# 国際シンポジウム「大学における研究データ管理の 意義と支援人材育成」

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# Big Data Use in the Humanities and Social Sciences:

# The Case of HathiTrust Research Center

Presented by
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University of Illinois Urbana-Champaign

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Twitter: @profdownie





# Takes many hands

This slide deck is a communal effort of HTRC community members, and includes contributions from:

- Ryan Dubnicek
- Glen Layne-Worthey
- Niko Parulian
- Ming Jiang
- Ted Underwood
- Dan Evans
- Boris Capitanu
- David Bainbridge
- Peter Organisciak
- And, many, many other collaborators and colleagues



# **Challenges & opportunities**

How to manage the scale and scope of the HathiTrust Digital Library?

How to *leverage* its scale and scope?

How to facilitate research given significant copyright constraints?

How better to enable work with *non-English* (and non-Latin, and R-to-L, etc.) languages?

How to *curate* a collection like this, to make it more *diverse*, *inclusive*, & *reflective* of the world of knowledge and books?

How best to manage the data that is generated by scholarly exploration

# HathiTrust is...



A partnership of member libraries (Keio University the only one in Japan)

A collective digital library (the largest ever assembled)
(UIUC's books are among them:
about a million of them)

The fruits of large-scale digitization initiatives

A trusted digital repository

A provider of special services

(e.g., the "Emergency Temporary Access Service")

Host of the HathiTrust Research Center

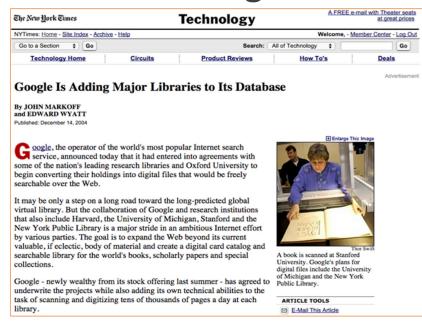


### **Currently Digitized:**

- 17,645,865 total volumes
- 8,484,623 book titles
- 469,920 serial titles
- 6,176,052,750 pages
- 791 terabytes
- 209 miles
- 14,337 tons
- 7,048,962 volumes(~40% of total) in the public domain



# **HathiTrust Origin stories**



December 2004:

Google & five research libraries announce massive book scanning project.

September 2005:

Authors Guild files a lawsuit against Google and the libraries for "massive copyright infringement."



# The HathiTrust Digital Library

Around 50% English (but including over 450 languages)

From the 15th to 21st century (but with a strong plurality from the 20th century)

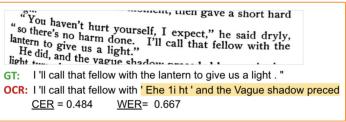
Around 61% in copyright or status unknown

Range of genres: fiction, history, science, government documents, and more

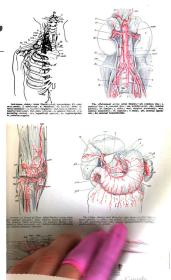
Items all contributed by HathiTrust member libraries

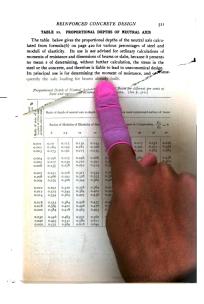
# But it's not perfect...

- ➤ Mass digitization
  - > (3+ billion page turns by thousands of scanning staff
- > Minimal curation
- Uncorrected OCR



Ming Jiang, et al. [Untitled manuscript], 2022





"Art of Google Books" http://theartofgooglebooks.tumblr.com/ Accessed October 25, 2018

# ...and it's not comprehensive or appropriately diverse

The HathiTrust Digital Library is wholly dependent on what member institutions provide, so...

...academic library acquisition patterns determine its content, e.g.,

A dearth of romance novels

See Katherine Bode, "Why you can't model away bias," MLQ, 2019

Substantial gaps in Black speculative fiction

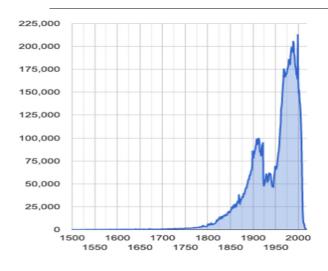
See Indiana University, "Corpus Completion Survey," https://sites.google.com/iu.edu/coverage/

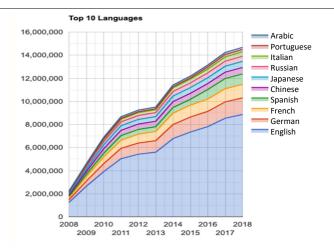
...echoing the strengths & weaknesses of academic libraries writ very large: lots of copies of lots of editions of lots of titles by Jane Austen

... but not nearly enough Octavia Butler



# HathiTrust collections over time





Publication dates of items in HathiTrust

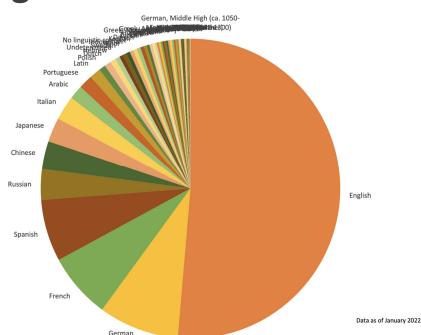
Languages in HathiTrust (by year of ingest, as of 2018)

HathiTrust languages

(as of 2022)

languages

Top 83 HathiTrust (>1,000 volumes each)



# HATHI TRUST Research Center

# HathiTrust Research Center

Co-hosted by Indiana University and the University of Illinois

Computational analysis of the HathiTrust Digital Library collections

Development and promotion of *non-consumptive research* 

- Collection-building
- Data crunching
- Creation of new text-mining interfaces and tools
- Deeper understanding and enhancement of our collections

https://analytics.hathitrust.org/



# Non-consumptive\* research

From the rejected "Authors Guild v. Google Books" Settlement:

"'Non-Consumptive Research' means research in which computational analysis is performed on one or more Books, but not research in which a researcher reads or displays substantial portions of a Book to understand the intellectual content presented within the Book."

## What and why?

- Complies with copyright law
- Foundation of HTRC work
- Related term: \*non-expressive use

### How?

- Partial Access
- Transformative Access
- Capsule Access



# Non-consumptive research examples

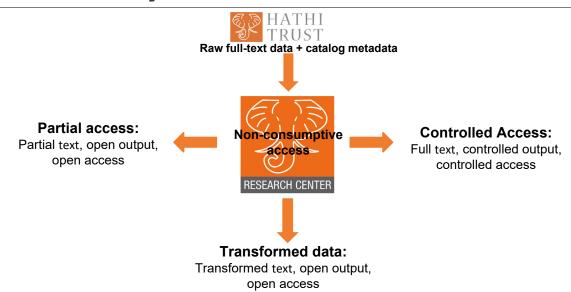
- Machine learning and AI
- Text extraction
- Textual analysis and information extraction
- Linguistic analysis
- Image analysis
- File manipulation
- OCR correction
- Indexing and search

More here: <a href="https://www.hathitrust.org/htrc">https://www.hathitrust.org/htrc</a> ncup



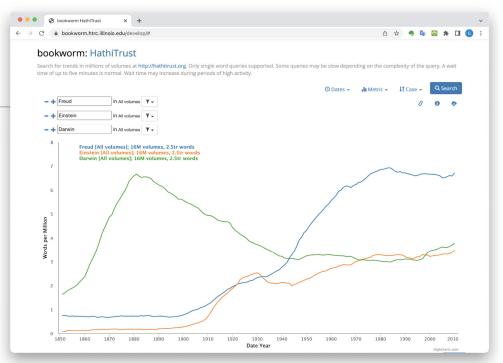
# **HTRC Analytics**

# https://analytics.hathitrust.org





# **Bookworm**



https://bookworm.htrc.illinois.edu

# "Gendered Characterizations" visualization

Word usage in English-language fiction over time...

...in descriptions or in dialogue

...comparing the **genders** both of a word's associated fictional

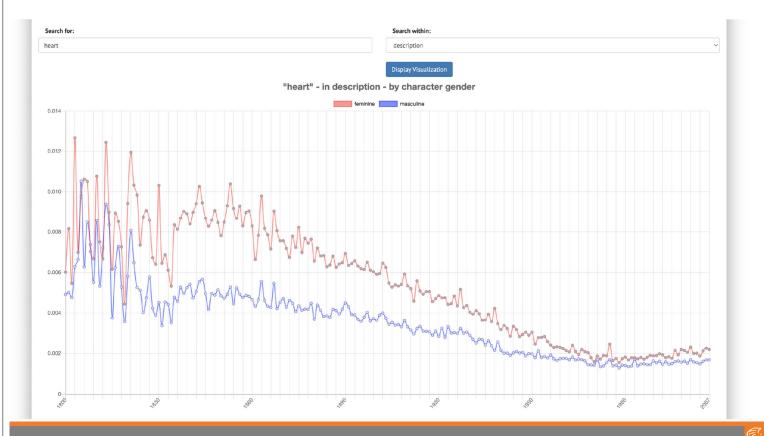
# characters

...and of its authors.

Example: "heart"

https://tools.htrc.illinois.edu/genderviz/







# **HTRC Extracted Features**

- A dataset derived from the entire HathiTrust corpus
- Volume-level, page-level, word-level data
- JSON format: structured data
- Copyright-free (downloadable in part or as a whole by anybody anywhere!)

https://analytics.hathitrust.org/datasets





# HTRC Extracted Features Dataset

Page-level features from 17.1 million volumes [v.2.0]

# **Attribution**

Jacob Jett, Boris Capitanu, Deren Kudeki, Timothy Cole, Yuerong Hu, Peter Organisciak, Ted Underwood, Eleanor Dickson Koehl, Ryan Dubnicek, J. Stephen Downie (2020). *The HathiTrust Research Center Extracted Features Dataset (2.0)*. HathiTrust Research Center.https://doi.org/10.13012/R2TE-C227

This feature dataset is free is released under a Creative Commons Attribution 4.0 International

License. 2.906.81

17,123,746

10,550,952 6,572,794

6,221,631,336

2,906,819,723,689



# **Extracted Features dataset**

```
body": {
  "tokenCount": 433,
  "lineCount": 89,
  "emptyLineCount": 22,
  "sentenceCount": 13,
  "tokenPosCount": {
    "1": {
    "CD": 2
}
                                            },
"2": {
"CD": 2
                                            },
"3": {
"CD": 2
 1265
1266
                                            },
"4": {
"CD": 1
 1267 -
 1268
 1269
 1270 -
1271
                                            },
"6": {
"CD": 1
 1272
},
"est": {
   "NN": 3
                                             },
"entirely": {
"RB": 1
                                             },
"quality": {
"NN": 1
                                             },
"clich6": {
"FW": 1
```

# **Other Derived Datasets**

# **Word Frequencies in English-Language Literature, 1700-1922** (Ted Underwood)

- Contains word frequencies for all English-language volumes of fiction, drama, and poetry
- Contains other volume metadata

# Geographic Locations in English-Language Literature, 1701-2011

(Matthew Wilkens and Guangchen Ruan)

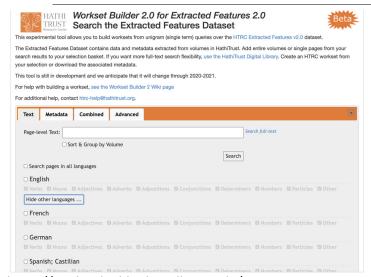
- Contains volume metadata as well as geographical locations
- Based on similar set of volumes in Ted Underwood's derived dataset

https://analytics.hathitrust.org/datasets

1



# Workset Builder: fine-grained search



https://worksetbuilder.htrc.illinois.edu/

AN OPEN DATA APPROACH TO REVEALING INDIGENOUS TEXTS IN LARGE-SCALE DIGITAL REPOSITORIES: A CASE-STUDY OF LOCATING PAGES OF MĀORI TEXT IN THE HATHITRUST

XMI

### 1. ABSTRACT

In this case study we report on our experiences in locating pages of M?ori text in the HathiTrust Digital Library (HTDL). Using traditional biographic metadata, i.e., the language field, only 182 items were returned out of HTDL's 17.1 million volumes. Our Open Data approach is based on the freely available HathiTrust Extracted Features Dataset. We establish a collection of high frequency terms in Te Reo Māori, which we iteratively use as search terms to identify a group of candidate texts. We then apply NLP analysis to verify those texts that contain substantial amounts of the Māori language. Using this approach we were able to increase the number of volume returned to 598. This positive result suggests that scholars who want to analyse other low-resourced languages should be able to adopt our workflow to reveal otherwise hidden texts in their desired languages.

David Bainbridge (davidb@waikato.ac.nz), University of Waikato, New Zealand, J Stephen Downie (jidownie@illinois.edu), University of Illinois USA and Hemi Whaanga , University of Waikato, New Zealand



# Worksets

Workset = a **custom dataset** of HathiTrust volumes for analysis in HTRC

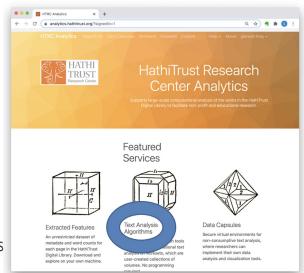
A simple list of volume IDs selected for analysis

- Transferable across HathiTrust systems
- Doesn't include the text

Citable, shareable, reproducible research

Analyze data from the HTDL without access to full text from HTDL

Researchers see only their workset and analyses



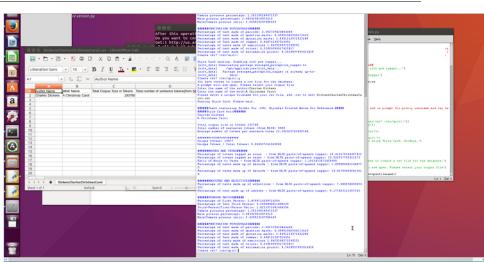
23



# Data Capsules for controlled full-text access

Secure analysis
environments
Linux virtual desktop

Protocols for data import & export



https://analytics.hathitrust.org/staticcapsules



# Working on challenges



# HathiTrust languages, a deeper dive: Japanese

• Deliberately targeting difficult cases: the page-level (NLP) metadata identified the text as Japanese But the the volume-level (human catalog) metadata was anything BUT Japanese

From manual classification of 400 randomly sampled pages:

- 6% were front cover images
- 6% were handwriting
- 19% were blank containing (some dirt marks)

### Additionally:

- 19% of pages were horizontal in orientation, 59% vertical, remainder could not be determined
- 46% of the sampled pages contained Kanji script
- However only 1 of them was found to be in the Japanese language

David Bainbridge, Genna Hilbing, Ming Jiang, Yuerong Hu, Glen Layne-Worthey, J. Stephen Downie, A Study on the Accuracy of OCR- and NLP-based Detection of Japanese Text in the HathiTrust Extracted Features v2.0 Dataset, DH2022 (Tokyo, Japan)



# **Identifying front-matter pages** algorithmically

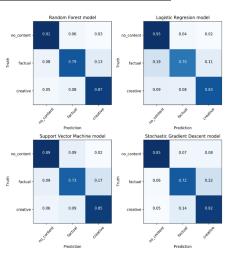
Identifying Creative Content at the Page Level in the HathiTrust Digital Library Using Machine Learning Methods on Text and Image Features

Nikolaus Parulian & Glen Worthey HathiTrust Research Center School of Information Sciences, University of Illinois Urbana-Champaign

iConference 2021







Benchmarking NLP performance

on uncorrected OCR

# A Prototype Gutenberg-HathiTrust Sentence-level Parallel Corpus for OCR Error Analysis: Pilot Investigations

### **ABSTRACT**

This exploratory study proposes a prototype sentence-level parallel corpus to support for the study of optical character recognition (OCR) quality in curated digitized library collections. Existing data

resources, such as I ally aligned content document-based or lenge of studying OC tic features like sente

The massive digitization of physical prints through machine scanning and optical character recognition (OCR) is of crucial importance to cultural heritage, knowledge preservation and general

1 INTRODUCTION AND BACKGROUND

I stepped forward and looked through a chink between the logs . OCR: I stepped forward an 1 r'ooked throuj^di a chink I)ctween tlie logs, I- stepped forward and l----ooked through---- a chink be---tween th--e logs

Figure 2: Visualization of GT-OCR character sequence alignment for token-based error detection. Symbol "-" denotes a gap character in the returned alignment. "|" denotes the position of aligned characters.

Ming Jiang, et al., JCDL 2022

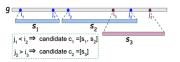


Figure 1: Aligned OCR candidate preparation, where g is a GT sentence,  $\{s_1, s_2, s_3\}$  are OCR snippets, i, j denote the starting

and ending pos	ition of any $g$ – $s$ common string, respectively					
Algorithm 1:	Sentence Alignment					
Input: G: GT	sentence list, <u>H</u> : OCR snippet list					
<b>Output:</b> <u>P</u> =[(	g, h)]: Sentence pair list					
1 for each sente	ence g in G <b>do</b>					
2 Retrieve a	sub-list of OCR snippets $H'$					
3 for each s	nippet $s$ in $H'$ <b>do</b>					
4 Comp	ute similarity score $Sim_g = \frac{ g \cap s }{ g }$					
5 Comp	ute similarity score $Sim_s = \frac{ g \cap s }{ s }$					
6 Find t	he longest common string $g[i:j]$					
7 end	end					
8 Get candi	Get candidate pair list $C = [(c, \sum Sim_q, \{Sim_s\})]$ (see					
Figure 1)	Figure 1)					
9 Find the o	Find the optimal candidate with $Max(Sim_q)$ from $C$					
	if $Max(Sim_q) > threshold A$ and $Avg[Sim_s] > threshold$					
B then						
11 Add th	nis candidate into P					
12 end						
13 end						



# What about under-represented literatures?





### Scholar-Curated Worksets for Analysis, Reuse and Dissemination (SCWAReD)

A 3-year Mellon Foundation-funded project to collaborate with scholars to assemble and document worksets of traditionally under-resourced and/or marginalized textual communities.

Flagship partners The Project on the History of Black Writing at the University of Kansas, led by co-PI Dr. Maryemma Graham, will workset of all African-American fiction in HTDL, based on manually-verified list of Black authors and works at Project HBW.

### Other partner projects:

- Mining the Native American Authored Works in HathiTrust for Insights Dr. Kun Lu, Dr. Raina Heaton (University
  of Oklahoma), and Dr. Raymond Orr (Dartmouth)
- The Black Fantastic: Curated Vocabularies, Artifact Analysis and Identification Dr. Clarissa West-White (Bethune-Cookman University) and Dr. Seretha Williams (Augusta University)
- Creating Period-Specific Worksets for Latin American Fiction Dr. José Eduardo González (University of Nebraska, Lincoln)
- The National Negro Health Digital Project: Recovering and Restoring a Black Public Health Corpus Dr. Kim Gallon (Brown University)





### **General SCWAReD workflow:**

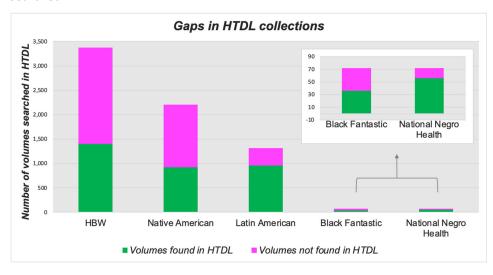
- Work with expert scholar collaborators to identify or generate a list of volumes relevant to their domain
- Search HTDL for these volumes using computational methods:
  - o Author-title search of metadata records using regular expressions
  - Keyword analyses
  - o Training machine learning classifier (more later!)
- Create workset of resultant volumes, including:
  - Workset inclusion rationale
  - Documentation of search methods, found and missing volumes
- Conduct exploratory data analysis of initial worksets:
  - Keyword + context analysis
  - Sentiment analysis
  - Entity extraction
  - o Topic modeling
- Share results





### **SCWAReD** worksets

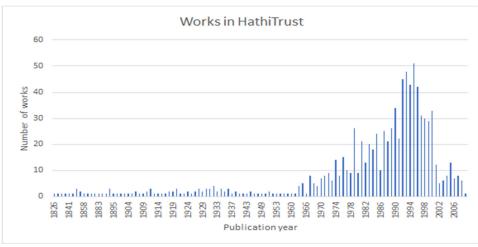
Searches have yielded the following found volumes from author-title lists that were searched:



Work is ongoing to pilot how to fill gaps where possible and to do final verification on found volume lists. An announcement will come with final numbers and metadata!

### **Native Authored Works - Early Exploratory Analysis**

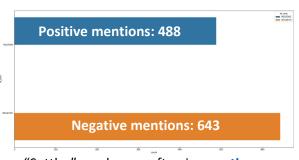
- The vast majority of works about Native peoples are not written by Native peoples (compare to 35,445 HT records for "American Indian")
- Native authorship was very limited prior to 1960
- Most Native authors in North America represented in HTDL are writing in English
- Found authors from ~207 federally recognized tribes represented in the database (compare with ~1,208 recognized Native Nations and First Nations in the US and Canada)



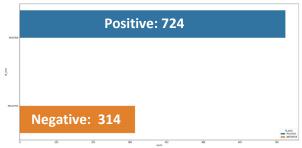
### **Native Authored Works - Early Exploratory Analysis**

Sentiment discrepancy between "settler" and "pioneer" when extracted within their contexts (sentence level).

Still needs to be explored in-depth by domain experts!



"Settler" used more often in **negative** context



"Pioneer" used more often in positive context

### **Black Fantastic - Early Exploratory Analysis**

### **Black Fantastic** literature:

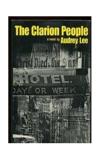
- Is written by artists of the African Diaspora
- Can be defined using Richard Iton's definition: fiction engaged with the intersections of race and technology, as well as transcultural iterations of world-building <sup>3</sup>
- Is sometimes referred to as "Afrofuturism"

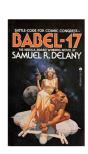


Richard Iton









### **Black Fantastic - Early Exploratory Analysis**

"Time" and time words are key terms in BF texts:

title	keyword	word_vol_rank	vol_words	tf-idf normalized	
Night studies : a novel /	time	1	6426	0.04255064906	
Over Edom, I lost my shoe /	time	1	3466	0.01673340132	
I want a black doll.	time	2	3117	0.01154263193	
Kindred /	time	3 2128		0.01891898843	
The survivors	time	3	2311	0.01379651864	
The salt eaters /	time	3	3137	0.01202072911	
The landlord.	time	3	3002	0.01017663998	
Tragic magic : a novel /	time	3	1239	0.009220445623	
The militants /	time	3	1799	0.008947247234	
Sweet whispers, Brother Rush /	time	4	1526	0.0122939275	
The spook who sat by the door : a novel /	time	4	2073	0.009220445623	
Light ahead for the Negro /	time	4	1040	0.004439473819	

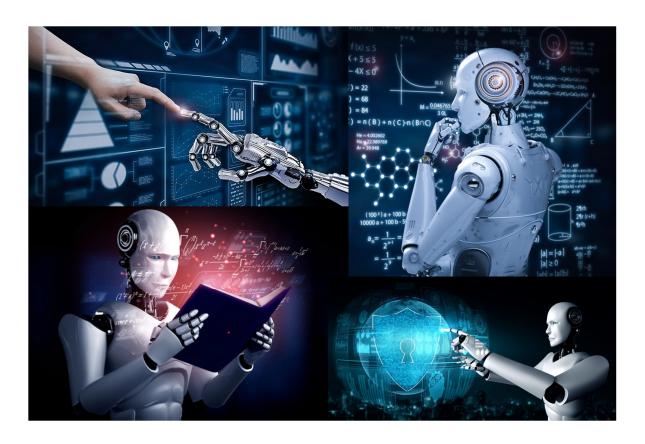
### **Black Fantastic - Early Exploratory Analysis**

Black Fantastic as a genre is complex and encompasses many different styles and themes.

A.K.A. a number of our exploratory ideas didn't yield significant results

- Topic modeling was inconclusive
- Term analysis only surfaced strong links to time terms
- Black Fantastic genre was complex enough to fool a classifier (more on this later!)

But there remains a lot of data to be analyzed and questions to be asked!



HTDL's massive amount of varied data presents possibilities for testing machine learning (ML) approaches for practical library tasks

Two projects at HTRC leveraging machine learning:

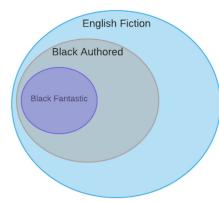
- Uncovering the Black Fantastic
  - Led by Nikolaus Nova Parulian, and in conjunction with Dr. Clarissa
     West-White and Dr. Seretha Williams as part of SCWAReD project
  - Can ML identify genre fiction (using Black Fantastic as a case study) within larger sets of fiction?
- Extending NovelTM Datasets for English-Language Fiction
  - Working with Dr. Ted Underwood, co-author, with Patrick Kimutis and Jessica Witte, of the initial NovelTM dataset
  - Seeks to improve upon initial classification process and extend the dataset to include items added since initial dataset was created

### The challenges of genre fiction:

- Variation between genres could/will be minute
- Genre is not something about which even humans agree
  - o Complicates creating/assembling training data

### Classification goal:

- Test results of training a classifier to differentiate between English-language fiction, Black-authored fiction and Black Fantastic fiction
- Benchmark performance of 4 predictive models:
  - Support Vector Machine (SVM)
  - Random Forest (RF)
  - Logistic Regression (LR)
  - Stochastic Gradient Descent (SGD)



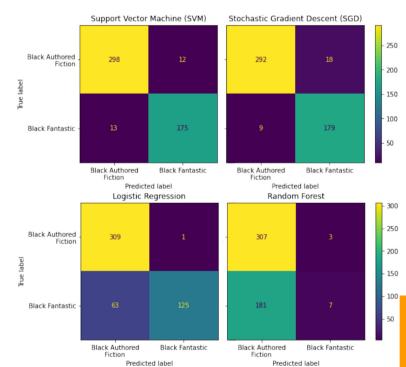
### **Uncovering Black Fantastic - The Data**

We train the classifier using HTRC EF data for each volume that is then vectorized (converted to numerical representations of relationships between words at the volume level) using TF-IDF

Workset Type	# # Tokens Volumes		# Unique tokens	Average. volume occurrence count for token	
Black Authored (BA) Fiction	1,556	325,214	213,547	12.21	
General English-Language (EL) Fiction	1,524	255,053	141,443	11.14	
Black Fantastic (BF) Fiction	37	94,034	28,204	4.33	

## **Uncovering Black Fantastic - The Results**

- SVM performed the best
- However, overall there is not enough data for a task like this to be useful from a practical standpoint of finding unknown Black Fantastic volumes in the HTDL
- More work to be done with more training data



# **Uncovering Black Fantastic - Word Feature Topic Clustering**



# UNCOVERING BLACK

**UNCOVERING BLACK** 

### **Extending NovelTM Datasets for English-language fiction**

The initial NovelTM project used ML to find English-fiction in the HTDL

### Why?

- The HTDL is massive, and metadata is missing, inconsistent, and incomplete.
- Even if metadata is complete:
  - standards have changed and evolved over time, leading to inconsistent metadata
  - we did not start explictly cataloging fiction as a genre until the 1990s\* and still today we often do not do so with sufficient granularity (e.g. to find "sci-fi" or "romance" easily)
- These problems are significantly exacerbated by scale, especially that of the massive HTDL

974 UU #bTXU #cTXU #d 20190701 #s google #u txu.059173024387941 #y 1952 #ric #q bib #t non-US bib date1 >= 1926

### The satanic verses 00630nam a22002171a 4500

# | Michael | Mich 001073126 MiAaHDL 035 U U \$\frac{1}{2}a (MiU)990015225940106381 040 ☐☐ ‡c SAZTEC 100 ☐☐ ☐☐ ‡a Borges, Jorge Luis, ‡d 1899-1986.

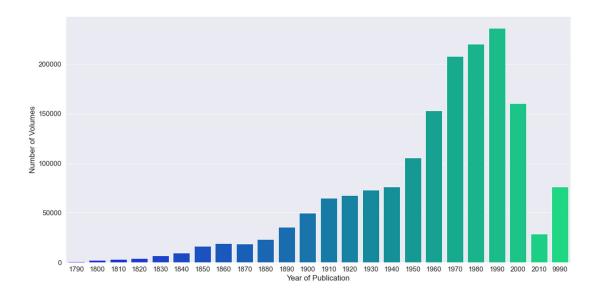
### Madame Bovary; mœurs de province

		LDR 00751nam a22002291 4500						
		001 001015020						
		003 MiAaHDL						
		005 20211114000000.0						
		006 m d						
		oo7 cr bnauaua						
		008 890208s1857 fr 00000 fre d						
		010 LLL ‡a 12027102						
-El	Aleph	035 LLL ‡a (MiU)990010150200106381						
	i ii opii	035 LLL ‡a sdr-miu.990010150200106381						
I DE	00521	035						
LDF		035 LLL ‡a (SAZTEC)095307540						
001		035 ULL ‡a (CaOTULAS)176049956						
- 003		035 UU \$\frac{1}{2}(MiU)Aleph001015020						
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<sup>\*</sup> Miller, David P. "Out from Under: Form/Genre Access in LCSH." Cataloging & Classification Quarterly 29, no. 1–2 (June 1, 2000): 169– 88. https://doi.org/10.1300/J104v29n01\_12.

### **Extending NovelTM Datasets - But Why Now?**

About 1.6 million new English-language items that might be fiction have been added to HTDL since the first NovelTM dataset was generated



### **Extending NovelTM Datasets - The Data**

We use a sample of volumes to train and test the classifier, and we benchmarked 3 different samples:

- **Sample 1:** 10,108 random volumes, matching distribution of items added to HTDL since 2016, by decade, yielding 1,605 fiction, 8,503 non-fiction.
- **Sample 2:** 9,969 random volumes, with the same selection logic as sample 1, but incorporating as many manually-verified fiction vols from NovelTM dataset as possible, yielding 1,580 fiction, 8,389 non-fiction volumes
- Sample 3: 9,061 volumes, including 53 F and 211 NF volumes for every decade represented in items added to HTDL since 2016, creating a train/test with equal volumes for each decade, yielding 1,328 fiction, 7,733 non-fiction volumes

We also benchmarked 3 different models:

- Logistic regression
- Support vector machine
- Random forest (120 trees)

Each model was implemented using scikit-learn in Python

### **Extending NovelTM Datasets - The Results**

Each model and sample did well, but the best overall (F1) was LR on the corrected Sample 3.

	Logistic Regression			Support Vector Machine			Random Forest		
	Р	R	F1	Р	R	F1	Р	R	F1
Sample 1	0.7838	0.9755	0.8692	0.8384	0.9205	0.8776	0.8665	0.8930	0.8795
Sample 2	0.8589	0.9470	0.9008	0.8885	0.9238	0.9058	0.8824	0.8940	0.8882
Sample 3	0.8804	0.9199	0.8997	0.9286	0.8750	0.9010	0.9697	0.8889	0.9275
Sample 3 - Corrected	0.9249	0.9506	0.9376	0.9702	0.9043	0.9361	0.9689	0.8642	0.9135
Mean values	0.8620	0.9483	0.9018	0.9064	0.9059	0.9051	0.9219	0.8850	0.9022

### **Extending NovelTM Datasets - The Results**

While reviewing errors, 4 main types were identified:

- Incorrect ground truth: volumes incorrectly tagged as fiction or not fiction in their metadata. Examples include Stephen Crane's The Red Badge of Courage or Wuthering Heights by Emily Bronte incorrectly being marked as not fiction.
- Volumes that blur the lines of fiction, such as memoir, biography, or travel narrative: volumes that look a lot like fiction or not fiction, but are the inverse. Examples include Daniel Defoe's Robinson Crusoe or John Hanning Speke's Journal of the Discovery of the Source of the Nile—the former being a "fake" travel narrative and the latter purporting to be authentic.
- Non-prose fiction: volumes that looked like fiction, but are a form separate from prose, and thus not correct for our dataset. Examples include books of poetry and dramas.
- True errors: the least frequent errors—volumes the model just got plain wrong. Examples include annotated volumes, containing fore- and/or afterwords that can influence the model, like The Works of Dr. Jonathan Swift from 1751. Other examples are Ward Greene's collection of prominent historical news stories, Star Reporters and 34 of Their Stories, or a bound anthology of Frank Leslie's Lady's Magazine.



# Tools for Open Research and Computation with HathiTrust: Leveraging Intelligent Text Extraction (TORCHLITE)

- A 2-year NEH-funded project to develop API infrastructure for access HTRC's Extracted Features (EF) Dataset and a web-based dashboard of visualization widgets
- Work is ongoing on dashboard and widget design, API development
- Will include incrementally updated EF data, easier mode of access to HTRC EF data, and the ability to better create custom visualizations of HathiTrust and HTRC data both inside the browser and in custom code.

# DATASET

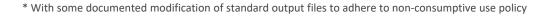
### **BookNLP dataset for English-language fiction**

Detailed derived data and metadata for each of ~213,000 English-language fiction volumes (from NovelTM set), in a fully open and public domain dataset.

### Data included for each volume:

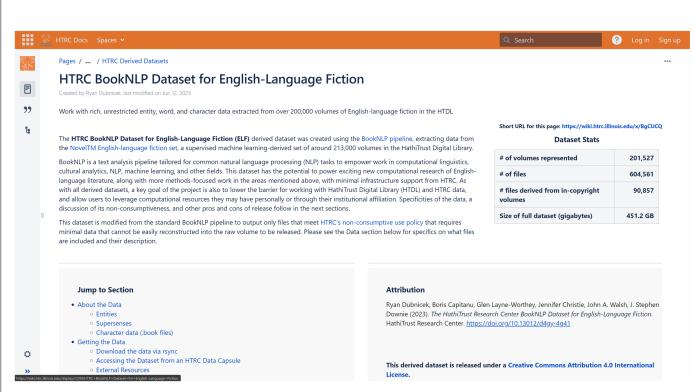
- Tokens and metadata about tokens \*
- Entities—people, places, organizations—in the text
- Quotations, speakers and metadata
- Supersense tags advanced semantic and entity tags based on WordNet
- .book file with the following info about every character mentioned 2+ times:
  - proper/common/pronominal references
  - referential gender
  - actions for the which they are the agent and patient
  - objects they possess
  - modifiers

BookNLP is a DH tool for book-length documents by David Bamman, Ted Underwood and Noah Smith. More info about BookNLP: <a href="https://github.com/booknlp/booknlp">https://github.com/booknlp/booknlp</a>





**David Bamman** 



Big Release Announcement

Short URL for this page: https://wiki.htrc.illinois.edu/x/BgCUCQ

# **Summary and conclusions**

- · Working at Hathitrust scale is a real big data challenge
- Combination of human expertise and AI/ML required to
  - Discover gaps
  - Build worksets
  - Correct metadata
  - Extract features
  - Overcoming OCR errors
  - Identify copyright issues
  - Analyze data
  - Curate and disseminate data and results



# ありがとう

Thank you