

Outlier Detection on Financial RDF Data

Mentor: Christiane Engels

Group: Zuhair Almhithawi, Nayef Roqaya, Berivan Ekmez

Final Presentation





Outline

- 1) Motivation
- 2) Project overview
- 3) Requirements
- 4) System architecture
- 5) Results
- 6) Demo





Motivation

- Open Government and Data Transparency initiatives => increasing number of datasets
- automatically analyze data sets.
- One aspect of analyzing data sets is finding unusual values, i.e. outliers

Project overview

Outlier Detection on Financial RDF Data

Objectives:

- ✓ Provide more insight into financial data by finding unusual values, i.e. outliers or anomalies:
 - > irregular behavior (corruption, fraud, . . .)
 - regions of special interest that e.g. require more subsidies
- ✓ Compare results of different outlier detection methods.

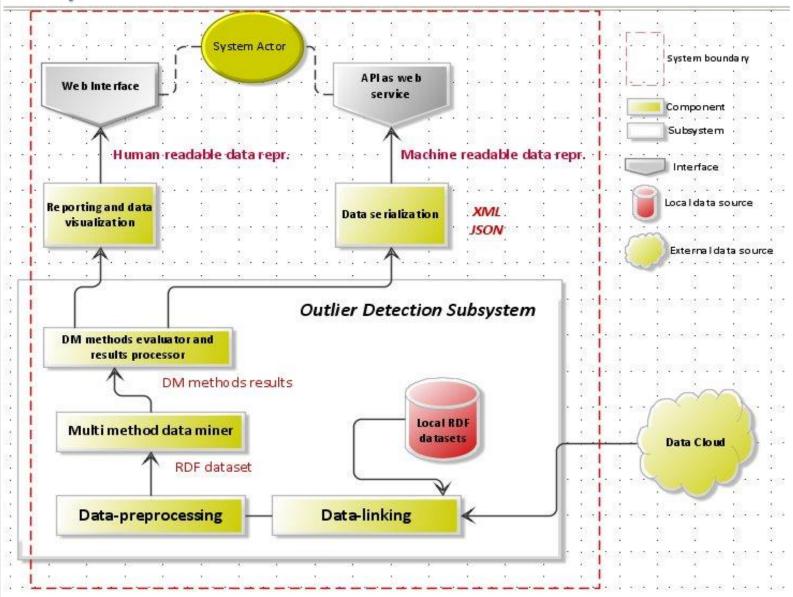


Requirements

- User can import RDF dataset(s):
 - Data Cube vocabulary
- Link local datasets to external dataset cloud:
 - DBPedia
- Apply and compare different outlier detection methods:
 - K-means
 - Chauvenet's Criterion
- Visualize the results:
 - Google Maps



System architecture



System architecture — data-linking module

Enrichment:

Why?

More accurate subpopulation outcomes.

How?

- DBPedia as an external data repository.
- SPARQL queries performed on Fuseki Jena server.

System architecture — data-linking module

• Before enrichment: Subject Observation **Predicate** Economic **Budget Phase** Amount Classification Object Draft 1154.5

System architecture — data-linking module

• After enrichment: Subject **Predicate** Observation **Budget Phase** Economic Amount Classification Object City Population Country GDP Draft 1154.5 Low High

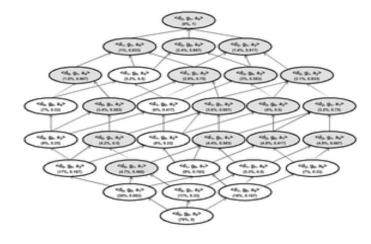
Master Computer Science

System architecture – data pre-processing module

• Sub population:

Why?

• Run outlier detection on original dataset !!?



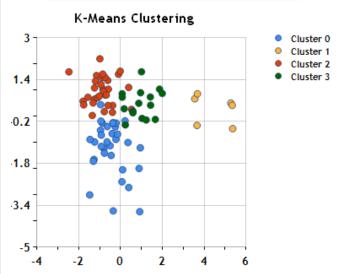
- Separation of original dataset into fragments gives more precise outlier results.
- Pruning when: No reducing or low KLV or number of instances is low



System architecture – outlier detection module

• Applied outlier detection methods:







Chauvenet's Criterion

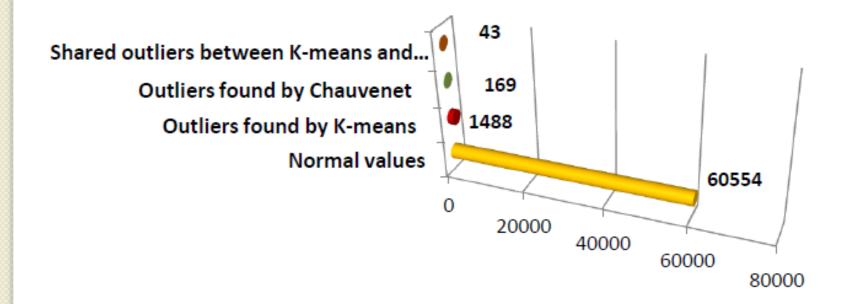




Results

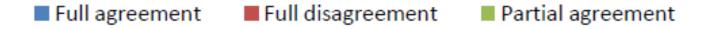
Percentage of outlier values

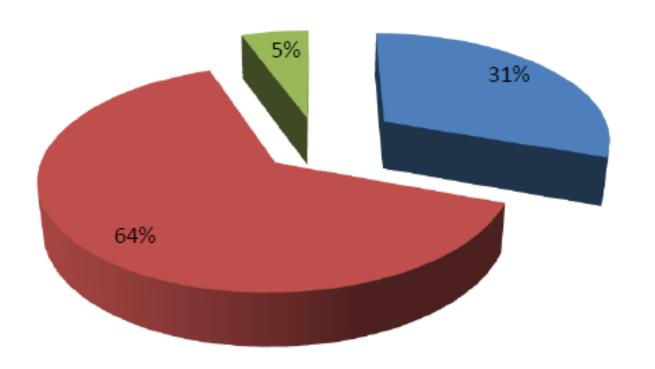
- Normal values
- Outliers found by K-means
- Outliers found by Chauvenet
- Shared outliers between K-means and Chauvenet



Results

Agreement degree





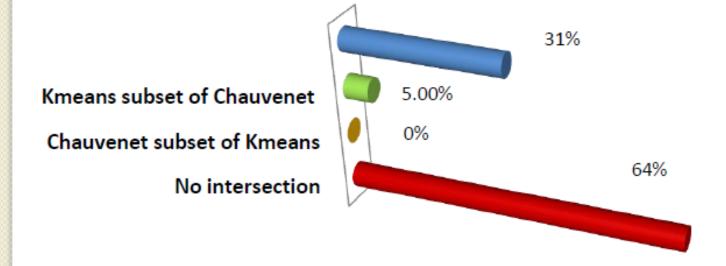
Results

Intersection of outlier results

■ No intersection

Chauvenet subset of Kmeans

- Kmeans subset of Chauvenet
 Kmeans subset of Chauvenet
 - Kmeans subset of Chauvenet & Chauvenet subset of Kmeans



Demo

