A Generic and Automated Staff Scraping Tool for School Webpages

Sara Alaoui and Haley Hunter-Zinck

Center for Optimization and Data Science (CODS) U.S. Census Bureau

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Project Motivation

- Data acquisition from the Web may offer several advantages when combined or even compared with vendor supplied data
 - 1. Control over timing
 - 2. Transparency
 - 3. Customizability
 - 4. Enhanced coverage



Project Motivation

- The National Teacher and Principal Survey (NTPS), sponsored by the National Center for Education Statistics (NCES), surveys K-12 schools and their staff
 - In order to sample teachers, the NTPS collects teacher rosters from sampled schools
 - Currently, schools submit a list of their school staff, verify a list of teachers obtained from commercial sources, or (if necessary) teachers are sampled from commercial sources
 - Sampled teachers are asked to complete a Teacher Questionnaire
- Alternate sources of data could augment, validate, and update schoolsubmitted rosters
 - Vendor supplied data
 - Data scraped from the Web



The NTPS web scraper consists of three major steps to generate the final payload





Find school websites via Google Places API to provide a starting website for each school

- Query with school name and address
 - Return most relevant Google Place
 - Request associated websites, address, and name annotations
- Some data quality concerns
 - No website
 - Broken links
 - Incorrect websites
 - District websites



Sampled private schools are more findable than public schools on the web via Google places API

Metric	Public	Private
Number of schools	10,000	3,100
Schools in sample with a returned website	90.4%	85.3%
Website accessible and relevant*	85.0%	96.7%
Estimated return rate	76.9%	82.5%



Step 2: crawl | Use the returned websites to locate and download directory pages



Staff directory pages are usually linked from the school's homepage with an intuitive label





We identify potential directory pages using a string similarity score



Finally, we construct a function that uses both the known expressions and their frequencies to estimate the likelihood that a page contains faculty directories

Expression	Frequency
Staff Directory	150
Staff	60
Faculty & Staff	20
Faculty and Staff	<15
Our Staff	<15
Teachers	<15
Faculty Directory	<15
Faculty	<15
Faculty & Staff Directory	<15
Teachers & Staff	<15
School Staff	<15

Curated list of expressions describing faculty directory pages for public schools



How well do we detect directory pages?

Directory detection: capture > 90% of directory pages with ~10% false positive rate at a string similarity threshold of ~ 0.9 for public and private schools





Downloading webpages - dynamic content

Before page loading:

School Staff

Dynamic content rendered upon user interaction



After page loading:

School Staff

1 2 3 4 ... > showing 1-4 of 65 staff

Teacher name Titles: Second grade teacher Emails: teacher@school.edu Phone number: 000-000-0000

Teacher name

Titles: Second grade teacher Emails: teacher@school.edu Phone number: 000-000-0000

Teacher name

Titles: Second grade teacher Emails: teacher@school.edu Phone number: 000-000-0000

Teacher name

Titles: Second grade teacher Emails: teacher@school.edu Phone number: 000-000-0000



Downloading webpages - pagination

Detecting pagination

- Look for typical patterns
 - A B C ... Z
 - 123...
- Extract links or xpaths from paginator elements
- Add to the download queue

Example of a paginated page

School Staff

1 2 3 4 ... > showing 1-4 of 65 staff

Teacher name Titles: Second grade teacher Emails: teacher@school.edu Phone number: 000-000-0000

Teacher name

Titles: Second grade teacher Emails: teacher@school.edu Phone number: 000-000-0000

Teacher name

Titles: Second grade teacher Emails: teacher@school.edu Phone number: 000-000-0000

Teacher name

Titles: Second grade teacher Emails: teacher@school.edu Phone number: 000-000-0000



Step 3: extract | Text from the HTML of potential directory pages is processed to retrieve school staff information



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Staff directory pages have different content and formats

	• • •	••••	• • • • • • • • • • • • •	• • • • • • • •
chool Staff			Position	<u>Title</u>
2 3 4 > showing 1-4	4 of 65 staff		Teacher 1 Teacher 2	Title A Title B
eacher name tles: Second grade teacher mails: teacher@school.edu hone number: 000-000-0000	Teacher name Titles: Second grade teacher Emails: teacher@school.edu Phone number: 000-000-0000		Teacher 3	Title C
		Ĩ.	Teacher 4	Title D
eacher name itles: Second grade teacher mails: teacher@school.edu hone number: 000-000-0000	Teacher name Titles: Second grade teacher Emails: teacher@school.edu Phone number: 000-000-0000		Teacher 5	Title E
			Teacher 6	Title F
			Teacher 7	Title G
		I	Teacher 8	Title H
			Teacher 9	Title I



First grade

Teacher name

Teacher name

Second grade

Teacher name

Teacher name

Teacher name

Teacher name

Teacher name

Third grade

Extract – named entity recognition (NER) We apply pretrained named entity recognition (NER) models in the CoreNLP page's HTML to highlight potential people, email, and title values	package on text extrac	cted from a
CoreNLP		
version 4.5.4		
Jane Doe math teacher janedoe@school.com John Doe science teacher johndoe@school.com		
Annotations — Inamed entities x	— Language — English	Submit
Named Entity Recognition:		
PERSON TITLE EMAIL PERSON TITLE EMAIL ORDINAL 1.0 NATIONALITY Jane Doe math teacher janedoe@school.com John Doe science teacher johndoe@school.com First Last English Mttps://corenlp.run/	TITLE EMAIL teacher firstlast@school.com	

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Extract – custom teacher title tagger

Motivation

- Pretrained NER models for TITLE entities from packages such as CoreNLP miss many teacher titles
- We manually curated teacher titles from school roster webpages and trained a custom teacher title tagger
- We represented each candidate text element with a features derived from the text and HTML element structure

Performance

•		••••••	
Precision	Recall	F1	
0.88	0.87	0.87	
	• • • • • • • • •	• • • • • • • • • • •	•••
	• • • • • • • •	• • • • • • • • • • •	••
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	• • • •		
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Extract – Parser Progress

Parsed payload for identified staff directory pages reveal smaller counts for private schools

Total count of parsed entities



Average count per page



Extract – Relationship Extraction



Represent the HTML content as a graph and traverse the HTML elements to find the shortest path between names and titles or other elements on a page.



Final Payload

In our final payloads, we extract more teacher names, titles, and emails from public than private school websites but capture more complete entries for private schools.





Conclusions

Limitations

- Scraping payloads limited by the ability to find a school website
- Pre-trained NER models built for full document text rather than webpage text
- Identifying pages with staff from multiple schools is difficult

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- We have developed an end-to-end web scraping and extraction pipeline for public and private school rosters
- The web scraping pipeline generalizes to many website formats
- Web scraped data can provide an alternative data source to augment traditional survey data collection methods



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NTPS • Maura Spiegelman Questions? sara.alaoui@census.gov		•	•••	• •	•	•••	•	•••	•	•••	•	•••		•	•••	
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 Maura Spiegelman Questions? sara.alaoui@census.gov 	<u>NTPS</u>	• • •	•••	• •	•	•••	•	•••	•	•••	•	•••	•	•	•••	
Questions? sara.alaoui@census.gov	Maura Spiegelm	nan	•••	• •	•	•••	•	•••	•	•••	•	•••	•	•	•••	
Questions? sara.alaoui@census.gov		• • •	•••	• •	•	•••	•	•••	•	•••	•	• •	•	•	• •	
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Supplemental Slides

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Extract – custom TEACHER_TITLE tagger

Methodology

- 3 discriminative features:
 - Structural information
 - Phrase embedding
 - Semantic distances
- Element-level predictions
- Hand + Machine-generated training data
- XGBoost model (based on cross validated model selection)
- ~ 2000 training examples ~ 400 websites





Extract – parser performance

We manually curated approximately 30 pages and assessed the overlap of values per page between the parsed data and curated data

Public schools



Private schools









Total runtime for the full public school sample of 9,627 pages is ~4.6 days

Step label	Step description	Runtime (s)
etl	Extract transform load for google results	0
setup	Database setup and initial Google results load	2
link	Link harvesting of subpage links from in-sample website URLs	~6103
directory	Directory page detection with gazetteer	17342
dynamic	Dynamic scraping, pagination detection, and page download	230675
text	Convert HTML download files to text strings	337
ner	Conduct named entity recognition (NER) on text strings	73744
parse	Parse names, titles, and emails from the HTML downloads	4340
post	Post-process the parsed results to clean and filter	18241
interim	Construct an interim payload of school names to teacher names	254
rel	Perform relation extraction to related names, titles, and emails	~50000
final	Construct the final payload relating school names to related names, titles, and emails	38
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Schools in sample drop off when websites are not found or no information is extracted



