

Code2Vec: Meaningful Embeddings of Medical Data

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DESCRIPTION

Grouping related disease incidents is a fundamental problem for healthcare and insurance providers. Our research creates a meaningful way to learn numerical representations (called *embeddings*) of diagnosis and procedure codes that (1) preserves medical meaning, including ostensibly subtle disease patterns and (2) provides a more practical framework for grouping disease incidents. This allows for unsupervised detection of disease comorbidities.

HOW EMBEDDINGS WORK

Vector space embeddings create represent non-numerical data (e.g. words, diagnoses, medications, etc.) that preserves real world meaning.

We use the natural language processing algorithm **GloVe** to find embeddings based on the context in which a particular diagnosis or procedure occurs in an individual's medical history [1].

GloVe leverages the *Distributional Hypothesis:* similar words (or codes) appear in similar sentence contexts (or disease history contexts).

Example (Words):

- "The weather forecast says it will rain today."
- "The weather forecast says it will snow today
- Based on the Distributional Hypothesis, "rain" and "snow" are treated as semantically-related words.

Calculating Code Embeddings:

Word/code embeddings are calculated by minimizing:

$$J = \sum_{i,j=1}^{V} f(X_{ij})(w_i^T w_j + b_i + b_j - log(X_{ij}))^2$$

where X_{ij} is the co-occurrence count of word i with word j, w_i and w_j are the embedded representations of words i and j, f is a weighting function, and b_i and b_j are bias terms.

CLUSTERING RESULTS

- We obtained 50-dimensional embeddings for 20,612
 ICD-9 diagnosis and CPT procedure codes.
- We grouped codes into 279 groups using k-means clustering with initial centroids chosen as the means of the corresponding Clinical Classification Software (CCS) disease groupings.

2D t-SNE Projections Projection of Code Embeddings

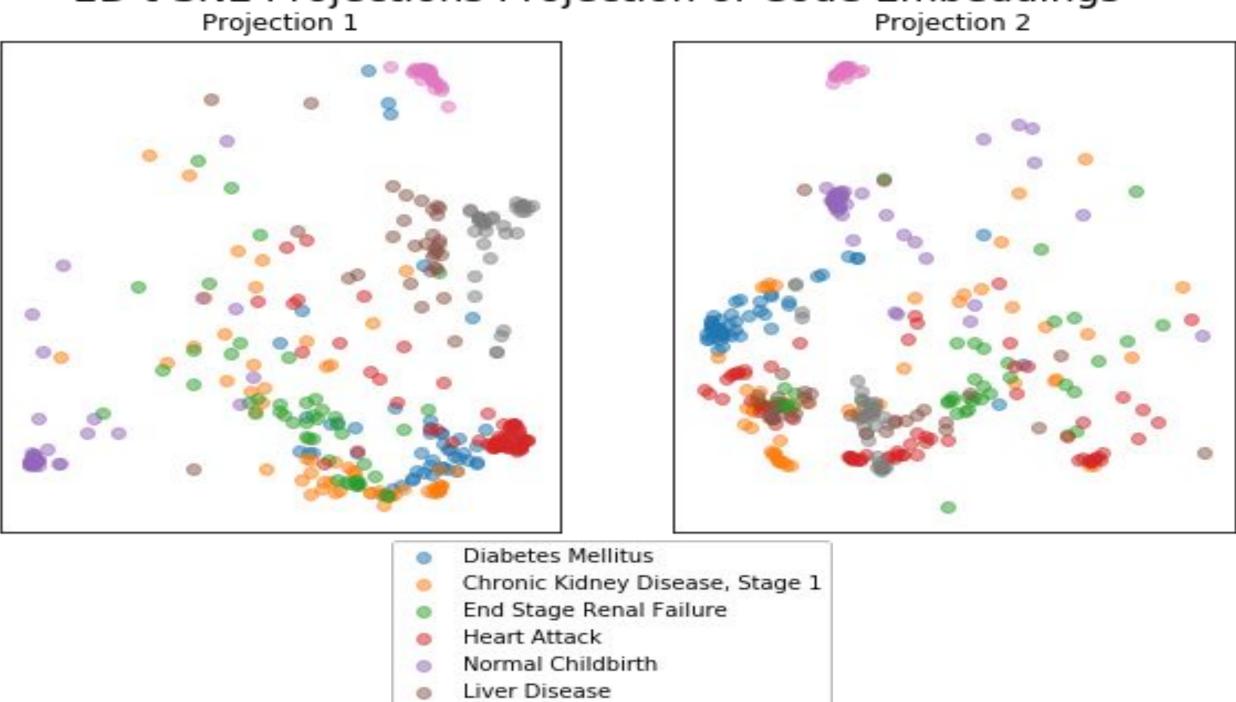


Figure 1: t-SNE Projections of Selected Code Clusters

Breast Cancer

The above plot shows two separate 2-dimensional projection of our code embeddings using t-SNE, which is designed to preserve both local clusters and global separation when reducing dimensionality. We show multiple random projections to better understand our highly dimensional embeddings. Note:

- Disease codes that cluster together occur in the same context
- Diabetes is clustered near heart attack, as well as certain cases of CKD and end-stage renal failure.
- The depression cluster (which contains a variety of mental illnesses) and the childbirth cluster are quite separated from the chronic diseases in both projections.
- Breast cancer and liver disease clusters are located near one another in both plots, potentially reflecting metastatic spread of the cancer to the liver [8].

CLUSTERING RESULTS

	End-Stage Renal Disease Related Code	Code Type	CCS Group
1	Hemodialysis procedure requiring repeated evaluation(s) with or without sub- stantial revision of dialysis prescription	Procedure	58
2	Malnutrition of moderate degree	Diagnosis	52
3	renal dialysis status	Diagnosis	158
4	Conditions due to anomaly of unspecified chromosome	Diagnosis	217
5	End-stage renal disease (ESRD) related services per full month; for patients twenty years of age and older	Procedure	58
6	Osteomalacia, unspecified	Diagnosis	52
7	Renal osteodystrophy	Diagnosis	161
8	End-stage renal disease (ESRD) related services monthly, for patients 20 years of age and older; with 4 or more face-to-face physician visits per month	Procedure	227
9	Secondary hyperparathyroidism (of renal origin)	Diagnosis	161
10	Encounter for extracorporeal dialysis	Diagnosis	158
11	Other and unspecified intracranial hemorrhage following injury without men- tion of open intracranial wound, with no loss of consciousness	Diagnosis	233
12	Serum screening for cytotoxic percent reactive antibody (PRA); standard method	Procedure	235

Table 1: Subsample of Stage V Renal Disease (ESRD) Cluster

Because of the distributional hypothesis, we expect codes in the same cluster to appear in the same context. Note the following about the above:

- Malnutrition is likely associated protein-energy malnutrition, a side effect of dialysis [3].
- Single gene chromosome anomalies have been shown to be a cause of polycystic kidney disease [4].
- Osteomalacia is caused by inability to absorb vitamin D, a symptom of kidney disease [5].
- Intracranial hemorrhage has been associated with polycystic kidney disease [6]. Note that the type our model identified is specifically without an intracranial wound.
- Cytotoxic panel-reactive antibody screenings are performed to assess an individual's suitability for an organ transplant. Kidneys are by far the most frequently transplanted organ [7].

SOURCES

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