Classy Data Analysis

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Challenge

We have a 1-D understanding of NPOs...

- More often than not, the services of an NPO span multiple sectors, e.g Health, Education, etc.
- Financially speaking, small local charities operate very differently from multi-million dollar organizations.

We need a more complex solution to find like-minded organizations.









Solution

... but NPOs are multi-faceted.

- Lay out a common "social space", where the organizations that drive social change and potential donors can connect, find organizations and be offered recommendations, and where discovery of new causes - and events within causes (i.e. fundraisers) - could be facilitated.
- Use combination of government IRS form 990 (returns for nonprofits) data along with external textual information (i.e. social media) to create a robust semantic space.





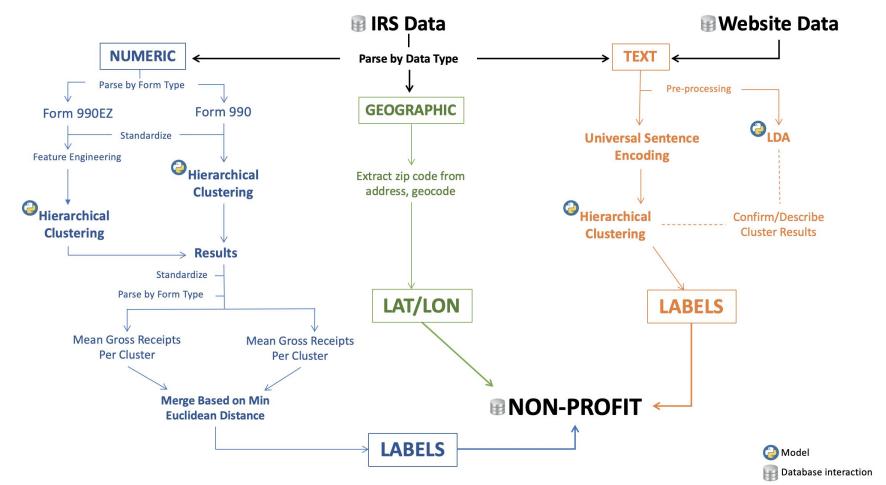




Agenda

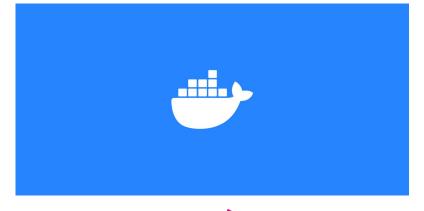
- 1. Introduction
- 2. Solution
- 3. Pipeline
- 4. Communication Plan
- 5. Visualization
- 6. Next Steps

Pipeline



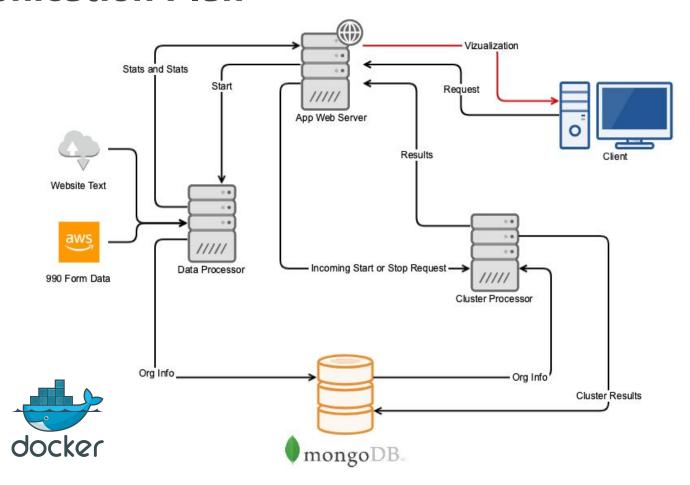
Docker Everything

- Created a complete docker environment for each module
- Allowed our code to be platform independent and easily scalable.





Communication Plan



MOTIVATION:

A **single**, easy-to-understand and **interactive** visual that answers the question:

"Which organizations are most similar to an organization that I care about?"

While tying together the three main descriptive categories of nonprofit organizations (text, finance, geography)

A Story of Usage

A very passionate supporter of the World Wide Fund for Nature (WWF) wants to reach out to other like-minded nonprofits to open discussions about creating a powerful "alliance" for nature conservancy. She desires to reach out to similarly well-funded organizations like the WWF and plans on personally meeting with the heads of such organizations.

- Text Data 200 features after embedding + PCA
- Financial Data 34 features
- Geographical Data 2 features (latitude / longitude)

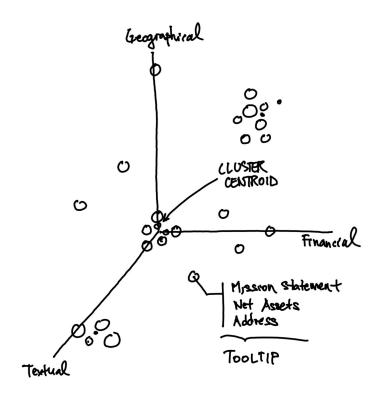
To represent these 3 categorical dimensions in a human-interpretable space, we collapse these 3 data sources into a single feature, each:

Distance to centroid!

How it Works:

- 1. Select an organization (Org. A)
- 2. Identify Org. A's cluster in "text space" and find all samples belonging to that cluster (Set A)
- 3. Identify Org. A's cluster in "financial space", find all samples belonging to that cluster (Set B), and create the union of Set A and Set B (Set C)
- 4. For all samples in Set C:
 - a. Calculate their distance in "text space" to the cluster centroid of Org. A
 - b. Calculate their distance in "financial space" to the cluster centroid of Org. A
 - c. Calculate their geographical distance from the location of Org. A
- 5. Plot distances on a 3-axis visualization (each axis for the 3 metrics in Step 4)

- Organizations close to the origin are similar to Org. A
- Organizations close to each other that are far away from the origin are close to each other and represent pockets of similar organizations that are loosely related to Org. A



Visualization Demo

Next Steps

- Finish visualization dashboard web application
 - Integrate tooltip interactivity
 - Build feature for selective filtering along descriptive dimensions
- Build out docker container that will include the entire data pipeline and code base

Questions?