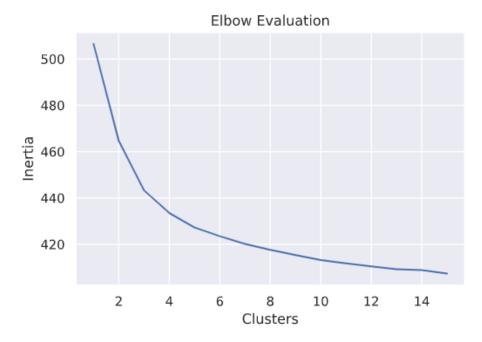
Clustering Patients Based on Similar Treatment Patterns

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Model

A KMeans model was built using 5 clusters. The number of clusters was chosen based on an elbow graph. Where the Y axis represents the mean sum squared distance of all samples to their closest cluster. The X axist represents the number of clusters.



This model clusters patients with similar treatment history. If more context is given to the patients within clusters then the model could be used to make recommendations and assumptions about patients.

Example: If patients in cluster x historically ended up being diagnosed with some sort of auto-immune disease (λ) then if a new patient is falls into cluster x they could be recommended to be tested preemptively for that disease λ .

The initial model that is included with this project had a sum of squared distances of all samples to their nearest cluter cnetroid of 427.3

Usage

Requirements

Ensure that you have all the required dependencies by running:

\$ python -m pip install -r requirements.txt

Training KMeans | training.py

The training.py script trains and saves a KMeans model. If **no** command line arguements are specified it will attempt to use the patients.csv and encounters.csv files found in the data/ directory.

Different datasets can be specified as such

python training.py <path to patients data> <path to encounters data>

Applying Model | apply_model.py

apply_model.py inputs datasets in the same format as those that were used to train the initial models and returns the cluster predictions for all the patients. The results are then written to a .csv file.

By default the provided datasets are used and the pretrained model is used. But specific models and datasets can be specified as such

python apply_model.py <patients.csv> <encounters.csv> <path to model>

Data

- SSN: Patient social security number.
- encounter id: Unique identifier representing a visit to a medical provider.
- line_number: The number of the medical proceduce/diagnosis from a resulting encounter.
- icd10_code: Unique code representing some sort of medical procedure/diagnoses

Notes:

- 1. Data is fully anonymized
 - 1. Social security numbers are hashed
 - 2. ICD10 codes are hashed
- 2. Unordered
 - 1. There is no indication of time.
 - 2. A patient with multiple encounters could have had them in any order.
 - 3. No assumptions about line number ordering is made.

Areas of Improvement

- More robust the CLI argument handling

 - Improve error handlingMeaningful error messages