

Data Analytics in Support of Effective Workforce Training

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Indiana University, Bloomington, IN, USA



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Data Analytics in support of effective Workforce Training/Katy Börner

Speakers



Katy Börner

Data Analytics, Indiana University

KATY BÖRNER is the Victor H. Yngve Distinguished Professor of Engineering and Information Science in the Departments of Intelligent Systems Engineering and Information Science, School of Informatics, Computing, and Engineering; core faculty of the Cognitive Science Program; and founding... [Read More](#) →

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Virtual

● [Advanced Analytics](#)

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Overview

Identify and Overcome Skill Discrepancies

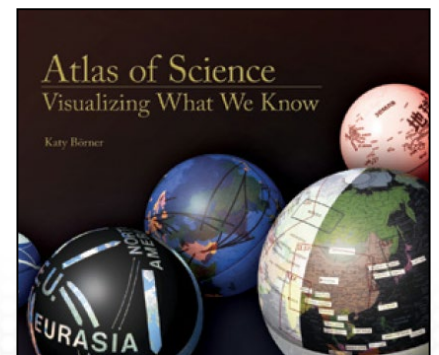
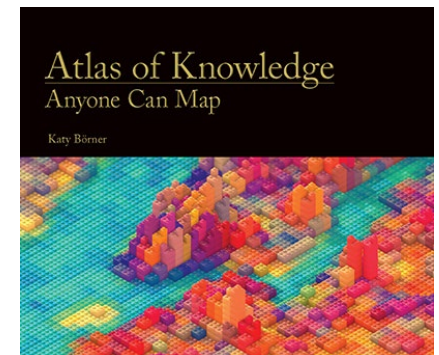
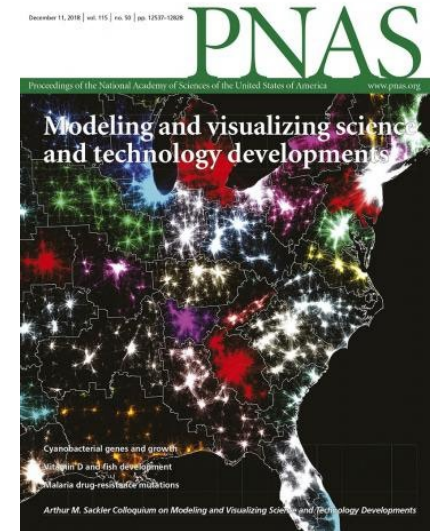
- Börner, Katy, Olga Scrivner, Michael Gallant, Shutian Ma, Xiaozhong Liu, Keith Chewning, Lingfei Wu, and James Evans. 2018. 2018. "[Skill Discrepancies Between Research, Education, and Jobs Reveal the Critical Need to Supply Soft Skills for the Data Economy](#)". *PNAS* 115 (50): 12630-12637. doi: 10.1073/pnas.1804247115.

Job Postings in The Substance Use Disorder Treatment Sector

- Scrivner, Olga, Thuy Nguyen, Kosali Simon, Esmé Middaugh, Bledi Taska, and Katy Börner. 2020. "[Job postings in the substance use disorder treatment related sector during the first five years of Medicaid expansion](#)". *PLOS One* 15 (1): e0228394. doi: 10.1371/journal.pone.0228394.

Empower Yourself and Others! Take the Visual Analytics Course: (<https://visanalytics.cns.iu.edu>)

- Börner, Katy, Andreas Bueckle, and Michael Ginda. 2019. [Data visualization literacy: Definitions, conceptual frameworks, exercises, and assessments](#). *PNAS*, 116 (6) 1857-1864.
- Börner, Katy. 2015. [Atlas of Knowledge: Anyone Can Map](#). Cambridge, MA: The MIT Press.
- Börner, Katy. 2010. [Atlas of Science: Visualizing What We Know](#). Cambridge, MA: The MIT Press.



Map of Scientific Collaborations from 2005-2009

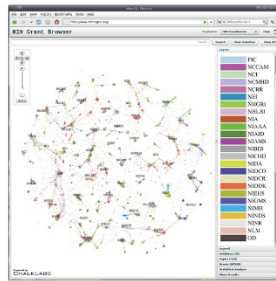


Computed Using Data from Elsevier's Scopus

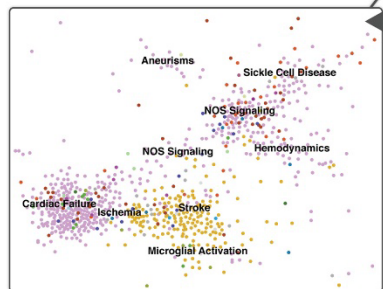
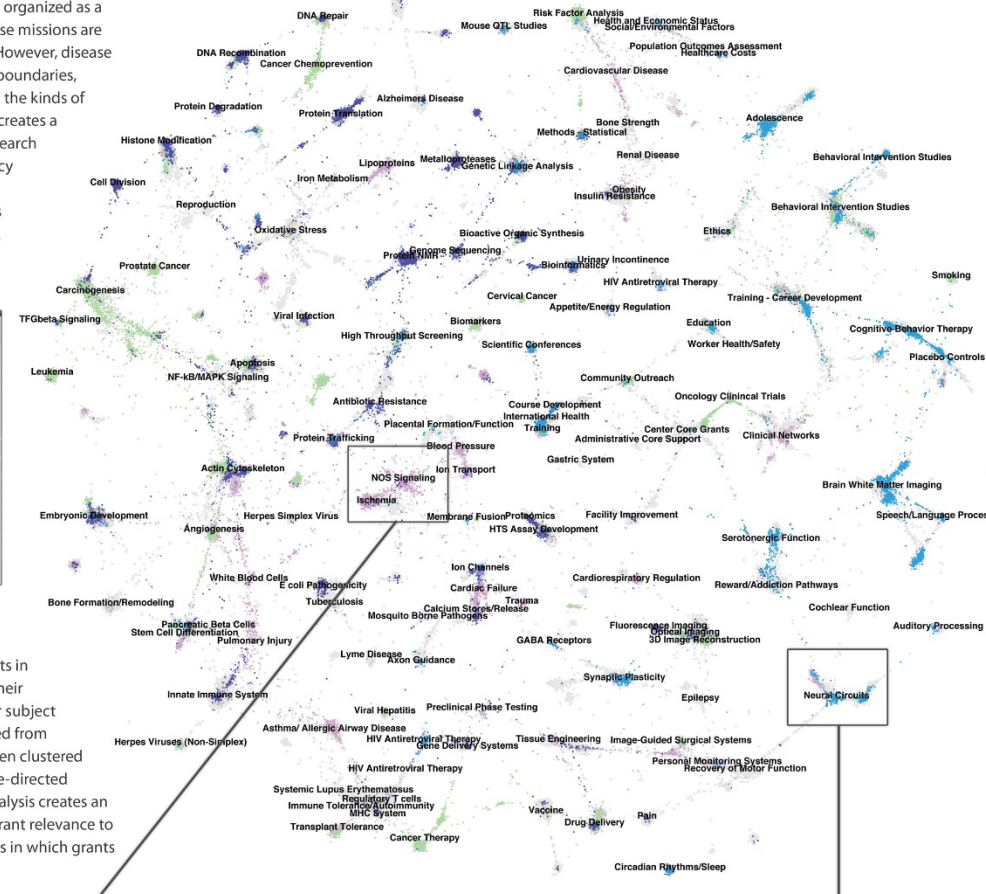
A Topic Map of NIH Grants 2007

Bruce W. Herr II (Chalklabs & IU), Gully Burns (ISI), David Newman (UCI), Edmund Talley (NIH)

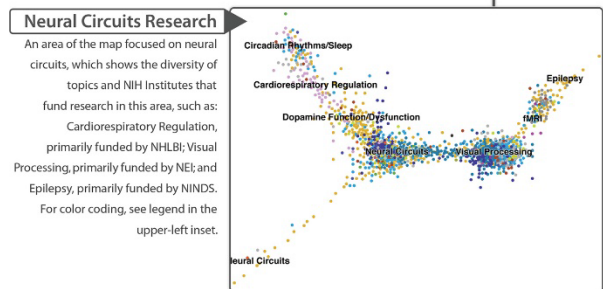
The National Institutes of Health (NIH) is organized as a multitude of Institutes and Centers whose missions are primarily focused on distinct diseases. However, disease etiologies and therapies flout scientific boundaries, and thus there is tremendous overlap in the kinds of research funded by each Institute. This creates a daunting landscape for decisions on research directions, funding allocations, and policy formulations. Shown here is devised an interactive topic map for navigating this landscape, online at www.nihmaps.org. Institute abbreviations can be found at www.nih.gov/icd.



Topic modeling, a statistical technique that automatically learns semantic categories, was applied to assess projects in terms used by researchers to describe their work, without the biases of keywords or subject headings. Grant similarities were derived from their topic mixtures, and grants were then clustered on a two-dimensional map using a force-directed simulated annealing algorithm. This analysis creates an interactive environment for assessing grant relevance to research categories and to NIH Institutes in which grants are localized.



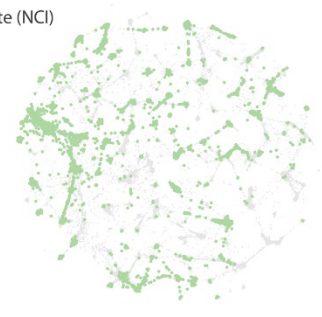
Cardiac Diseases Research
An area of the map focused on cardiovascular function and dysfunction. Cardiac Failure (primarily funded by NHLBI) is typically clustered next to Stroke (NINDS), since these are the two major medical emergencies associated with ischemia, which results from a restricted blood supply. Also localized in this area are grants focused on Nitric Oxide (NOS) Signaling, a major biochemical pathway for vasodilation, and grants on Hemodynamics, Sickle Cell Disease, and Aneurysms.



Neural Circuits Research
An area of the map focused on neural circuits, which shows the diversity of topics and NIH Institutes that fund research in this area, such as: Cardiorespiratory Regulation, primarily funded by NHLBI; Visual Processing, primarily funded by NEI; and Epilepsy, primarily funded by NINDS. For color coding, see legend in the upper-left inset.

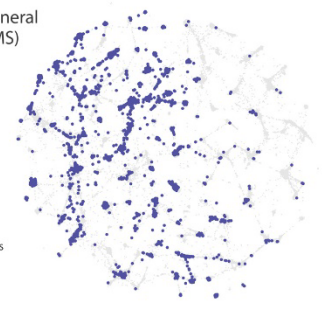
National Cancer Institute (NCI)

- TOP 10 TOPICS
- 1 Oncology Clinical Trials
 - 2 Cancer Treatment
 - 3 Cancer Therapy
 - 4 Carcinogenesis
 - 5 Risk Factor Analysis
 - 6 Cancer Chemotherapy
 - 7 Metastasis
 - 8 Leukemia
 - 9 Prediction/Prognosis
 - 10 Cancer Chemoprevention



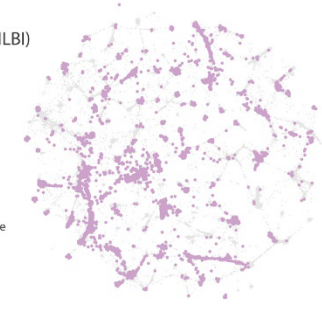
National Institute of General Medical Sciences (NIGMS)

- TOP 10 TOPICS
- 1 Bioactive Organic Synthesis
 - 2 X-ray Crystallography
 - 3 Protein NMR
 - 4 Computational Models
 - 5 Yeast Biology
 - 6 Metalloproteases
 - 7 Enzymatic Mechanisms
 - 8 Protein Complexes
 - 9 Invertebrate/Zebrafish Genetics
 - 10 Cell Division



National Heart, Lung, and Blood Institute (NHLBI)

- TOP 10 TOPICS
- 1 Cardiac Failure
 - 2 Pulmonary Injury
 - 3 Genetic Linkage Analysis
 - 4 Cardiovascular Disease
 - 5 Atherosclerosis
 - 6 Hemostasis
 - 7 Blood Pressure
 - 8 Asthma/ Allergic Airway Disease
 - 9 Gene Association
 - 10 Lipoproteins



National Institute of Mental Health (NIMH)

- TOP 10 TOPICS
- 1 Mood Disorders
 - 2 Schizophrenia
 - 3 Behavioral Intervention Studies
 - 4 Mental Health
 - 5 Depression
 - 6 Cognitive-Behavior Therapy
 - 7 AIDS Prevention
 - 8 Genetic Linkage Analysis
 - 9 Adolescence
 - 10 Childhood

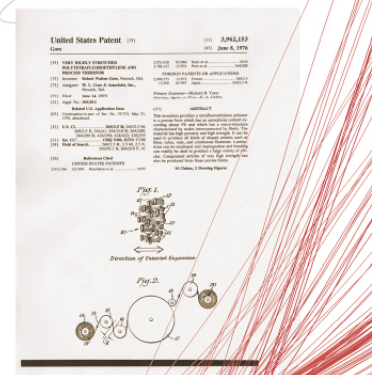
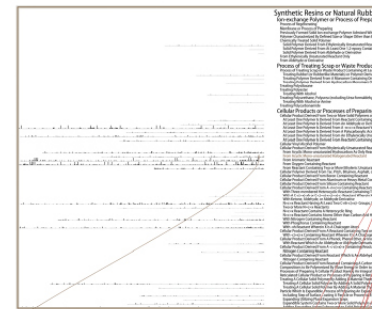


Impact

The United States Patent and Trademark Office does scientists and industry a great service by granting patents to protect inventions. Inventions are categorized in a taxonomy that groups patents by industry or use, proximate function, effect or product, and structure. At the time of this writing there are 160,523 categories in a hierarchy that goes 15 levels deep. We display the first three levels (13,529 categories) at right in what might be considered a textual map of inventions.

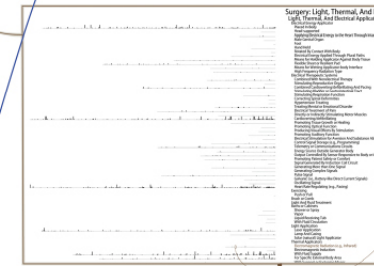
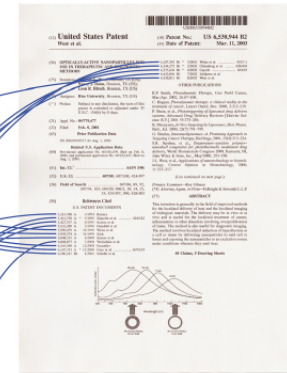
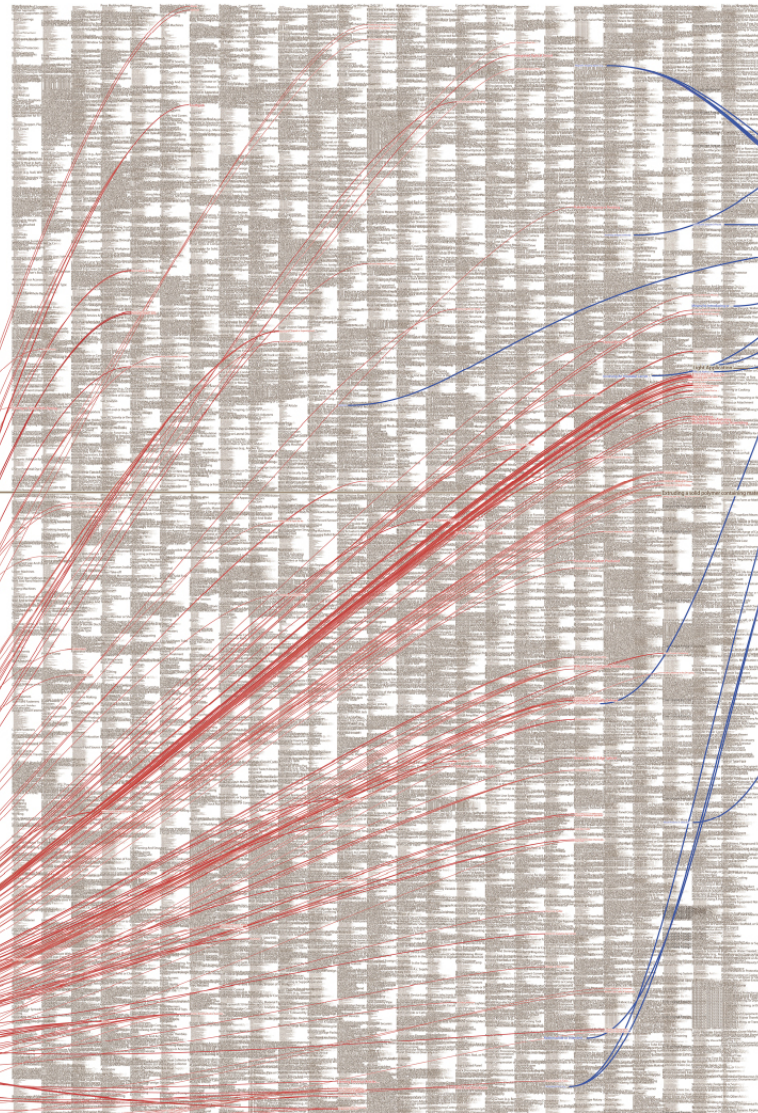
Patent applications are required to be unique and non-obvious, partially by revealing any previous patents that might be similar in nature or provide a foundation for the current invention. In this way we can trace the impact of a single patent, seeing how many patents and categories it affects.

The patent on Goretex—a lightweight, durable synthetic fiber—is an example of one that has had significant impact. The box below enlarges the section of the hierarchy where it is filed, and the red lines (arranged to start along a time line from 1981 to 2006) point to the 130 categories that contain 182 patents, from waterproof clothing to surgical cosmetic implants, that mention Goretex as "prior art."



The US Patent Hierarchy

Prior Art

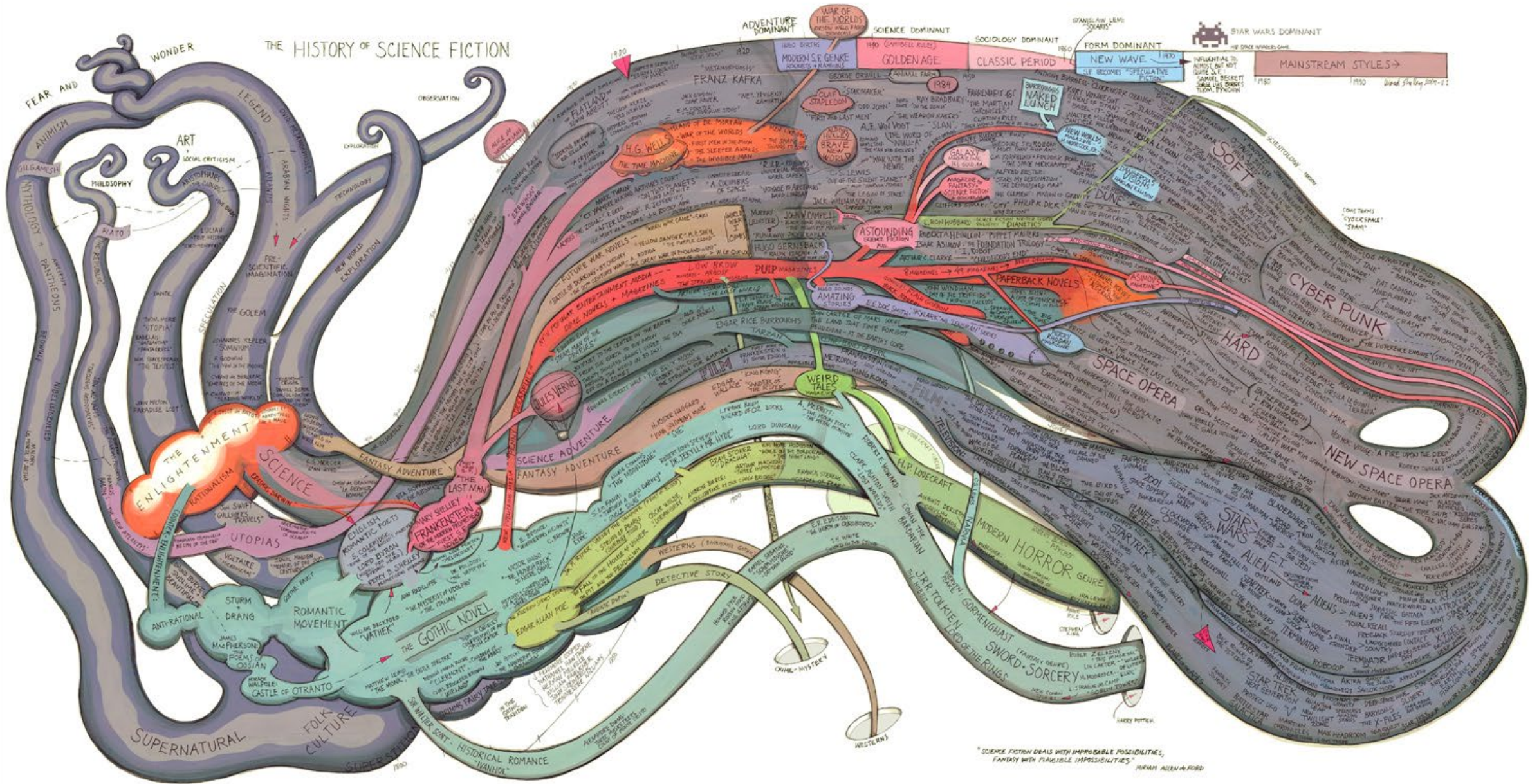


New patents often build on older ideas from many different categories. Here, blue lines originate in the sixteen categories that contain patents cited as prior art for a patent on "gold nanoshells." Gold nanoshells are a new invention: tiny gold spheres (with a diameter ten million times smaller than a human hair) that can be used to make tumors more visible in infrared scans; they have even helped cause complete remission of tumors in tests with laboratory mice. The blue lines show that widely separated categories provided background for this invention.

Keeping categories understandable is an important part of maintaining any taxonomy, including the patent hierarchy. Categories are easier to understand, search, and maintain if they contain elements that comfortably fit the definition of the category. The box above shows tiny bar charts, part of a *Taxonomy Validator* that reveals whether elements fit their categories. Categories may need to be redefined, and sometimes need to be split when they get too vague or large; a problem shared by many classification systems in this information-rich century. But how can we tell which ones to eliminate, add or revise—or how to revise them—in the complex, abstract sociolinguistic spaces we partition into ontologies?

Something as simple as a bar chart helps people see how entities in a category relate to that category. Here, each bar encodes a "distance to prototype": how much each patent differs from an idealized "prototype patent" for that category. A measure like this can be based on statistics, computational linguistics, or even human insight. Thus a category with mostly small bars is a good one, and a generally ragged one needs scrutiny or reorganization; but one that has only two or three tall bars may mean that only those few elements don't belong.

Even simple visuals can make thinking easier by providing better distilled data to the eye: vastly more data than working memory can hold as words. They focus people on exactly the right issues, and support them with the comprehensive overviews they need to make more informed judgements.



VII.10 History of Science Fiction - Ward Shelley - 2011

Check out our **Zoom Maps** online!

VII.10
History of Science Fiction, by Ward Shulman

BROOKLYN, NY, 2011
Courtesy of Ward Shulman Studio

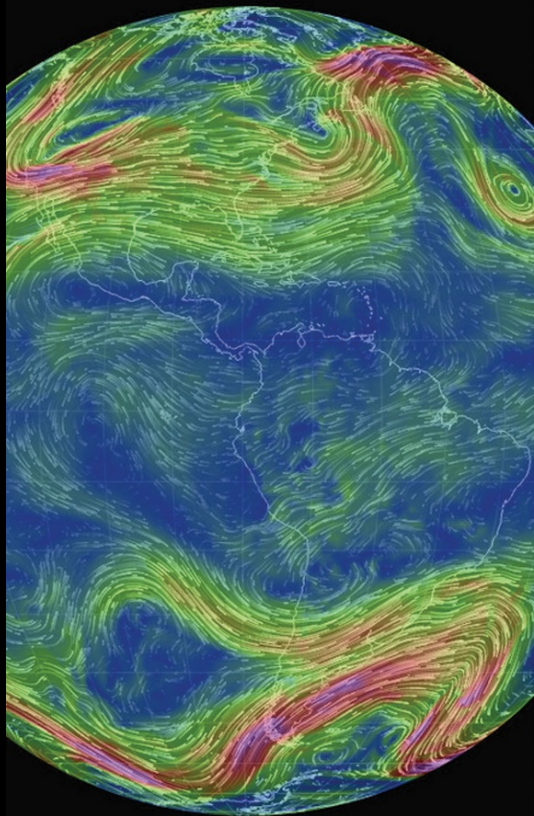
Ward Shulman is an artist identified with the Williamsburg scene in Brooklyn, New York. This map plots the science fiction literary genre from its nascent beginnings in the late 18th century, through the Victorian era, to the modernist and postmodernist eras. Emerging out of the data, here the narrative structure precedes and organizes the data. The map's structure is like a tree, tracing roots to pre-historical sources and whose body, the branches, are like a tree's canopy, showing the progression of the genre through a number of distinct periods, which are charted, citing hundreds of authors and works.

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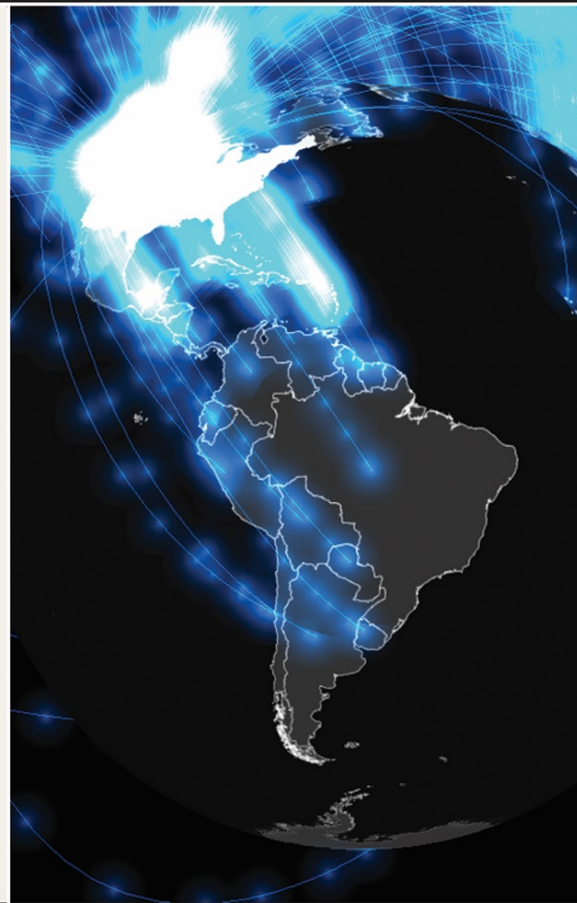
Earth

Weather on a worldwide scale



AcademyScope

Exploring the scientific landscape



Mapping Global Society

Local news from a global perspective

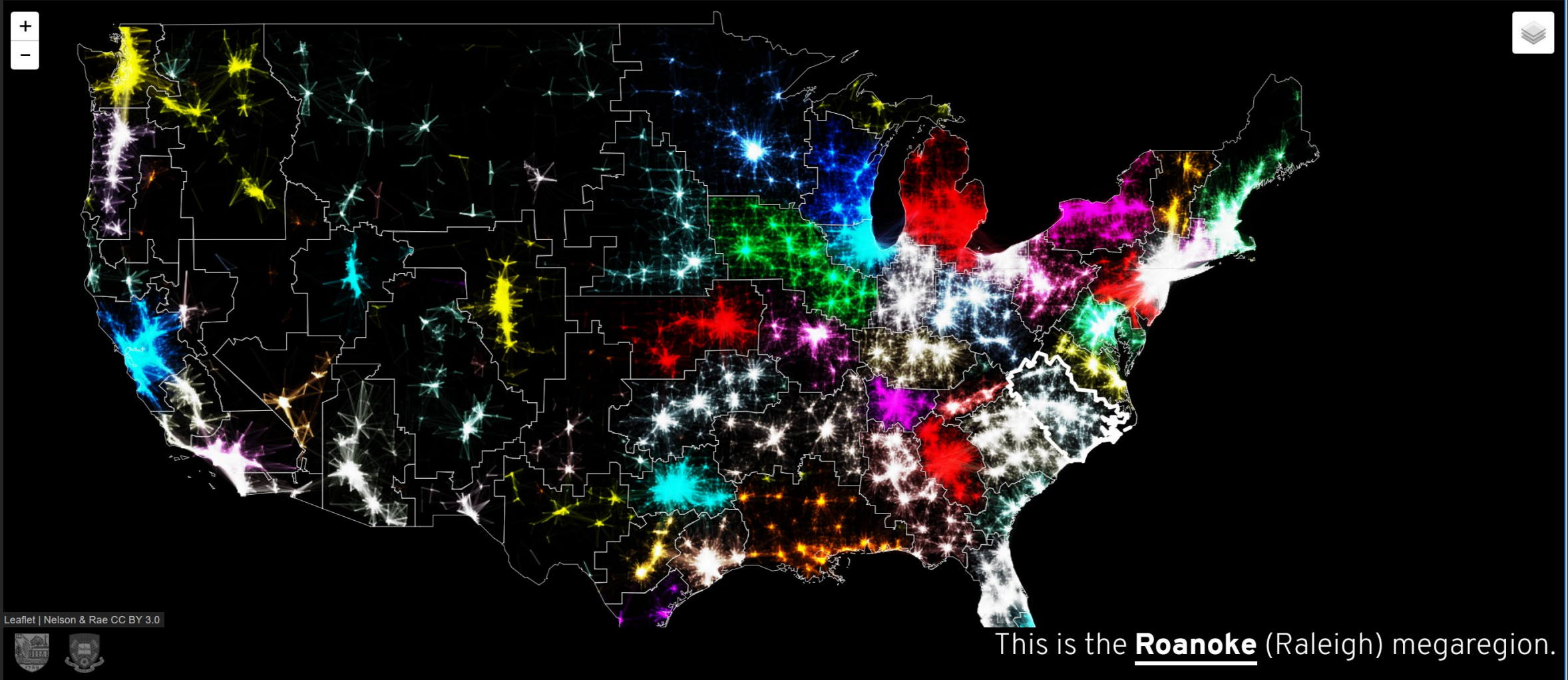


Charting Culture

2,600 years of human history in 5 minutes

THE MEGAREGIONS OF THE US

Explore the new geography of commuter connections in the US.
Tap to identify regions. Tap and hold to see a single location's commuted.





Oct 1-Dec 3, 2020: Exhibit on display at the Dimension Mill in Bloomington, IN on <https://dimensionmill.org>

Identify and Overcome Skill Discrepancies

Börner, Katy, Olga Scrivner, Michael Gallant, Shutian Ma, Xiaozhong Liu, Keith Chewning, Lingfei Wu, and James Evans. 2018. 2018. "[Skill Discrepancies Between Research, Education, and Jobs Reveal the Critical Need to Supply Soft Skills for the Data Economy](#)". *PNAS* 115 (50): 12630-12637. doi: 10.1073/pnas.1804247115.

See also <https://www.pnas.org/modeling>

Study the (mis)match and temporal dynamics of science and technology (S&T) progress, education and workforce development options, and job requirements.

Challenges:

- Rapid change of STEM knowledge
- Increase in tools, AI
- Social skills (project management, team leadership)
- Increasing team size

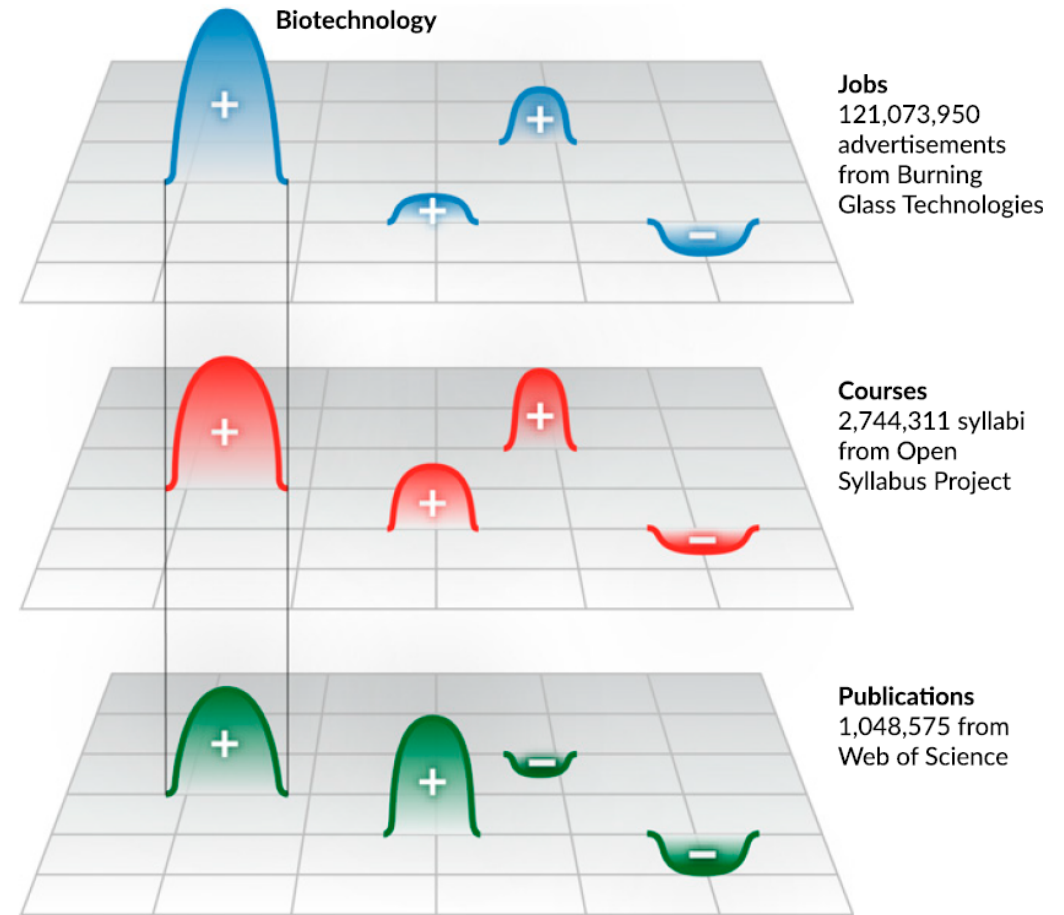
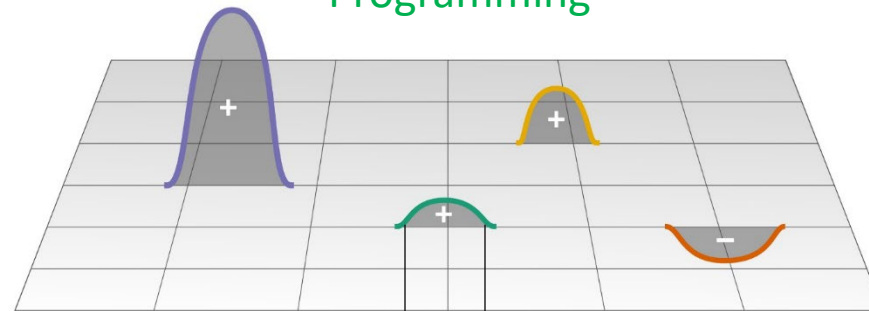
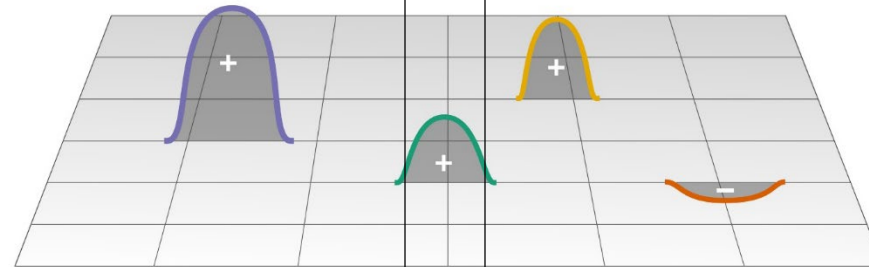


Fig. 1. The interplay of job market demands, educational course offerings, and progress in S&T as captured in publications. Color-coded mountains (+) and valleys (-) indicate different skill clusters. For example, skills related to Biotechnology might be mentioned frequently in job descriptions and taught in many courses, but they may not be as prevalent in academic publications. In other words, there are papers that mention these skills, but labor demand and commercial activity might be outstripping publication activity in this area. The numbers of jobs, courses, and publications that have skills associated and are used in this study are given on the right.

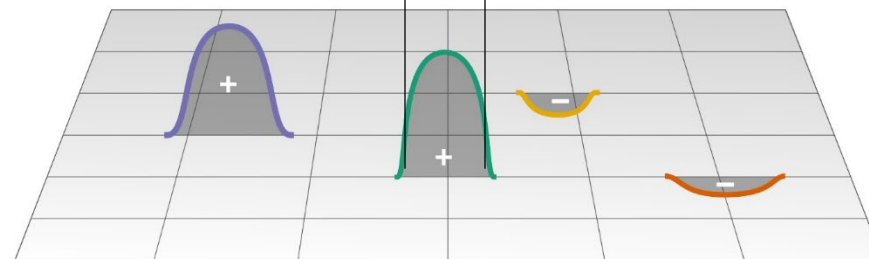
Programming



Jobs

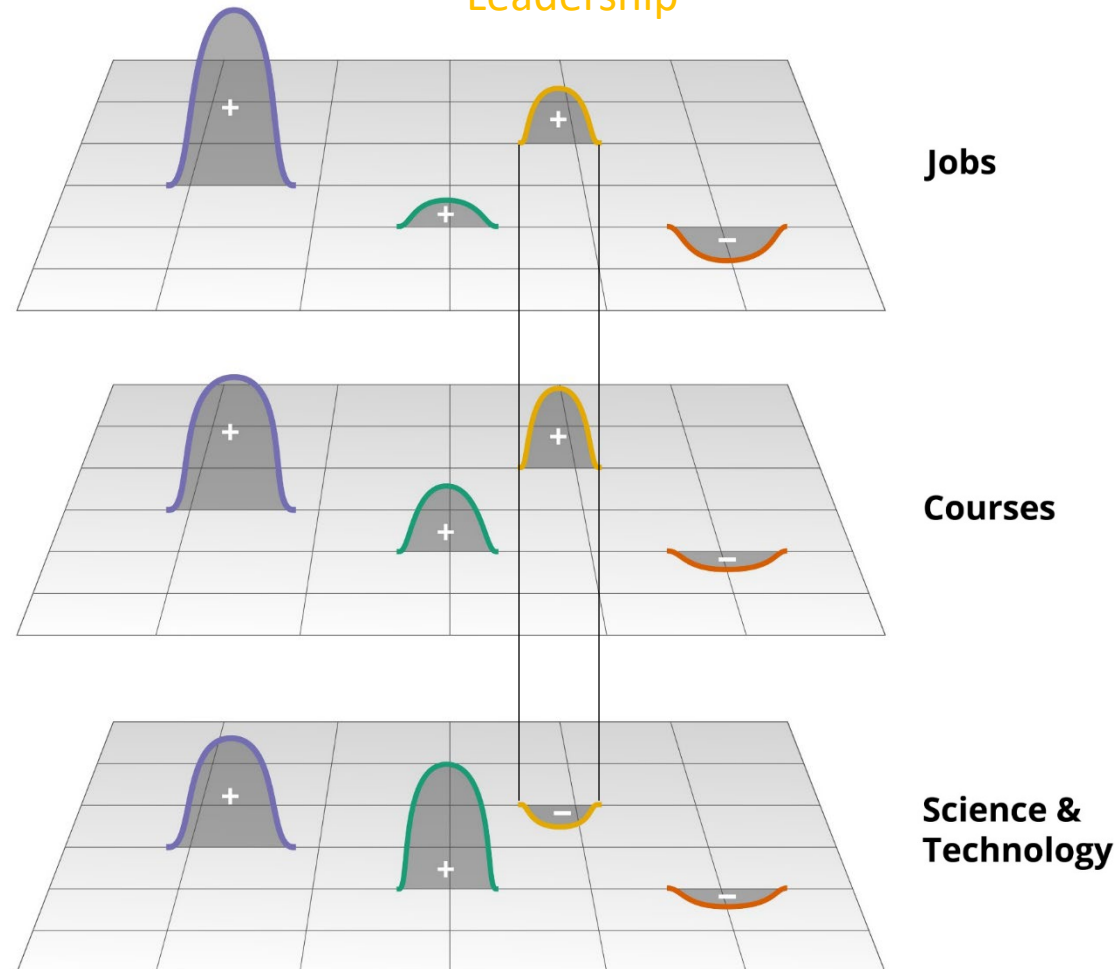


Courses



Science & Technology

Leadership

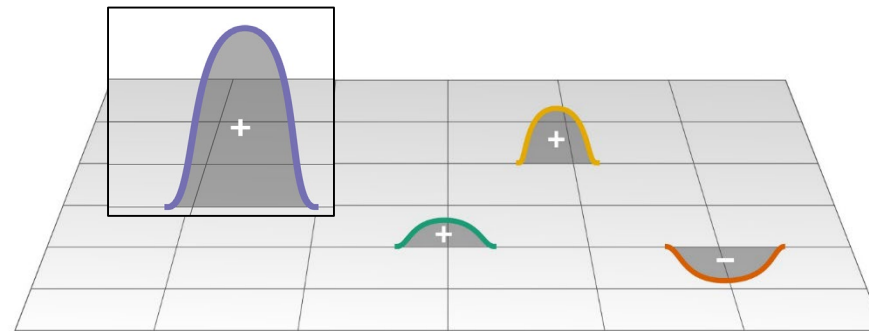


Jobs

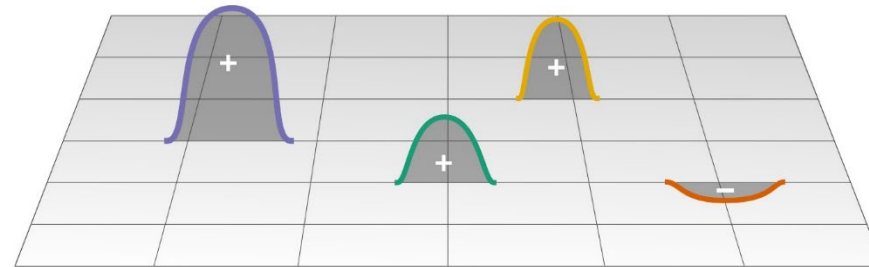
Courses

**Science &
Technology**

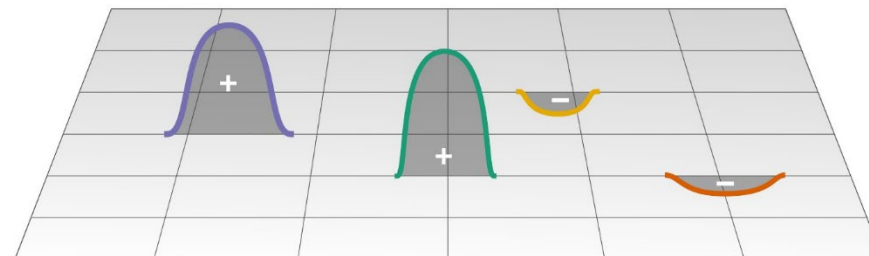
Biotechnology



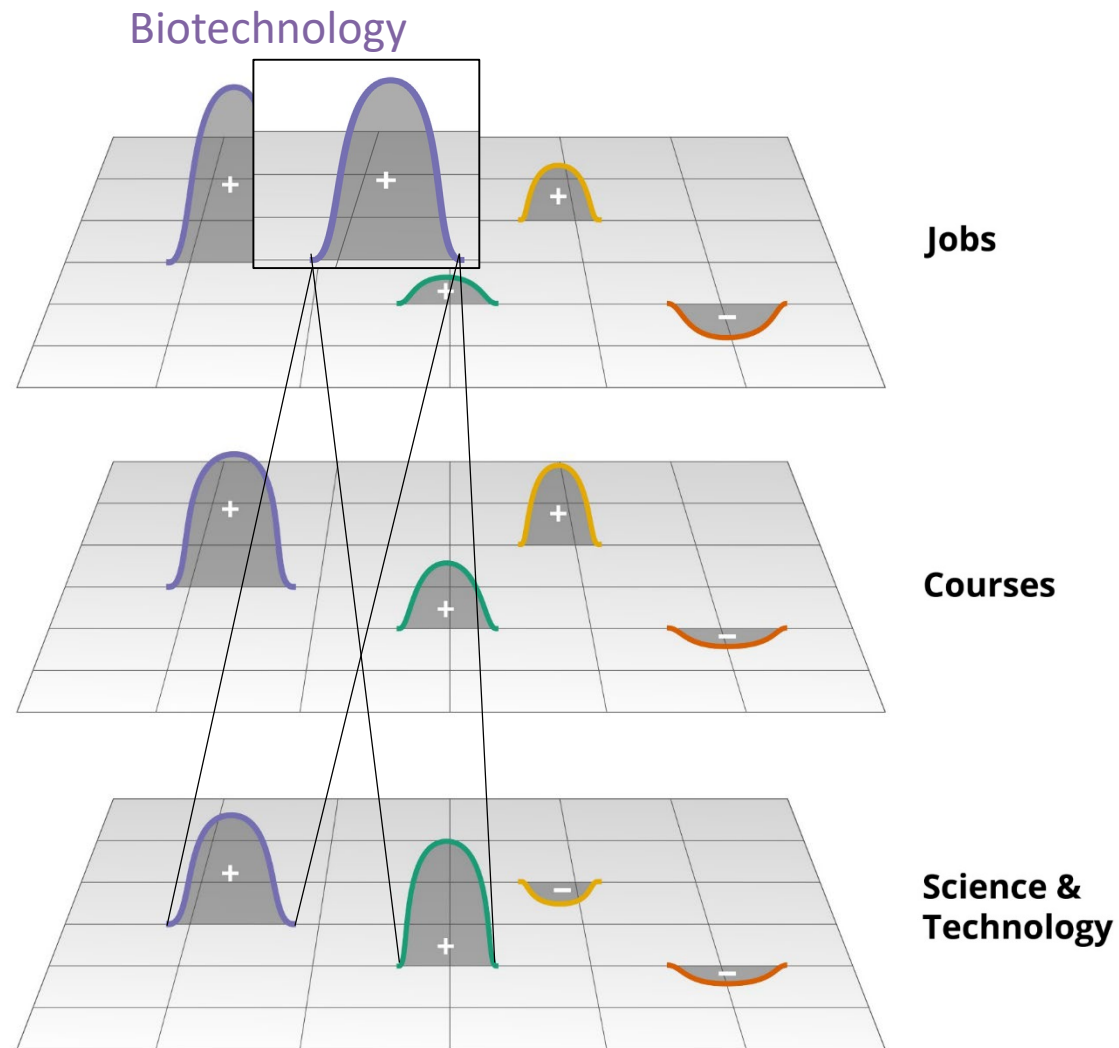
Jobs



Courses



**Science &
Technology**



Stakeholders and Insight Needs

- **Students:** What jobs will exist in 1-4 years? What program/learning trajectory is best to get/keep my dream job?
- **Teachers:** What course updates are needed? What balance of timely and timeless knowledge (to get a job vs. learn how to learn) should I teach? How to innovate in teaching and maintain job security or tenure?
- **Universities:** What programs should be created? What is my competition doing? How do I tailor programs to fit local needs?
- **Science Funders:** How can S&T investments improve short- and long-term prosperity? Where will advances in knowledge also yield advances in skills and technology?
- **Employers:** What skills are needed next year and in 5 and 10 years? Which institutions produce the right talent? What skills does my competition list in job advertisements?
- **Economic Developers:** What critical skills are needed to improve business retention, expansion, and recruitment in a region?

What is ROI of my time, money, compassion?

Urgency

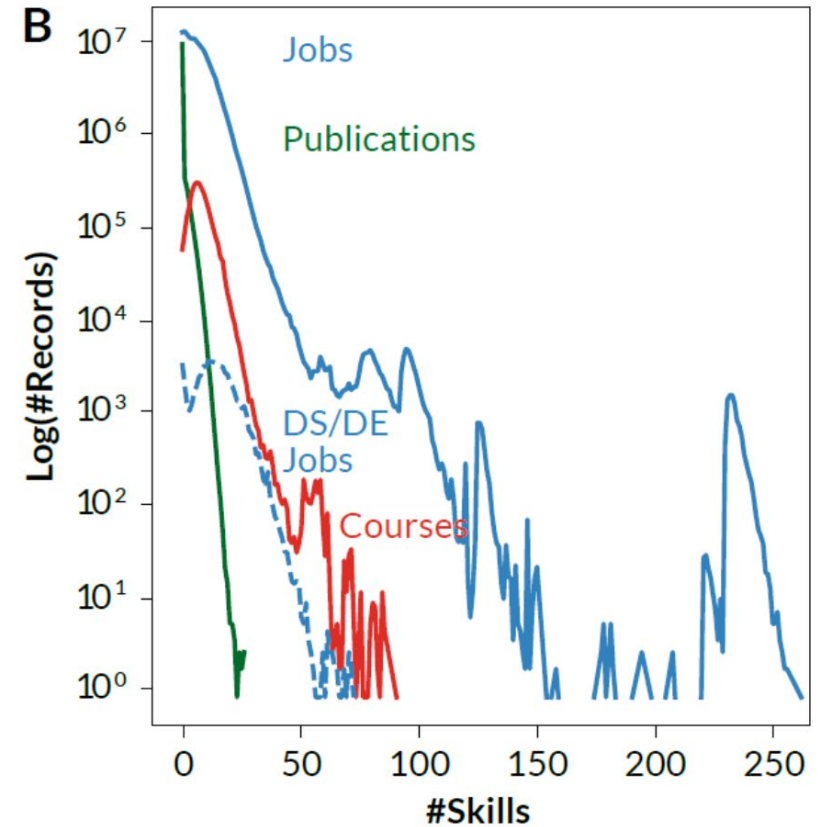
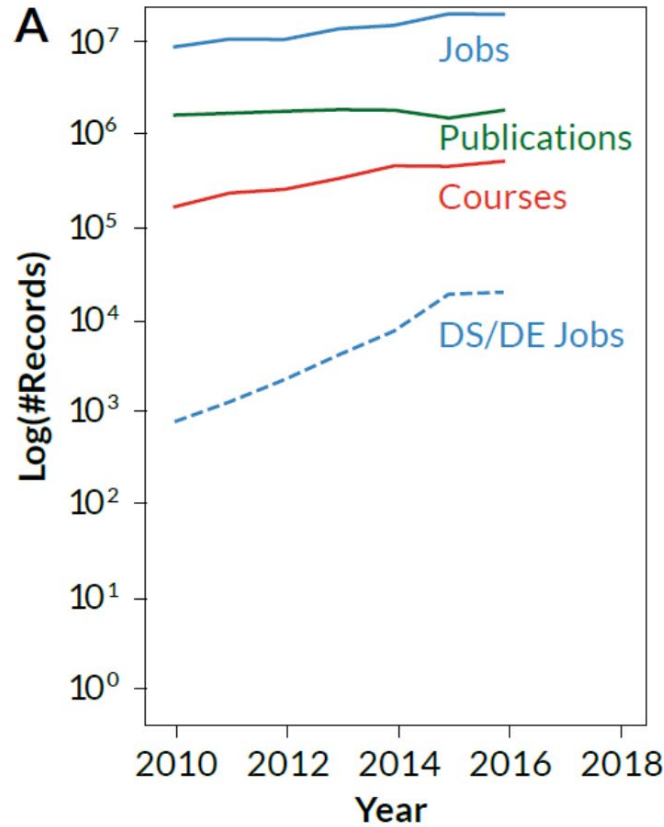
- 35% of UK jobs, and 30% in London, are at high risk from automation over the coming 20 years.
<https://www2.deloitte.com/content/dam/Deloitte/uk/Documents/uk-futures/london-futures-agiletown.pdf>
- The rise of artificial intelligence will lead to the displacement of **millions of blue-collar as well as white-collar jobs** in the coming decade.
Auerswald PE (2017) The Code Economy: A Forty-thousand-year History; Beyer D (2016) The future of machine intelligence: Perspectives from leading practitioners ; Brynjolfsson E, McAfee A (2014) The second machine age: Work, progress, and prosperity in a time of brilliant technologies; Ford M (2015) Rise of the Robots: Technology and the Threat of a Jobless Future.
- The pandemic is speeding up automation, and 85 million jobs are on the line.
<https://www.cnn.com/2020/10/20/business/wef-future-of-jobs-report/index.html>

Datasets Used

Job advertisements by Burning Glass posted between Jan 2010-Dec 2016.

Web of Science publications published Jan 2010-Dec 2016.

Course descriptions from the Open Syllabus Project acquired in June 2018 for courses offered in 2010-2016.



Data Type	#Records	#Records with skills	#Records without skills
All Courses	3,062,277	2,744,311	54,733
All Jobs	132,011,926	121,073,950	10,937,976
DSDE Jobs	69,405	65,944	3,461
All Publications	15,691,162	1,048,575	14,642,587
DSDE Publications	1,048,575	807,756	240,819

Fig. 2. Basemap of 13,218 skills. In this map, each dot is a skill, triangles identify skill clusters, and squares represent skill families from the Burning Glass (BG) taxonomy. Labels are given for all skill family nodes and for the largest skill cluster (NA) to indicate placement of relevant subtrees. Additionally, hard and soft skills are overlaid using purple and orange nodes, respectively; node area size coding indicates base 10 log of skill frequency in DS/DE jobs. Skill area computation uses Voronoi tessellation.

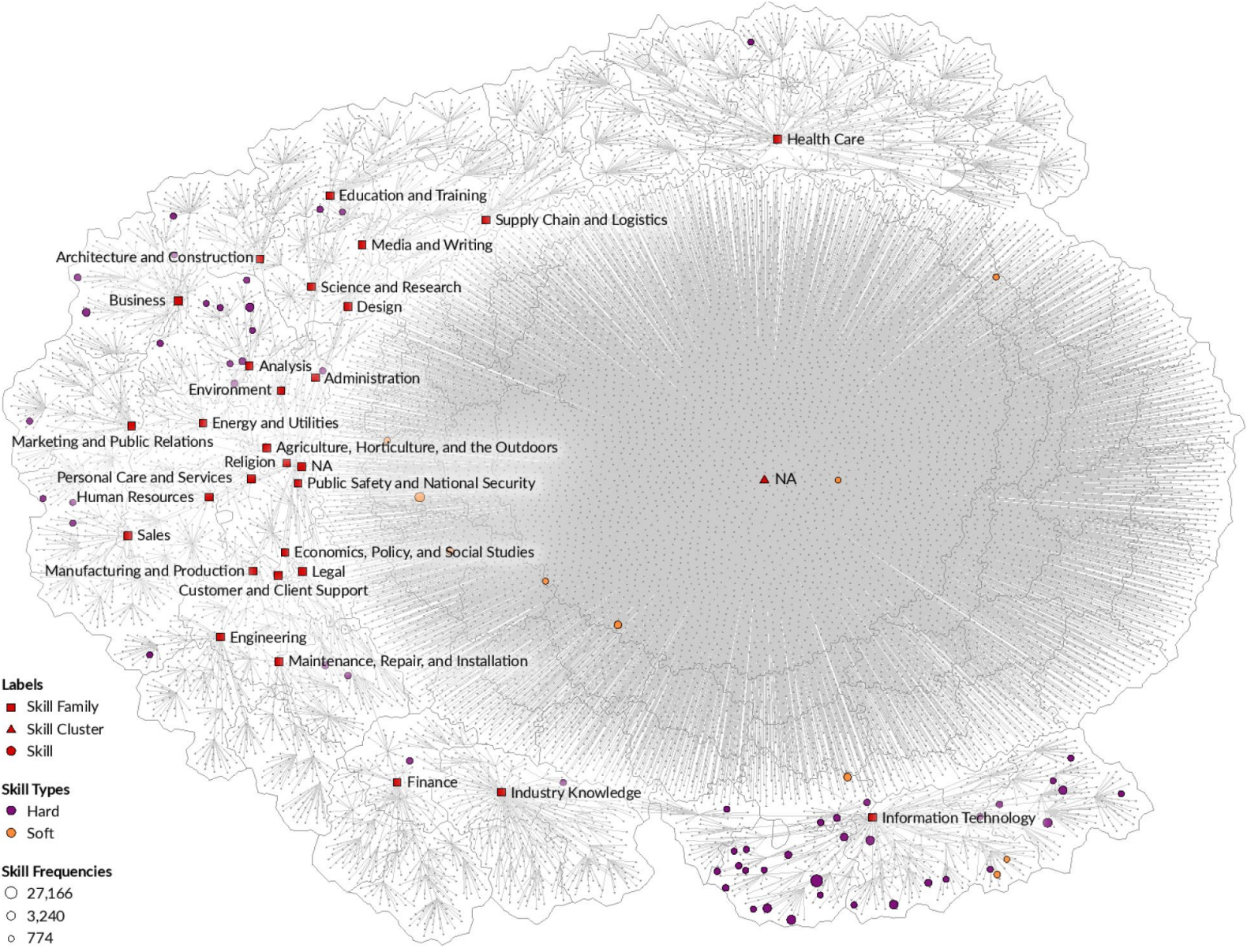
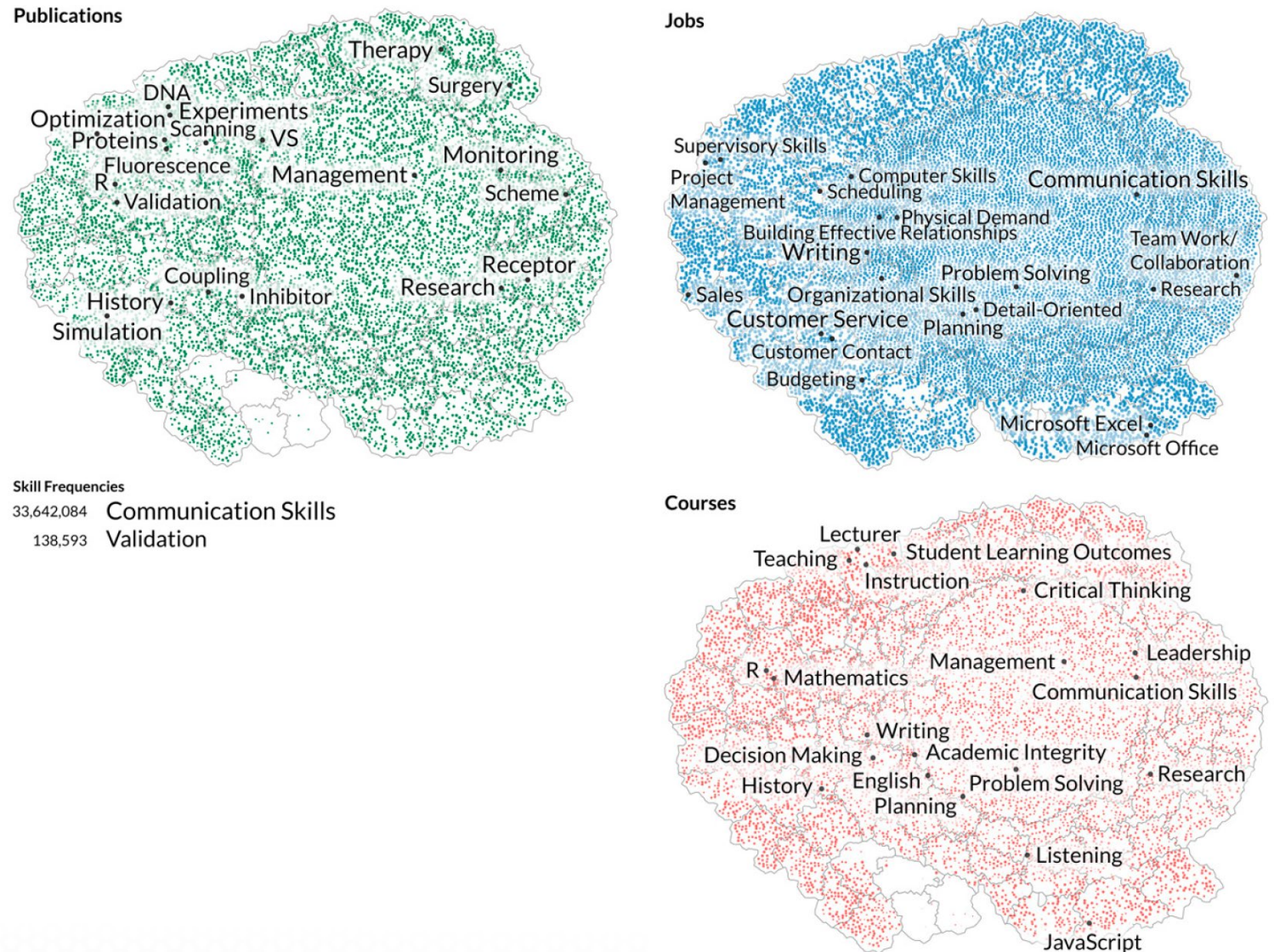


Fig. 3. Basemap of 13,218 skills with overlays of skill frequency in jobs, courses, and publications. This figure substantiates the conceptual drawing in Fig. 1 using millions of data records. Jobs skills are plotted in blue, courses are in red, and publications are in green. Node area size coding indicates base 10 log of skills frequency. The top 20 most frequent skills are labeled, and label sizes denote skill frequency.



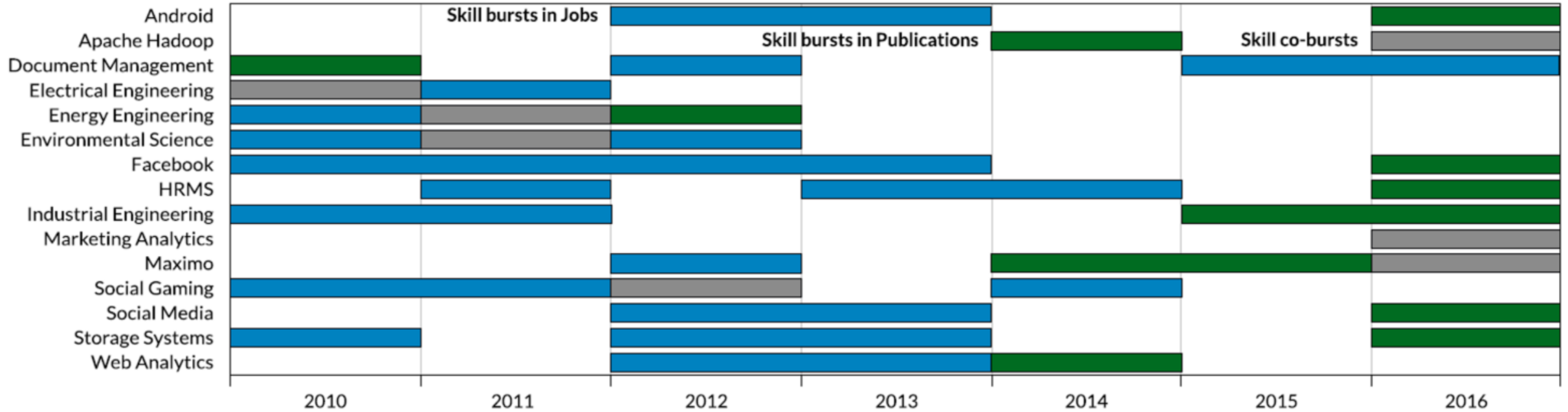


Fig. 4. Burst of activity in DS/DE skills in jobs and publications. Each burst is rendered as a horizontal bar with a start and an end date; skill term is shown on the left. Skills that burst in jobs are blue; skills bursting in publications are green. Seven skills burst in both datasets during the same years and are shown in gray. HRMS stands for human resources management system, and Maximo is an IBM system for managing physical assets.

Kullback-Leibler Divergence

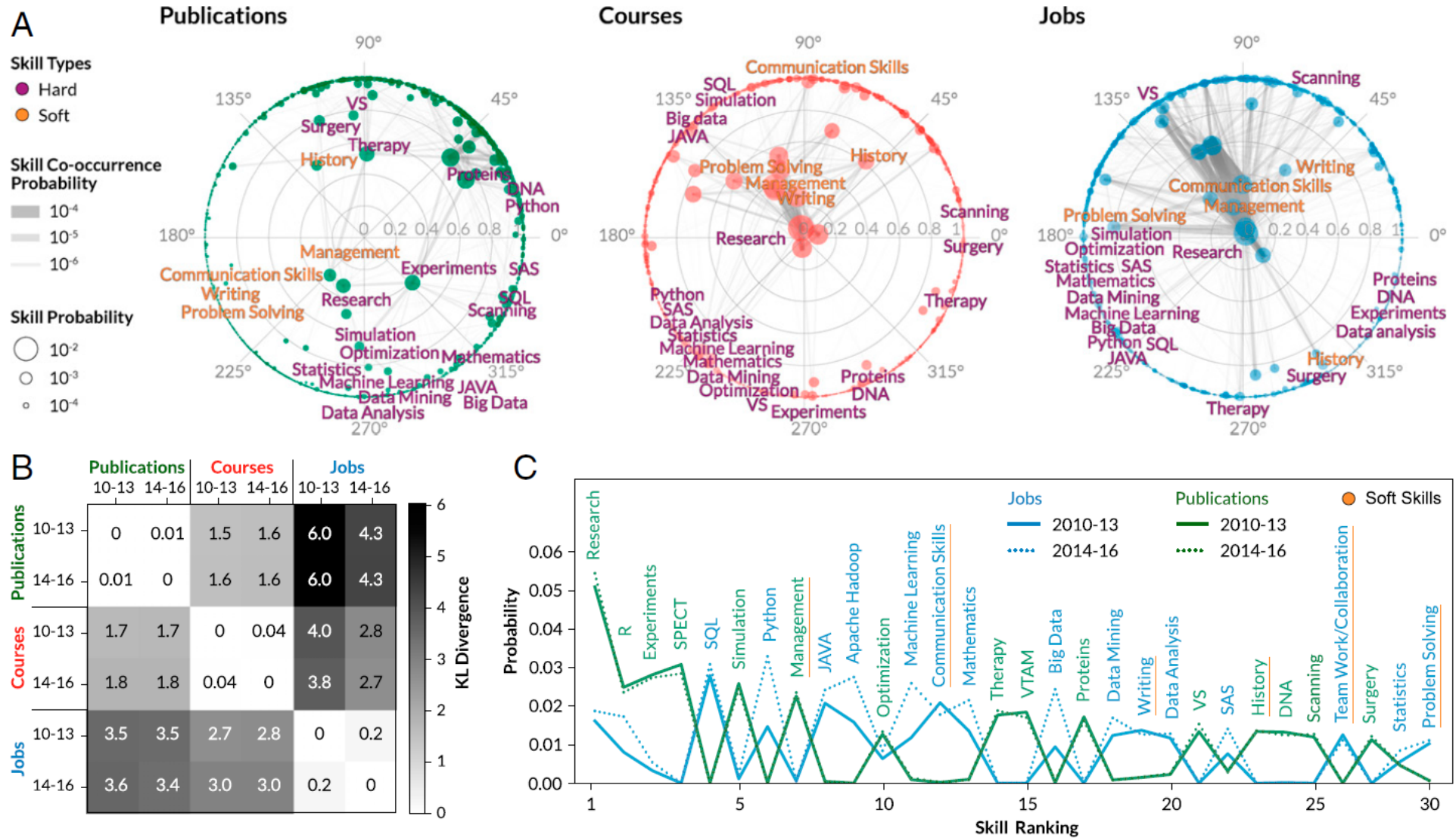


Fig. 5. Structural and dynamic differences between skill distributions in jobs, courses, and publications for 2010–2013 and 2014–2016. (A) Poincaré disks comparing the centrality of soft skills (orange) and hard skills (purple) across jobs, courses, and publications. (B) KL divergence matrix for jobs, courses, and publications in 2010–2013 and 2014–2016. (C) The most surprising skills in publications and jobs; *R* is a scripting language, VTAM refers to the IBM Virtual Telecommunication Access Method application, VS is the integrated development environment Visual Studio, and SAS is a data analytics software.

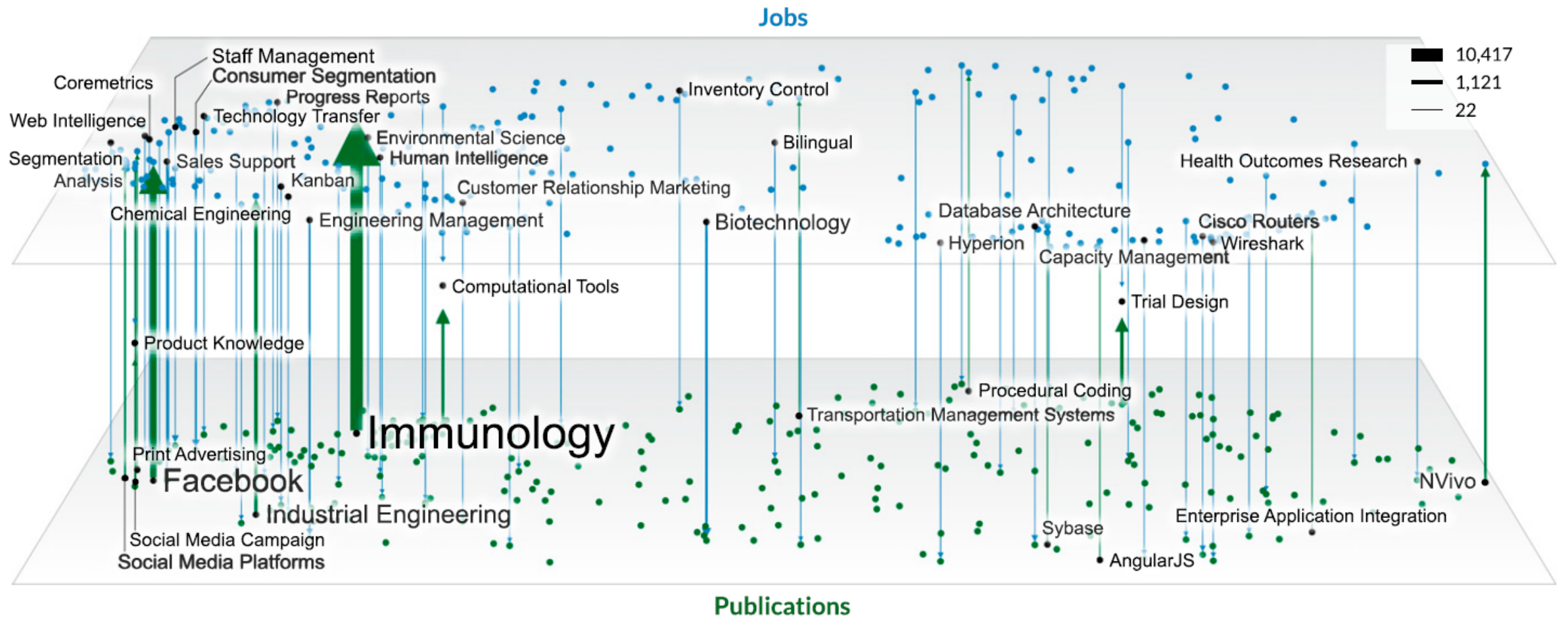


Fig. 6. Strength of influence mapping. Top 200 most frequent skills in jobs (blue) and in publications (green) plotted on the skills basemap from Fig. 2. Arrows represent skills with significant Granger causality (P value < 0.05). Line thickness and label size indicate skill frequency. The direction and thickness of each arrow indicate the F -value strength and direction.

Fig. 7. Multivariate Hawkes Process influence network of DS/DE skills within job advertisements 2010–2016. Each of the 45 nodes represents a top-frequency skill (29 soft and 16 hard skills) with a strong influence edge from/to other skill(s) in job advertisements between 2010 and 2016. Node and label size correspond to the number of times that the skill appeared in a job advertisement. Thickness of the 75 directed edges indicates influence strength.

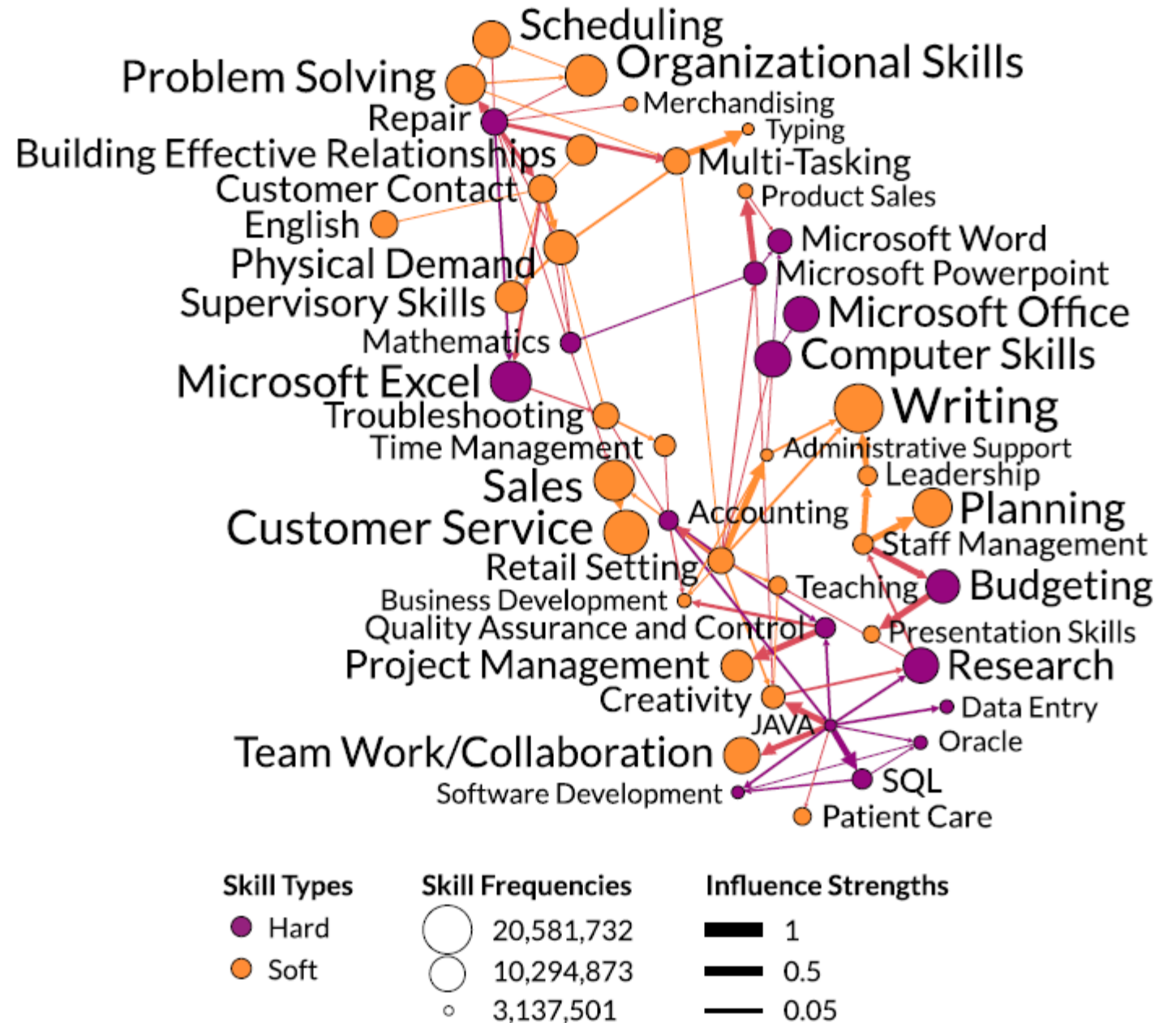
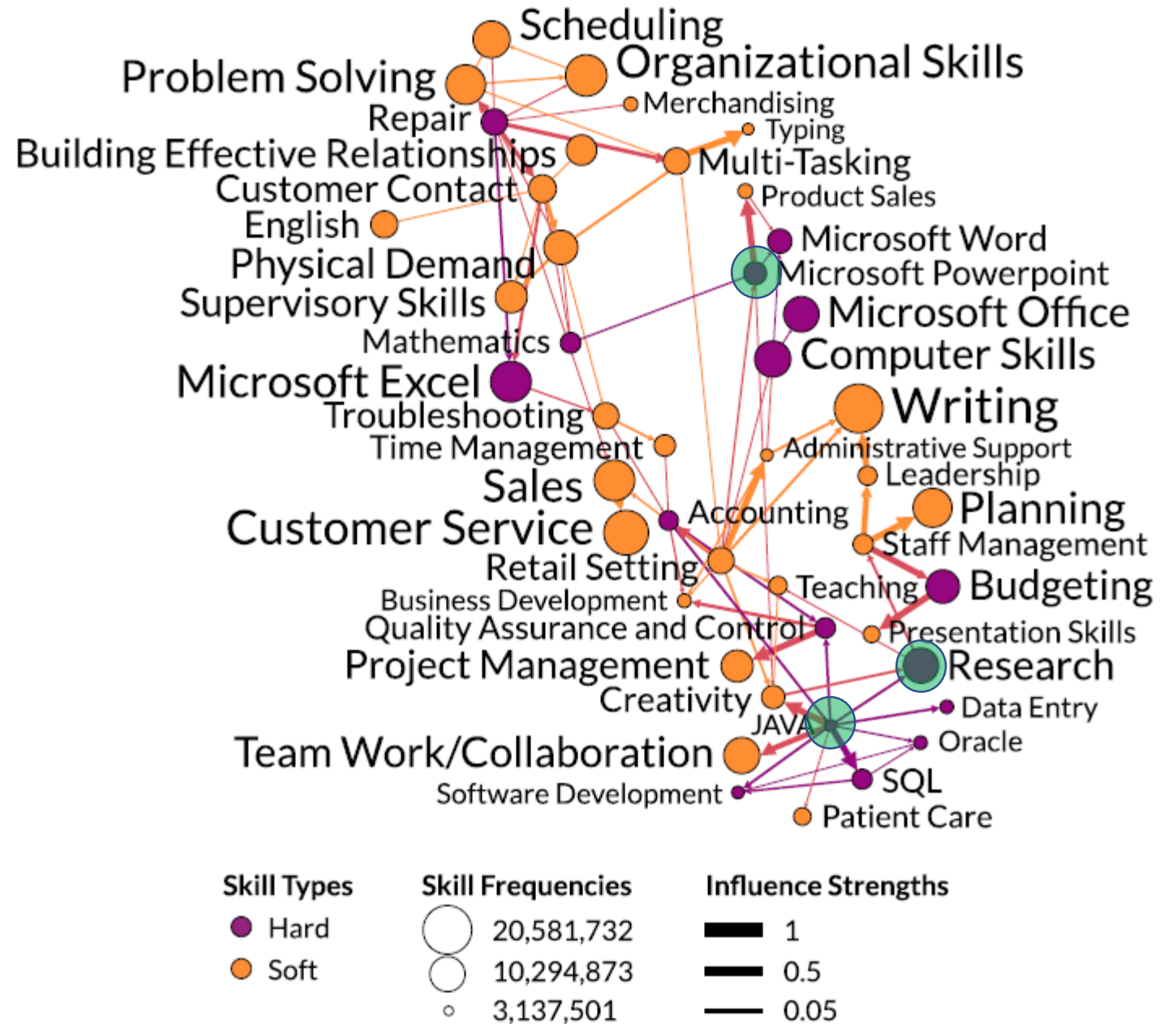


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Results

- Novel cross-walk for mapping publications, course offerings, and job via skills.
- Timing and strength of burst of activity for skills (e.g., Oracle, Customer Service) in publications, course offerings, and job advertisements.
- Uniquely human skills such as communication, negotiation, and complex service provision are currently underexamined in research and undersupplied through education for the labor market in an increasingly automated and AI economy.
- The same pattern manifests in the domain of DS/DE where teamwork and communication skills increase in value with greater demand for data analytics skills and tools.
- Skill demands from industry are as likely to drive skill attention in research as the converse.

Job Postings in The Substance Use Disorder Treatment Sector

Scrivner, Olga, Thuy Nguyen, Kosali Simon, Esmé Middaugh, Bledi Taska, and Katy Börner. 2020. "[Job postings in the substance use disorder treatment related sector during the first five years of Medicaid expansion](#)". *PLOS One* 15 (1): e0228394. doi: 10.1371/journal.pone.0228394.

Background

Effective treatment strategies exist for substance use disorder (SUD), however severe hurdles remain in ensuring adequacy of the SUD treatment (SUDT) workforce as well as improving SUDT affordability, access and stigma. Although evidence shows recent increases in SUD medication access from expanding Medicaid availability under the Affordable Care Act, it is yet unknown whether these policies also led to a growth in hiring in the SUDT related workforce, partly due to poor data availability. Our study uses novel data to shed light on recent trends in a fast-evolving and policy-relevant labor market, and contributes to understanding data sources to track the SUDT related workforce and the effect of recent state healthcare policies on the supply side of this sector.

Methods and data

We examine hiring attempts in the SUDT and related behavioral health sector over 2010-2018 to estimate the causal effect of the 2014-and-beyond state Medicaid expansions on these outcomes through “difference-in-difference” econometric models. We use Burning Glass Technologies (BGT) data covering virtually all U.S. job postings by employers.

Findings

Nationally, we find little growth in the sector’s hiring attempts in 2010-2018 relative to the rest of the economy or to health care as a whole. However, this masks heterogeneity in the bimodal trend in SUDT job postings, with some increases in most years but a decrease in 2014 and in 2017, as well as a shift in emphasis between different occupational categories. Medicaid expansion, however, is not associated with any statistically significant change in overall hiring attempts in the SUDT related sector during this time period, although there is moderate evidence of increases among primary care physicians.

RESEARCH ARTICLE

Job postings in the substance use disorder treatment related sector during the first five years of Medicaid expansion

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Abstract

Background

Effective treatment strategies exist for substance use disorder (SUD), however severe hurdles remain in ensuring adequacy of the SUD treatment (SUDT) workforce as well as improving SUDT affordability, access and stigma. Although evidence shows recent increases in SUD medication access from expanding Medicaid availability under the Affordable Care Act, it is yet unknown whether these policies also led to a growth in hiring in the SUDT related workforce, partly due to poor data availability. Our study uses novel data to shed light on recent trends in a fast-evolving and policy-relevant labor market, and contributes to understanding data sources to track the SUDT related workforce and the effect of recent state healthcare policies on the supply side of this sector.

Methods and data

We examine hiring attempts in the SUDT and related behavioral health sector over 2010-2018 to estimate the causal effect of the 2014-and-beyond state Medicaid expansions on these outcomes through “difference-in-difference” econometric models. We use Burning Glass Technologies (BGT) data covering virtually all U.S. job postings by employers.

Findings

Nationally, we find little growth in the sector’s hiring attempts in 2010-2018 relative to the rest of the economy or to health care as a whole. However, this masks heterogeneity in the bimodal trend in SUDT job postings, with some increases in most years but a decrease in 2014 and in 2017, as well as a shift in emphasis between different occupational categories. Medicaid expansion, however, is not associated with any statistically significant change in overall hiring attempts in the SUDT related sector during this time period, although there is moderate evidence of increases among primary care physicians.

OPEN ACCESS

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Data Availability Statement: Job Postings data are available from Burning Glass Technologies upon request. An interested researcher should send a request to info@burning-glass.com. The remaining

Introduction & Motivation

- Worldwide, the direct burden of illicit drug dependence increased to 20 million disability-adjusted life years in 2010. Examples of these illicit drugs are opioids, cocaine, amphetamines, and cannabis.
- In the US, mental health and SUD together became the leading cause of disease burden in 2015, while nearly 3% of Americans aged 12 years or older reported SUDs in the same year.
- The most effective SUD treatment (SUDT) is a combination of long-acting medications (usually methadone or buprenorphine) administered as part of a cognitive behavioral approach (such as counseling, family therapy, and peer support programs).
- In 2017, there were 13,857 treatment facilities in the U.S. with over 1,356,015 clients enrolled (The National Survey of Substance Abuse Treatment Services; NSSATS).
- The SUDT workforce is deemed inadequate by almost any measure. Workforce shortages and barriers have played a prominent role in limiting treatment access among those suffering from SUDs.

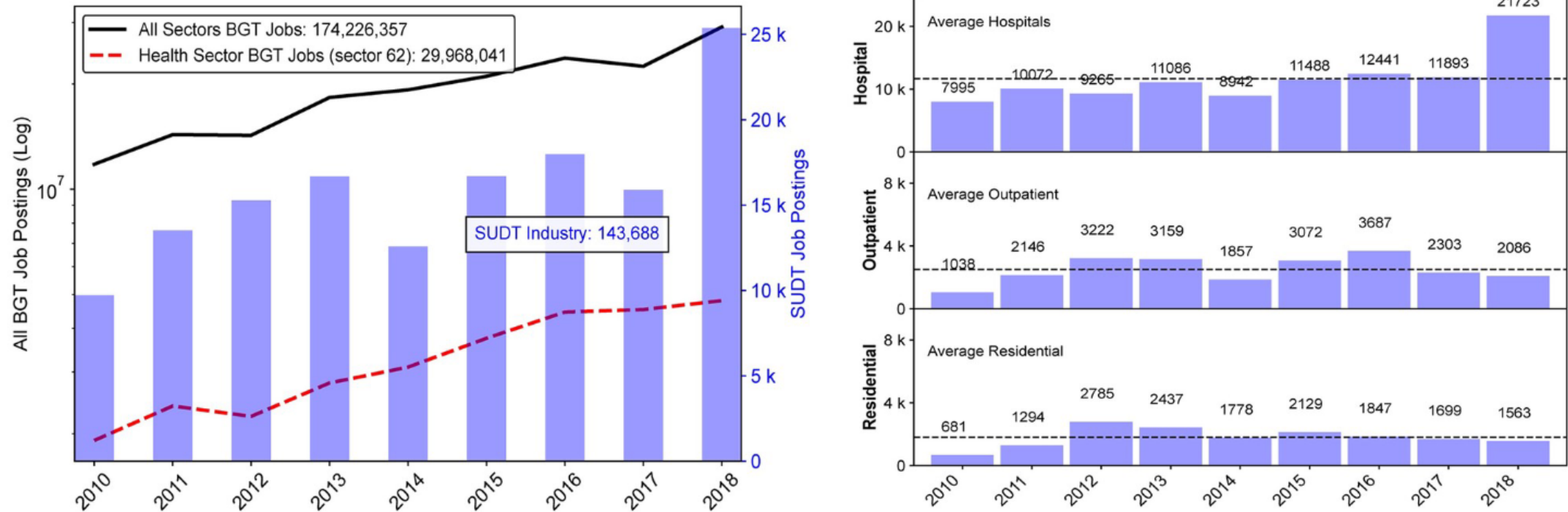


Fig 1. BGT online job postings. (A) BGT Job postings for all industries (black), Healthcare industry (red) and SUDT industries (light blue). The aggregated amount for all job postings is calculated for the period from 2010 through 2018. The healthcare sector is identified by the NAICS code ‘62’. The SUDT facilities are identified by three NAICS codes ‘6222’, ‘6214’, ‘6232’ filtered at 6-digit level. The left y-axis corresponds to the logarithmic trend lines for the total of all BGT job postings (black solid line) and the total of BGT healthcare sector (red dashed line). The y-right axis represents the SUDT sector values, shown as bar graphs. (B) Break down of job postings for three SUDT sectors. Three SUDT sectors are represented by their number of annual online job postings. Average line is calculated for each SUDT sector. Data Source: Burning Glass Technologies. 2019.

<https://doi.org/10.1371/journal.pone.0228394.g001>

Outpatient SUDT Centers

Residential SUDT Facilities

SUDT Hospitals

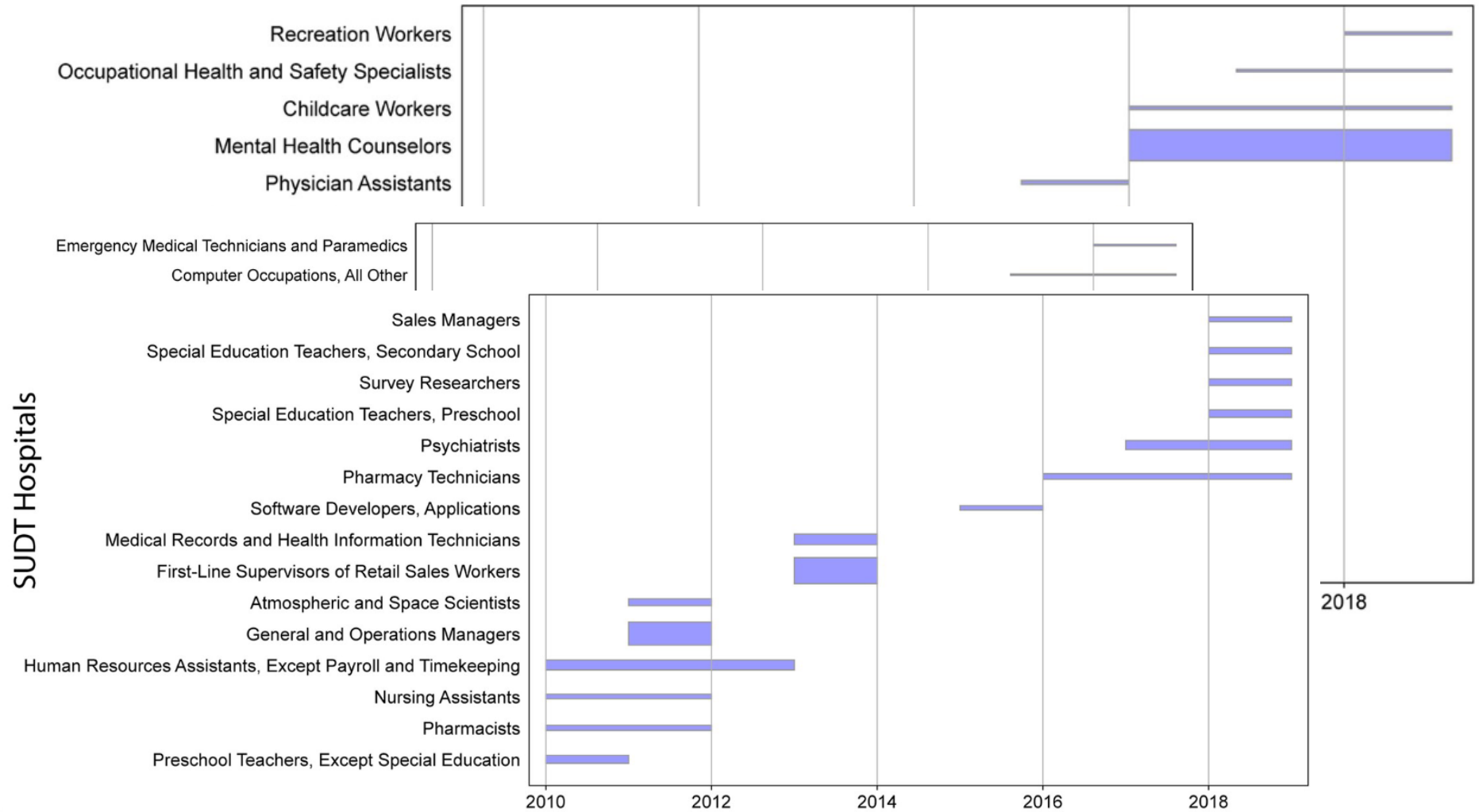
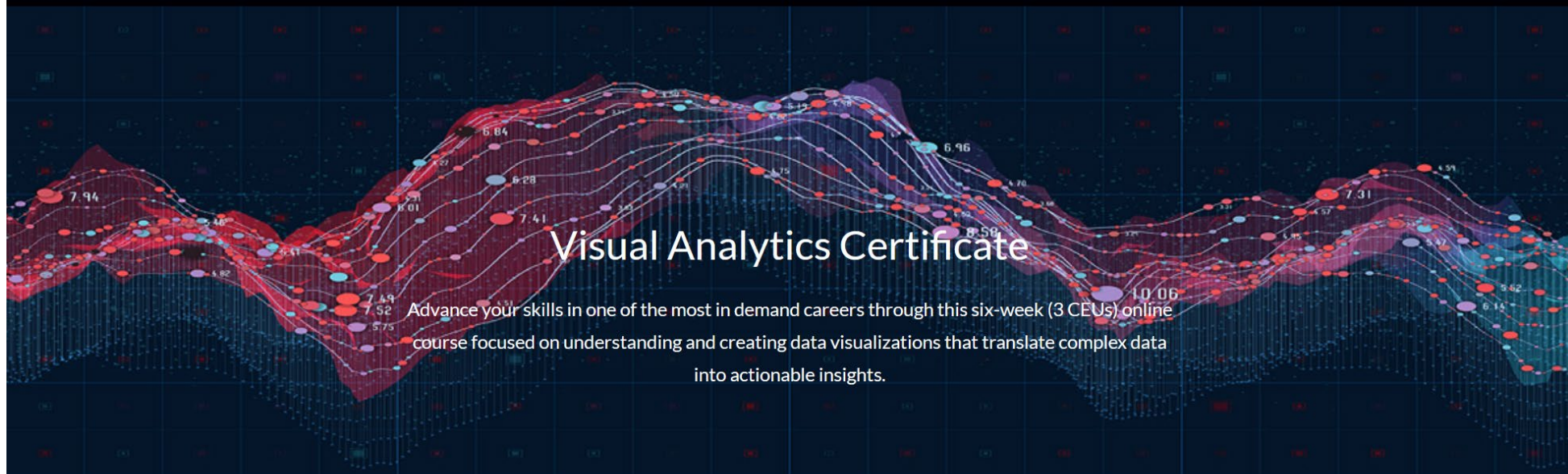


Fig 2. Bursting top-15 SUDT occupations during 2010–2018. Each spike in demand is shown as a horizontal bar with a start and an end date. The length of the bar corresponds to the duration of the hiring burst, the width of the bar shows the burst strength, measured as weight (e.g., in the top panel, the Mental Health Counselor occupation has the strongest and the longest burst in the years 2016–2018).

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Börner, Katy, Andreas Bueckle, and Michael Ginda. 2019. Data visualization literacy: Definitions, conceptual frameworks, exercises, and assessments. *PNAS*, 116 (6) 1857-1864.



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