

**Team - AngryNerds** 

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### Introduction

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- Contracts are usually dispersed over different websites - tons of contracts and sessions that are hard to navigate.
- Many citizens are left in dark about how to approach police reforms

#### **Motivation**

- Socially Meaningful
- Interesting
- Powerful
- decrypted

### What we are trying to achieve?

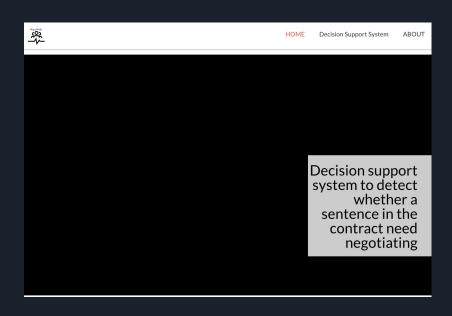


Fig 1. Website For Decision Support System

- Analyze the contracts from police departments.
- Discover problematic sentences and clauses.
- Categorize the problematic sentences.
- Build a website gives users a way to search for a sentence or select the sentence they feel confused about in a user-friendly way and displays the category of the sentence.

### Libraries and Data we used

Main Libraries: stringdist, RTextTools (A supervised learning package for text classification)<sup>[1]</sup>, R Shiny<sup>[2]</sup>

Data: Contracts data from "Police Union Contract Project" project<sup>[3]</sup>.

- 87 police contracts from the 100 largest U.S. cities.
- Human-annotated data<sup>[4]</sup> as the ground truth to train the model.

| Sentence                                    | Category                                    |
|---|---|
| Human Resources Department files are a      | Erases misconduct records                   |
| If the questioning is mechanically recorded | Gives officers unfair access to information |
|   |   |

# **Data Exploration**

Number of words increase :

Longest Contract had 83,670 terms.

Shortest Contract: 427 words

Mostly in range - 15,000-40,000 words

• Number of stop words also increase nearly.

Some contracts had more than 20,000 stopwords.

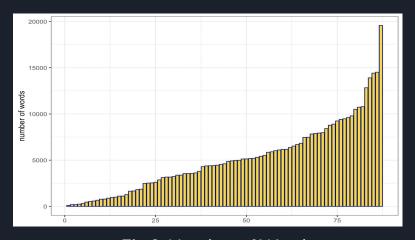


Fig 2. Number of Words

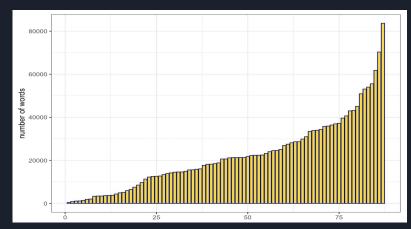


Fig 3. Number of Stop Words

# Data Preprocessing(1)

#### Step-1: Convert all the pdf file into txt file

- Using Adobe Acrobat to convert into docx
- Using online converter tool to convert docx into .txt files

#### Step-2: Read all the contracts into the R memory.

- Read a txt file one by one.
- Store sentences of length range(60, 500) in 'text' vector.

# Data Preprocessing(2)

#### Step-3: Load the human-annotated data file (Ground Truth File)

 Get the `language` and `category` columns to define labels for sentences in `text` vector.

#### **Step-4: Define the labels for problematic sentences**

- Define a `label` vector to hold labels/ categories for sentences in `text` vector.
- `stringsim` function of `stringdist` library is used to calculate similarity score between sentences in `text` vector and `language` vector.
- If similarity score is greater than 0.85, the corresponding category is assigned to sentence otherwise it is a problematic sentence.

## Data Preprocessing(3)

**Step-5: Text Preprocessing in one Step-**

Using `RTextTools` library: create\_matrix() function to create term-document matrix

- Remove stop words
- Remove Numbers
- Stem the words
- Remove the sparse terms

### Challenges in Data Preprocessing

- Labelling the Ground Truth (Human Annotated Data) with the text corpus
  - -> Used `stringsim` function of `stringdist` library to calculate similarity score and decide

# **Data Modeling**

### Supervised

We have the ground truth

### Classification

• We need to classify the sentences into different problem categories

#### So what we got?

**SVM** (Support Vector Machines)

SLDA (Supervised Latent Dirichlet Allocation)

Boosting, Bagging, Random Forest, Neural Network

**Decision Tree** 

## Workflow

Tidy the data

Build the models

Evaluate the performances

# Tidy the data

Problem 1: Highly imbalance

1:63

2:55

3: 143

4:303

5: 20952

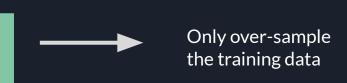
6:95

7:294

Random over-sampling<sup>[4]</sup>

# **Tidy the data**

Problem 2: Test data sampling



### **Build the models**

Problem: Too large dataset

Removing sparse terms: Allow 0.998 sparsity:

## **Performance Evaluation(1)**

| Model          | Precision | Recall    | R-score   |
|----------------|-----------|-----------|-----------|
| SVM            | 0.5400000 | 0.6585714 | 0.5857143 |
| SLDA           | 0.2171429 | 0.6185714 | 0.2628571 |
| Boosting       | 0.1571429 | 0.3557143 | 0.1400000 |
| Bagging        | 0.5414286 | 0.6257143 | 0.5742857 |
| Forest         | 0.6357143 | 0.6600000 | 0.6371429 |
| Tree           | 0.1442857 | 0.1714286 | 0.1457143 |
| Neural Network | 0.2800000 | 0.5057143 | 0.3185714 |

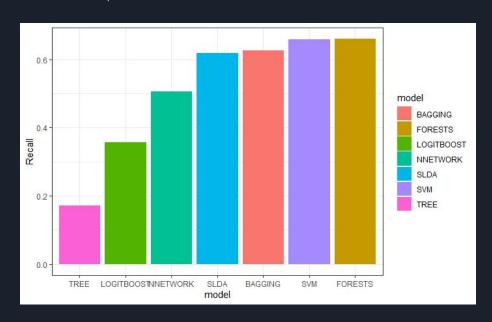
Table 1: Performances of all the models

## Performance Evaluation(2)

|      | Coverage | Recall |
|------|----------|--------|
| n>=1 | 1.00     | 0.81   |
| n>=2 | 1.00     | 0.82   |
| n>=3 | 1.00     | 0.82   |
| n>=4 | 0.99     | 0.83   |
| n>=5 | 0.96     | 0.87   |
| n>=6 | 0.85     | 0.98   |
| n>=7 | 0.51     | 0.99   |

Table 2: Ensemble agreement coverage and recall<sup>[5]</sup>

### Performance Evaluation(3)



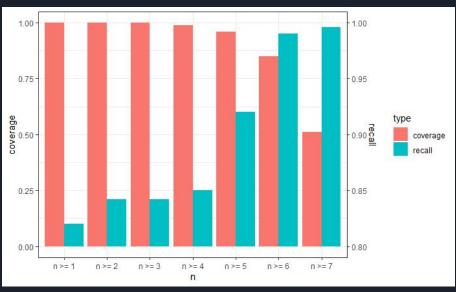


Fig 4: Recall of the models

Fig 5: Coverage and recall at several ensemble cut-points

# Decision Support System(1)

- We use text mining and model training to categorize our document sentences
- put data in the website, provide users two ways (select or search box) to use the system to predict problematic sentences.
- CLICK get started to get start!

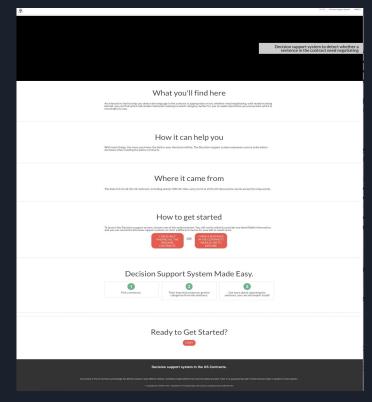


Fig 6: Home page of Decision Support system

# Decision Support System(2) - Steps

- Step 1: Input a sentence or select the sentence which you want
- Step 2: Given the sentence, system predicts the category of the sentence
- Step 3: Ask for help based on the category

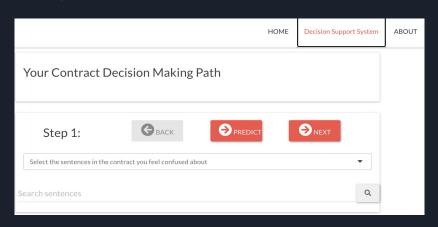


Fig 7: Decision Support System Page - Step 1

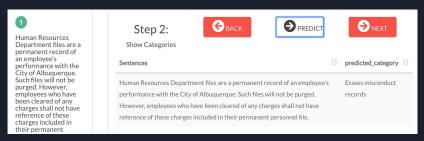


Fig 8: Decision Support System Page - Step 2

## **Our Creative Approach**

- ★ Deal with the real world problems USE Shiny app to create UI system, providing a user-friendly way to solve the problem
- ★ Text mining based on all the pdf contracts, create a way to deal with the pdf contracts
- ★ Model training can get high recall results, can predict any sentence regarding the police contracts and output good results

