Can Wikipedia Come to Al's Rescue (Again)?

Brent Hecht

Microsoft / Northwestern

20 June 2024

brenthecht.com



A bit about me



Associate Professor

Language models and AI, in particular the "content ecosystems" they require

August 2019

"Come help redefine the future of work"



Director of Applied Science & Partner

Experiences and Devices,
Microsoft

Increasing the pace, impact, and responsibility of research



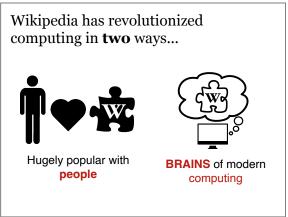
Hard to imagine a higher-Theory of change impact topic in this intersection than the subject of this talk **Our RAI** Stuff that is great Stuff that is great for the world mission for Microsoft

A brief outline

- Looking back: Wikipedia's central role in the development of modern AI (case study from my career)
- **The present conundrum**: The dominant LLM paradigm threatens Wikipedia, large portions of the content ecosystem, and ultimately itself.
- Looking forward: What Wikipedia can do to help itself in the LLM era, and make the LLM era much better in the process

Simply stated: Modern AI does not exist without the "Wikipedia dataset"

- Strong argument it was the single most important dataset for AI research since about 2005
 - When the stochastic turn in AI needed data, Wikipedia was there to provide it.
 - Used as the research dataset of first resort in many areas for two decades
- Semantic relatedness ("proto-LLMs"), knowledge graphs, information retrieval, information extraction...
- Core to basically every LLM training dataset from the beginning up through now
- Of course not just academia and LLMs, also many other commercial applications



Slide from WikiSym 2012 Keynote







Mining meaning from Wikipedia

Olena Medelyan*, David Milne, Catherine Legg, Ian H. Witten

University of Waikato, Knighton Road, Hamilton 3216, New Zealand

Received 23 February 2009; received in revised form 14 April 2009; accepted 1 May 2009

Communicated by e. motta

Available online 29 May 2009

Abstract

Wikipedia is a goldmine of information; not just for its many readers, but also for the growing community of researchers who recognize it as a resource of exceptional scale and utility. It represents a vast investment of manual effort and judgment: a huge, constantly evolving tapestry of concepts and relations that is being applied to a host of tasks.

Medelyan et al. 2009

I used to try to keep track of all the ways Wikipedia was supporting very successful AI products...

strengths of the Satori Ontology is the ability to manage multiple source schemas and cross domain linkages."

- Excel
- Flight Simulator

▼ Apple

- ▼ Siri
- Examples via vandalism
- Examples are easy to come by
- "Who is Grover Cleveland?"
- "Tell me something about the New York Times"
- A few moderately useful references
- https://techcrunch.com/2013/06/10/apple-updates-siri-with-twitter-wikipedia-bing-integration-new-commands-and-male-and-female-voice/
- https://www.thesun.co.uk/tech/4224107/what-is-an-indian-apples-siri-has-a-racist-and-offensive-answer-to-this-question/
- ▶ They call it "Siri" knowledge
- Apple Maps

▼ Google

- Google Search
- Massively important relationship for both Google and Wikipedia
- A detailed overview can be found in this paper: https://www.aaai.org/ocs/index.php/ICWSM/ICWSM17/paper/download/15623/14799
- ▼ Wikipedia 2x important as deep learning to Google's search results performance

- ▼ BERT
- omnioutliner:///open?row=ke2RXiyn5BN
- YouTube
- ▼ Context for "fake news"
- https://www.vanityfair.com/news/2018/03/youtube-wikipedia-conspiracy-theory-video-problem
- https://twitter.com/krmaher/status/973791488583888896
- ▼ Big Google datasets
- KELM: Integrating Knowledge Graphs with Language Model Pre-training Corpora

Wolfram Alpha

- "Since the inception of WolframlAlpha, Wikipedia has held a special place in its development pipeline. We
 usually use it not as a primary source for data, but rather as an essential resource for improving our natur
 language understanding, particularly for mining the common and colloquial ways people refer to entities
 and concepts in various domains."
- http://blog.wolfram.com/2015/03/20/new-in-the-wolfram-language-wikipediadata/

▼ Facebook

- Providing context for news
- https://newsroom.fb.com/news/2018/04/news-feed-fyi-more-context/

Helping People Better Assess the Stories They See in News Feed I F...

https://newsroom.fb.com/news/2018/04/news-feed-fyi-more-context.

Search in newsroom

Q

Home News Company Info News Feed FYI Directory Media Gallery

Investor Relations

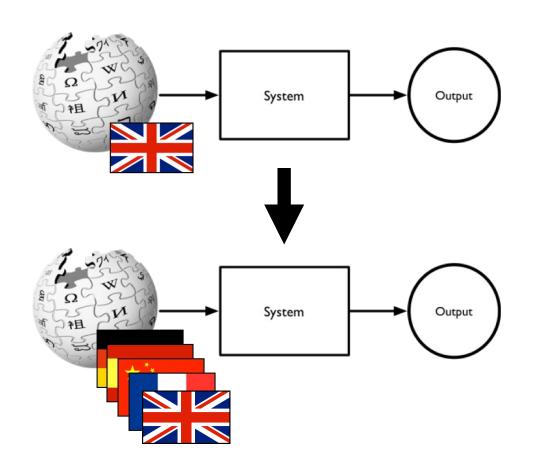
...but it got to be way too much.

To provide some color to the importance of Wikipedia, let's go back to around 2008 or so...





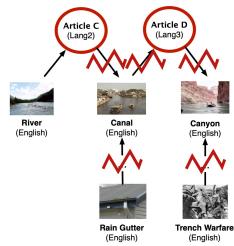
A failed attempt at improving semantic relatedness algorithms

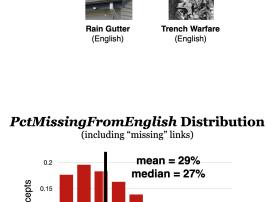


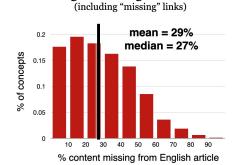


Wikipedia and uncovering "algorithmic bias"

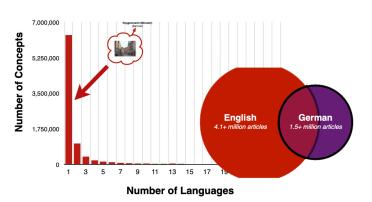
- One way I'm in debt to Wikipedia is it helped us uncover what we now know as "algorithmic bias"
- Wikipedia's role here should be better known
- Key papers: Hecht and Gergle 2010, Bao et al. 2012

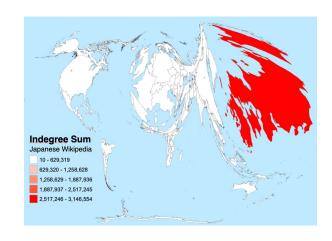






Distribution of Languages Per Concept





Wikipedia and uncovering "algorithmic bias"

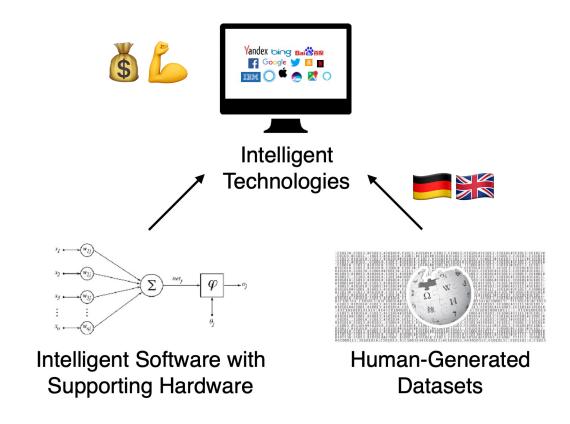
SRenglish(Concept A, Concept B)
=?
SRGERMAN(Concept A, Concept B)
=?
SRSPANISH(Concept A, Concept B)
=?

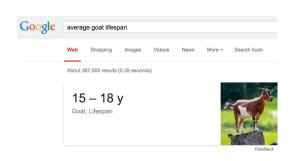
Cross-I	anguage
---------	---------

SR Measure	Mean r _s	Max r _s	
WikiRelate	0.33	0.52	
MilneWitten	0.47	0.55	
OutlinkOverlap	0.36	0.48	
WAGDirect	0.41	0.54	
Explicit Semantic Analysis	0.41	0.58	

From studying cultural biases to working to address harmful power imbalances

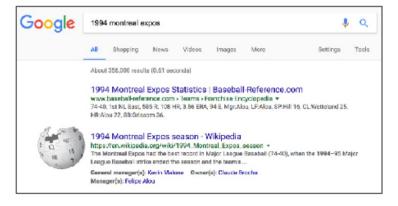
- First: Wikipedia as a way to advance Al
- Then: Concerned about power relationships between Wikipedia and AI technologies
- Finally: All data/content creators, not just Wikipedia

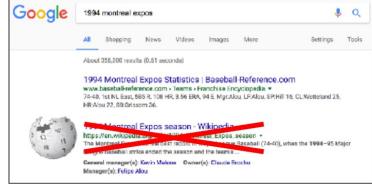




(Taraborelli 2015)

Wikipedia links increase clickthrough rates by 80%...

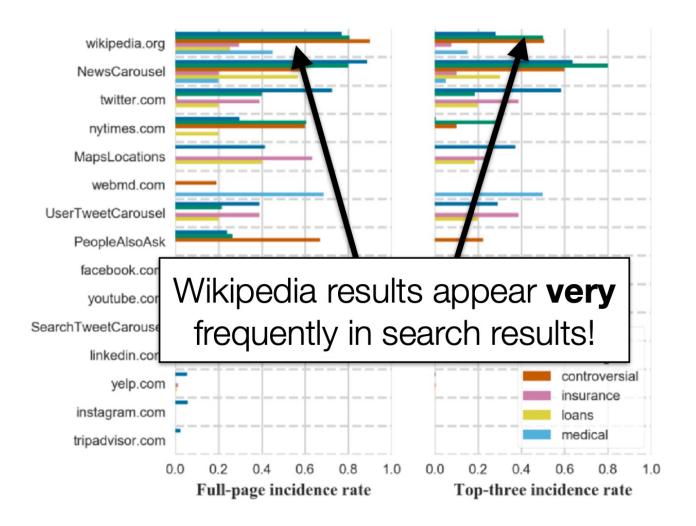




Wikipedia Links Present 618 queries CTR = **26.1%** Wikipedia Links Removed
387 queries
CTR = 14.0%

Context: A CTR improvement of 1% is a big deal

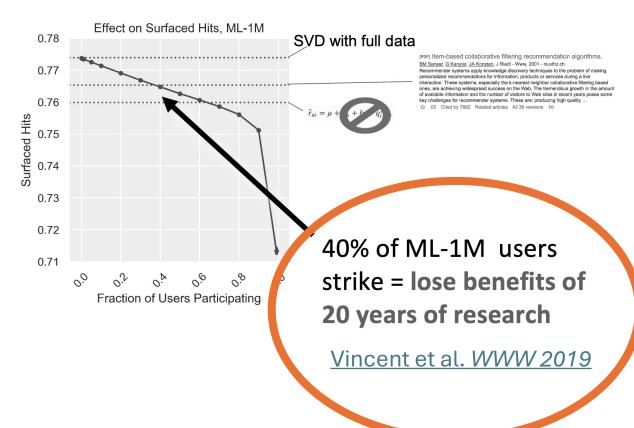
Johnson et al. ICWSM 2017



Vincent et al. ICWSM 2019; Vincent et al. CSCW 2021

Key concept: The "data strike"

- "Data strikes" (Vincent et al. 2019)
 occur when a significant proportion of
 content producers for a given AI/ML
 system stop producing that content in
 order to force a change in behavior by
 the system owner.
- Data strikes have always been a substantial lurking force in AI and the computing industry
- Main things holding them back was information asymmetry and implicit non-compete between AI/ML system owner and content producer. Those are now gone/eroded.



Data Leverage: A Framework for Empowering the Public in its

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ABSTRACT

Many powerful computing technologies rely on implicit and explicit data contributions from the public. This dependency suggests a potential source of leverage for the public in its relationship with technology companies: by reducing, stopping, redirecting, or otherwise manipulating data contributions, the public can reduce the effectiveness of many lucrative technologies. In this paper, we synthesize emerging research that seeks to better understand and help people action this data leverage. Drawing on prior work in areas including machine learning, human-computer interaction, and fairness and accountability in computing, we present a framework for understanding data leverage that highlights new opportunities to change technology company behavior related to privacy, economic inequality, content moderation and other areas of societal concern. Our framework also points towards ways that policymakers can bolster data leverage as a means of changing the balance of power between the public and tech companies.

CCS CONCEPTS

 Human-centered computing → Collaborative and social computing theory, concepts and paradigms.

KEYWORDS

data leverage, data strikes, data poisoning, conscious data contribu-

ACM Reference Format:

Nicholas Vincent, Hanlin Li, Nicole Tilly, Stevie Chancellor, and Brent Hecht. 2021. Data Leverage: A Framework for Empowering the Public in its Relationship with Technology Companies. In Conference on Fairness, Accountability, and Transparency (FAccT '21), March 3–10, 2021, Virtual Event, Canada. ACM, New York, NY, USA, 13 pages. https://doi.org/10.1145/342188.348885

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FAccT '21, March 3–10, 2021, Virtual Event © 2021 Association for Computing Machinery ACM ISBN 978-1-4503-8309-7/21/03...\$15.00 https://doi.org/10.1145/3442188.3445885 Northwestern University bhecht@northwestern.edu

1 INTRODUCTION

In August 2020, the most valuable five technology companies had a total market cap of US\$7 trillion [98]. This valuation is driven in part by large models that use data generated by the public to recommend content, rank search results, and provide many other services [25, 90, 101, 136]. More generally, lucrative technologies used by many companies rely on data generated by large groups of people to fulfill critical customer needs [8, 25, 101, 124] and drive decision-making [26].

The reliance of powerful technologies (and thus powerful companies) on "data labor" [8, 101, 123] by the general public presents an enormous opportunity for the public to gain more power in its relationship with tech companies. People perform data labor when they engage in the multitude of interactions with technology that generate data for firms (e.g. liking, clicking, rating, posting). By leveraging tech companies' reliance on their data labor, the public could demand changes on pressing issues [60, 73], such as diminished privacy [24, 40], the reinforcement of problematic societal biases by AI systems [5, 7, 44, 73, 95], eroded labor rights [59, 97], environmental harms [108], content moderation challenges [52], and the current imbalance in how profits from data-driven technologies are distributed between tech operators and data contributors [25, 101, 126]. Armed with the knowledge of the importance of data contributions and the tools to action this knowledge, the public could potentially interfere with recommender systems, search engines, image classifiers, and other technologies until tech companies made changes related to these issues.

To capture the power inherent in the public's data labor, this paper introduces the concept of "data leverage" and discusses how the concept can be made operational. Simply put, data leverage refers to influence that members of the public have over tech companies because important computing technologies rely on the public's data contributions. Data leverage catalyzes power achieved by harming data-dependent technologies as well as power achieved by improving alternative data-dependent technologies and thereby creating increased competition [121]. The concept of data leverage highlights an emergent theme in the FAccT community and related areas, including human-computer interaction (HCI), social computing, society and technology studies (STS), machine learning (ML), and particularly ML research that seeks to advance fairness, justice, and a human-centered perspective (e.g. [5, 20, 29, 54]). This paper shows that this interdisciplinary lens can provide a structure for understanding and actioning an almost entirely untapped source of power that can advance a wide variety of pro-social goals. Our

The Dimensions of Data Labor: Activists, and Pone, Labor: Empower Data Producers

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> Stevie Chancellor steviec@umn.edu University of Minnesota Minneapolis, MN, USA

ABSTRACT

Many recent technological advances (e.g. ChatGPT and search engines) are possible only because of massive amounts of usergenerated data produced through user interactions with computing systems or scraped from the web (e.g. behavior logs, user-generated content, and artwork). However, data producers have little say in what data is captured, how it is used, or who it benefits. Organizations with the ability to access and process this data, e.g. OpenAI and Google, possess immense power in shaping the technology landscape. By synthesizing related literature that reconceptualizes the production of data for computing as "data labor", we outline opportunities for researchers, policymakers, and activists to empower data producers in their relationship with tech companies, e.g advocating for transparency about data reuse, creating feedback channels between data producers and companies, and potentially developing mechanisms to share data's revenue more broadly. In doing so, we characterize data labor with six important dimensions - legibility, end-use awareness, collaboration requirement, openness, replaceability, and livelihood overlap - based on the parallels between data labor and various other types of labor in the computing literature.

CCS CONCEPTS

 \bullet Human-centered computing \rightarrow HCI theory, concepts and models

KEYWORDS

user-generated data, empowerment, data leverage

ACM Reference Format:

Hanlin Li, Nicholas Vincent, Stevie Chancellor, and Brent Hecht. 2023. The Dimensions of Data Labor: A Road Map for Researchers, Activists, and Policymakers to Empower Data Producers. In 2023 ACM Conference on Fairness, Accountability, and Transparency (FAccT '23), June 12–15, 2023,

Preprint for FAccT '23, June 12-15, 2023, Chicago, IL, USA

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Chicago, IL, USA. ACM, New York, NY, USA, 11 pages. https://doi.org/10.1145/3593013.3594070

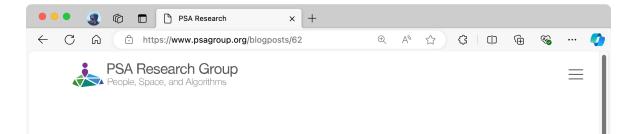
1 INTRODUCTION

Technology users generate large troves of data in their daily interactions with computing systems, e.g. behavior logs, content, and personal information. Currently, this data primarily benefits just a small set of technology organizations that are equipped with the means and resources to collect, process, and model data at scale for their own benefits (e.g. insights, models, sales of services and advertisements). For example, publicly available texts and artwork enabled the creation of generative AI models like ChatGPT and Dall-E because model developers were able to scrape and process data from billions of web pages 1. Conversely, data producers like artists, writers, and users have little to no power in deciding how their data is used or who it benefits [4, 7, 42, 63]. This power imbalance between data producers and technology operators has manifested in public outcries about industry practices in the tech sector. For example, emerging generative AI models such as Stable Diffusion, Dall-E, and GitHub Copilot have sparked extensive criticism among artists and programmers because of these models' unapproved reuse of their work and implications on future employment opportunities [79-81]. More broadly, social media users have long protested the monetization of user data and the corporate surveillance practices that tend to go with it [45].

Given data producers' lack of power over the data they generate, researchers, policymakers, and activists have advocated for a new producer-oriented paradigm shift to increase the voice of the data-generating public – understanding data generation as a form of labor, or "data labor" [7]. Supporters of this approach have argued that treating data as an outcome of social labor instead of "exhaust" will pave the way for more broadly distributing the power and benefits of data [86], and scholars have addressed what this may look like in practice. Initial (yet abstract) proposals include supporting "data unions" [63] or "mediators of individual data" [42] that negotiate data use terms with technology firms on behalf of their data-producing "union" members [63], farfting legislation that would grant users greater control over the data they produce [1, 76], and creating tools to support user-driven collective action [20, 86].

^{*}Chancellor completed much of this work while at Northwestern University

https://commoncrawl.org/2022/10/sep-oct-2022-crawl-archive-now-available/



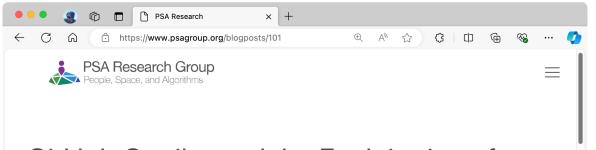
Don't give OpenAl all the credit for GPT-3: You might have helped create the latest "astonishing" advance in Al too

By Nicholas Vincent on 2020-09-22

The much-celebrated GPT-3 that can answer questions, write poems, and more wouldn't be possible without content written by millions of people around the world. Shouldn't they get some credit?

You may have heard about OpenAl's "GPT-3"— an "astonishing" machine learning system that can produce impressive poems, code and op-eds. However, it wasn't just OpenAl that built this: it was also millions of people writing, posting, and voting on content. In fact, it's possible that you played a role in creating GPT3!

Indeed, content and data you helped generate may have been used to build AI systems like GPT-3 in the past. It's even more likely your data will be used to build AI systems in the future. Do you deserve some of the credit for AI's success? What about the profits (for instance, through a "data dividend")? What might you do if you aren't



GitHub Copilot and the Exploitation of "Data Labor": A Wake-Up Call for the Tech Industry

By psagrp on 2021-07-08

This post was jointly written by the PSA Research Group. Points of contact: Nick Vincent and Hanlin Li.

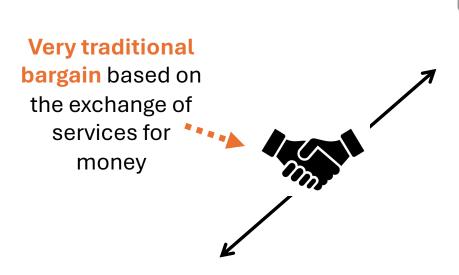
The release of GitHub Copilot is causing quite the outcry in the tech industry, and for good reason. Copilot is a big leap forward in the effort to automate programming, and perhaps eventually programming jobs. However, we in the tech industry should realize that Copilot is merely a taste of our own medicine; we've been creating similar problems for people outside the tech industry for years.

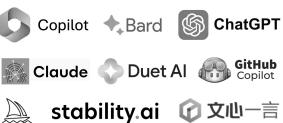
GitHub Copilot is an AI system that uses a statistical model learned from "billions of lines of public code" on GitHub. This code has been written by millions of programmers around the world, and many of those programmers are not happy. They argue that GitHub should

A brief outline

- Looking back: Wikipedia's central role in the development of modern Al
- The present conundrum: The dominant LLM paradigm threatens Wikipedia, large portions of the content ecosystem, and ultimately itself.
- Looking forward: What Wikipedia can do to help itself in the LLM era, and make the LLM era much better in the process

Two Grand Bargains that Make LLMs Possible





Model Builders

(and builders of LLM-based applications)









Code

The web







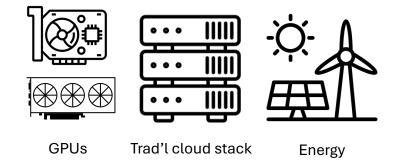
User-generated content

Scientific papers

Social media

Content Producers

("data labor")

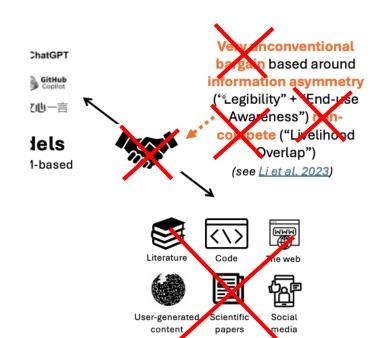


Cloud Infra Providers

(Hardware, Energy, etc)

Flash forward to 2022...

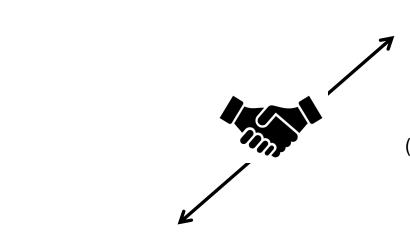
- Information asymmetry heavily eroded by ChatGPT, GitHub Copilot
 - People know what we're doing and how we're doing it
- Non-compete heavily eroded by LLM-based businesses
 - Promise of LLMs is largely in the same information work domains from which they get their content (law, medicine, coding, science...)
 - Understandable given the nature of the technology: They naturally compete with their content supply chain (within-domain vs. outside domain performance)

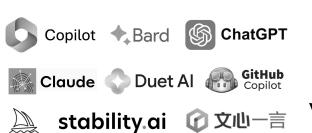


Content Producers

("data labor")

We need a new Grand Bargain with content producers

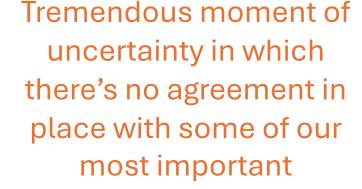




Language Models

(and Copilots + other LLM-based applications)

We need a new Grand Bargain with content producers.



suppliers.







The web





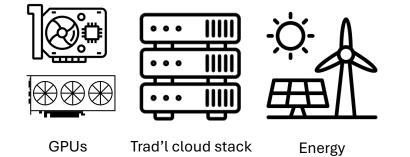
User-generated content

Scientific papers

Social media

Content Producers

("data labor")



Cloud Infrastructure

(Hardware, Energy, etc)

Two ways current LLM paradigm is in trouble from lack of grand bargain

- Capabilities: Al significantly less competent than it could be
- Societal impacts: Potential harms very substantial

Misalignment of incentives leads to a "data strike era"



Journalism Data Strike



General Data Strike



Twitter/X Data Strike

Well-executed data strikes likely highly effective against LLMs

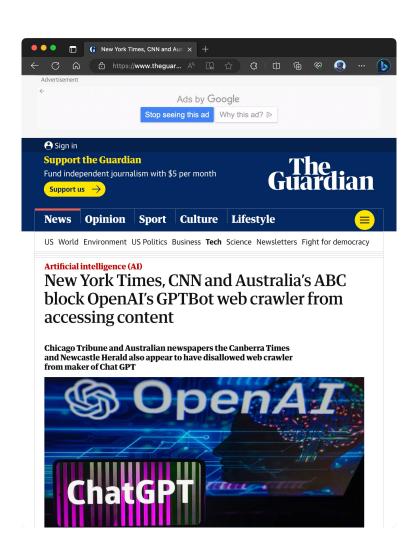
Eval data	<u>PD</u>	<u>PD</u> SW	PDSWBY	Pythia
FreeLaw	5.3	5.7	6.5	5.6
Gutenberg	15.2	12.5	14.1	13.1
HackerNews	38.0	13.7	14.5	13.3
Github	13.5	2.7	2.8	2.4
NIH ExPorter	28.2	19.2	15.0	11.1
PhilPapers	31.7	17.6	15.0	12.7
Wikipedia	28.9	20.3	11.3	9.1
CC News	34.0	23.3	21.2	12.0
BookCorpus2	25.3	19.2	19.6	13.2
Books3	27.2	19.3	18.6	12.6
OpenWebText2	37.8	21.1	18.8	11.5
Enron Emails	18.6	13.2	13.5	6.9
Amazon	81.1	34.8	37.0	22.9
MIMIC-III	22.3	19.0	15.5	13.1
Average	29.1	17.3	16.0	11.4

Significant drops in perplexity when forced to only train on data with explicit consent.

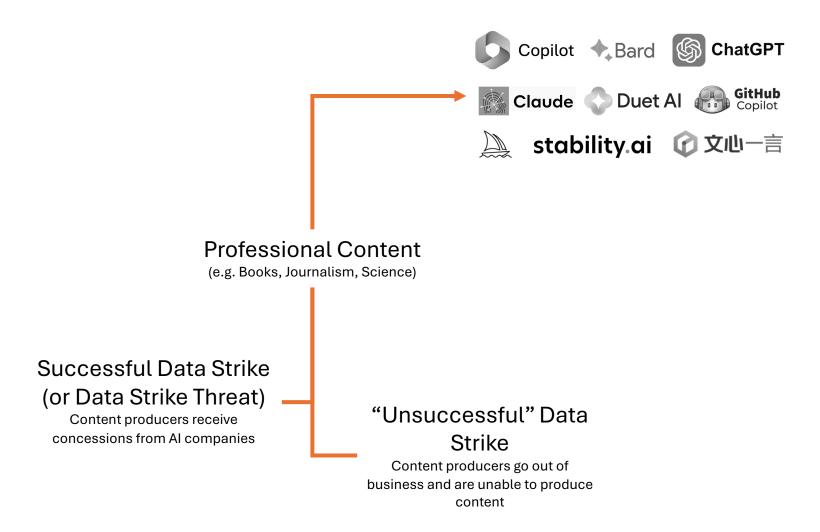
Suggests data strikes along professional domains would work well.

What do LLM data strikes look like?

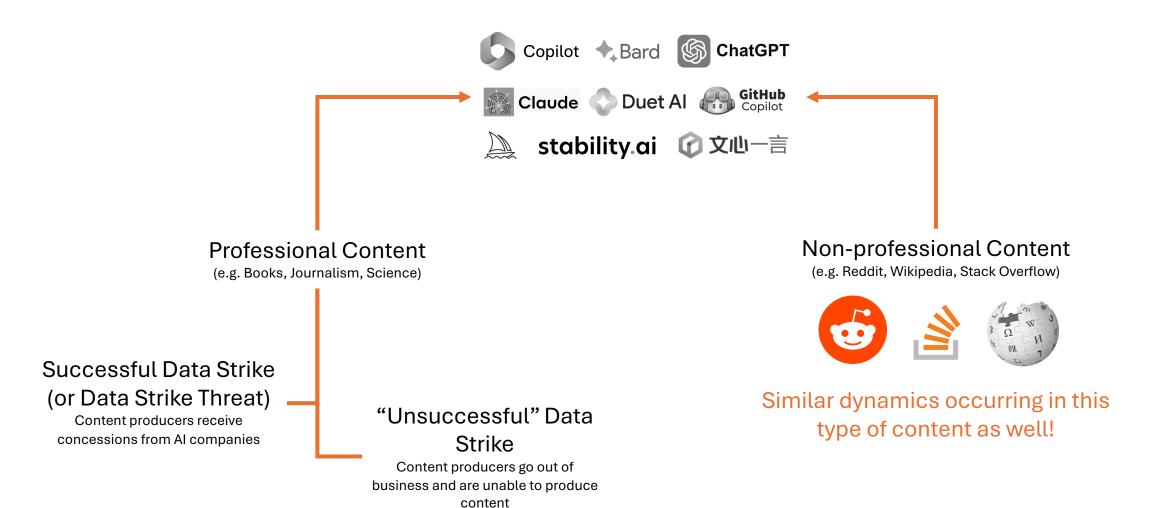
- Low effort: robots.txt, IP blocking
 - Easy to give up Common Crawl and Bing traffic in particular
- Medium effort: Putting everything behind a paywall or registration wall
- Higher effort: Work stoppages, class action lawsuits
- Highest effort: Going out of business



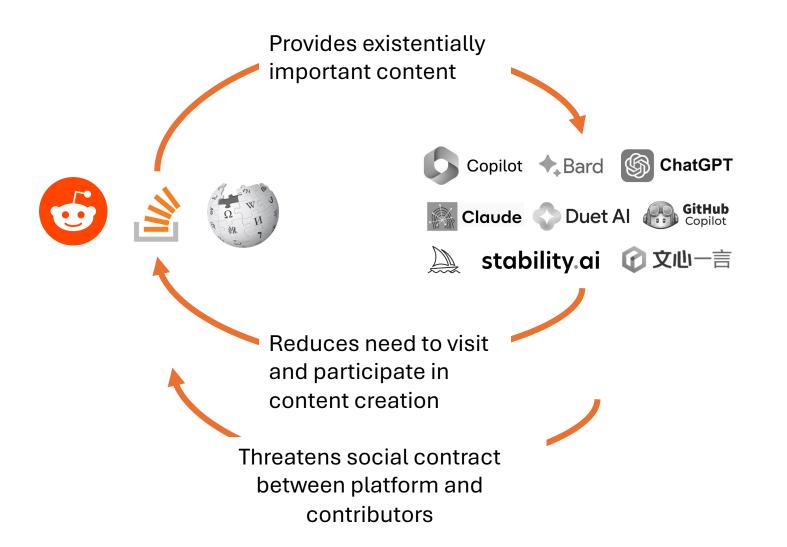
Are data strikes guaranteed?



Are data strikes guaranteed?



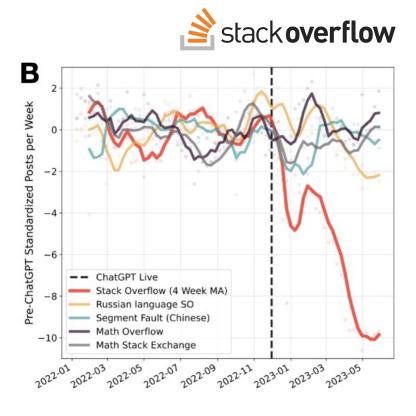
The hypothesized LLM-UGC "Doom Loop"



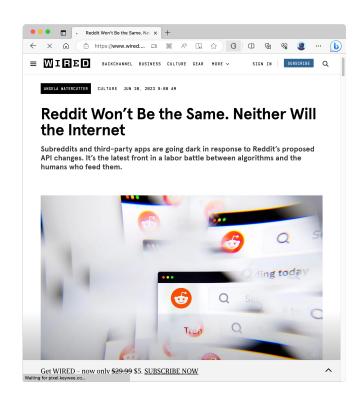
Aka the "Paradox of Reuse"

(<u>Taraborelli 2015;</u> <u>McMahon et al. 2017;</u> <u>Vincent 2022</u>)

Some evidence the doom loop has begun

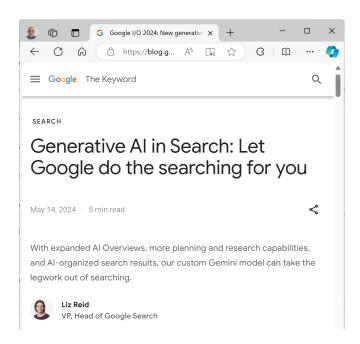


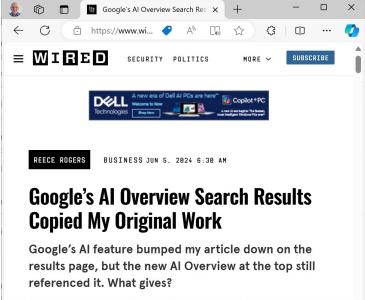
Rio-Chanona et al. 2023

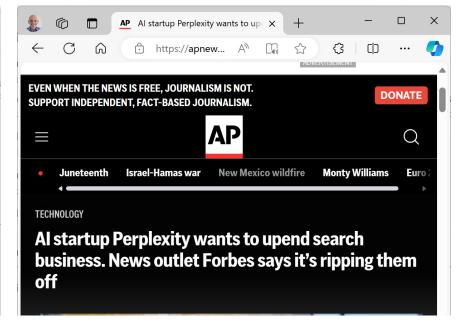


Reddit moderator strike caused by LLM-induced change to Reddit's business model

Recent developments likely to accelerate "doom loop" trends







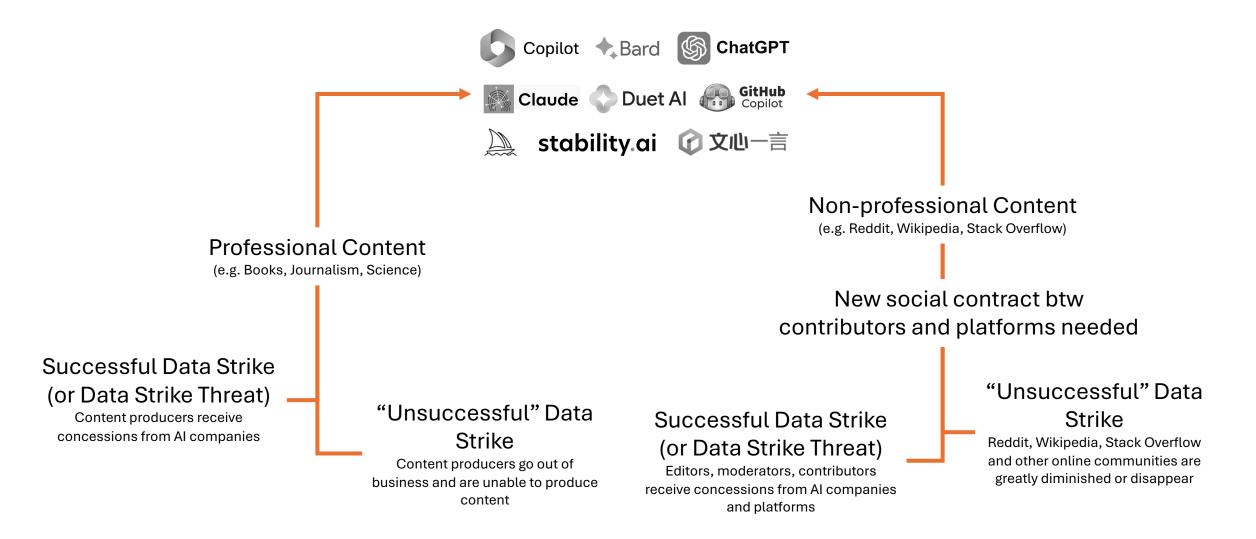
The "Paradox of Reuse" of Everything



Feedback

(Taraborelli 2015)

Are data strikes guaranteed?



Two ways current LLM paradigm is in trouble from lack of grand bargain

- Capabilities: Al significantly less competent than it could be
- Societal impacts: Potential harms very substantial

Risk of large-scale labor disruptions

- More complex than commonly understood, can be overblown
- Critical to leverage history of technology and productivity
- There is unmet demand in so many places that can absorb productivity increases
- What AI community works on matters; are we building substitutional or augmentative technologies? ("Turing Trap")
- But elephant in the room: Goal of some Al companies vs. current capabilities
 - Effectively will create infinite productivity increases that no amount of increased demand can absorb



The Turing Trap: The Promise & Peril of Human-Like Artificial Intelligence



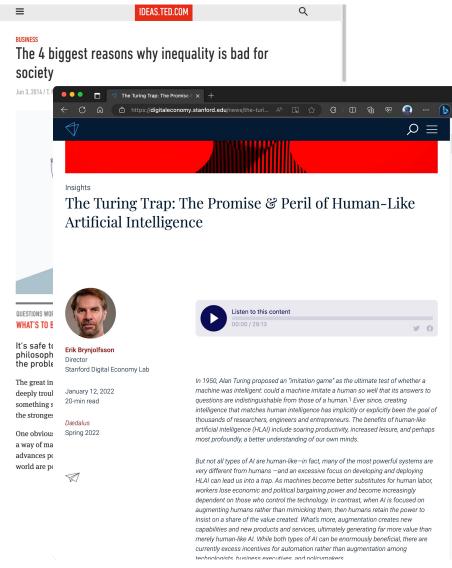
OpenAl's mission is to ensure that artificial general intelligence (AGI)—by which we mean highly autonomous systems that outperform humans at most economically valuable work—benefits all of humanity. We will attempt to directly build safe and beneficial AGI, but will also consider our mission fulfilled if our work aids others to achieve this outcome. To that end, we commit to the following principles:

Significant drop in "labor share of revenue"?

- "The fraction of economic output that accrues to workers as compensation in exchange for their labor" (<u>US BLS 2017</u>)
- Once thought to be relatively stable, has been declining in recent decades, at least in part due to computing
 - 1% increase in IT intensity is associated with a 0.1% decline in labor share of revenue (Brynjolfsson et al. 2023)
- Reasonable argument is this is because we're just not paying most of our labor, e.g. done by our quality and relevance assessors, reviewers, etc.
 - Reddit moderators save Reddit at least 3% of revenue per year (<u>Li et al. ICWSM</u> 2022)
- Serious risk of massive reduction in labor share of revenue if current LLM paradigm continues

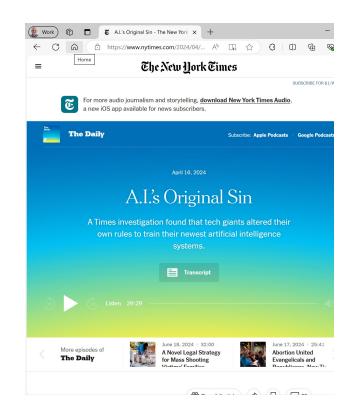
Effects of computing-induced wealth and power concentration

- Excessive wealth concentration widely recognized as non-optimal for many societal goals that have wide consensus
- **Economic productivity** e.g. GDP (which is also really important to Microsoft's business!)
- Failure of democratic institutions "We may have democracy, or we may have wealth concentrated in the hands of a few, but we can't have both." (Louis Brandeis)
- Obsoletion of most ethical frameworks: If all humans have equal value, why should a few have 10^9 more resources?



Risk of legitimacy crisis in Al

- What happens if the dominant technology has such a wide mismatch between value accretion and value creation?
 - Can the market still work? Will people still believe in it?
- Property dispossession perceived as an irremovable "origin sin" (NY Times) without changes?
- History of scaled property dispossession is terrifying, including those justified by technological advancement
 - English Enclosures a strong analogy
 - Led to hundreds of years of riots, and likely (in part) the beheading of King Charles I.
 - Diggers and Levellers, revisited -> data poisoning, data center attacks?





Demographic dimensions to property dispossession

- New paper at CHI: "American Jews May Be Disproportionately Harmed by Intellectual Property Dispossession in LLM Training"
- "Systemic property dispossession from minority groups has often been carried out in the name of technological progress. In this paper, we identify evidence that the current paradigm of LLMs likely continues this long history."
- Parallels historical property dispossession because of causal relationship to history; reason why certain demographics invested more in intellectual property

A Canary in the AI Coal Mine: American Jews May Be Disproportionately Harmed by Intellectual Property Dispossession in Large Language Model Training

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> Training. In Proceedings of the CHI Conference on Human Factors in Computing Systems (CHI '24), May 11–16, 2024, Honolulu, HI, USA. ACM, New York, NY, USA, 17 pages. https://doi.org/10.1145/3613904.3642749

ABSTRACT

Systemic property dispossession from minority groups has often been carried out in the name of technological progress. In this paper, we identify evidence that the current paradigm of large language models (LLMs) likely continues this long history. Examining common LLM training datasets, we find that a disproportionate amount of content authored by Jewish Americans is used for training without their consent. The degree of over-representation ranges from around 2x to around 6.5x. Given that LLMs may substitute for the paid labor of those who produced their training data, they have the potential to cause even more substantial and disproportionate economic harm to Jewish Americans in the coming years. This paper focuses on Jewish Americans as a case study, but it is probable that other minority communities (e.g., Asian Americans, Hindu Americans) may be similarly affected and, most importantly, the results should likely be interpreted as a "canary in the coal mine"

1 INTRODUCTION

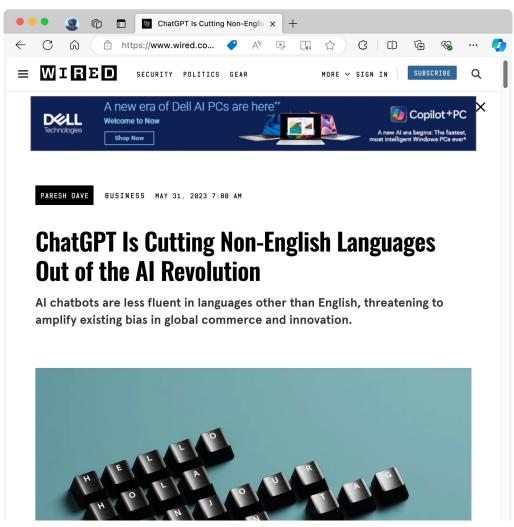
One of the most prominent critiques of large language models (LLMs) is that they train on massive amounts of content without the consent of the authors of that content [8, 59, 73, 91, 98, 103, 109, 124–126]. This concern is exacerbated by one of the core promises of LLMs: their ability to use patterns in their training data to substitute for the paid labor of those who created said data. People in a wide range of professions (e.g., fiction-writing and journalism) are now accusing language modeling companies of not only stealing their content (e.g., novels and news stories), but also of using this very content to put them out of a job (e.g., 158, 98, 105, 116), Indeed,

Dataset	% IP with DJN author	% IP with U.S. Jewish author	% Expected IP with U.S. Jewish author	Relative Dispossession Magnitude
PubMed Central	0.19	1.39-1.91	0.5-0.7	2.02-3.71 X
Books3	0.98	7.01-9.64	1.8-2.4	2.92-5.36 X
ArXiv	0.28	2.01-2.77	0.5-0.7	3.07-5.63 X
GitHub	0.29	2.08-2.86	0.4-0.6	3.53-6.46 X
FreeLaw	0.93	6.65-9.14	1.8-2.4	2.77-5.08 X
Total	0.54	3.83-5.26	1.0-1.3	2.86-5.25 X
Weighted Total	0.37	2.63-3.61	0.8-1.0	2.46-4.51 X

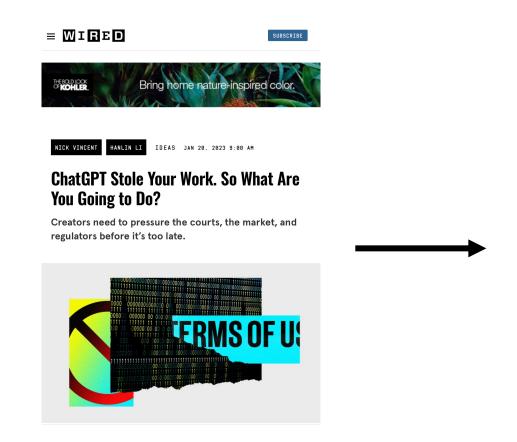
Precel et al. 2024

"Data colonialism" as a real risk for LLM content creation efforts

- Need to make sure strong grand bargain is in place before sourcing/creating content for LLMs in low-resource languages
- Else there's a risk of replicating very problematic value transfers



Opportunity: "They took my stuff" to "We went to the moon

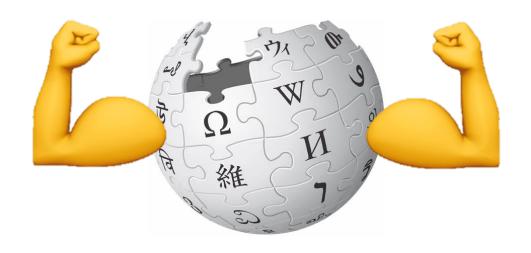




How do we go from this...

...to this?

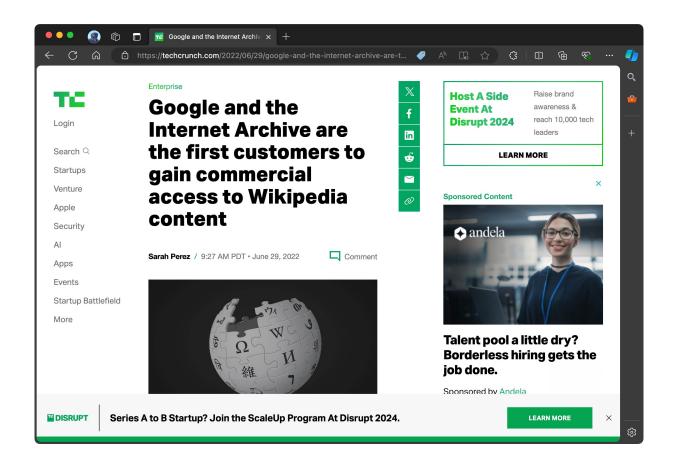
What should the Wikipedia community do in this moment?



First proposition: Understand it has a good deal of leverage

Maximized if acting as a community, partnering with other open data communities

Successes in the past!

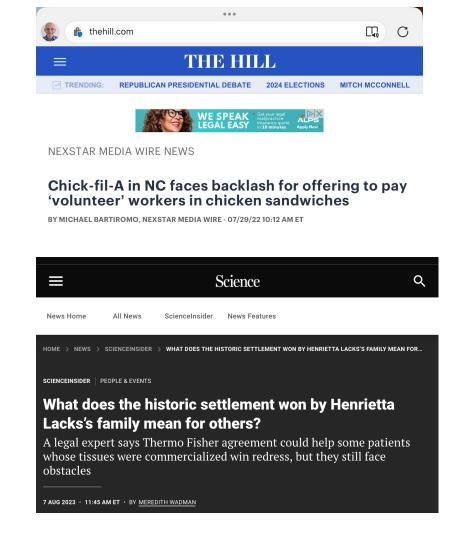


How: The four sources of content leverage

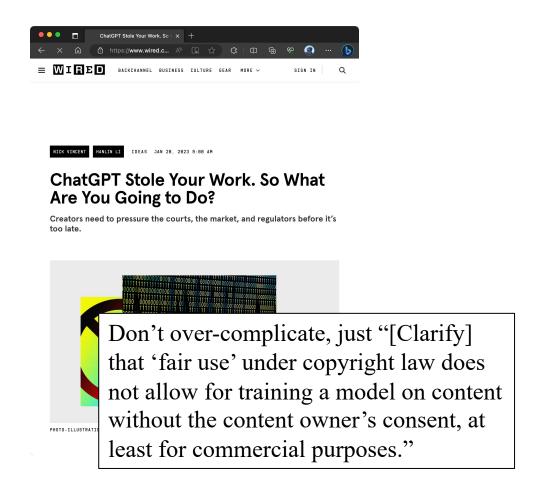
- Legal leverage
- Policy leverage
- Reputational leverage
- "Data leverage"
 - "Conscious Data Contribution"
 - "Data Strikes"

Legal leverage: Some interesting legal research directions

- Naturally cannot comment on any active lawsuits
- Copyright law is only a tiny part of the picture in the literature
 - Labor law, contract law, unjust enrichment, privacy law, huge number of jurisdictions (countries, first-order admins)
- LLMs produce content, so laws that reduce content protections can undermine value of LLMs



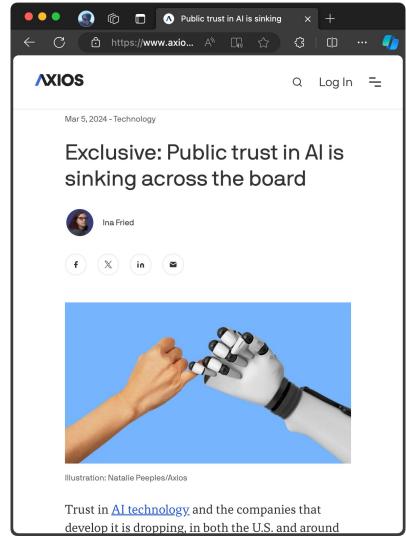
Policy leverage: Simple, realistic policy changes can reorder the LLM world overnight





Reputational leverage: Already having a significant effect

- Statements with legitimacy from trusted content institutions will matter
- Proposed solutions to bring trust up to higher levels will matter
- Lots of opportunity here for win-wins



Data leverage: Strikes and CDC







Data Strikes



Data Poisoning



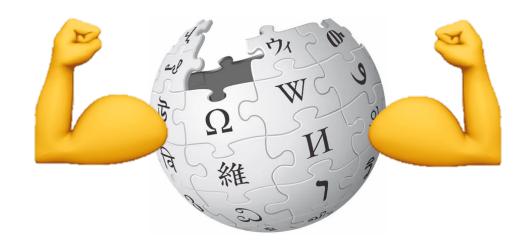
Conscious Data Contribution (CDC)



The current LLM paradigm is not the only one

- Parametric uses of content
 - Collective bargaining with content producer unions
 - Sustaining content deals with large content providers
- Non-parametric uses of content (the future?)
 - Just-in-time content use (e.g. Min et al. 2023)
 - Style markets (<u>Crawford 2022</u>)
 - Provenance tracking (<u>Lanier 2023</u>)
 - Per-use or per-time subscription models
 - Cloud platforms can implement "model + content buffet"-style offerings

Use leverage takes effort! ("no pain no gain")



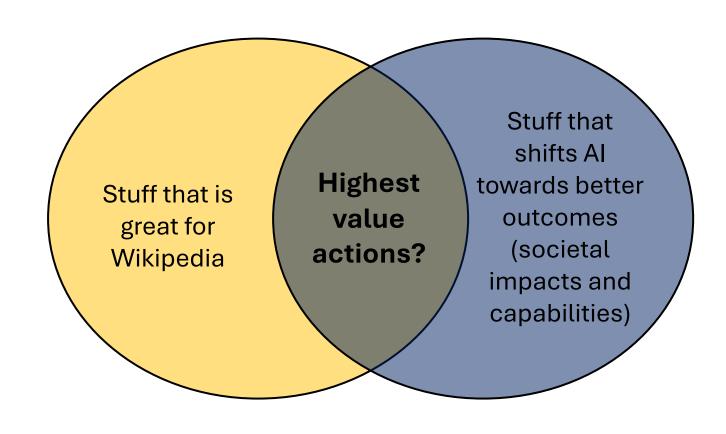
This logo didn't get these muscles overnight!

How: Values vs. tactics

- Some things that are values of the community look more like tactics through an AI lens, in particular "open"
 - LLMs should encourage us to ask what do we want to accomplish with "open"?
 - "Sum of all knowledge" → ensuring equal access to the benefits of open knowledge?
- What does "open" as a value vs. a tactic look like in the AI era?
 - Behavioral licenses like "Responsible AI Licenses" interesting to consider
 - Licensing likely only part of any solution here, and enforcement essential
 - Other types of leverage also needed

How should Wikipedia use its leverage?

- For itself
 - Ensure Wikipedia stays healthy in the AI era
 - Incentivize AI systems that use Wikipedia to reflect Wikipedia values
- For others: Wikipedia likely one of the best opportunities to shift the AI market towards better outcomes more generally



Wikipedia values can help AI thrive

- Free and accessible repository of human knowledge: Wikipedia faces similar challenges to all other open content producers
 - Solve it for Wikipedia, solve it for everyone?
- No original research: Wikipedia has strong stake in high-quality globally accessible information ecosystem
 - Need people to be incentivized to share knowledge in the open
- Citation/verifiability: Provenance a key step in passing value to content producers
 - Current LLM paradigm does not prioritize citation/provenance, but there are technical paths forward
- What others? For discussion!

A brief outline

- Looking back: Wikipedia's central role in the development of modern AI (case study from my career)
- The present conundrum: The dominant LLM paradigm threatens Wikipedia, large portions of the content ecosystem, and ultimately itself.
- Looking forward: What Wikipedia can do to help itself in the LLM era, and make the LLM era much better in the process

Thank you! Questions?

- Reach out at <u>brent.Hecht@Microsoft.com</u> or bhecht@northwestern.edu
- Links to papers available at my website: brenthecht.com
- Special thanks to three former students who led much of the work discussed: Isaac Johnson (WMF), Nick Vincent (Simon Fraser), Hanlin Li (U Texas)
- Deck available at: <u>brenthecht.com/wikiworkshop2024</u>