# **Automated Fraud Detection**

Using the Enron Email Corpus to Train Fraud Detection Models



## The Question

### **Email Corpus**

Remaining emails between 1997-2001 made public by Federal Energy Regulatory Commission during its investigation.

- 150 users, mostly senior managers
- Over 500,000 emails

### **Enron Scandal**

At the time represented the largest bankruptcy in American history

- Bankruptcy declared in 2001
- Lead to de facto fall of Arthur Anderson
- Sarbanes-Oxley Act

### The Question

Given the far reaching implications of the Scandal, could we learn from the mistakes of the past?

 Can we use the enron emails to train a model to find those committing fraud?

## **Technical Details**

### Scrub and Explore

### **Initial Wrangling**

Exploring the dimensions of the dataset, reviewing formatting for the future model, visualizing complexity of the problem

### Unsupervised

### **K-Means Clustering**

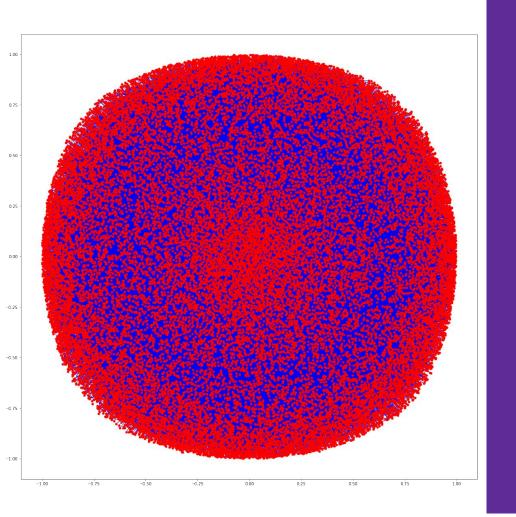
Using a 2 Cluster Model

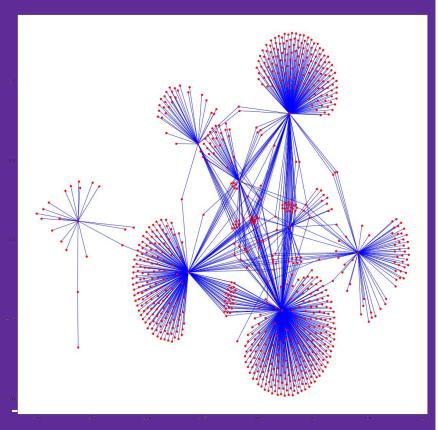
 Trained to find people and important language

### Supervised

### **K-Nearest Neighbors**

Over 90 % accuracy in appropriately classifying the emails in the clusters.





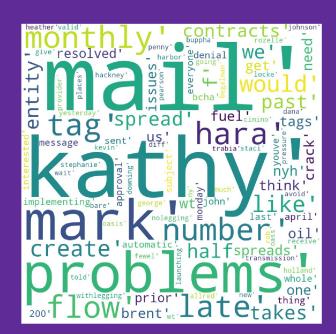
# Solution

Efficiency in Resource Management

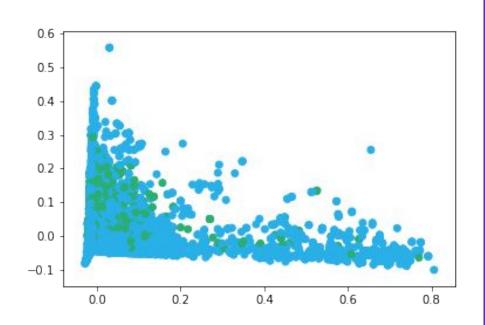
Created a model that could detect individuals who may require additional scrutiny based on their email traffic.

# Important Words

Clear implications

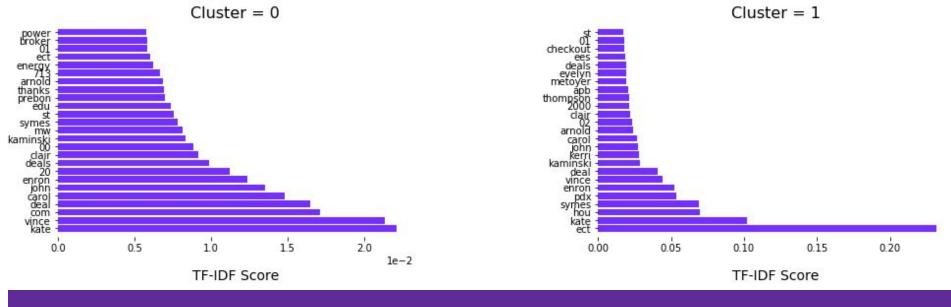


**Results and Recommendations** 

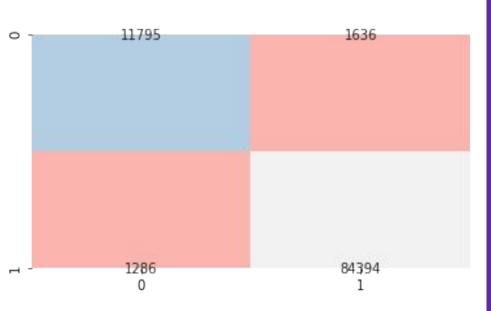


# K-Means Clustering

2 Cluster Model



# Top Words and Top People



# Accuracy

People were implicated in the scandal, and the words in their emails were indicative of ongoing fraud.

# **Insights and Recommendations**

- Efficient First Line of Defense
  - Utilizing Modern Machine Learning as the first indicator of red flags
- Reallocation of resources
  - Fraud Analysts and Risk Managers can reorganize their time to focus on the flags produced by the model capturing the same risks with fewer man hours
- Two-step Model
  - Easily calibrated with additional text bodies
  - Ultimately creates a computationally efficient paradigm



## **Future Model Extensions**

Data

**Model Update** 

**Additional Tuning** 

### **Big Data**

Additional resources allocated to run model on the full dataset

 Limitations for my personal machine and Kaggle Remote servers

### **Clustering Steps**

First Cluster on People

 Then Cluster and Classify on Language used by people

### **GridSearch for Params**

Try additional Clusters and additional Neighbors for further model tuning.

Utilizer GridSearch

## **Questions?**

