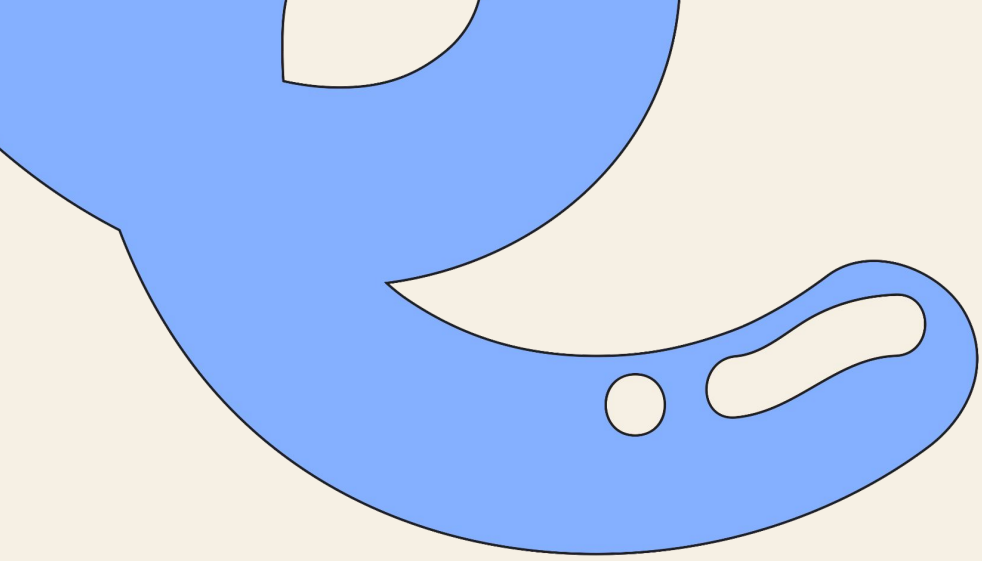




Improving Amtrak's FOIA Process

Team 25063



- 01 Introduction
- 02 Problem
- 03 Solution
- 04 Analysis of Results
- 05 Prototype Demo

Agenda

Introduction



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Use of AI

Generative AI was heavily used throughout this project

Platform	Google AI Studio. Gemini 2.0 Flash model
How it was used	Explain concepts concisely. How to generate a chart using Altair. Debug through errors. Determine where a keyword should go based on a department's name
Learning points	Filtering dataframes. Applying functions to each row of a dataframe. What stop words are in the context of keyword extraction. Libraries to use



Problem

The way Amtrak processes FOIA requests is inefficient and contains a single point of failure. This can introduce bottlenecks, and create confusing processes.



Proposed Solution

**Use of cosine similarity to compare
texts, and good ol' algorithms!**

Data Cleanup

There are FOIAs that contain the following keywords:

- “Duplicate”
- “Not a proper FOIA request”
- “unclear”

These FOIAs will only pollute our keyword analysis and slow down the mapping process

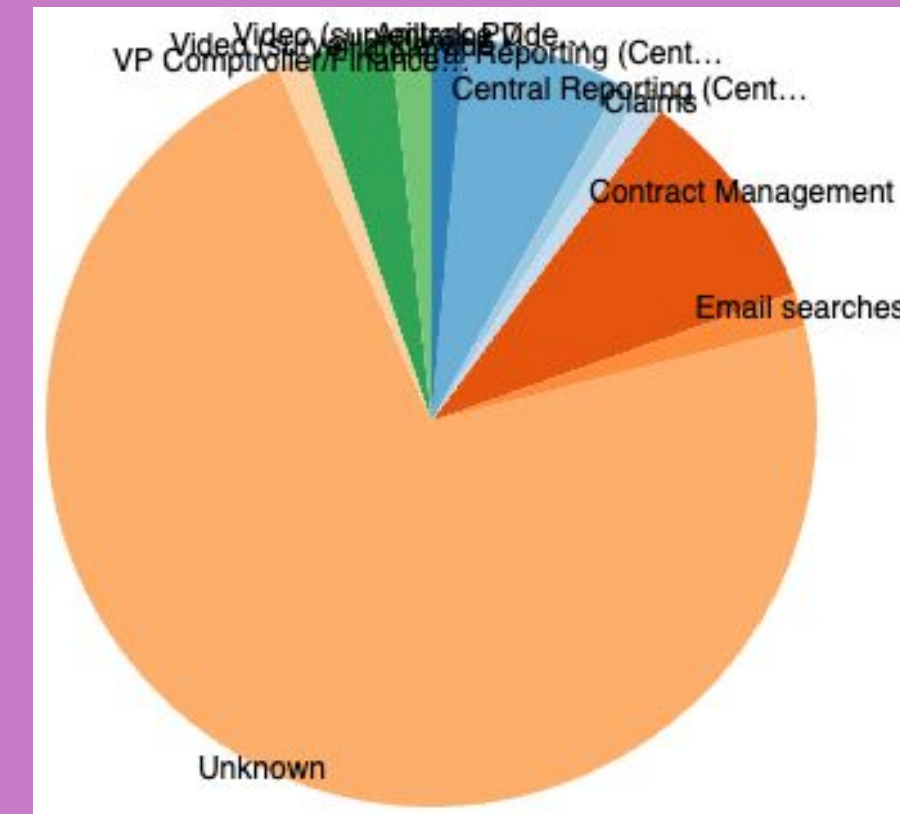
Classifying FOIAs

- Our initial solution was to create an algorithm that will look at each FOIA in the database, and categorize them based on *exact* keywords that are matched to a department.
- Can slow down with a larger data set
 - ... takes 2 seconds with ~4,000 requests!

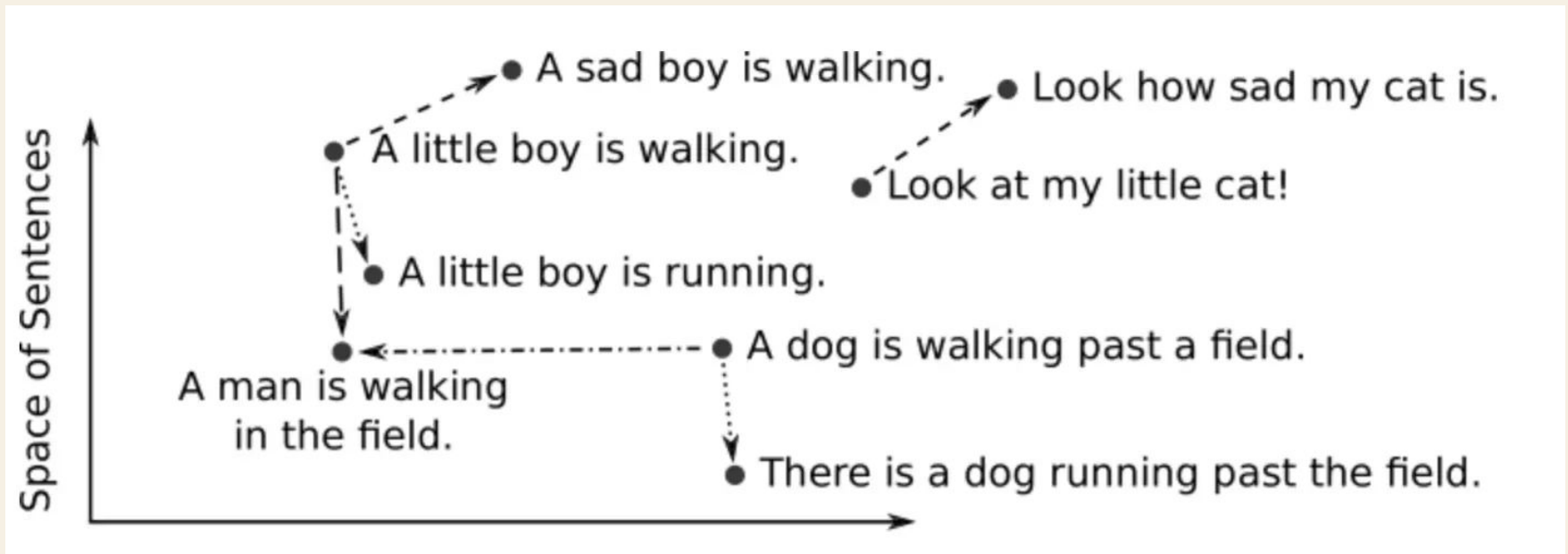
```
a ~/D/C/I/2/C/IC25-2025 main λ time python3 src/classification.py
```

```
-----  
Executed in 2.35 secs fish external  
usr time 2.25 secs 0.08 millis 2.25 secs  
sys time 0.07 secs 1.26 millis 0.06 secs
```

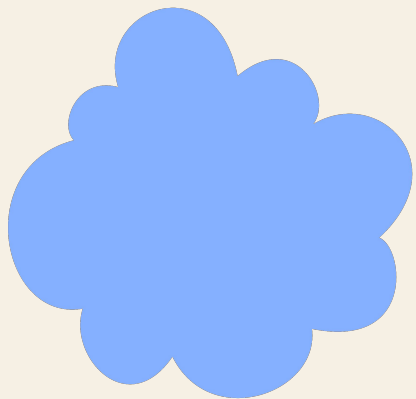
Timing stats, takes ~2s to run.



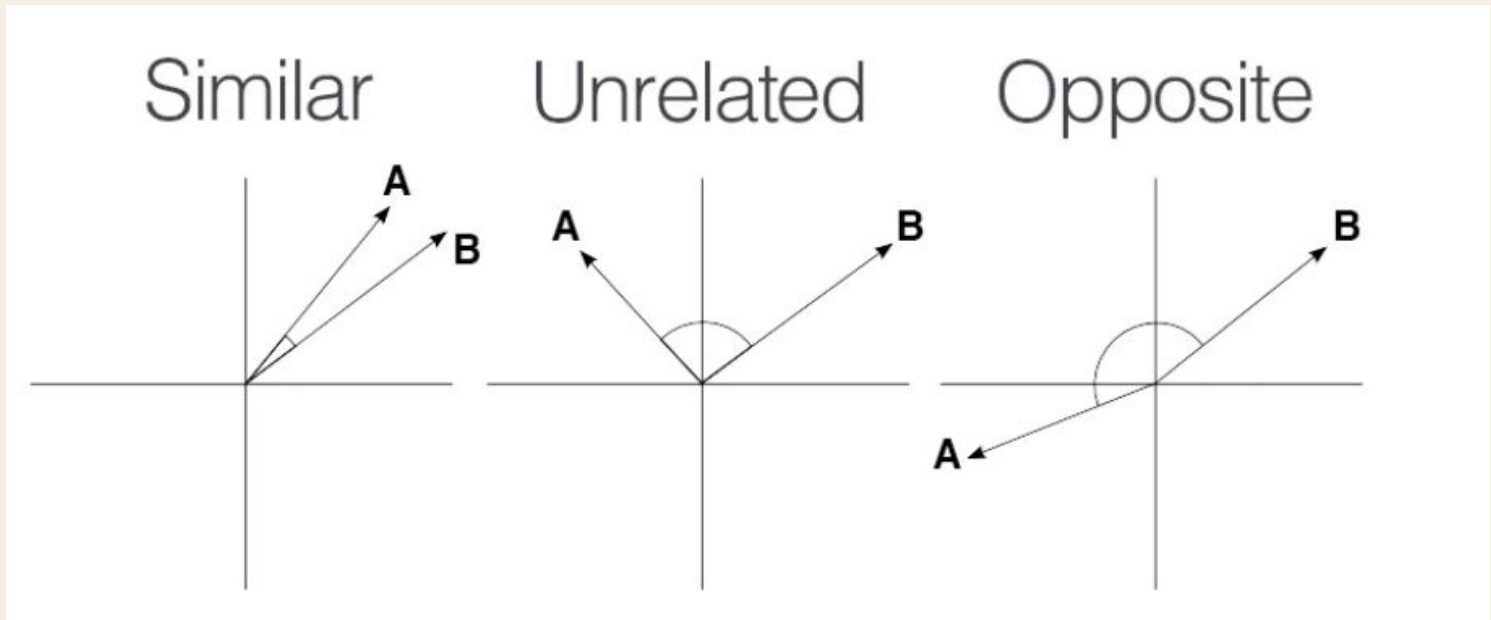
Final results. Too many unknowns :(



Encode department name plus keywords into a vector



Encode FOIA description into a vector

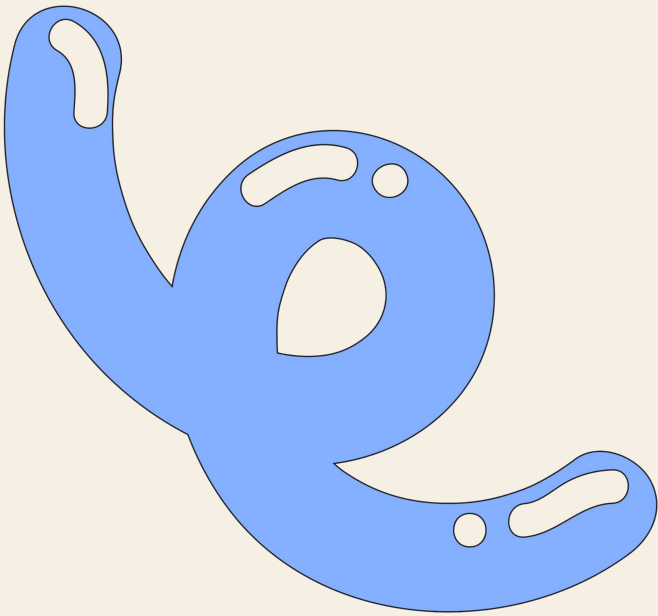


Map FOIAs by the department whose similarity score is above .50, otherwise, set as unknown

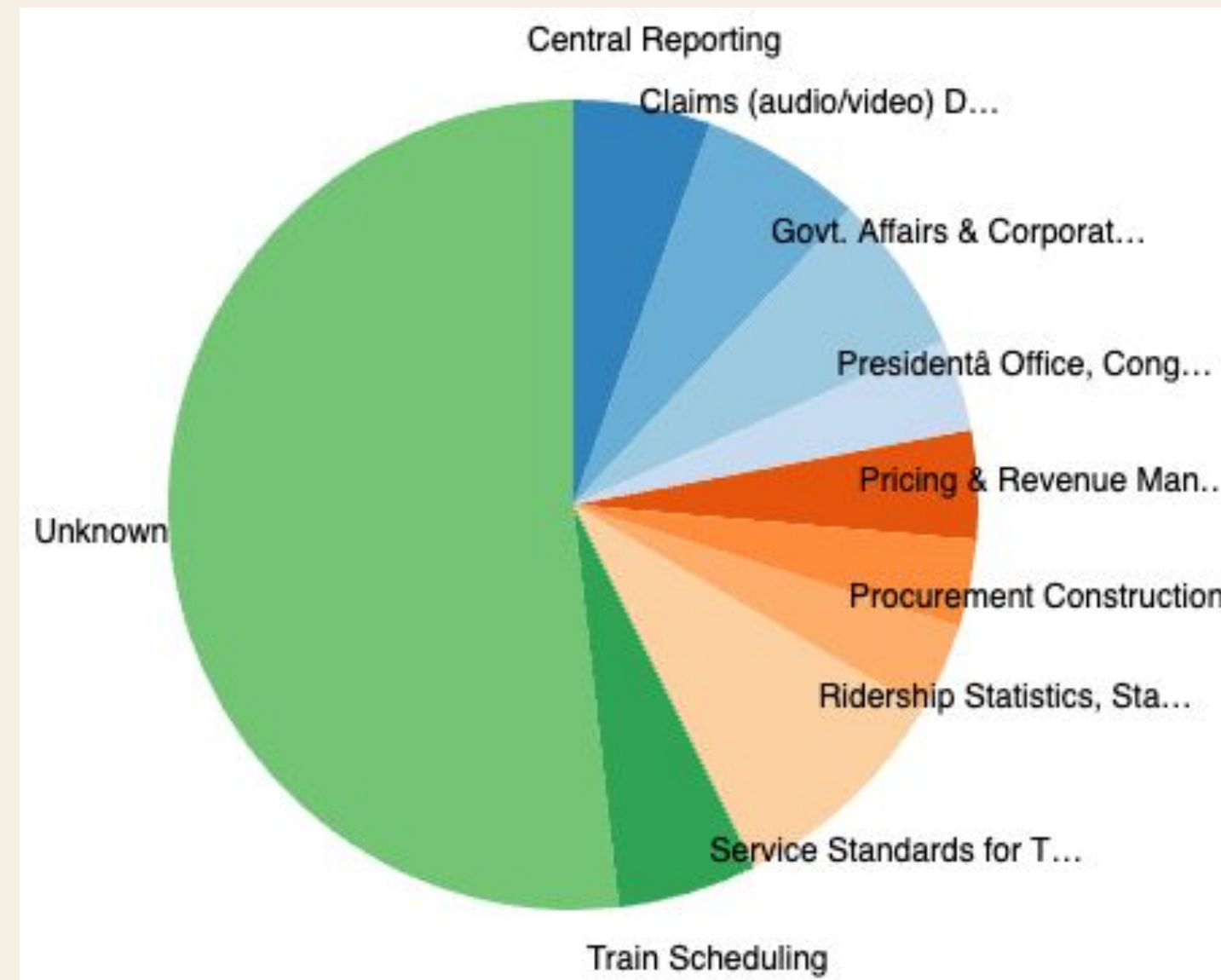
Trick to Remember Trigonometric Table



	0°	30°	45°	60°	90°
SIN	0	$\frac{1}{2}$	$\frac{1}{\sqrt{2}}$	$\frac{\sqrt{3}}{2}$	1
COS	1	$\frac{\sqrt{3}}{2}$	$\frac{1}{\sqrt{2}}$	$\frac{1}{2}$	0
TAN	0	$\frac{1}{\sqrt{3}}$	1	$\sqrt{3}$	Not Defined



- Only 85 out of 4,119 FOIAs have a similarity score above 60%!
- Lots of unknowns because of missing keyword matches. Not good! :(

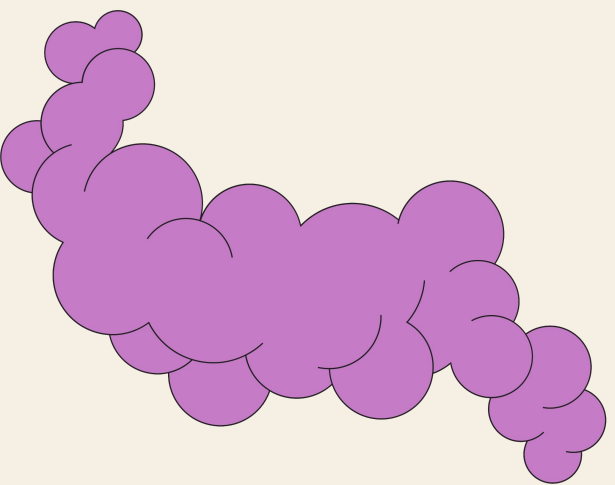


Little over ~1,770 unknown FOIAs w/
cosine similarity!

Analysis of Results

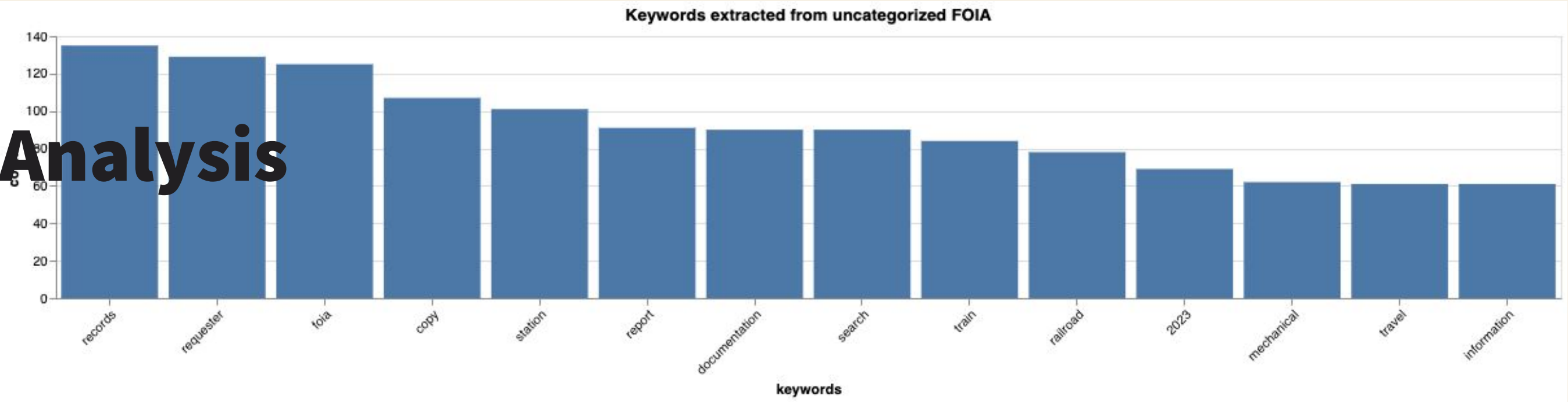
How can we improve this?

- Departmental data provided by Amtrak could include key contextual information to help us create matching keywords.
 - Performed our own research on common keywords such as “RFP (Request for Proposals)”, and “OIG (Office of the Inspector General)”
- Filter out irrelevant keywords from FOIA requests, such as common English words, or keywords that have already been matched.
 - Machine learning opportunity!

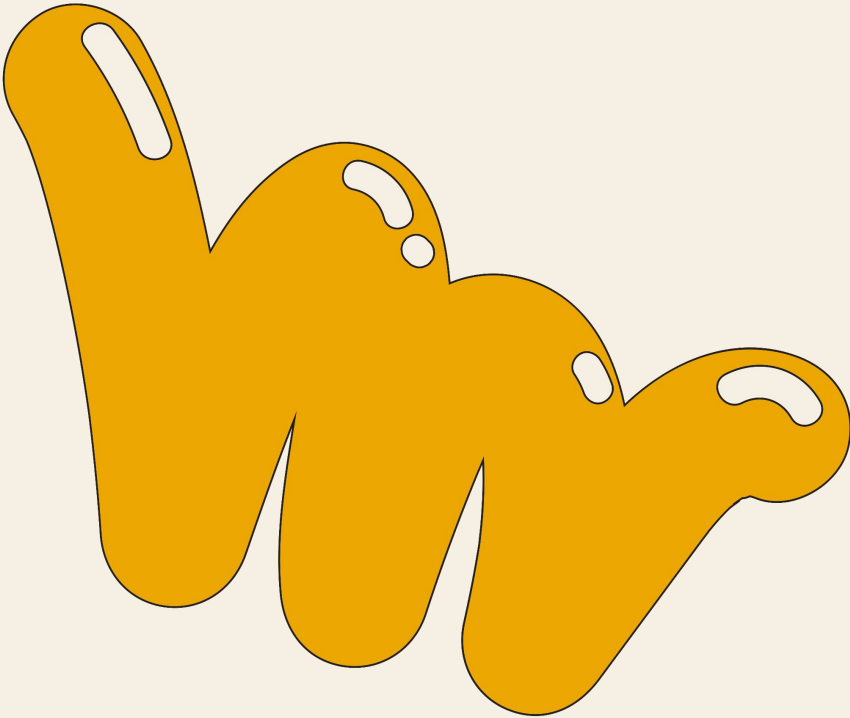
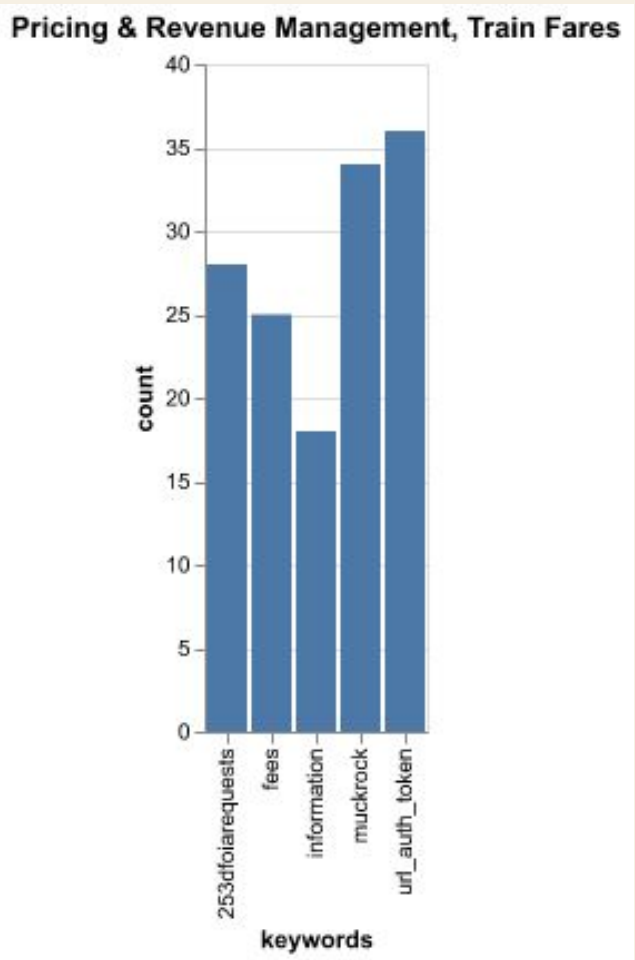
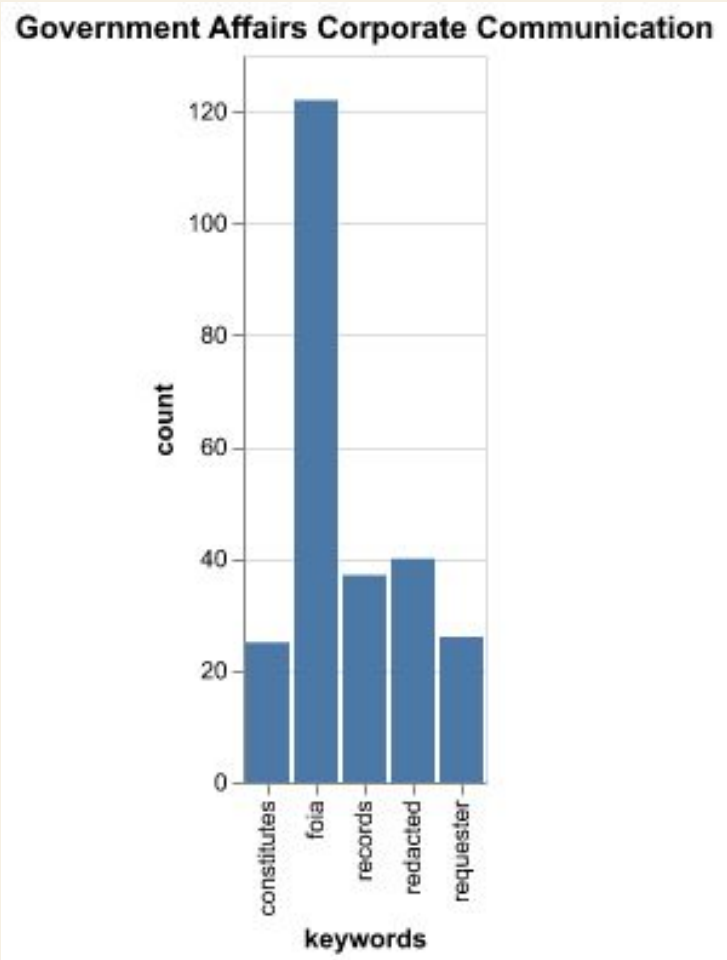
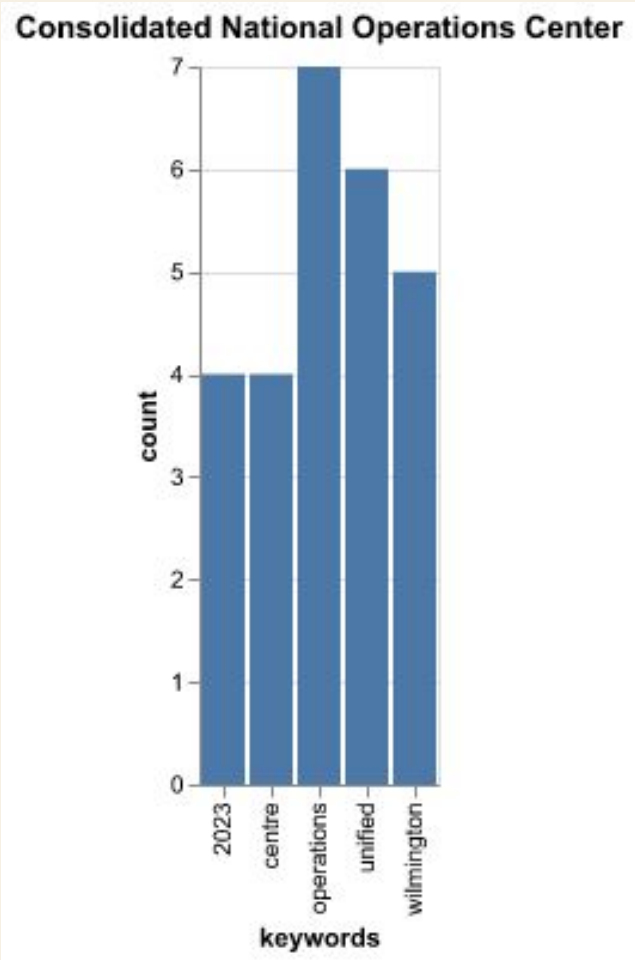
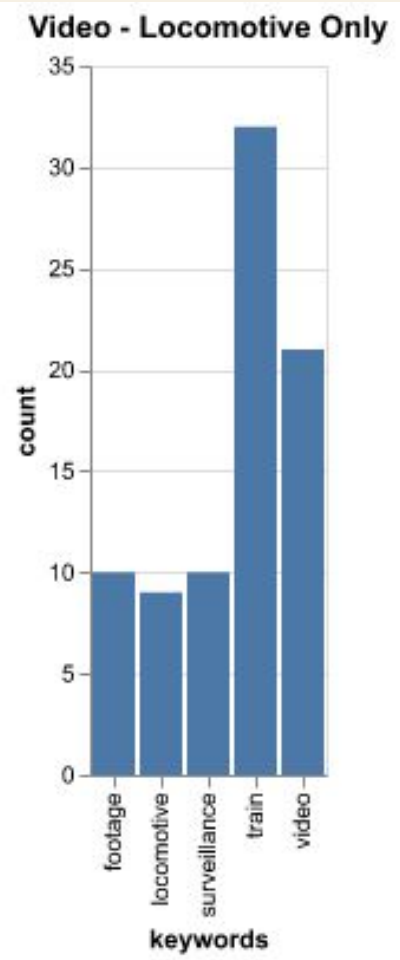


Keyword

Analysis



Extracted keywords to understand why cosine similarity scores are low...



Conclusion

- Although the results of cosine similarity look poor, the accuracy will only improve as more relevant keywords are added to a department.
- We believe this solution is scalable as we'll show in our prototype demo
- Introducing a more sophisticated traditional algorithm could also be beneficial in the long term.
 - ... or a combination of both solutions!





Prototype Demo



THANK YOU!

Any

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