The Erdős Institute

# Foursquare Location Matching

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#### 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**

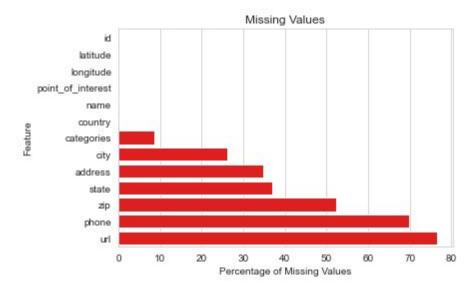
Foursquare is a location technology company which supplied our data via Kaggle.

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

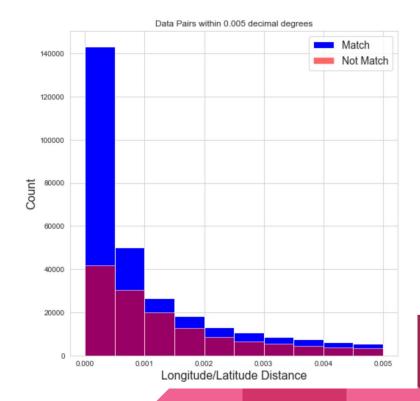
The above two data points represent the same POI.

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

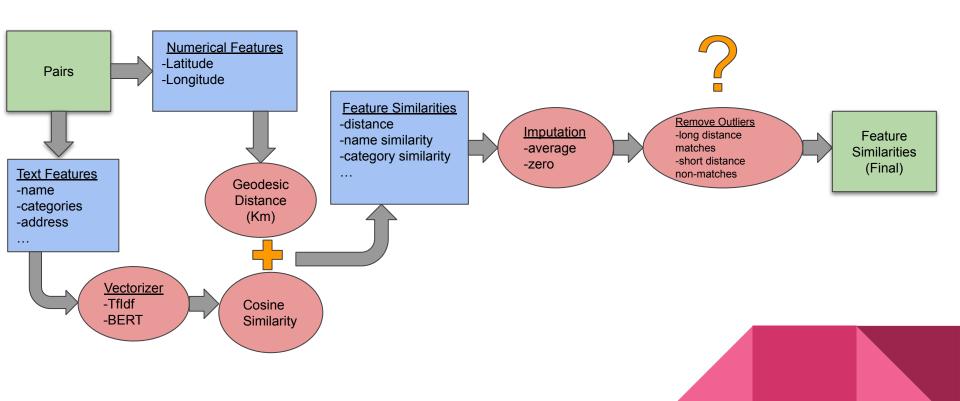
# **Exploratory Data Analysis**



- 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
Feed-Forward 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.
- Missing data was imputed using mean values (+2% accuracy for XGB)

- \* "1 Towne Centre Blvd #2800" and "1 Towne Centre Blvd" have a Tfldf 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
- Analyze reasons behind false positives and false negatives
- Better zip-code processing



Thank you to Akul and the Erdos Institute!