

Government websites as data: A methodological pipeline with application to the websites of municipalities in the United States

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Overview

Home Mayor » City Council » Departments » About Erie » Important Documents

Welcome to the City of Erie, Pennsylvania.



OUR VISION FOR ERIE

Erie is a **Community of Choice**.
We celebrate our:

- Diverse cultures,
- Welcoming, vibrant neighborhoods,
- World-class downtown and bayfront,
- Schools of excellence, and
- Abundance of family-sustaining jobs.

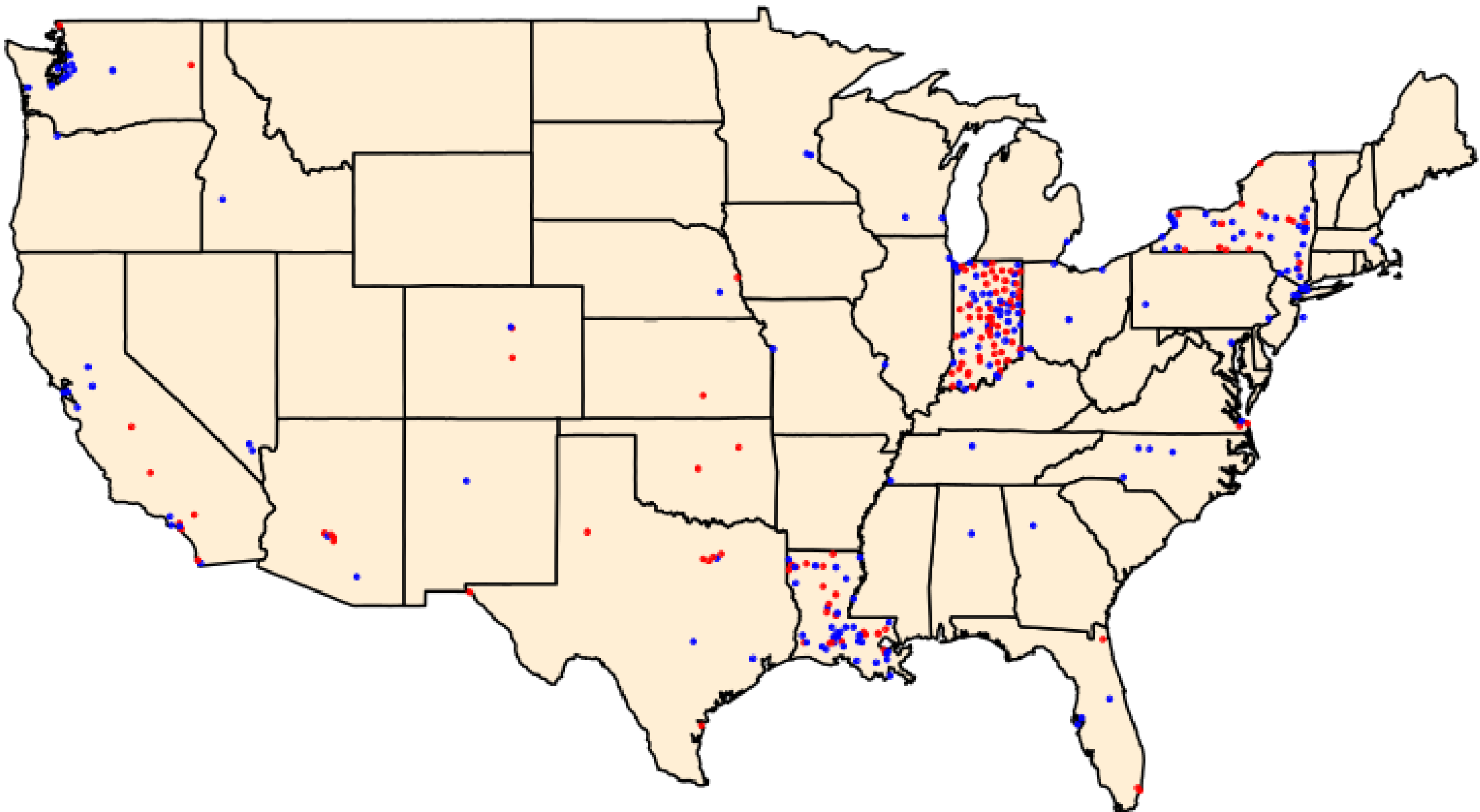
Contact the Mayor's Office

Joe Schember
JOSEPH V. SCHEMBER
Mayor, City of Erie

- ▶ Government websites contain important information
- ▶ Research has largely relied on manual coding
- ▶ Our contribution: pipeline for automated analysis
- ▶ Application: websites of partisan municipalities

Data Collection

- ▶ City URLs are scraped from Wikipedia and the GSA



Site to Text Conversion

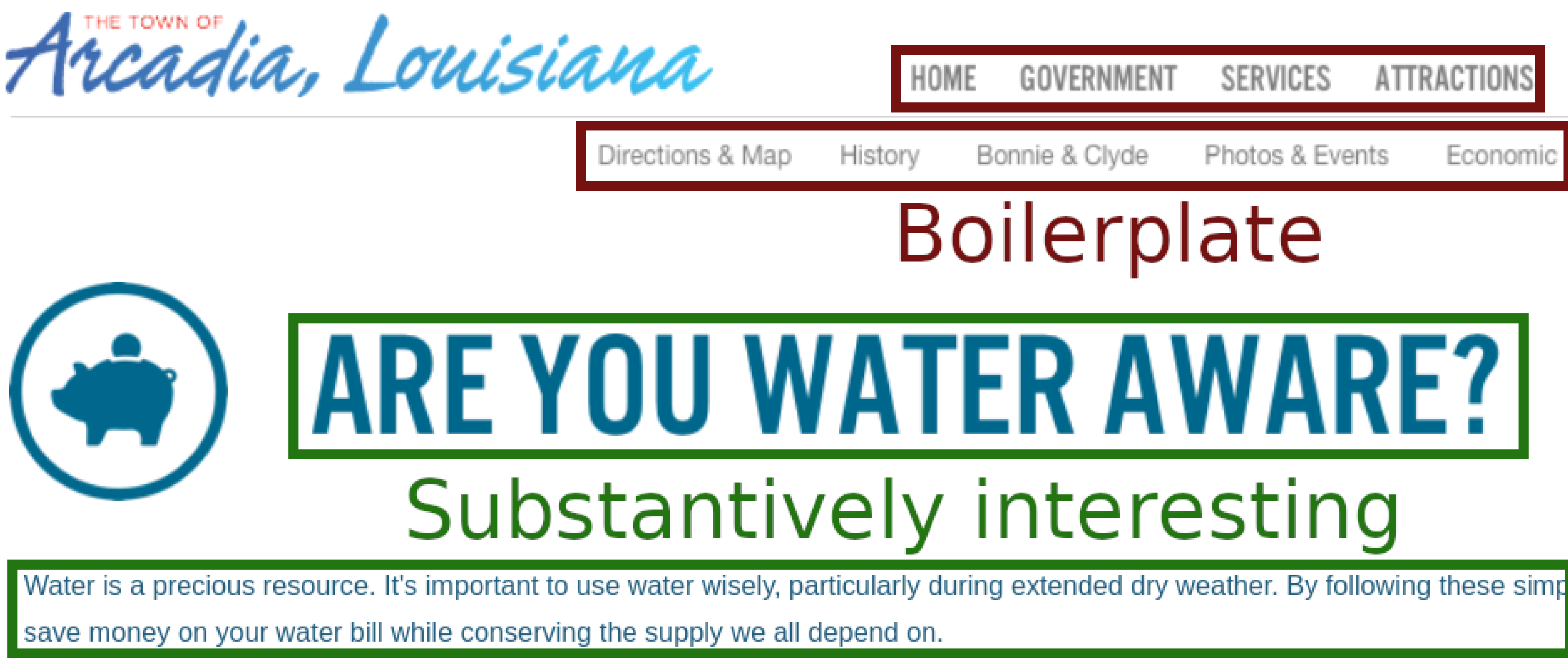
| Filetype | Occurances Before | Occurances After |
|----------|-------------------|------------------|
| html | 211682 | 887362 |
| pdf | 464842 | 638802 |
| jpg | 0 | 36958 |
| xml | 0 | 29638 |
| Other | 162681 | 9475 |
| ics | 435 | 8950 |
| png | 0 | 8863 |
| doc | 6972 | 8430 |
| txt | 317 | 6025 |
| | 793990 | 5234 |
| docx | 3137 | 4319 |
| TOTAL | 1644056 | 1644056 |

Table: Number of files per type, before and after detecing them via their magic number. The table shows that a lot of files originally have the wrong type, and that converting them correctly has a large impact on how many of them end up being usable.

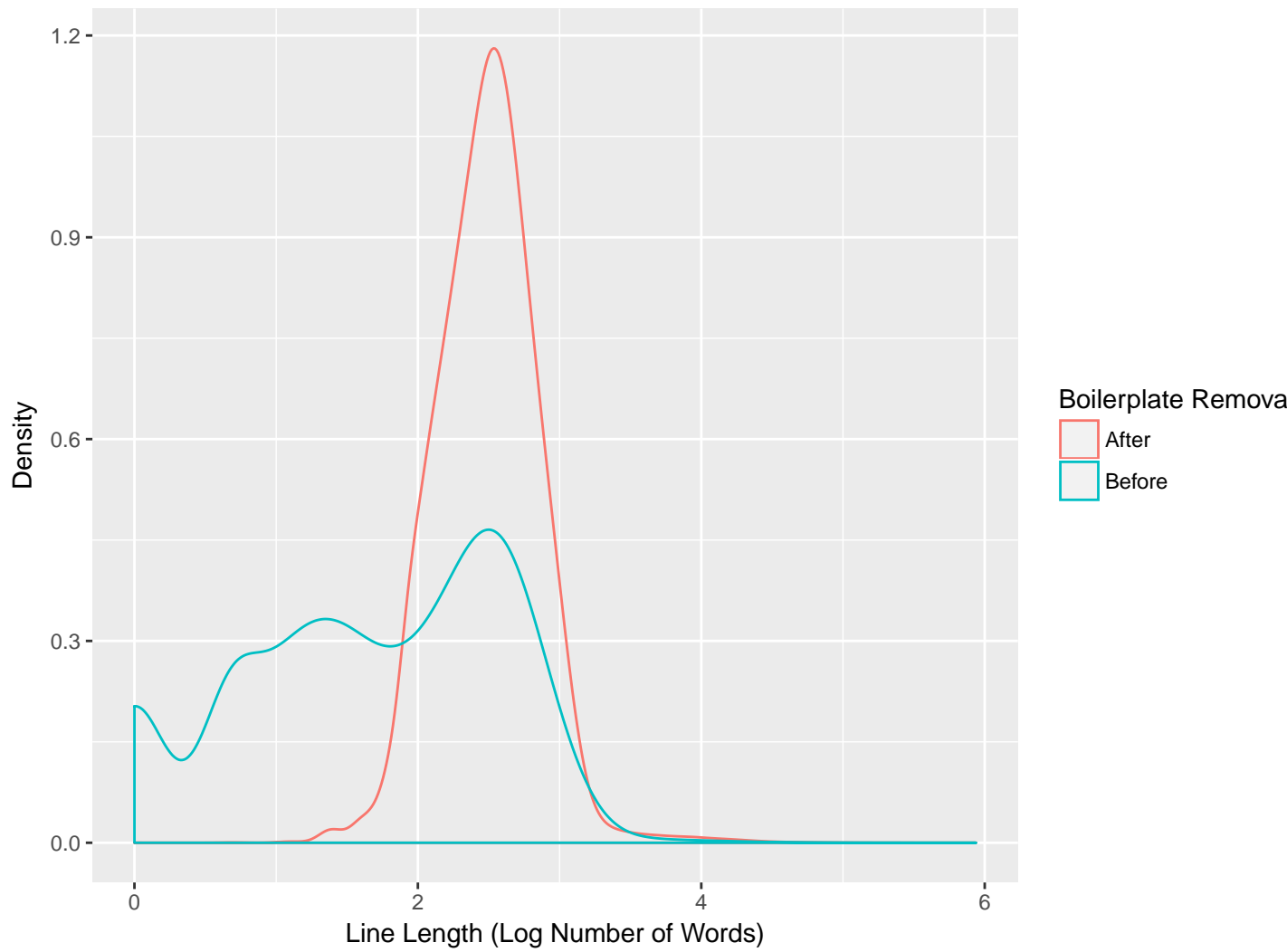
- ▶ The file endings from city websites are sometimes wrong
- ▶ We use file signatures to identify the correct type
- ▶ Preprocessing: lowercase; removal of punctuation, numbers, dates, etc.; removal of non-English words; lemmatization

Boilerplate Removal

- ▶ Websites contain a lot repetitive and uninformative content



- ▶ A classifier (random forest) is trained to identify and remove boilerplate lines



| | Value |
|-----------------------------|-------|
| Percent Correctly Predicted | 0.87 |
| Precision | 0.87 |
| Recall | 0.91 |
| F1-Score | 0.89 |

Table: Performance metrics for random forest boilerplate classifier.

Analysis

- ▶ Structural topic model with 60 topics
- ▶ Covariates: Party, state, population, median income

| # | Top Word 1 | Top Word 2 | Top Word 3 | Top Word 4 | Top Word 5 | Top Word 6 | Tokens assigned |
|----|-------------|------------|--------------|-------------------|----------------|----------------|-----------------|
| 42 | breastfeed | vaccine | infection | symptom | asthma | mosquito | 2497 |
| 17 | alarm | disaster | fire | rescue | preparedness | evacuation | 989 |
| 53 | drinking | wastewater | water | pipeline | pump | disinfection | 461 |
| 49 | bin | recycling | garbage | recyclables | recyclable | bag | 1791 |
| 7 | energy | garland | renewable | solar | electricity | climate | 742 |
| 23 | bid | proposer | bidder | contractor | subcontractor | contract | 447 |
| 57 | duct | conduit | bolt | splice | valve | fitting | 1373 |
| 13 | server | wireless | software | telecommunication | subscriber | desktop | 1092 |
| 1 | storm | runoff | infiltration | discharge | drainage | drain | 516 |
| 38 | youth | student | parent | teacher | immigrant | literacy | 714 |
| 59 | sampling | sample | analytical | concentration | hydrocarbon | toxicity | 1241 |
| 45 | premise | licensee | violation | license | permit | inspection | 509 |
| 60 | exhaust | fugitive | aircraft | airport | aviation | diesel | 731 |
| 4 | facade | awning | porch | roof | balcony | exterior | 1108 |
| 15 | complainant | defendant | allegation | complaint | allege | discrimination | 1384 |
| 14 | yes | agency | federal | recipient | compliance | entity | 205 |
| 25 | cannabis | marijuana | senate | dispensary | ballot | cultivation | 1188 |
| 47 | employee | overtime | sick | wage | grievance | bargaining | 511 |
| 37 | audit | auditor | internal | procedure | accountability | oversight | 420 |
| 21 | housing | affordable | homeless | homelessness | affordability | landlord | 318 |
| 34 | comment | draft | feedback | stakeholder | suggest | discussion | 289 |
| 19 | debt | bond | governmental | obligation | financial | accounting | 251 |
| 40 | bicycle | bike | lane | crosswalk | pedestrian | bicyclist | 574 |
| 32 | budget | revenue | expenditure | appropriation | fund | million | 242 |
| 48 | absent | aye | khan | nay | berry | voting | 528 |
| 31 | chair | agenda | commission | speaker | chairperson | committee | 314 |
| 55 | robbery | homicide | arrest | sergeant | suspect | burglary | 1395 |

Conclusion

- ▶ Republican cities feature information on basic utilities and protection from natural disasters
- ▶ Democratic cities feature information on policy deliberation, crime control and city budgeting
- ▶ Politics at the municipal level is not entirely non-partisan
- ▶ We plan to implement the pipeline in an R package