**Discovering Data Sets Through** Machine Learning: An Ensemble Approach to Uncovering the Prevalence of Government-**Funded Data Sets** 

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# Problem Setup

- Documents and labels from Coleridge Show US the Data competition: <u>Coleridge Initiative</u>
  <u>Show US the Data | Kaggle</u>
  - ♦ Over 14k documents with known data sets
  - ♦ Data sets are defined, but location within the document are not given
- ♦ We examine the first, second, and third place submissions the competition
  - ♦ Each submission took a unique approach to discovering data sets
  - ♦ We further develop and refine these approaches

### Data Set Format

Data Set Name

- Rural or urban residence was defined using the US Department of Agriculture's (USDA) 2003 Rural-Urban Continuum Codes.
- The county-level irrigated statistics provided by the USDA Census of Agriculture provided the spatial area target for the MIrAD-US model.
- Perhaps the best known is the USGS North American Breeding Bird Survey (http://www.mbr-pwrc.usgs.gov/), a continuing effort of about 40 years' duration.

### Data Set Format

- Data Set Name
- ♦ ["/']Data Set Name[,]["/']

- Data for total confirmed cases per capita were derived from the 'Our World in Data', which is a platform for global data on a broad range of conditions.
- According to the National Science Foundation "Survey of Earned Doctorates," 40% of doctoral degrees earned in political science in 2009 went to women.

### Data Set Format

- ♦ Data Set Name
- ♦ ["/']Data Set Name[,]["/']
- ♦ Data Set Name (DSN)

- We will draw an example from the Baltimore Longitudinal Study of Aging (BLSA), which is an ongoing research project of the National Institute on Aging of the NIH.
- Our analysis of the 2012 Program for the International Assessment of Adult Competencies (PIAAC) data indicated that, for ages 55-65, rates of AET participation are similar for both men and women.

### Data Set Format

- ♦ Data Set Name
- ♦ ["/']Data Set Name[,]["/']
- ♦ Data Set Name (DSN)
- ♦ Data Set (Alt-DSN) Name

- The Beginning Postsecondary Students (BPS) Longitudinal Study, in which you have been a continuing participant, is one of the major surveys used to provide this information.
- The NSF-NIH Survey of Graduate Students and Postdoctorates in Science and Engineering (GSS), which collects these data, was refined in 2007 to improve reporting.

### Data Set Format

- Data Set Name
- ♦ ["/']Data Set Name[,]["/']
- ♦ Data Set Name (DSN)
- ♦ Data Set (Alt-DSN) Name

- The Baccalaureate and Beyond Longitudinal Study (B&B) will provide information concerning education and work experiences after the bachelor's degree.
- A notable example of a project incorporating these techniques is the National Oceanic and Atmospheric Administration's (NOAA) Coastal Change Analysis Program (C-CAP).
- For understanding the SCS SST variability, the NOAA daily Optimum Interpolation Sea Surface Temperature (AVHRR OISSTv2) is used in this study.

### Data Set Format

- ♦ Data Set Name
- ♦ ["/']Data Set Name[,]["/']
- ♦ Data Set Name (DSN)
- ♦ Data Set (Alt-DSN) Name
- ♦ DSN

- These are derived from a stochastic set of wind fields, calculated using a model forced by historical observations of TCs from the IBTrACS v02r01 dataset (Knapp et al., 2010).
- Figure 3 shows examples of surges from simulations used to create the Reference Set of storms along the Mississippi coast using the SLOSH model.

# Approach 1: String Matching

## Submitted Approach

Given a collection of known data sets and a document to search:

- 1. Convert the data sets to lower case
- 2. Convert the document to lower case
- 3. For each data set, use substring matching to see if it is in the document

#### Advantages:

♦ Fast

#### **Disadvantages:**

- Must know all data sets and permutations
- ♦ Changing the case can lead to unwanted behavior in acronyms that are homographs (e.g., SLOSH, HERD → slosh, herd)

## **Updated Approach**

- Given a collection of known data sets and a document to search:
  - ♦ Convert the data sets into a regular expression:
    - Each word is searched for both upper and lower-case versions
    - ♦ All acronyms are searched only in upper-case form

#### Advantages:

- ♦ Fast, but not as fast as simple substring search
- Does not confuse acronyms with common homographs

#### Disadvantages :

• Must know all data sets and permutations

Data Set Name √["/']Data Set Name[,]["/'] √Data Set Name (DSN) √Data Set (Alt-DSN) Name √DSN √Data Set Name √["/']Data Set Name[,]["/'] √Data Set Name (DSN) √Data Set (Alt-DSN) Name √DSN √

# Approach 2: Entity Classification

## Submitted Approach

#### Given a document:

- 1. Find all instances of the pattern *Data Set Name* (*DSN*) in the text, these are entities.
- 2. For each entity, use a SciBERT-based classifier to classify the entity as a data set entity or not.

#### Advantages:

 Model learns what to separate data set like entities from non-data set like entities

#### **Disadvantages:**

 Only captures data sets that follow the name and abbreviation pattern

## **Updated Approach**

- ♦ Given a document:
  - 1. Extract *entities* using a regular expression (see article).
  - 2. For each entity, use a SciBERT-based classifier to classify the entity as a data set entity or not.

#### Advantages:

 Model learns what to separate data set like entities from non-data set like entities

#### Disadvantages:

 Only captures data sets that follow the regular expression pattern

Data Set Name X["/']Data Set Name[,]["/'] XData Set Name (DSN) √Data Set (Alt-DSN) Name √DSN XData Set Name √["/']Data Set Name[,]["/'] XData Set Name (DSN) √Data Set (Alt-DSN) Name √DSN X

# **Approach 3: Token Classification**

## Submitted Approach

Train a masked language-based model:

- 1. Find all sentences containing a data set reference.
- 2. For each sentence, mask out the data set and train a model to classify if the masked entity is likely a data set reference
- Store masked classification embeddings for inference

#### Advantages:

♦ Model learns how data sets are described

#### **Disadvantages:**

 Masked embedding vectors are randomly sampled for inference. Sampling from that distribution can produce inconsistent results

## **Updated Approach**

- ♦ Train a Named Entity Recognition (NER) model
  - 1. Train a NER model to predict whether or not each word is likely to be a dataset

#### Advantages:

♦ Learns to distinguish data sets at the word-level

#### Disadvantages :

 Predictions at the word-level could have odd false positives (e.g., "in Science", "Study")

Data Set Name √["/']Data Set Name[,]["/'] √Data Set Name (DSN) √Data Set (Alt-DSN) Name √DSN √Data Set Name √["/']Data Set Name[,]["/'] √Data Set Name (DSN) √Data Set (Alt-DSN) Name √DSN √

# Where do we go from here?

- ♦ <u>Openness</u> is key
- ♦ Code √:
  - o GitHub repository: <u>https://purl.archive.org/democratizing-data/code</u>
  - Develop new methods and compare them with methods in our work
- ♦ Data √:
  - Kaggle data: <u>https://www.kaggle.com/c/coleridgeinitiative-show-us-the-data/</u>
  - Dataset Mentions Detections Dataset: <u>https://doi.org/10.1162/tacl\_a\_00592</u>
- ♦ Benchmarking:
  - The initial Kaggle competition worked well for this.
  - Need a community-based approach:
    - Adding more documents to the data set
    - Improving current data set quality