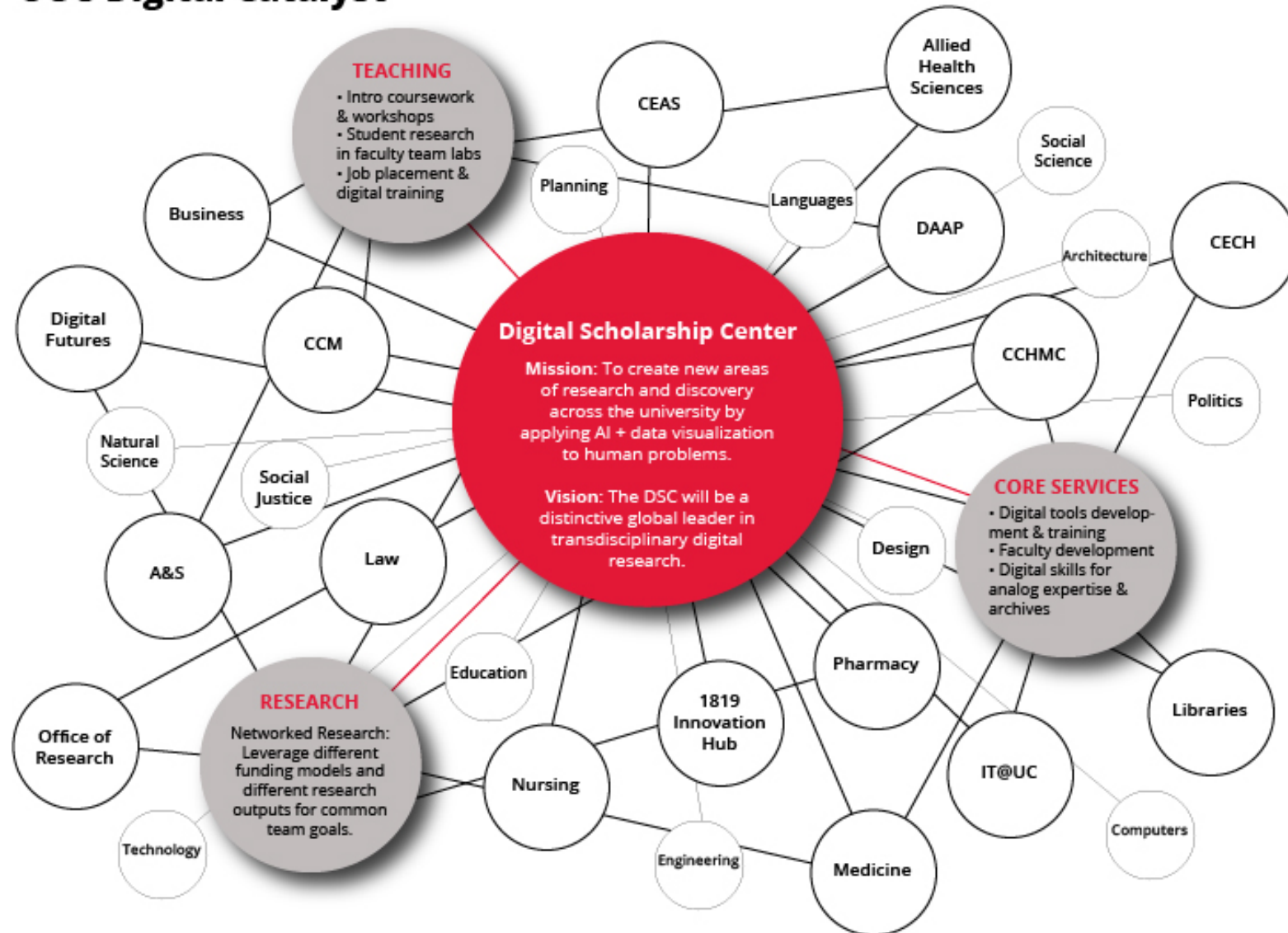


# DSC: What We Do

- We are a technical catalyst: technology to activate new research
- Machine Learning and Human-Interpretable Data Visualization on Large Unstructured Datasets (Text, Image, Sound, Video)
- We translate between disciplines that rarely interact in order to connect content experts with technical experts.
- We provide resources and infrastructure to nurture research questions and collaborations that slip between the cracks of colleges and funding agencies.



# UC's Digital Catalyst



# Digital Integration



**UC DIGITAL  
SCHOLARSHIP CENTER**

**R&DS**  
Research & Data Services

UC Libraries provides access to a wide range of Research Data and GIS services and resources for the campus community. Informationists and librarians are available to assist researchers in managing and preserving research data, finding and acquiring external data, and in utilizing GIS techniques and software. The library also provides a variety of computing and collaboration spaces to support researchers.

# DSC: Who

- **DSC**



James Lee, Assc Vice Provost for Digital Scholarship and Assc Dean of Libraries, Director of DSC



Kristen Burgess, Operational Manager



Lindsay Nickels, Program Coordinator



Ezra Edgerton, Data Visualization Developer



Erin McCabe, Digital Scholarship Library Fellow

- **RDS**



Amy Koshoffer, Asst Director of Research & Data Services



Tiffany Grant, Asst Director of Research & Data Services



Rebecca Olson, Business and Social Science Informationist



Don Jason, Health Informationist



Ted Baldwin, Director, Science & Engineering Libraries



Dorcas Washington, Data Analyst Specialist

- **Graduate Students (English, Computer Science, Business Analytics)**

# Digital Scholarship Center, Phase 1: From Initiative to Center



University of  
CINCINNATI

HOME ABOUT US WHAT IS DH NEWS VISUALIZATIONS PUBLICATIONS CONTACT US

## UC DIGITAL SCHOLARSHIP CENTER

A catalyst for collaborative, trans-disciplinary forms of research and teaching, bringing together humanistic methods with technical innovations.

↓

The DSC has assembled research groups that genuinely span multiple disciplines, with people trained to think very differently about every step in the research process.

Teams are composed of true partners across entire research lifecycle:

- o Formulation of research questions
- o Pitching grant proposals
- o Dataset cleanup and manipulation
- o Data analysis and visualization
- o Argument formation
- o Publication of findings

In 2017, the DSC received \$900,000 from the Andrew W. Mellon Foundation to expand this mission.





# Digital Scholarship Center, Phase 2: Center Expansion



## **UC awarded a \$700K grant from The Andrew W. Mellon Foundation**

The renewal grant will advance and expand the Digital Scholarship Center's "catalyst" model

# How?

1. Technical: Our platform adapts machine learning approaches to any text and image dataset for research projects. We apply these methods in a discipline-specific way.
2. Human: We assemble teams to nurture these unconventional transdisciplinary research questions and partnerships – every collaborator has different goals and culture.

Saba Mann  
University of Cincinnati  
Advised by  
Prof. Matthew Mizinsky

Keywords:  
exhibition design,  
self-advocacy,  
participatory design

This project has been supported in part by  
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Catalyst Award from the Digital Scholarship Center,  
University of Cincinnati

## Curatorial Groupware

**What is it?**  
A custom – but extensible – groupware for collaborative curation that extends the digital interfacing of the physical (and often laborious) exhibition design process.

**Platform**  
History Moves is a research collaborative that enables self-advocacy through participatory design of a public history exhibition titled “Women’s History of HIV in America”

HISTORY MOVES PROCESS (PARTICIPATORY DESIGN × PUBLIC HISTORY)

COLLECTION → INTERPRETATION → DESIGN → CURATION → DISSEMINATION

**Digital > Curatorial Groupware**

New might digitization impact the curation process of History Moves? 🔍

Remote collaboration | Synchronicity | Visualization | AR preview  
Programming | Site-specific curation | Touch-sensitive audio

**How might digital tools impact the curation process of community-built archives?**

A mobile and modular exhibition carries the stories of over 40 participating women – from disparate geographies of Chicago, Brooklyn, and N. Carolina – through a series of posters composed of their words, personal images, and relevant historical archival images and information. As the exhibition travels, the women’s narratives will be transported to diverse sites through visualized verbatims from long-form interviews. Curatorial Groupware furthers the participants’ agency to self-advocate in this process by allowing the participants to curate the content to be presented at each venue.

Digitization of this process enables real-time synchronous curation of the community-built archive. Project participants are able to work together remotely to curate site-specific exhibitions, building on their collective knowledge of sites and locations. The groupware mimics natural forms of in-person communication, extending the collective decision-making of a participatory design process.

**INSTRUCTION MANUAL**  
Venue Information  
Space Design  
Selected Content  
Curated Elevations  
Programming

**Physical > Curatorial Groupware**

**Which physical processes are integral in the digital system?**

## Data Management

- Open Science Framework
- Open Datasets
- Machine Actionable Data Management
- "Gentle Introduction to Data": Human Welcome



- High Performance Computing
- Storage
- Data Structures:
  - Format,
  - Fields,
  - Metadata,
  - Machine Readable

## Data Analysis: Digital Humanities to Bioinformatics

- Multimodal Deep Learning / Machine Learning
- Data Visualization Interfaces
- Mixed Methods: EDA + CDA
- Quantitative and Qualitative Data



- Team Science Culture
- Writing + Environmental Scans
- Method Translation
- Student Training and NextGen MA / PhD

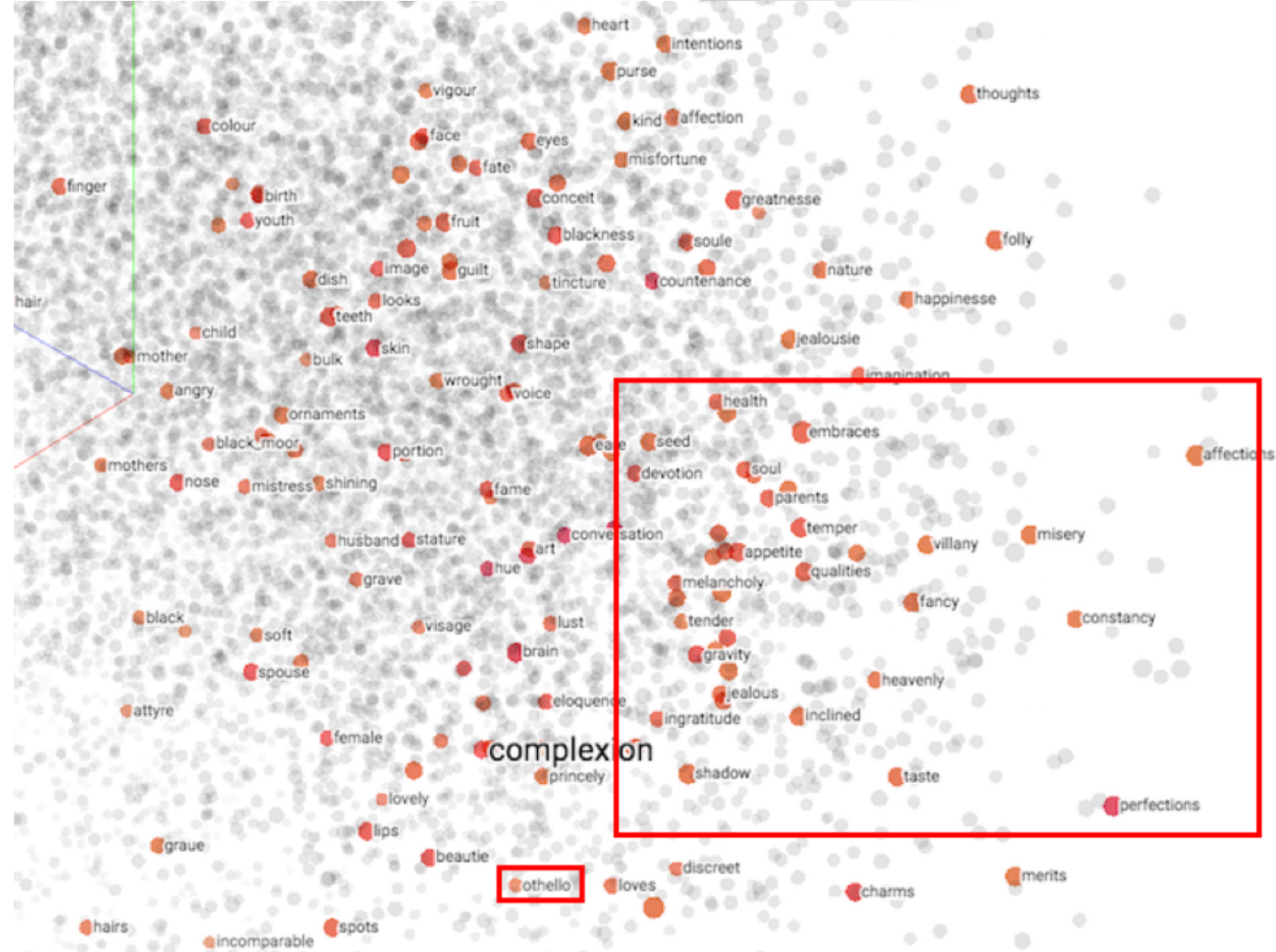
## Digital Outcomes and Products

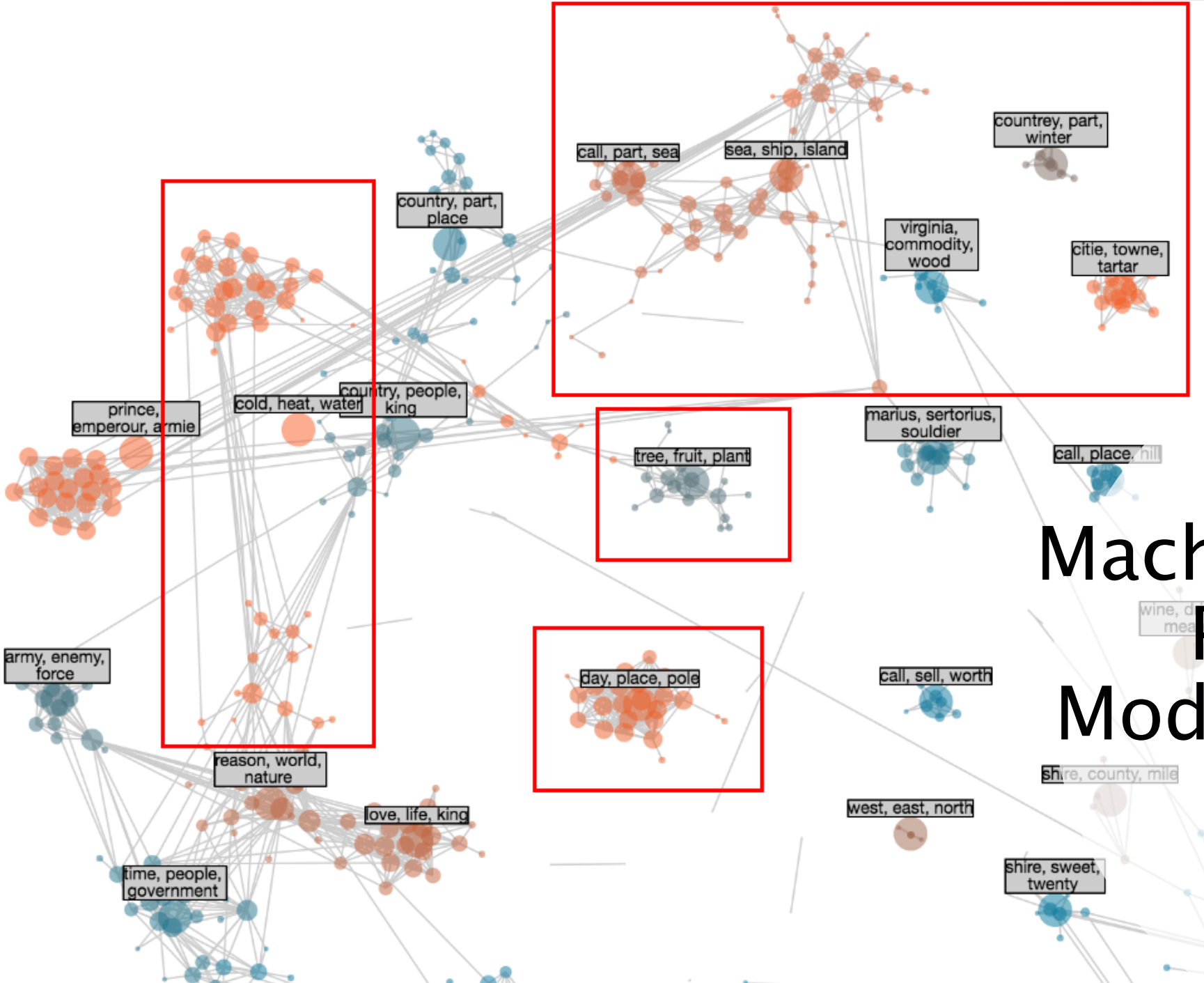
- Publications
- Conference Presentations
- Reference Datasets
- Project Websites / Apps
- Grants
- Machine Actionable Data Management



# Human Centered ML Needs Data Visualization

```
python run_lda_trainer.py -t 50 -p 100 -s temperate  
[DEBUG] [batch.clusters_lda] [2019-09-19 14:59:39,740] [179]: Wrote corpus file: /Users/james.jlee/RNLP-Data/LDA-models/eebo2/lda-temperate-bow_2019-09-19_14.48.04_50_100/corpus_data.json  
INFO:gensim.models.ldamodel:using symmetric alpha at 0.02  
INFO:gensim.models.ldamodel:using symmetric eta at 0.02  
INFO:gensim.models.ldamodel:using serial LDA version on this node  
INFO:gensim.models.ldamodel:running online LDA training, 50 topics, 100 passes over the supplied corpus of 8721 documents, updating model once every 2000 documents, evaluating perplexity every 8721 documents, iterating 50x with a convergence threshold of 0.001000  
INFO:gensim.models.ldamodel:PROGRESS: pass 0, at document #2000/8721  
DEBUG:gensim.models.ldamodel:performing inference on a chunk of 2000 documents  
DEBUG:gensim.models.ldamodel:132/2000 documents converged within 50 iterations  
DEBUG:gensim.models.ldamodel:updating topics  
INFO:gensim.models.ldamodel:merging changes from 2000 documents into a model of 8721 documents  
INFO:gensim.models.ldamodel:topic #19 (0.020): 0.006*man + 0.005*great + 0.004*nature + 0.004*time + 0.003*cold + 0.003*place + 0.003*onely + 0.003*oyle + 0.003*selfe  
INFO:gensim.models.ldamodel:topic #7 (0.020): 0.008*time + 0.007*man + 0.005*men + 0.005*great + 0.004*earth + 0.004*place + 0.004*day + 0.003*god + 0.003*selfe + 0.003*onely  
INFO:gensim.models.ldamodel:topic #41 (0.020): 0.005*day + 0.004*men + 0.004*faire + 0.004*great + 0.003*man + 0.003*god + 0.003*nature + 0.003*horse + 0.003*selfe + 0.003*onely  
INFO:gensim.models.ldamodel:topic #34 (0.020): 0.005*great + 0.005*man + 0.004*men + 0.003*called + 0.003*place + 0.003*onely + 0.003*selfe + 0.003*god + 0.003*time + 0.003*land  
INFO:gensim.models.ldamodel:topic #4 (0.020): 0.008*great + 0.007*men + 0.005*place + 0.005*man + 0.004*god + 0.004*time + 0.003*king + 0.003*selfe + 0.003*onely + 0.003*nature  
INFO:gensim.models.ldamodel:topic diff=16.752374, rho=1.000000  
INFO:gensim.models.ldamodel:PROGRESS: pass 0, at document #4000/8721  
DEBUG:gensim.models.ldamodel:performing inference on a chunk of 2000 documents  
DEBUG:gensim.models.ldamodel:373/2000 documents converged within 50 iterations  
DEBUG:gensim.models.ldamodel:updating topics  
INFO:gensim.models.ldamodel:merging changes from 2000 documents into a model of 8721 documents  
INFO:gensim.models.ldamodel:topic #49 (0.020): 0.006*men + 0.005*time + 0.005*man + 0.005*great + 0.004*called + 0.003*bloud + 0.003*onely + 0.003*vertue + 0.003*god + 0.003*reason  
INFO:gensim.models.ldamodel:topic #8 (0.020): 0.008*fleash + 0.006*man + 0.006*time + 0.006*men + 0.005*great + 0.004*place + 0.003*pole + 0.003*degrées + 0.003*onely + 0.003*air  
INFO:gensim.models.ldamodel:topic #46 (0.020): 0.006*man + 0.005*men + 0.005*flesh + 0.005*ounce + 0.004*great + 0.004*dram + 0.004*life + 0.003*nature + 0.003*age + 0.003*wana_ounce  
INFO:gensim.models.ldamodel:topic #6 (0.020): 0.010*man + 0.008*great + 0.006*men + 0.004*called + 0.004*life + 0.004*time + 0.004*king + 0.004*god + 0.003*onely + 0.003*people  
INFO:gensim.models.ldamodel:topic #17 (0.020): 0.008*god + 0.008*world + 0.006*man + 0.005*called + 0.005*earth + 0.005*men + 0.004*nature + 0.004*reason + 0.003*great + 0.003*circles  
INFO:gensim.models.ldamodel:topic diff=5.133218, rho=0.707107  
INFO:gensim.models.ldamodel:PROGRESS: pass 0, at document #6000/8721  
DEBUG:gensim.models.ldamodel:performing inference on a chunk of 2000 documents
```





There are 55 labels. Summary

Distance function: **sym\_kullback\_leibler: 0.3**

CCs with Ubiquity:  & Above

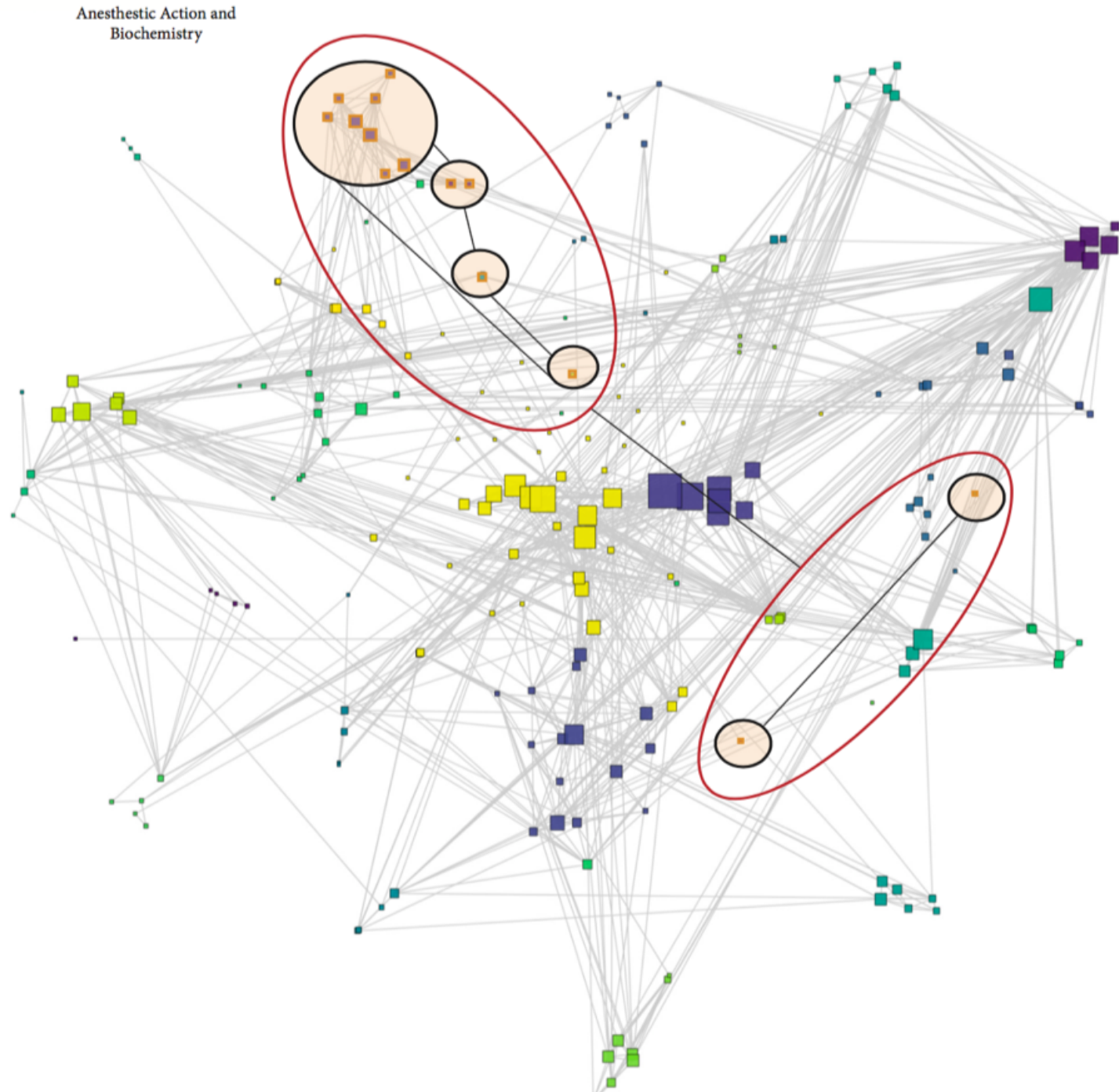
Show By Cc / By Node.

|            |     |    |   |
|------------|-----|----|---|
| 54 (1.1)   | 133 | 49 | cold, heat, water, air, part, body, hot, time, place, nature                  |
| 53 (1.2.1) | 63  | 47 | sea, ship, island, time, land, find, part, selfe, onely, place                |
| 42 (1.3)   | 175 | 30 | god, christ, lord, world, church, christian, word, heaven, spirit, sin        |
| 44 (1.7)   | 108 | 27 | country, part, place, soil, river, air, degree, lie, north, commodity         |
| 0 (3)      | 121 | 24 | day, place, pole, latitude, degree, parallel, hour, longest_day, earth, heure |
| 41 (1.0)   | 148 | 21 | prince, emperour, armie, send, force, king, army, place, fight, souldier      |
| 50 (1.4.0) | 118 | 20 | reason, world, nature, find, soul, opinion, life, mind, time, god             |
| 46 (1.9)   | 146 | 19 | time, people, government, prince, law, nation, person, state, king, subject   |
| 47 (1.10)  | 168 | 19 | tree, fruit, plant, grow, ground, earth, seed, root, garden, sort             |
| 48 (1.5.0) | 127 | 18 | country, people, king, sea, ship, place, island, part, time, land             |
| 51 (1.4.1) | 140 | 18 | love, life, king, leav, time, desire, day, find, spring, eat                  |
| 52 (1.2.0) | 183 | 10 | call, part, sea, country, place, east, west, south, north, river              |
| 5 (8)      | 100 | 8  | call, sell, worth, iron, brass, buy, people, piece, shore, run                |
| 6 (9)      | 101 | 8  | call, place, hill, ancient, shire, towne, city, time, castle, mountaine       |

# Machine Learning Platform. Model of Models (MoM)

# Machine Learning Platform: Model of Models (MoM)

- Two machine learning strategies used to observe the latent patterns in large corpora.
  - Topic modeling (Latent Dirichlet Allocation – LDA)
  - Word embeddings (word2vec, BERT)
- Aggregates multiple models in parallel to compare word usage across the parallel models.
- Clusters integrate topics from the 6 models into an aggregated “model of models”:
  - Confirm consistent topics across all models
  - Reveal underrepresented topics that may not have appeared in a single model representation.
- Distributed parallel approach increases user confidence and interpretability of our models by bringing the most stable topics to the top tier of the model results



# Machine Learning Platform: Trust the Human for Validation

- Internal validation: Topic Coherence
- External validation:
  - Subject matter expert tagging of randomized 20% (N>1000) corpus
  - Blind human coder panel – Percentage agreement
- Replicability:
  - Pattern recognition capabilities of NLP methods as an information retrieval – and not a black-box classification – approach.
  - Provide models capable of evaluation by our panel of multiple independent coders.
  - Parallel replicates in each model (6–20 runs)
- Hybrid ML approach:
  - Human judgment of subject matter experts to verify and tag the model result – outperforms a purely machine-based analysis.
  - Semi-supervised learning for classification.

# Interoperability with Datasets, Medium and Large

- HTRC Extracted Features, JSTOR Data for Research, Chronicling America, Text Creation Partnership, Harvard Case Law.
- PubMed + PubMed Central, US Patent Claims, EPIC EHR FHIR Data Structures (IRB Approved).
- Social Media (Twitter, Instagram, Reddit).
- Small corpora: we'll help you read them.

[model](#) | of | [models](#) | [Get Started](#) | [Learn](#) | [Analyze](#)

[Account](#) ▾

Select Database

Search Text

Filter Docs

Explore Docs

Select Vis

Set and Run Model

## Select your Database ⓘ

|   |   |  |
|---|---|--|
| <p>COVID-19 Articles: 13.2k docs</p> <p>1.27 s/100 docs</p>                                     | <p>Pubmed Abstract: 29.4M docs</p> <p>1.27 s/100 docs</p>         | <p>Pubmed Central: 2.15M docs</p> <p>2.36 s/100 docs</p>   |
| <p>Jstor Life Science: 825k docs</p> <p>28.6 s/100 docs</p>                                     | <p>CaseLaw: 325k docs</p> <p>18.4 s/100 docs</p>                  | <p>Archaeology: 2.39k docs</p> <p>41.7 s/100 docs</p>      |
| <p>Iowa Latin Canon: 2.98k docs</p> <p>No Time Data <input type="checkbox"/></p>                | <p>Ehealth Alzheimer: 129 docs</p> <p>1.30 s/100 docs</p>         | <p>Text Creation: 69.9k docs</p> <p>72.8 s/100 docs</p>    |
| <p>AC Justice: 297 docs</p> <p>No Time Data <input type="checkbox"/></p> <p>3.60 s/100 docs</p> | <p>Anesthesiology: 28.0k docs</p> <p><input type="checkbox"/></p> | <p>Ted Talks: 992 docs</p> <p><input type="checkbox"/></p> |

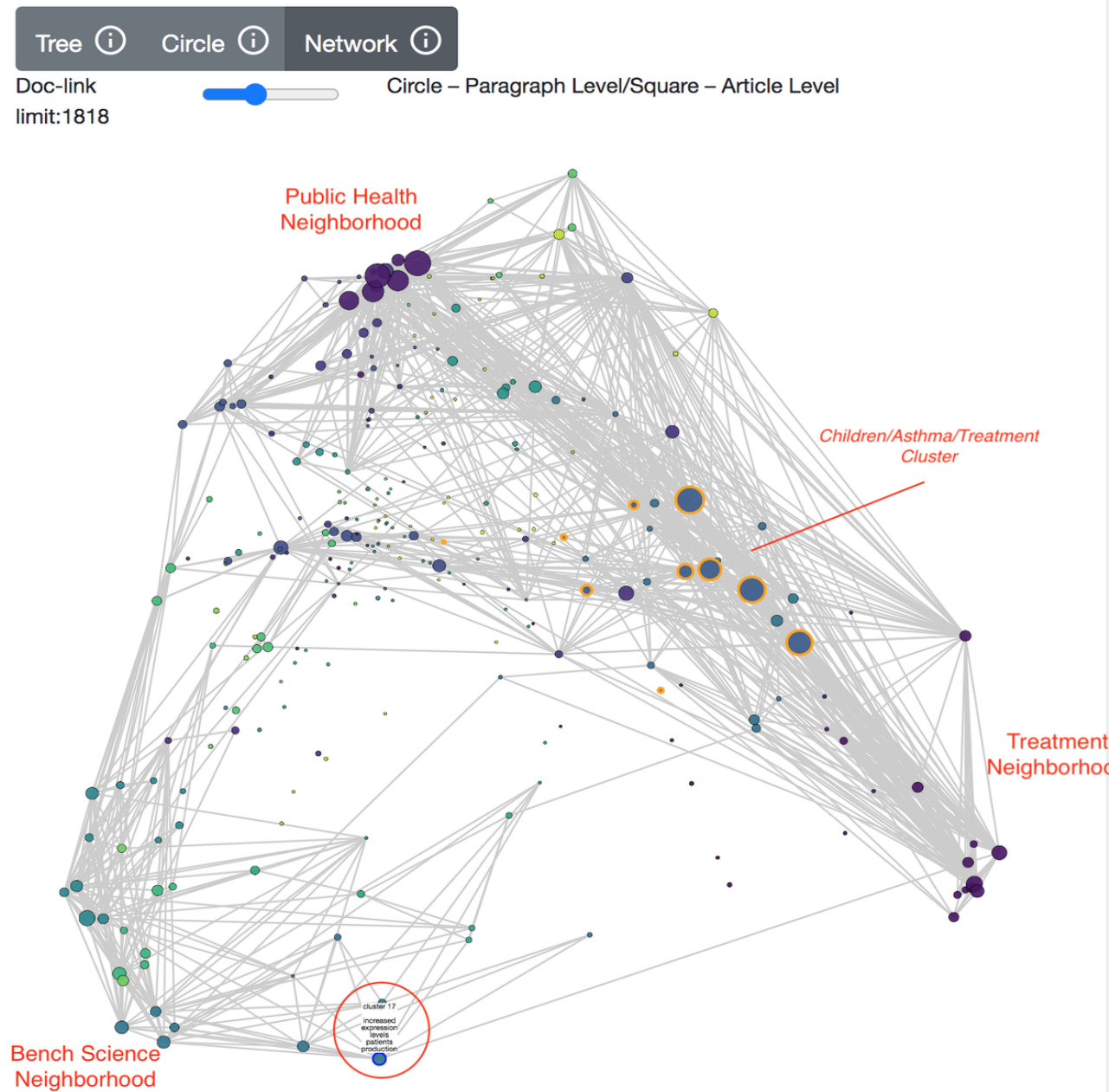


Documents ⓘ Clusters ⓘ

|   |                    |
|---|--------------------|
| The Impact of Respiratory Viral Infection on Wheezing Illnesses and Asthma Exacerbations  | 5.803449511528015  |
| Epidemiology of viral respiratory tract infections in a prospective cohort of infants and toddlers attending daycare                      | 5.163539171218872  |
| Population-based hospitalization incidence of respiratory viruses in community-acquired pneumonia in children younger than 5 years of age | 5.11656254529953   |
| Respiratory Syncytial Virus Coinfections With Rhinovirus and Human Bocavirus in Hospitalized Children                                     | 4.919344365596771  |
| Virus Etiology of Airway Illness in Elderly Adults  | 4.881579369306564  |
| Respiratory viral infections in a cohort of children during the first year of life and their role in the development of wheezing☆         | 4.8710708022117615 |

## The Impact of Respiratory Viral Infection on Wheezing Illnesses and Asthma Exacerbations

Overview ::: Respiratory viral-induced wheezing illnesses in young children Viral bronchiolitis is a LRTI typically associated with cough, tachypnea, retractions, and diffuse wheezing and rales [8], [9]. Bronchiolitis is a leading cause of hospitalizations in the first year of life, accounting for an estimated 120,000 infant hospitalizations annually [10]. In infants, the etiologic agents of bronchiolitis and other viral respiratory infections associated with wheezing include respiratory syncytial virus (RSV), rhinovirus, influenza, parainfluenza (PIV), adenovirus, and more recently identified viruses, such as human metapneumovirus (hMPV) and human boca virus (hBoV) [11], [12], [13], [14]. RSV causes epidemics of bronchiolitis and typically circulates in temperate climates during November to April with peaks in the winter months [11], [15], [16]. In tropical climates, peaks are related to temperature and level of rainfall [17]. RSV infects the majority of children during their first year of life and essentially all children show evidence of RSV infection by age 3 years [18]. The initial RSV infection is typically the most severe, causing lower respiratory tract disease, such as bronchiolitis, in 20% to 30% of infants [11], [18], [19]. Other viruses such as rhinovirus, PIV, and adenovirus circulate yearly year round with seasonal peaks of illness [10], [11], [19].

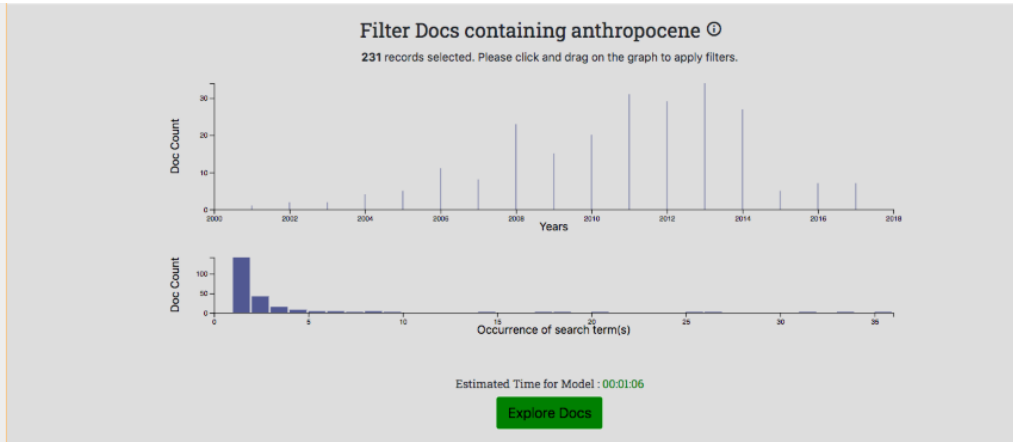


| cluster | #para, #docs | # topics | terms   |
|---------|--------------|----------|---|
| 12      | 0,7872       | 10       | children patients asthma rsv age                    |
| 2       | 0,7190       | 19       | patients patient cases days years                   |
| 3       | 0,7100       | 8        | information countries people health research        |
| 20      | 0,4774       | 13       | binding proteins protein peptides viruses           |
| 17      | 0,4586       | 10       | expression production cells activation increased    |
| 10      | 0,3609       | 7        | detection sensitivity samples detect positive       |
| 7       | 0,3257       | 8        | infected model individuals transmission parameters  |
| 29      | 0,3176       | 7        | cells expression infected incubated infection       |
| 23      | 0,2567       | 8        | participants studies respondents information people |
| 6       | 0,2566       | 8        | samples positive detected tested patients           |
| 11      | 0,2453       | 7        | human humans species viruses virus                  |
| 16      | 0,2350       | 5        | covid-19 patients                                   |

# Model of Models: User Interface

Adaptable visualization outputs based on a single underlying model.

- Select Database
- Search Text
- Filter Docs
- Explore Docs
- Select Vis
- Set and Run Model



### Explore 231 Documents

| Article Title  | date | score | included |
|--|------|-------|----------|
| The Anthropocene: From Global Change to Planetary Stewardship                          | 2011 | 35    | ✓        |
| Global assemblages, resilience, and Earth Stewardship in the Anthropocene              | 2013 | 35    | ✓        |
| Global Analysis of River Systems: From Earth System Controls to Anthropocene Syndromes | 2003 | 33    | ✓        |
| Observing changing ecological diversity in the Anthropocene                            | 2013 | 31    | ✓        |
| The Anthropocene: Are Humans Now Overwhelming the Great Forces of Nature?              | 2007 | 26    | ✓        |
| Detection of Natural and Anthropic Features on Small Islands                           | 2017 | 25    | ✓        |
| Curiosity and context revisited: crassulacean acid metabolism in the Anthropocene      | 2008 | 20    | ✓        |
| The Anthropocene Mass Extinction: An Emerging Contribution Theme for Science           | 2011 | 18    | ✓        |

### The Anthropocene: From Global Change to Planetary Stewardship

ID: 41417334  
 Journal: Ambio  
 Pub Date: 2011-01-03  
 Author: Will Steffen, Åsa Persson, Lisa Deutsch, Jan Zalasiewicz, Mark Williams, Katherine Richardson, Carole Crumley, Paul Crutzen, Carl Folke, Line Gordon, Mario Molina

AMBIO (2011) 40:739-761 DOI 10.1007/s13280-011-0185-x INVITED PAPER The Anthropocene: From Global Change to Planetary Stewardship Will Steffen, Åsa Persson, Lisa Deutsch, Jan Zalasiewicz, Mark Williams, Katherine Richardson, Carole Crumley, Paul Crutzen, Carl Folke, Line Gordon, Mario Molina, Veerabhadran Ramanathan, Johan Rockström, Marten Scheffer, Hans Joachim Schellnhuber, Uno Svedin Received: 29 June 2011 / Accepted: 29 June 2011 / Published online: 12 October 2011 Abstract Over the past century, the total material wealth of humanity has been enhanced. However, in the twenty-first century, we face scarcity in critical resources, the degradation of ecosystem services, and the erosion of the planet's capability to absorb our wastes. Equity issues remain stubbornly difficult to solve. This situation is novel in its speed, its global scale and its threat to the resilience of the Earth System. The advent of the Anthropocene, the time interval in which human activities now rival global geophysical processes, suggests that we need to fundamentally alter our relationship with the planet we inhabit. Many approaches could be adopted, ranging from geo-engineering solutions that purposefully manipulate parts of the Earth System to becoming active stewards of our own life support system. The Anthropocene is a reminder that the Holocene, during which complex human societies have developed, has been a stable accommodation

Estimated Time for Model : 00:01:06

231 of 231 Documents Selected

[Choose Visualization](#) [Results Page](#)

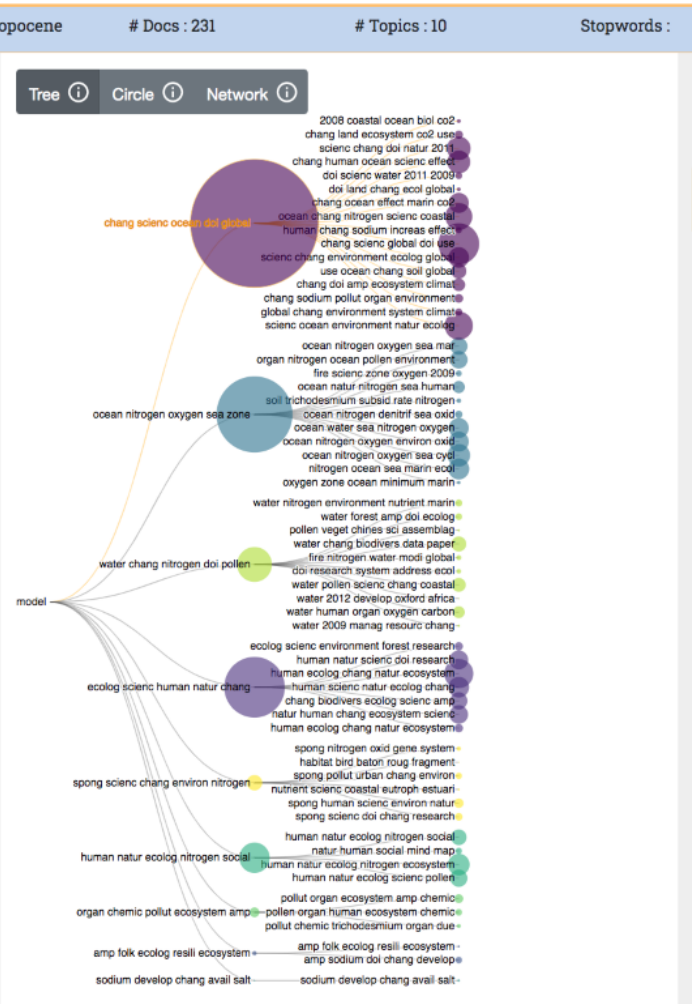
Corpus : JSTOR Term : anthropocene # Docs : 231 # Topics : 10

### Documents

| Title  | Score              |
|--|--------------------|
| Organic Pollutants in Coastal Waters, Sediments, and Biota: A Relevant Driver for Ecosystems During  | 10.233321070671082 |
| Planetary Stewardship in an Urbanizing World: Beyond City Limits                                     | 5.5185911655426025 |
| Elevated level of carbon dioxide affects metabolism and shell formation in oysters                   | 4.993282516138458  |
| How will Ocean Acidification Affect Baltic Sea Ecosystems? An Assessment of Plausible Impacts on Key | 4.991919696331024  |
| Public awareness, concerns, and priorities about anthropogenic impacts on marine environments        | 4.9840657114982605 |
| The Anthropocene: Are Humans Now Overwhelming the Great Forces of Nature?                            | 4.955211868285133  |

### Organic Pollutants in Coastal Waters, Sediments, and Biota: A Relevant Driver for Ecosystems During the Anthropocene?

Estuaries and Coasts (2011) 33:1-14 DOI 10.1007/s12237-009-9255-8 THE H.T. OUM SYNTHESIS ESSAY Organic Pollutants in Coastal Waters, Sediments, and Biota: A Relevant Driver for Ecosystems During the Anthropocene? Jordi Dachs · Laurence M. Æj a nel le Received: 11 November 2009 / Revised: 28 November 2009 / Accepted: 4 December 2009 / Published online: 9 January 2010 © Coastal and Estuarine Research Federation 2010 Abstract The total number of synthetic organic chemicals introduced to the environment by humans has never been quantified, but it is not lower than thousands. A fraction of these chemicals have toxic effects to coastal organisms and presumably affect ecosystems structure and function. During the last decades, some of the processes affecting the transport, degradation, and fate of a limited number of chemicals have been studied, and the rising concern for environmental risk of organic chemical has led to the regulation of a few of them by national and international organisms. However, the environmental inventory of organic pollutants is far from being quantified, and current methodologies used in most toxicological tests only allow to determine effects of individual chemicals to organisms. There are major limitations on appropriate methodologies to assess the effects of organic pollutants at population and ecosystem levels and the effects induced by complex mixtures of organic pollutants present in natural environments. The modification of the composition of the biosphere by a



### Clusters

| cluster | #para | # docs | # topics | terms                               |
|---------|-------|--------|----------|-------------------------------------|
| 0       | 183.0 | 16     | 16       | chang scienc ocean doi global       |
| 3       | 105.0 | 11     | 11       | ocean nitrogen oxygen sea zone      |
| 7       | 39.0  | 10     | 10       | water chang nitrogen doi pollen     |
| 1       | 73.0  | 7      | 7        | ecolog scienc human natur chang     |
| 8       | 21.0  | 6      | 6        | spong scienc chang environ nitrogen |
| 5       | 41.0  | 4      | 4        | human natur ecolog nitrogen social  |
| 6       | 17.0  | 3      | 3        | organ chemic pollut ecosystem amp   |
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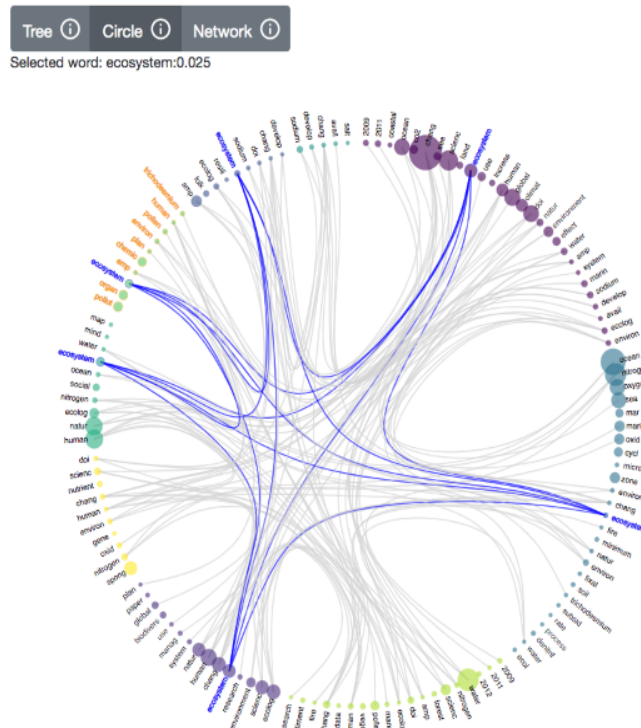
# Model of Models: User Interface

Adaptable visualization outputs based on a single underlying model.

Corpus : JSTOR    Term : anthropocene    # Docs : 231    # Topics : 10    Stopwords :    Years : year-year

Documents ⓘ

| Title  | Score              |
|--|--------------------|
| Organic Pollutants in Coastal Waters, Sediments, and Biota: A Relevant Driver for Ecosystems During  | 6.712952628731728  |
| Modern pollen and land-use relationships in the Taihang mountains; Hebei province, northern China—a  | 0.998788595199585  |
| Freshwater for Resilience: A Shift in Thinking   | 0.9986644983291626 |
| The integration of land and marine spatial planning  | 0.9986258149147034 |
| The Social Dimension in Ecosystem Management. Strengths and Weaknesses of Human-Nature Mind Maps     | 0.9982788562774658 |
| Annual Survival of Birds Captured in a Habitat Island Bordered by the Urban Matrix of Baton Rouge, L | 0.9972216871116638 |



Clusters ⓘ

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|---------|--------------|----------|---|
| 0       | 153.0        | 16       | chang scienc ocean doi global             |
| 3       | 105.0        | 11       | ocean nitrogen oxygen sea zone            |
| 7       | 39.0         | 10       | water chang nitrogen doi pollen           |
| 1       | 73.0         | 7        | ecolog scienc human natur chang           |
| 8       | 21.0         | 6        | spong scienc chang environ nitrogen       |
| 5       | 41.0         | 4        | human natur ecolog scienc nitrogen social |
| 6       | 17.0         | 3        | organ chemic pollut ecosystem amp         |
| 2       | 4.0          | 2        | amp folk ecolog resili ecosystem          |
| 4       | 1.0          | 1        | sodium develop chang evall salt           |

**Organic Pollutants in Coastal Waters, Sediments, and Biota: A Relevant Driver for Ecosystems During the Anthropocene?**

Estuaries and Coasts (2010) 33:1–14 DOI 10.1007/S12237-009-9255-8 THE H. T. ODUM SYNTHESIS ESSAY Organic Pollutants in Coastal Waters, Sediments, and Biota: A Relevant Driver for Ecosystems During the Anthropocene? Jordi Dauchs · Laurence M éj a neil Received: 11 November 2009 / Revised: 28 November 2009 / Accepted: 4 December 2009 / Published online: 9 January 2010 © Coastal and Estuarine Research Federation 2010 Abstract The total number of synthetic organic chemicals introduced to the environment by humans has never been quantified, but it is not lower than thousands. A fraction of these chemicals have toxic effects to coastal organisms and presumably affect ecosystems structure and function. During the last decades, some of the processes affecting the transport, degradation, and fate of a limited number of chemicals have been studied, and the rising concern for environmental risk of organic chemical has lead to the regulation of a few of them by national and international organisms. However, the environmental inventory of organic pollutants is far from being quantified, and current methodologies used in most toxicological tests only allow to determine effects of individual chemicals to

Corpus : TED    Term :    # Docs : 992    Stopwords :    Years : year-year

Selected word: parent

| title   | journal | date       |
|---|---------|------------|
| For parents, happiness is a very high bar                     |         | 2014-04-15 |
| Lessons from the longest study on human development           |         | 2013-10-02 |
| Human nature and the blank slate                              |         | 2008-09-26 |
| Rebuilding a neighborhood with beauty, dignity, hope          |         | 2008-01-20 |
| The forgotten history of autism                               |         | 2019-06-17 |
| 3 fears about screen time for kids – and why they're not true |         | 2013-10-12 |
| My wish: Once Upon a School                                   |         | 2009-03-18 |
| Teaching with the World Peace Game                            |         | 2011-04-20 |
| How acetaminophen can grow old better                         |         | 2013-11-29 |
| How to raise a black son in America                           |         | 2015-04-23 |

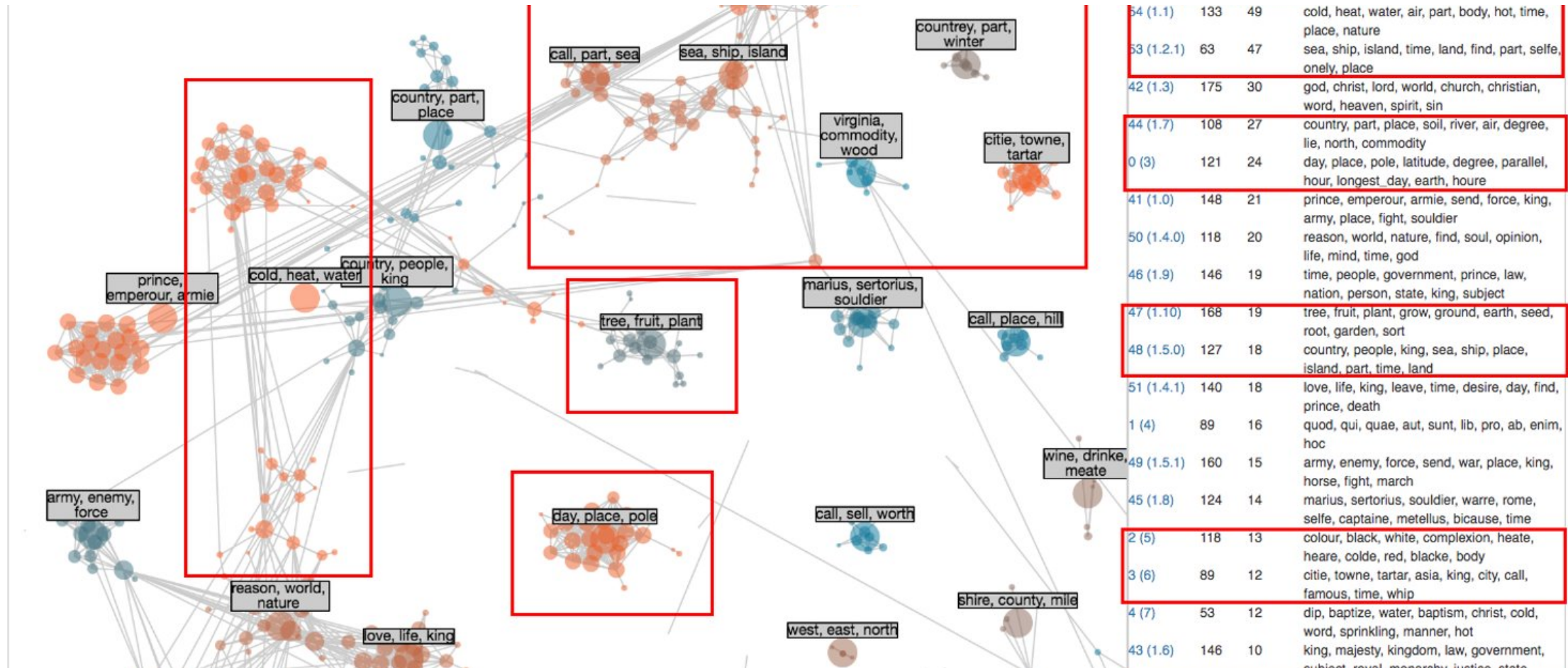
**Lessons from the longest study on human development**

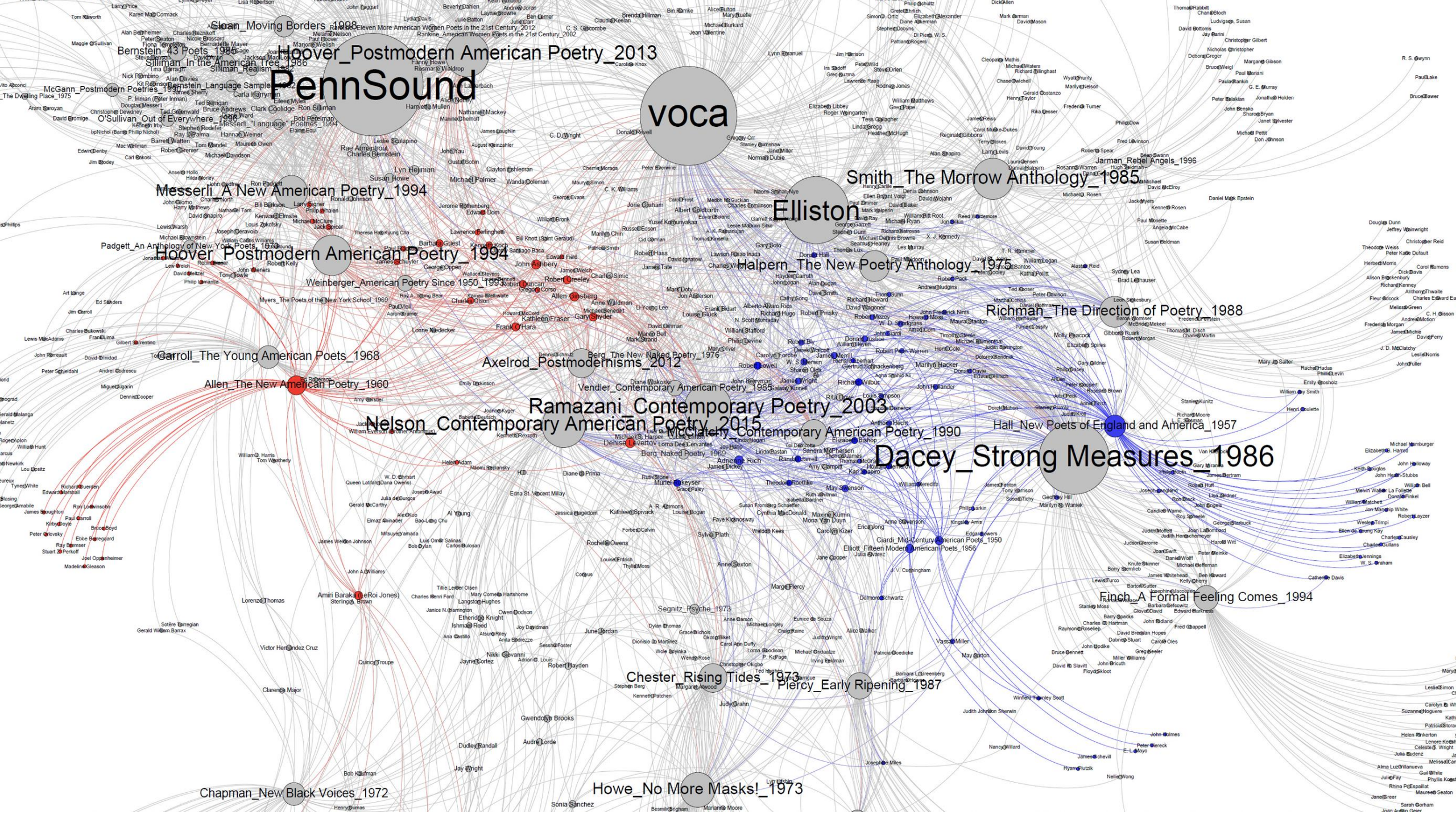
Today I want to confess something to you, but first of all I'm going to ask you a couple of questions. How many people here have children? And how many of you are confident that you know how to bring up your children in exactly the right way? (Laughter) OK, I don't see too many hands going up on that second one, and that's my confession, too. I've got three boys; they're three, nine and 12. And like you, and like most parents, the honest truth is I have pretty much no idea what I'm doing. I want them to be happy and healthy in their lives, but I don't know what I'm supposed to do to make sure they are happy and healthy. There's so many books offering all kinds of conflicting advice. It can be really overwhelming. So I've spent most of their lives just making it up as I go along. However, something changed me a few years ago, when I came across a little secret that we have in Britain. It's helped me become more confident about how I bring up my own children, and it's revealed a lot about how we as a society can help all children. I want to share that secret with you today. For the last 70 years, scientists in Britain have been following thousands of children through their lives as part of an incredible scientific study. There's nothing quite like it anywhere else in the world. Collecting information on thousands of children is a really powerful thing to do, because it means we can compare the ones who say, do well at school or end up healthy or happy or wealthy as adults, and the ones who struggle much more, and then we can sift through all the information we've collected and try to work out why their lives

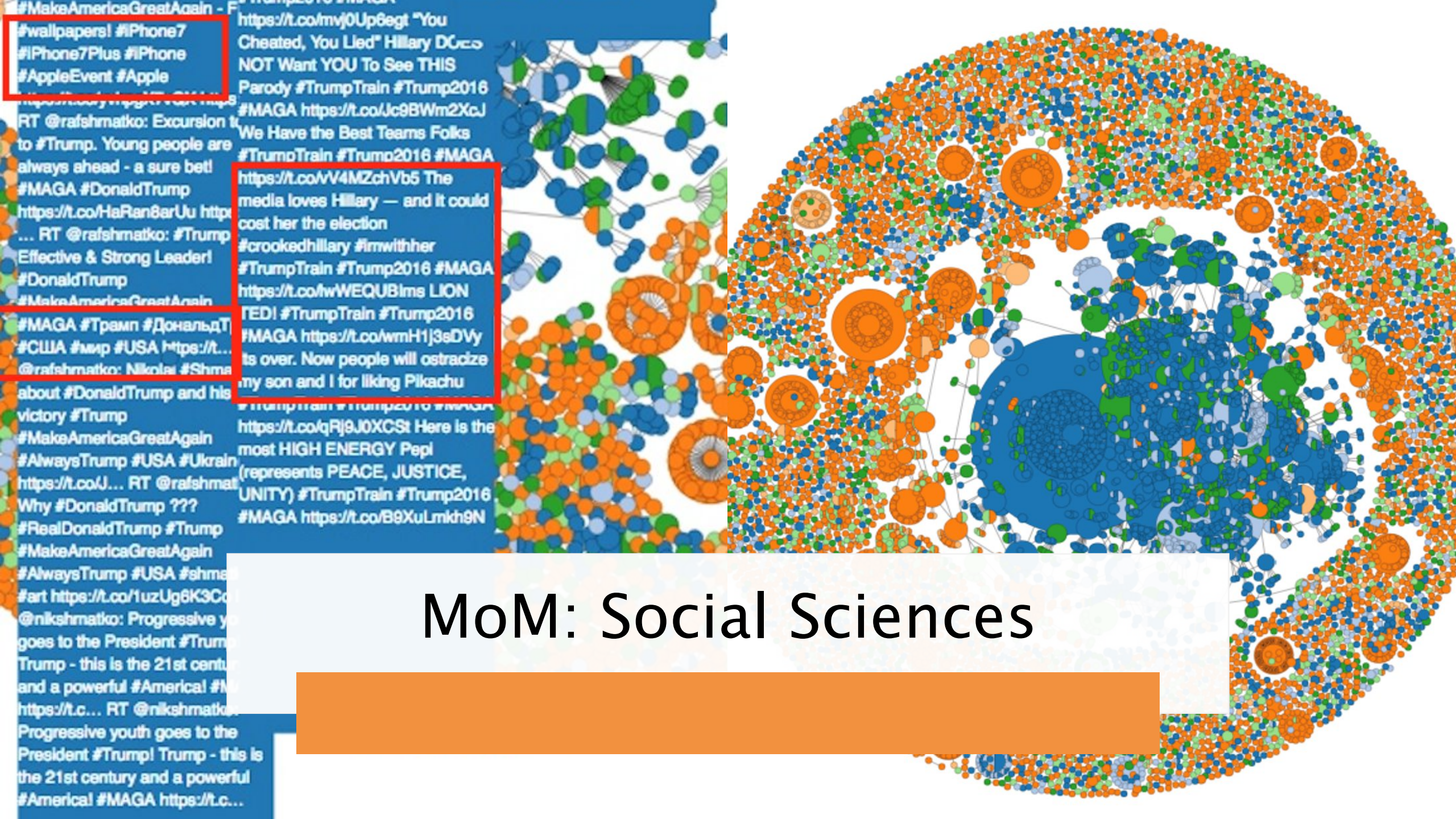


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| 71      | 39   | human-2188, exen-1289, natur-1111, become-1103, untrans-993,         |
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| 60      | 35   | put-2054, turn-1442, picture-888, run-834, head-840,                 |
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| 84      | 31   | today-1414, bring-739, job-641, free-430, plan-388,                  |
| 10      | 30   | place-1590, away-943, line-693, across-605, top-525,                 |
| 46      | 27   | creat-1467, system-1454, data-922, gene-918, inform-860,             |
| 75      | 27   | work-4141, idea-2055, part-1543, learn-1458, whole-985,              |
| 33      | 26   | know-5818, war-4477, do-3822, some-3043, know-5818,                  |
| 37      | 26   | phone-443, full-393, reach-377, cut-378, ground-345,                 |
| 63      | 25   | think-5916, well-3277, happen-2342, mean-2187, night-1288,           |
| 28      | 24   | almost-642, ahead-540, within-523, past-499, third-395,              |
| 57      | 23   | like-8094, size-5247, look-4768, move-1282, open-778,                |
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| 12      | 22   | get-2722, end-1424, see-652, record-430, get-2722,                   |
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| 54      | 20   | grow-675, disease-532, cause-508, cancer-468, pattern-371,           |
| 72      | 19   | high-904, less-645, area-563, food-555, level-544,                   |
| 90      | 19   | need-2703, care-689, social-572, society-484, issue-443,             |
| 81      | 19   | technology-1408, power-1285, develop-934, effect-575, product-547,   |
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| 65      | 15   | five-896, half-737, hour-708, six-582, 100-544,                      |
| 88      | 14   | light-913, water-875, sit-469, size-389, surface-382,                |
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# MoM: Digital Humanities





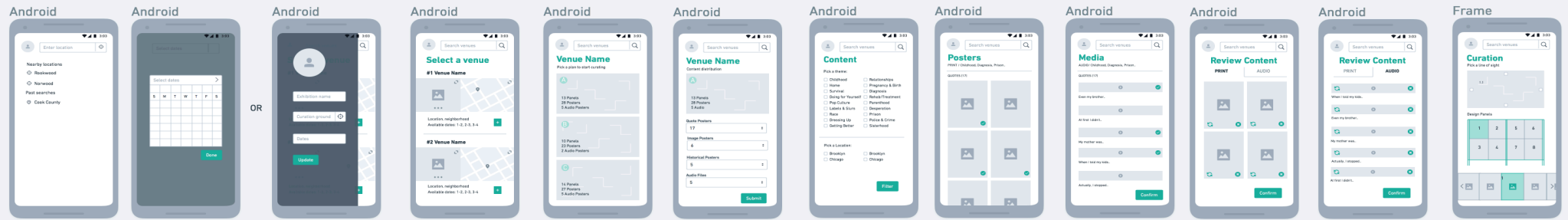


#MakeAmericaGreatAgain - F  
#wallpapers! #iPhone7  
#iPhone7Plus #iPhone  
#AppleEvent #Apple  
RT @rafshmatko: Excursion to  
to #Trump. Young people are  
always ahead - a sure bet!  
#MAGA #DonaldTrump  
https://t.co/HaRan8arUu http  
... RT @rafshmatko: #Trump  
Effective & Strong Leader!  
#DonaldTrump  
#MakeAmericaGreatAgain  
#MAGA #Трамп #ДональдТ  
#США #мир #USA https://t.c  
@rafshmatko: Nikolai #Shma  
about #DonaldTrump and his  
victory #Trump  
#MakeAmericaGreatAgain  
#AlwaysTrump #USA #Ukrain  
https://t.co/J... RT @rafshmat  
Why #DonaldTrump ???  
#RealDonaldTrump #Trump  
#MakeAmericaGreatAgain  
#AlwaysTrump #USA #shma  
#art https://t.co/1uzUg6K3Cc  
@nikshmatko: Progressive y  
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Trump - this is the 21st centu  
and a powerful #America! #M  
https://t.c... RT @nikshmatko  
Progressive youth goes to the  
President #Trump! Trump - this is  
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#America! #MAGA https://t.c...

https://t.co/mvj0Up8egt "You  
Cheated, You Lied" Hillary DC  
NOT Want YOU To See THIS  
Parody #TrumpTrain #Trump2016  
#MAGA https://t.co/Jc9BwM2XcJ  
We Have the Best Teams Folks  
#TrumpTrain #Trump2016 #MAGA  
https://t.co/VV4MZchVb5 The  
media loves Hillary — and it could  
cost her the election  
#crookedhillary #imwithher  
#TrumpTrain #Trump2016 #MAGA  
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FED! #TrumpTrain #Trump2016  
#MAGA https://t.co/wmH1j3sDVy  
its over. Now people will ostracize  
my son and I for liking Pikachu  
#TrumpTrain #Trump2016 #MAGA  
https://t.co/qRj9J0XCSt Here is the  
most HIGH ENERGY Pepl  
(represents PEACE, JUSTICE,  
UNITY) #TrumpTrain #Trump2016  
#MAGA https://t.co/B9XuLmkh9N

# MoM: Social Sciences

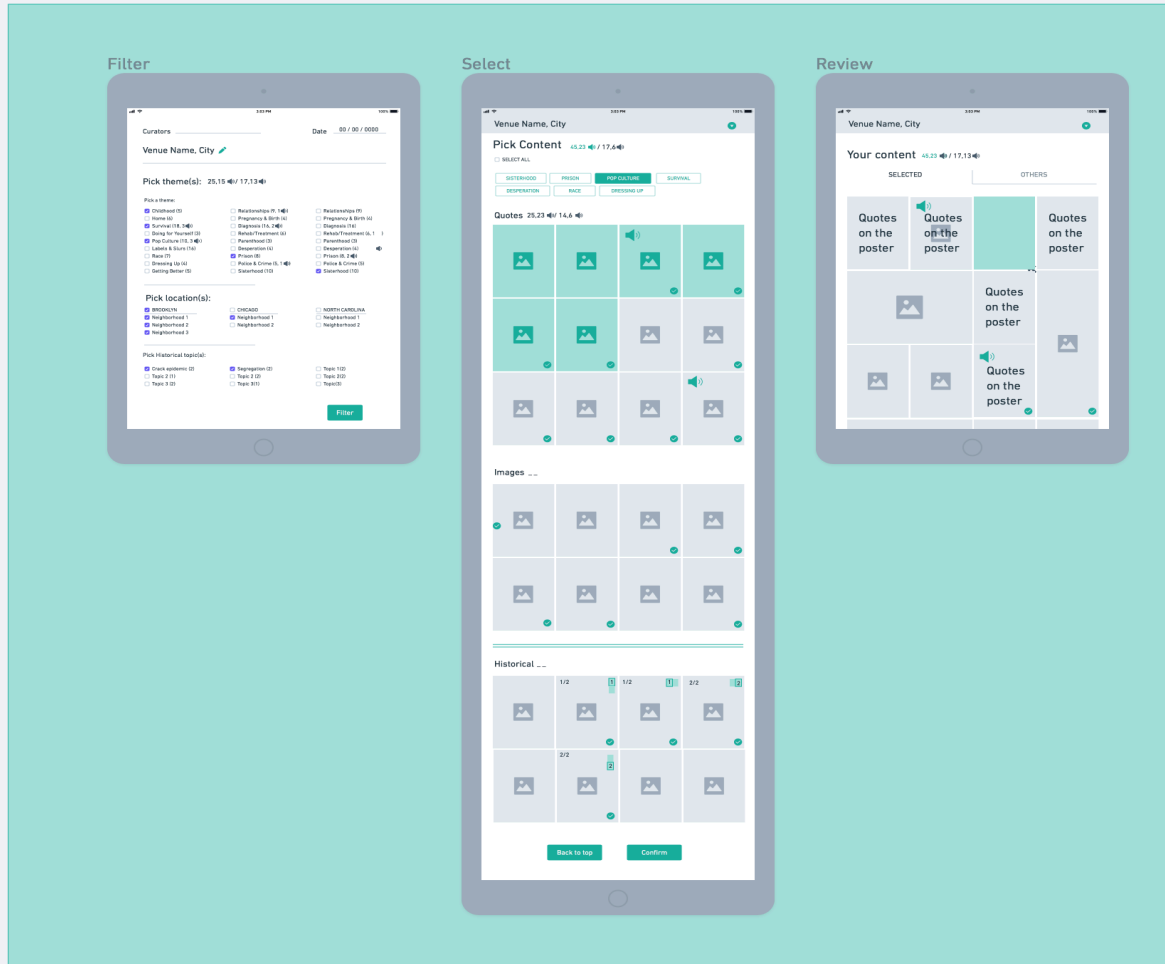




FOR JONATHAN

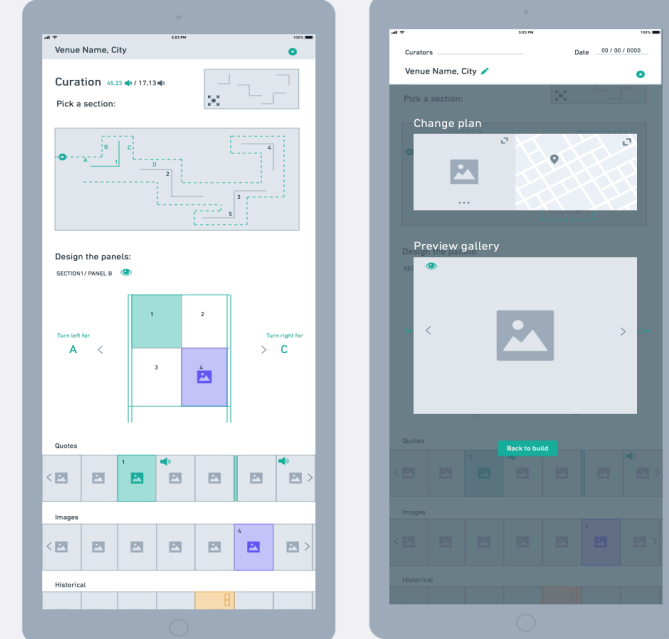
FOR JONATHAN

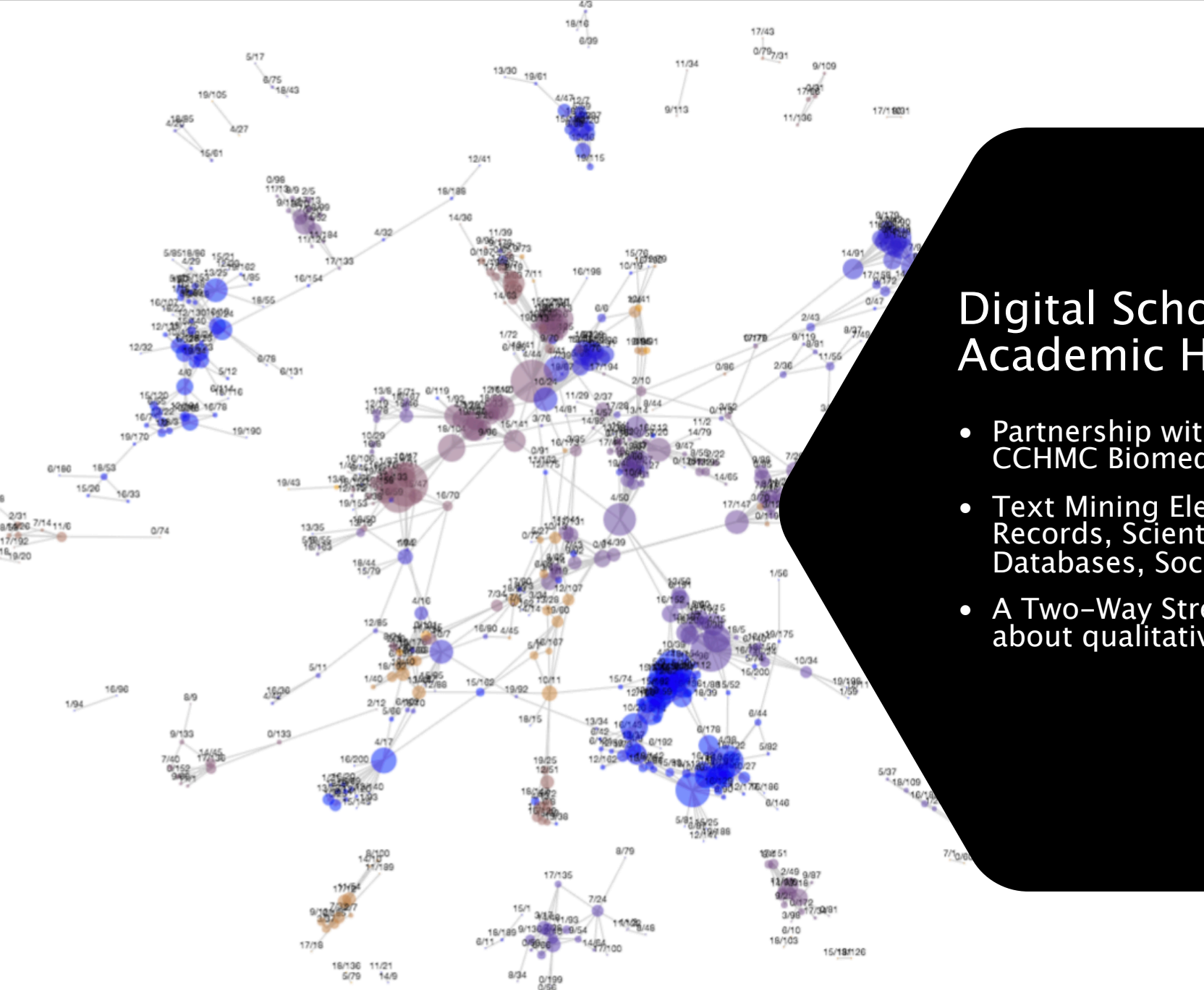
FOR JONATHAN



Curate

Preview





There are 69 labels, including -1.  
 Selected label: 2.0; # Edges: 43  
 Show By Cc / By Node.

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| 1.0 | 18/48       | evolution 0.03, specie 0.02  |
| 1.0 | 18/48       | apoptosis 0.02, <b>expression 0.02</b> , mutant 0.02, <b>function 0.02</b> , control 0.06, control 0.03, <b>arabidopsis 0.03</b> , yeast 0.02, signal 0.02                         |
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| 1.0 | 18/48       | ance 0.03, motif 0.03, mutation 0.03, sequence 0.03  |
| 1.0 | 18/48       | me 0.03, stress 0.03   |
| 1.0 | 18/48       | study 0.02, <b>arabidopsis 0.02</b> , study 0.02, <b>arabidopsis 0.02</b> , complex 0.02, <b>show 0.02</b> , sequence 0.02   |
| 1.0 | 18/48       | cell 0.03, <b>expression 0.02</b> , structure 0.03   |
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| 1.0 | 18/48       | <b>protein 0.03</b> , unit 0.02, mammalian 0.02, mutant 0.04, <b>function 0.03</b> , development 0.03, regu  |
| 1.0 | 18/48       | stress 0.02  |
| 1.0 | 18/48       | chain 0.02, <b>gene 0.04</b> , angiosperm 0.02, family 0.02  |
| 1.0 | 18/48       | <b>expression 0.06, show 0.03, function 0.03</b> , express 0.03, <b>arabidopsis 0.03</b> , control 0.02, role 0.02   |
| 1.0 | 5/100       | <b>gene 0.09</b>   |
| 1.0 | 15/159      | <b>protein 0.26, domain 0.08</b>   |
| 1.0 | 19/113      | <b>protein 0.02, yeast 0.02</b> , show 0.02  |
| 1.0 | 10/27       | <b>protein 0.04, domain 0.03, family 0.02, sequence 0.02</b>   |
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| 1.0 | 16/98       | <b>protein 0.13</b>  |

# Digital Scholarship and Academic Health

- Partnership with College of Medicine, CCHMC Biomedical Informatics.
- Text Mining Electronic Health Records, Scientific Literature, Grant Databases, Social Media, Imaging.
- A Two-Way Street: Teaching STEM about qualitative data.



## Uncertain Diagnosis Project

CCHMC Hospital Medicine Division

### MetaMap

classify based on...

- symptoms
- disorders/diagnoses
- tests/labs
- specialties
- treatments

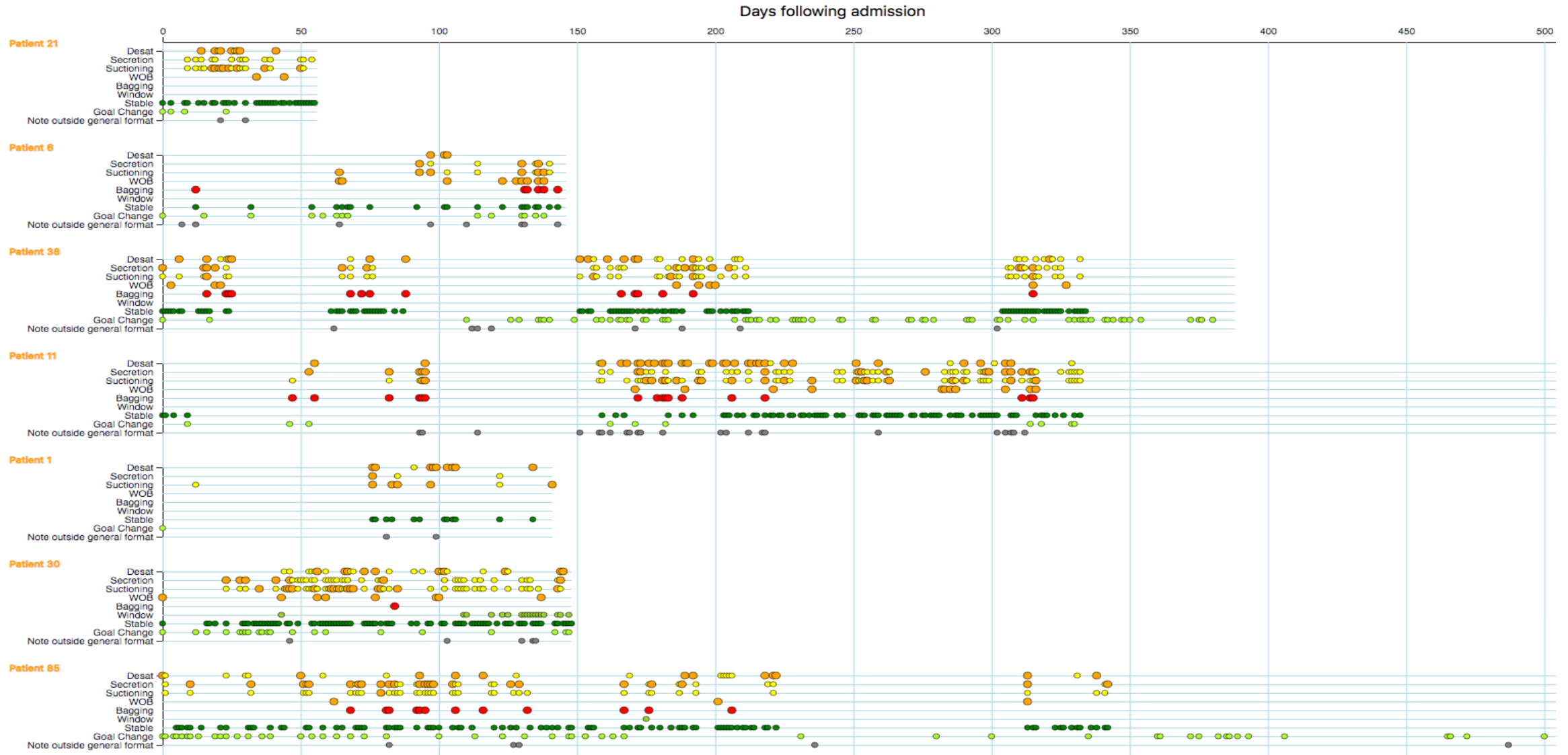
| Tier 1                   | Tier 2                 |
|--------------------------|------------------------|
| differential             | if                     |
| etiology                 | likely                 |
| uncertain                | consult*               |
| unclear                  | could                  |
| possib*                  | may                    |
| consider                 | unknown                |
| vs                       | although               |
| abdominal/abdominal pain | suggest*               |
|                          | however                |
|                          | elevated               |
|                          | broad                  |
|                          | further workup/work up |

### Uncertain Case

Unclear diagnosis at this time, but differential would include post-viral gastroparesis/ileus, although severe intermittent abdominal pain would not be consistent with that diagnosis. Could have intermittent intussusception or volvulus with a lead point of an enlarged lymph node in the setting of recent viral gastroenteritis. Renal colic is a possibility with the description of “writhing” in pain, but pain is not localized to the back or flanks and no blood of other abnormality seen on UA. Biliary colic could be considered, although would be unusual in her age range and without associated with food. Appendicitis remains on the differential, as was not visualized on ultrasound, but exam findings not consistent with the diagnosis.

# Using Linguistic Trends to Improve Patient Outcomes

## Vent Notes Prototype



# Human Expression and the Social Determinants of Health

Term : Adverse\_childhood\_experiences

# Topics : 10

tfidf : False

Passes : 20

## Documents

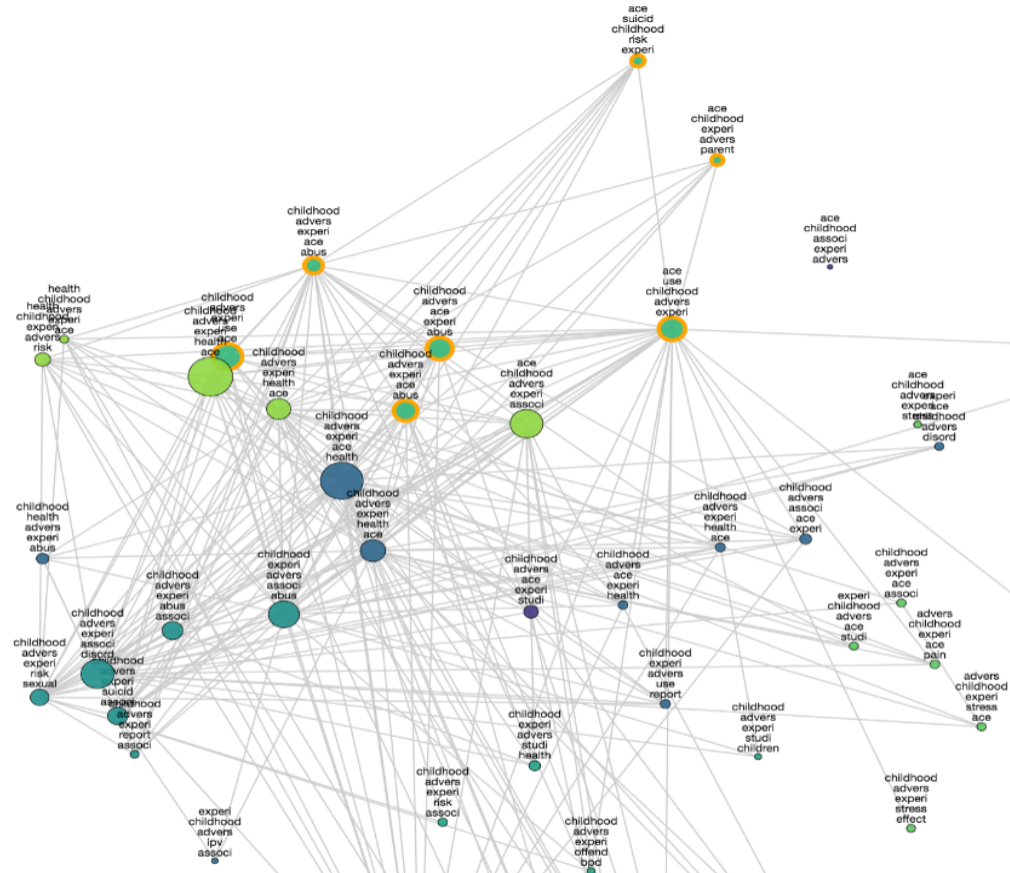
| Title  | Score              |
|--|--------------------|
| Growing up with parental alcohol abuse: exposure to childhood abuse, neglect, and household dysfunt  | 4.791228473186493  |
| Child maltreatment; parent alcohol and drug-related problems; polydrug problems; and parenting pract | 4.61366169154644   |
| Adverse childhood experiences and personal alcohol abuse as an adult.                                | 4.612937569618225  |
| DOES FATHERS' PRENATAL MENTAL HEALTH BEAR A RELATIONSHIP TO PARENTING STRESS AT 6 MONTHS?            | 4.468125253915787  |
| Adverse childhood experiences; alcoholic parents; and later risk of alcoholism and depression.       | 4.165758430026472  |
| Adverse childhood experiences and frequent headaches in adults.                                      | 3.9799956157803535 |
| Associations between adverse childhood experiences; psychological distress; and adult alcohol proble | 3.9198680222034454 |
| Adverse childhood experiences and the association with ever using alcohol and                        | 3.9084124863147736 |

Tree Circle Network

KL limit:0.39



Circle – Paragraph Level/Square – Article Level



## Clusters

| cluster | #para, #docs | # topics | terms                                  |
|---------|--------------|----------|--|
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| 8       | 1142,0       | 8        | ace childhood experi advers abus       |
| 9       | 425,0        | 8        | advers ace childhood experi stress     |
| 2       | 345,0        | 6        | ace childhood advers experi use        |
| 6       | 812,0        | 6        | childhood advers experi associ abus    |
| 11      | 206,0        | 6        | childhood advers experi ace use        |
| 10      | 783,0        | 5        | childhood advers experi health ace     |
| 5       | 555,0        | 4        | ace health childhood experi advers     |
| 7       | 269,0        | 4        | childhood experi advers studi health   |
| 0       | 20,0         | 1        | sexual report childhood telomer advers |
| 1       | 22,0         | 1        | use ace drug alcohol initi             |
| 3       | 10,0         | 1        | hpa cortisol matern symptom relat      |
| 12      | 63,0         | 1        | resili children ace advers childhood   |

## Relative impact of adverse events and screened symptoms of posttraumatic stress disorder and depression among active duty soldiers seeking mental health care.

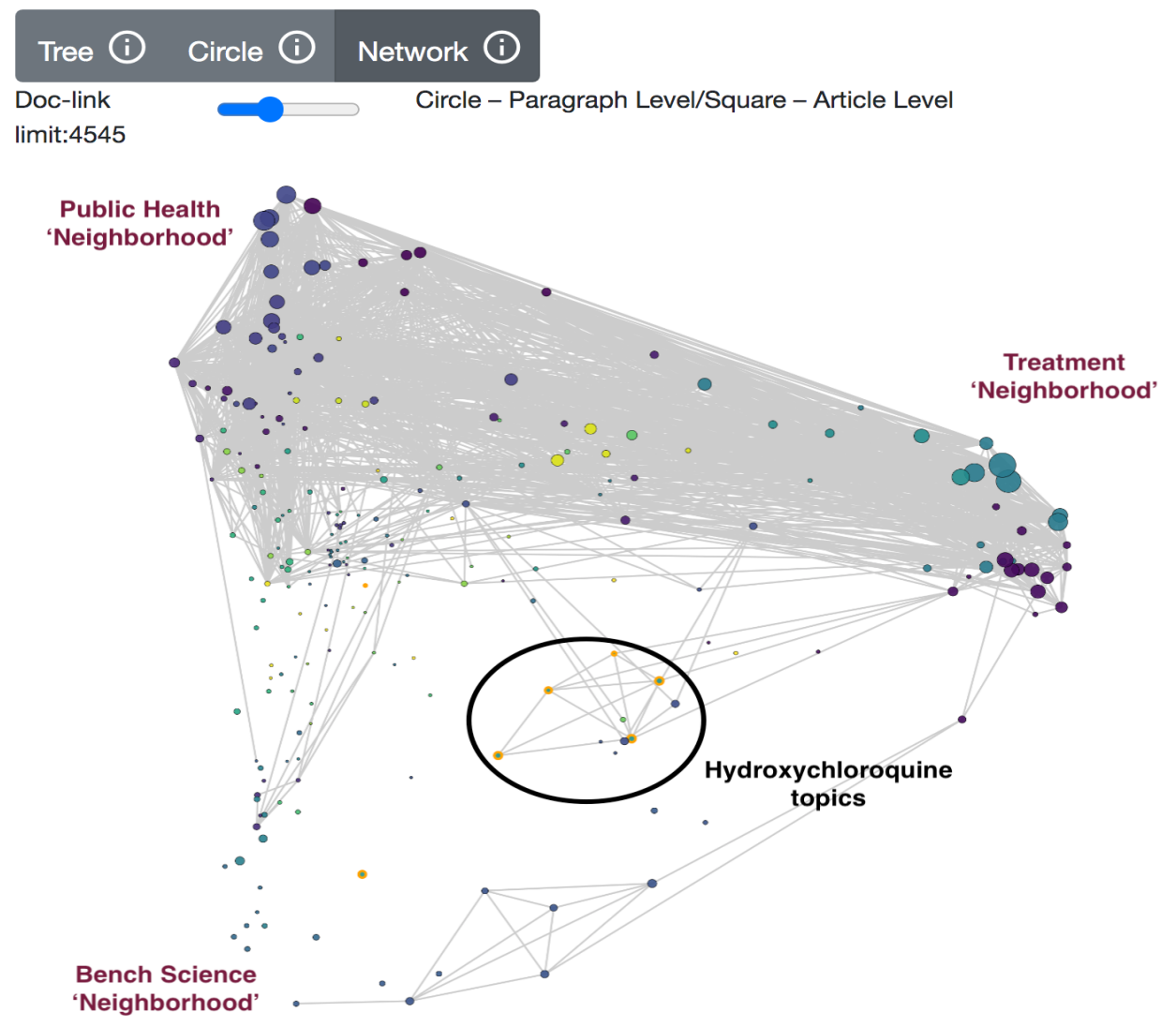
Symptoms of depression and posttraumatic stress are among the most studied psychological difficulties among soldiers. Such symptoms have been linked to a history of adverse events among both civilians and combat veterans. There is a paucity of research on this topic that can be applied to an active duty clinical population. Intake screening data were reviewed for 1,626 soldiers presenting to an outpatient mental health clinic to identify variables, including history of potentially traumatic experiences, associated with screened symptoms of posttraumatic stress disorder (PTSD) and depression. Demographics such as age, gender, and military rank, as well as number of adverse childhood experiences were significant predictors of screened PTSD and depression. A history of deployment to a combat zone predicted screened PTSD, but not depression. The role of childhood abuse as a risk factor is discussed and highlighted in the etiology of symptoms for soldiers seeking

Corpus : Covid      Term : severe\_acute\_respiratory\_syndrome\_coronavirus\_2-OR-covid-19-OR-sars\_cov\_2      # Docs : 31,818      # Topics : 40      Stopwords :      Years : -

Documents ⓘ

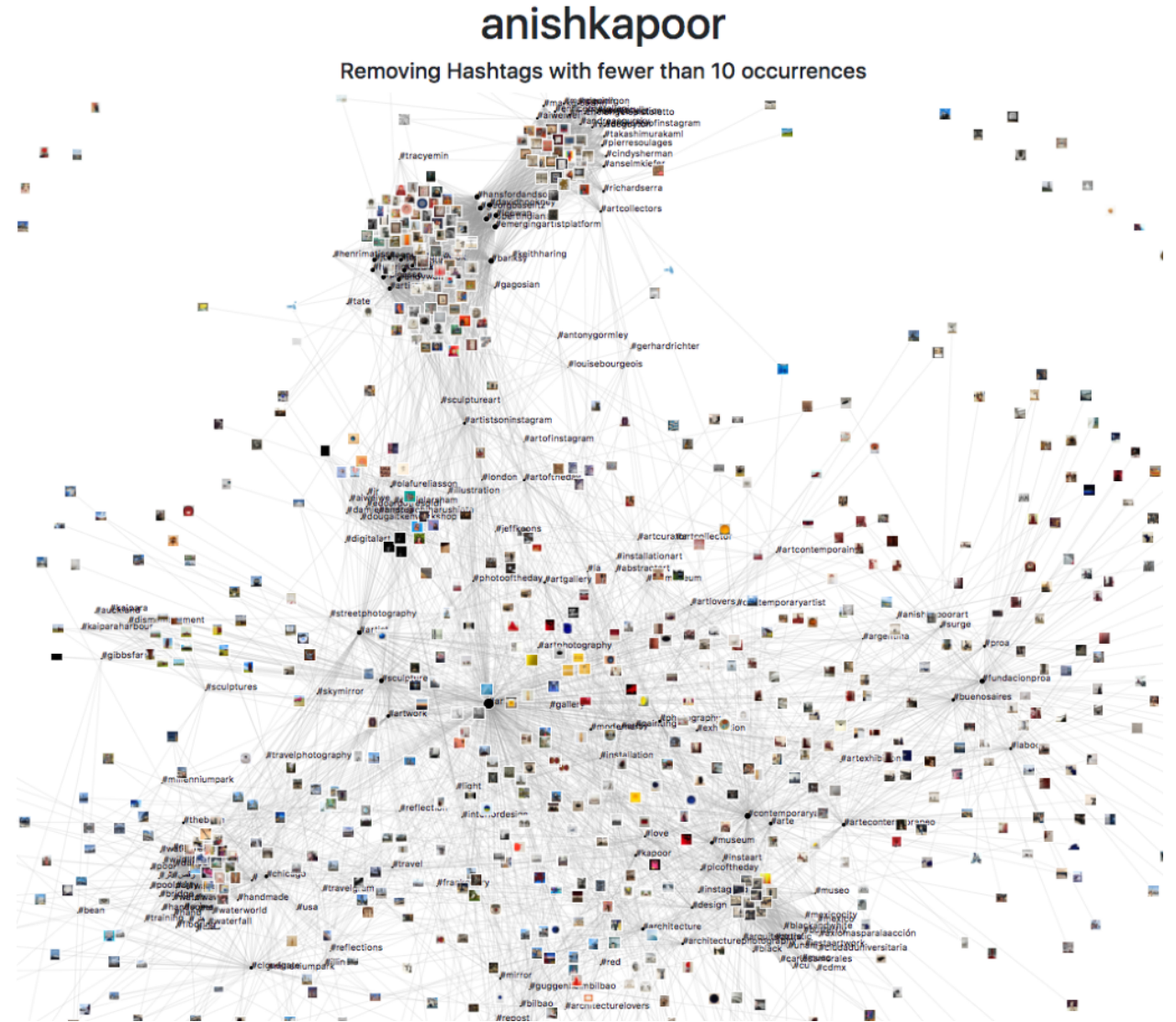
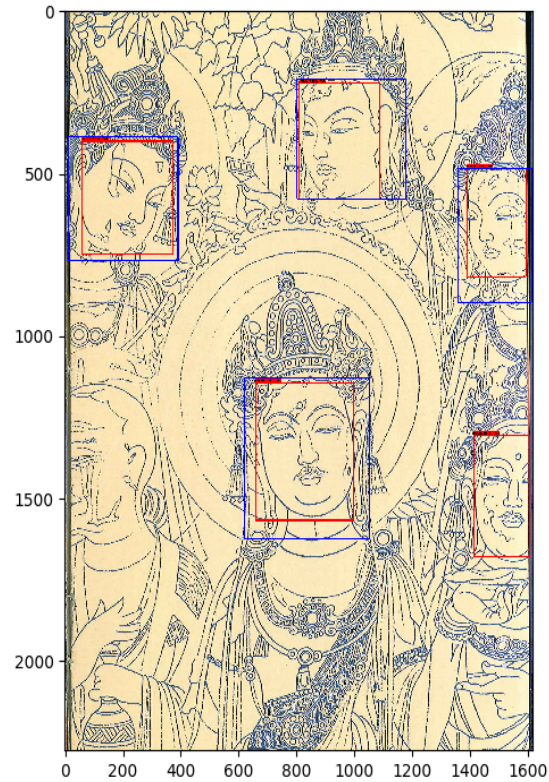
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|--|---------------|
| Response to  | 10.3162498474 |
| Combating Devastating COVID -19 by Drug Repurposing                              | 4.48861750960 |
| Favipiravir versus Arbidol for COVID-19: A Randomized Clinical Trial             | 3.87015905976 |
| Title: What do we know about remdesivir drug interactions?                       | 3.79288643598 |
|  | 3.75096887350 |
| Interferon beta-1a for COVID-19: critical importance of the administration route | 3.61853513121 |
| Chloroquine and hydroxychloroquine for COVID-19: A word of caution               | 3.58210587501 |

Doc View



| cluster | #para, #docs | # topics | terms  |
|---------|--------------|----------|--|
| 16      | 0,23418      | 15       | patients patient pandemic resources severe           |
| 1       | 0,18905      | 17       | patients severe study treatment cases                |
| 7       | 0,16931      | 13       | pandemic people crisis impact students               |
| 8       | 0,16355      | 10       | cases data infected model population                 |
| 0       | 0,8368       | 6        | cases china outbreak transmission infected           |
| 19      | 0,6891       | 7        | cases patients patient diagnosis images              |
| 10      | 0,4979       | 9        | patients increased il-6 inflammation cytokine        |
| 37      | 0,4670       | 5        | ppe patients risk patient masks                      |
| 3       | 0,4287       | 7        | positive patients negative samples days              |
| 20      | 0,4140       | 7        | treatment drugs patients efficacy hydroxychloroquine |

# Second Phase: Multimodal ML



# Second Phase: Jupyter Notebooks Pipeline

## Organization of Data

What do you notice about the format of the data above?

Each sentence is already *tokenized* split into a series of word and punctuation strings, with whitespace removed. This saves a lot of time having to do this work ourselves, manually.

To start to organize our data, let's put these sentences into a pandas DataFrame, an object which has a format very similar to an Excel spreadsheet. We will first make two spreadsheets (one for news, and one for romance), and then combine them into one. We will also add the category each sentences came from, which will be our labels for each sentence and its associated feature representation (which we will build ourselves).

```
In [ ]: ndf = pd.DataFrame({'sentence': news_sent,  
                        'label': 'news'})  
rdf = pd.DataFrame({'sentence': romance_sent,  
                    'label': 'romance'})
```

```
In [ ]: # combining two spreadsheets into 1  
df = pd.concat([ndf, rdf])
```

Let's see what this DataFrame looks like

```
In [ ]: df
```

```
In [ ]: df.head()
```

## So how many texts are there of each type?

```
In [ ]: df['label'].value_counts()
```

## What if we want to visualize that information?

We first create a figure and axes on which to draw our charts using `plt.subplots()`. Each chart is one axes, and a figure can contain multiple charts. Our data is encapsulated in `df['label'].value_counts()`, which is itself a dataframe. We then tell the Pandas to visualize the dataframe as a bar chart using `.plot.bar(ax=ax, rot=0)`. The `ax` keyword tells Pandas which chart in the figure to plot, and the `rot` keyword controls the rotation of the x axis labels.

```
In [ ]: fig, ax = plt.subplots()  
_ = df['label'].value_counts().plot.bar(ax=ax, rot=0)  
fig.savefig("categories_counts.png", bbox_inches = 'tight', pad_inches = 0)
```

We have slightly more news texts than romance texts, which we should keep in mind as we go ahead with classification.

## Extracting Features

### Defining Features

What should we use as features for the dataset? What did we use for the fruit example before?

| Object | Height | Width | Color | Mass | Round? |
|--------|--------|-------|-------|------|--------|
| Apple  | 6cm    | 7cm   | Red   | 330g | TRUE   |

# Let's Work Together

- Model of models platform: <https://modelofmodels.io>
- DSC website: <http://dsc.uc.edu>

## Second Grant Objectives:

- 15 Subgrants through Mellon Foundation Award using our technologies.
- Expand use cases and projects deploying MoM and Jupyter library.
- Experimental use of MoM for data services.
- Partnerships with external collaborators engaged in digital scholarship.

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