

The  
Erdős  
Institute

# Foursquare Location Matching

Team Hopf Bundle:  
Halley Fritze, Jay Hathaway, Max Vargas

# Motivation and Problem Statement

## Motivation:

- ❖ Businesses require reliable location information to run locals ads or expand to new cities.

## Problem:

- ❖ These location data sets contain a lot of noise, unstructured information, and incomplete or inaccurate attributes.

## Goal:

- ❖ Match data points describing the same POIs using machine learning.
- 

# Data Extraction

Foursquare is a location technology company which supplied our data via Kaggle.

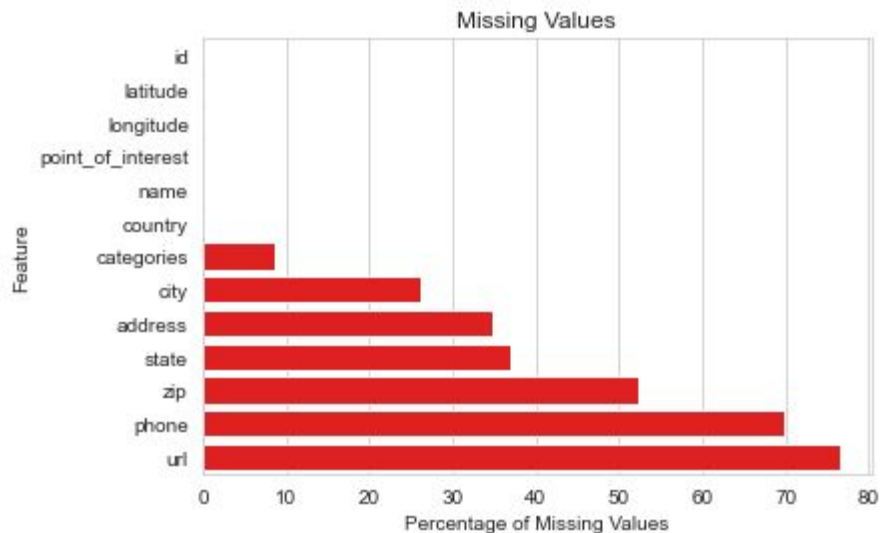
id	name	latitude	longitude	address	city	state	zip	country	url	phone	categories	point_of_interest
E_00001d92066153	Restaurante Casa Cofiño	43.338196	-4.326821	NaN	Caviedes	Cantabria	NaN	ES	NaN	NaN	Spanish Restaurants	P_809a884d4407fb
E_7e0d8e9138dd56	Casa Cofiño	43.338130	-4.326717	Barrio de los Caviedes s/n	Valdáliga / Cantabria	Spain	39593	ES	NaN	34942708046	Spanish Restaurants	P_809a884d4407fb

The above two data points represent the same POI.

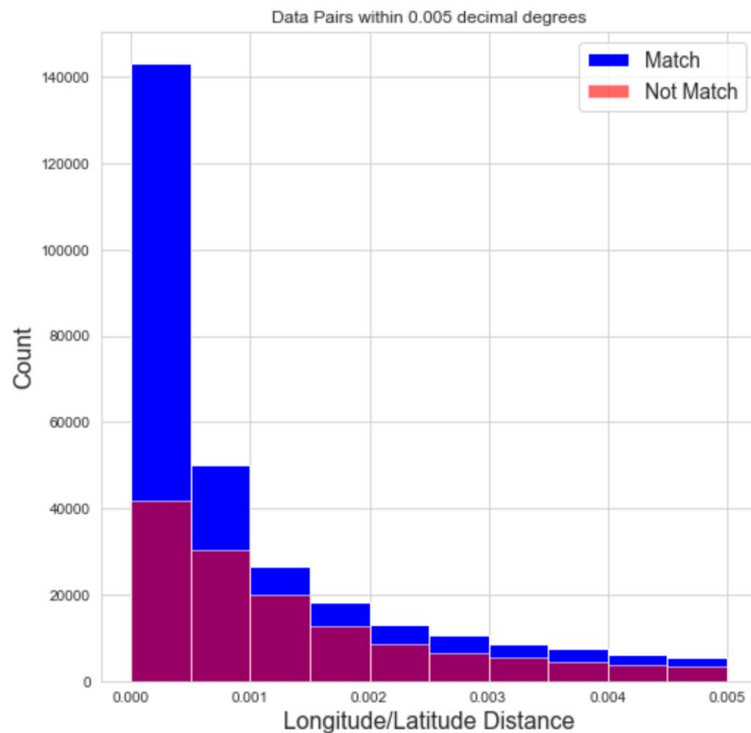
Additionally, they supplied a data set containing pairs of points with a boolean 'match' feature.



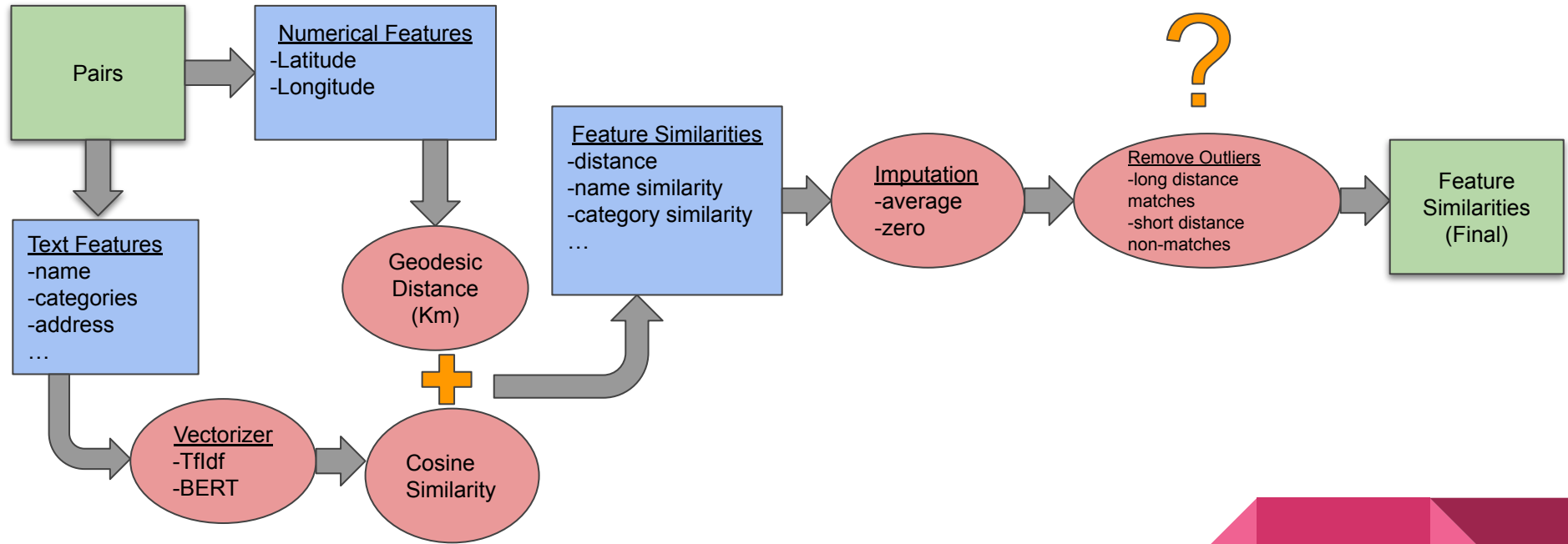
# Exploratory Data Analysis



- ❖ Many of our features are missing entries.
- ❖ Close distant pairs may not be the same POI.



# Feature Engineering



# Baseline Training

	Accuracy	Precision	Recall
Logistic Regression (Distance)	0.6889	1.0	0.6889
Logistic Regression (Category+Name)	0.7205	0.8994	0.7467
K Nearest Neighbors	0.7269	0.7612	0.8793
Feedforward NN	0.7259	0.9308	0.7390
Random Forest	0.7285	0.8644	0.7697

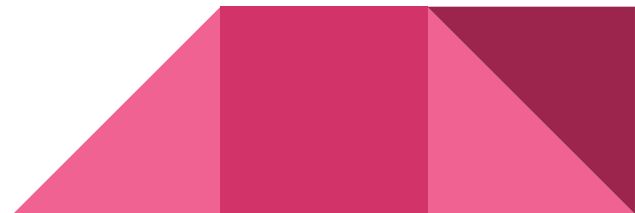
- ❖ Baseline models were trained on location, name, and category features.
- ❖ KNN, Neural Networks, and Random Forests achieved the best performances.
- ❖ Improvements depended on better data-cleaning techniques.

# Training with all features

	Accuracy	Precision	Recall
K Nearest Neighbors	0.7731	0.8010	0.8922
Feedforward NN	0.7771	0.9166	0.7924
XGBoost	0.7842	0.8998	0.8086

- ❖ Models were trained with all features in the dataset.
- ❖ XGBoost yields the highest accuracy. (learning\_rate=0.5, max\_depth=5, n\_estimators=200)
- ❖ Missing data was imputed using mean values (+2% accuracy for XGB)

- ❖ “1 Towne Centre Blvd #2800” and “1 Towne Centre Blvd” have a TF-IDF similarity of 0.8036.
- ❖ “400 Fairview Ave” and <NaN> have a similarity score of 0.5375 due to mean imputation.



# Further Improvements

- ❖ Consider using BERT as our vectorizer for feature engineering.
- ❖ Develop better cleaning techniques to improve accuracy
- ❖ Further hyperparameter optimization
- ❖ Analyze reasons behind false positives and false negatives
- ❖ Better zip-code processing

False Negative >

Frontier Baggage Claim	37.614955	-122.384861		NaN	San Francisco	CA	NaN	US	NaN	NaN	General Travel	P_7ec26b3743da93
Terminal 1 Baggage Claim	37.614645	-122.385662		Harvey Milk Terminal 1	San Francisco	CA	94128	US	NaN	NaN	Baggage Claims	P_7ec26b3743da93

Mann Center for the Performing Arts	39.983467	-75.221849	5201 Parkside Ave	Philadelphia	PA	19131	US	http://www.manncenter.org	2155467900		Performing Arts Venues, Music Venues, Outdoor ...	P_6f0f7249e54870
---	-----------	------------	-------------------------	--------------	----	-------	----	---------------------------	------------	--	--	------------------

Mann Center for the Performing Arts D Gate	39.983558	-75.223507	5201 Parkside Ave	Philadelphia	PA	19131	US		NaN	NaN	Performing Arts Venues	P_70d53bfea1ff14
---	-----------	------------	-------------------------	--------------	----	-------	----	--	-----	-----	------------------------------	------------------

< False Positive

Thank you to Akul and the Erdos Institute!