



Discovering Nulls and Outliers

DS-1004 Big Data

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MINIMIZERS

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https://github.com/ranamihir/big_data_project

INTRODUCTION

► Problem Statement:

1. Null Value Detection
2. Outlier Detection
 - Univariate outliers
 - Multivariate outliers

► Data Set Collection:

- NYC Open Data
- 50 data sets

PROBLEM FORMULATION

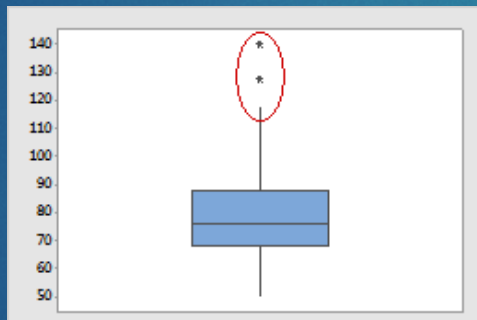
1. Data Cleaning

- "\$1.99" → 1.99, "1,000" → 1000, 10003 → "10003" (zip code)

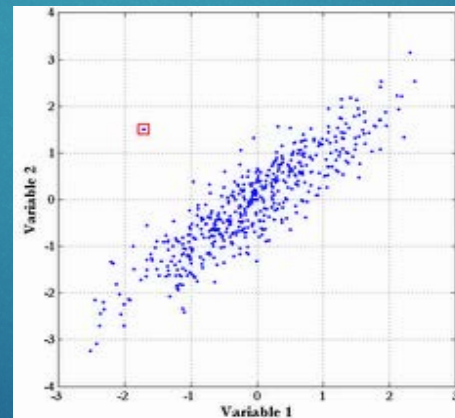
2. Missing Value Treatment

- "None", "N/A", " ", "-", "-999", "999", etc.

3. Outliers

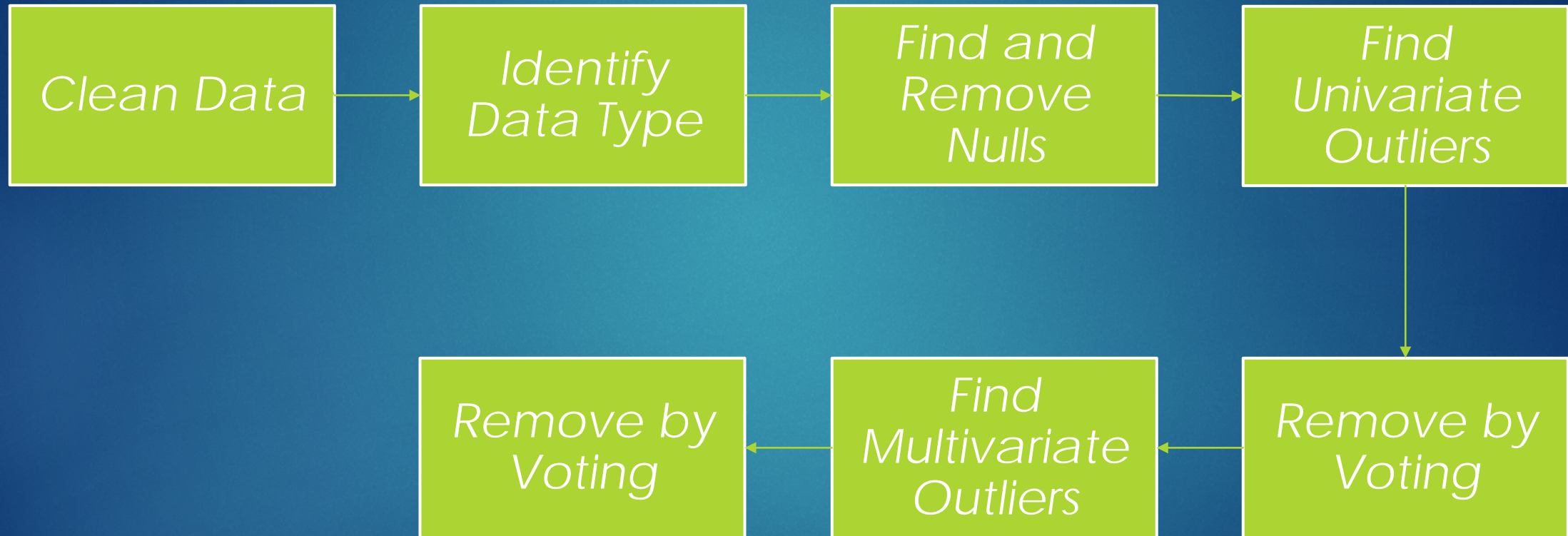


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METHODOLOGY



OUTLIER DETECTION

- ▶ *Nearest Neighbor based*
 - ▶ *DBSCAN*
- ▶ *Clustering based*
 - ▶ *k-Means*
- ▶ *Mixture of Parametric Distributions*
 - ▶ *Gaussian Mixture Models*

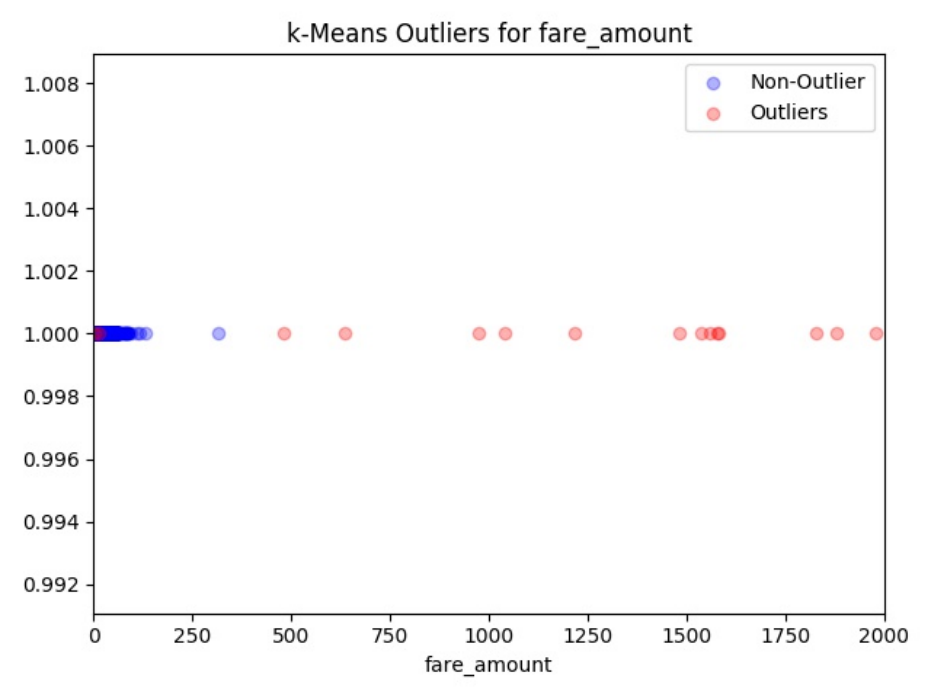
OUTLIER DETECTION

- ▶ *Non-Parametric*
 - ▶ *Histogram / frequency – based*
- ▶ *Statistical Anomaly based*
 - ▶ *Box plot Rule*
 - ▶ *Gaussian model based (z-score)*
 - ▶ *Other Probabilistic models (Beta, Gamma, etc.)*

KEY STRENGTHS

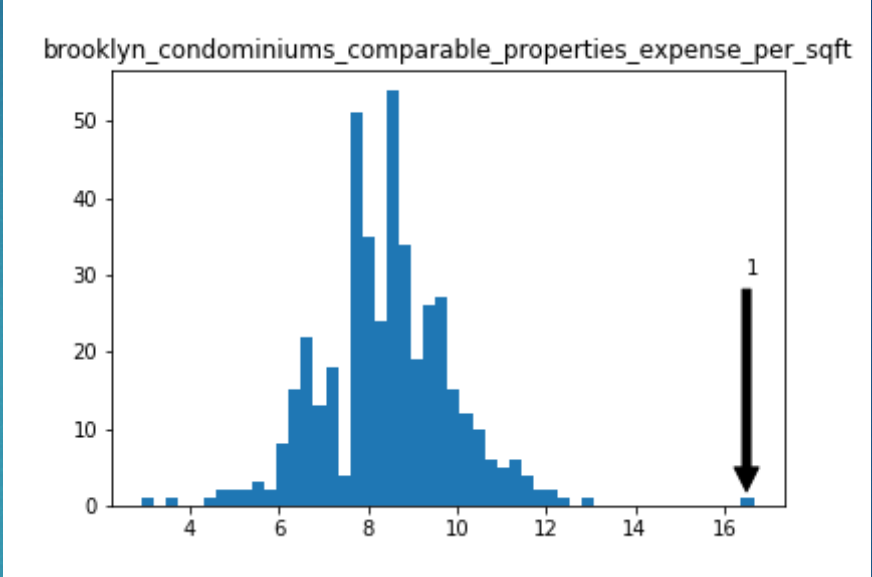
- ▶ *End-to-end automated framework*
- ▶ *Box plot Rule at core*
 - *No input specific to particular data set / column required*
- ▶ *Robust*
 - *Multiple techniques optimizing different metrics*
 - *Voting / Intersection of multiple similar techniques*
- ▶ *Efficient*
 - *Remove univariate outliers before finding multivariate ones*

RESULTS



Column	Value
brooklyn_condominiums_comparable_properties_address	"5 UNKNOWN"

Data: bss9-579f.tsv



Summary	num_level_3	rid	num_level_3
count	5302	847	1386.0
Mean	239.776	851	1434.0
stddev	267.144	963	1337.0
min	0.0	990	1344.0
max	2020.0	1493	1522.0

Data: usap-qc7e.tsv

REFERENCES

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4. Ming Hua and Jian Pei. 2007. Cleaning Disguised Missing Data: A Heuristic Approach. In *Proceedings of the 13th ACM SIGKDD International Conference on Knowledge Discovery and Data Mining (KDD '07)*. ACM, New York, NY, USA, 950–958. <https://doi.org/10.1145/1281192.1281294>
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