

Extracting semantic relations from unlabeled text

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Motivation

A 14yo bxr owned by a reputable breeder is being treated for IBD with pred.

Motivation

[A 14yo bxr]*ANIMAL* owned by [a reputable breeder]*HUMAN* is being treated for [IBD]*DISEASE* with [pred]*DRUG* . [4]

Why the problem is hard ?

- Huge amount of data available on the web
- No manual tags or labels available
- Don't know exactly how many types of entities are present
- Presence of many irrelevant relations as noise

Problem Definition

Given a corpus of data of extracted relational tuples of the form $r(a, b)$, clusters the data using their relationship and also determine the best match for a given relation.

Related Work

- TextRunner: identifies relational tuples in one pass of the web [3]
- Semantic Network Extractor: Jointly cluster relation and object string [5]
- Infinite Relational Model [1]

Algorithm Specification

- $P(z^1, \dots, z^n \mid R^1, R^2, \dots, R^m)$

- Generative Model

$$P(R^1, R^2, \dots, R^m, z^1, \dots, z^n) = \prod_{i=1}^m P(R^i \mid z^1, \dots, z^n) \prod_{j=1}^n P(z^j)$$

- $P(z^j)$ is calculated using Chinese Restaurant Process
- $R(i, j) \mid z, \eta(a, b)$ is calculated using Bernoulli Distribution
- Chinese Restaurant Process (CRP) also determines the number of clusters

Output Matrices

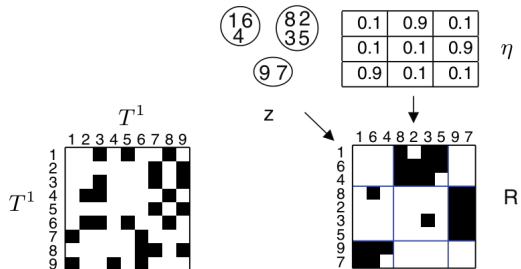


Figure: Output Matrices

References

- 1 Kemp Charles, Tenenbaum Joshua B, Griffiths Thomas L, Yamada Takeshi, and Ueda Naonori. Learning systems of concepts with an infinite relational model. 21(1):381, 2006.
- 2 Turney Peter D, Pantel Patrick, et al. From frequency to meaning: Vector space models of semantics. Journal of Artificial Intelligence Research, 37(1):141–188, 2010.
- 3 Banko Michele. Open information extraction for the web. PhD thesis, University of Washington, 2009.
- 4 Huang Ruihong and Riloff Ellen. Inducing domain specific semantic class taggers from (almost) nothing. Proceedings of the Association for Computational Linguistics (ACL), 2010.
- 5 Kok Stanley and Domingos Pedro. Extracting semantic networks from text via relational clustering. Proceedings of ECML, 2008.

Source code and Dataset

- The Source code for IRM is publicly available at <http://www.psy.cmu.edu/~ckemp/code/irm.html>
- The dataset is available at http://knight.cis.temple.edu/~yates/data/resolver_data.tar.gz

Questions !!

Formulae Specifications

- Generative Model

$$P(R^1, R^2, \dots, R^m, z^1, \dots, z^n) = \prod_{i=1}^m P(R^i \mid z^1, \dots, z^n) \prod_{j=1}^n P(z^j)$$

- Generating Clusters (CRP)

$$P(z_i = a \mid z_1, \dots, z_{i-1}) = \frac{n_a}{i-1+\gamma} \text{ if } n_a > 0$$

$$P(z_i = a \mid z_1, \dots, z_{i-1}) = \frac{\gamma}{i-1+\gamma} \text{ if } a \text{ is a new cluster}$$

- Generating Relations from clusters

$$z \mid \gamma \sim CRP(\gamma)$$

$$\eta(a, b) \mid \beta \sim Beta(\beta, \beta)$$

$$R(i, j) \mid z, \eta \sim Bernoulli(\eta(z_i, z_j))$$

- Inference

$$P(R \mid z) = \prod_{a, b \in N} \frac{Beta(m(a, b) + \beta), Beta(\bar{m}(a, b) + \beta)}{Beta(\beta, \beta)}$$

Semantic Network Extractor

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function  $SNE(S_r, S_x, S_y, R)$ 
  inputs:  $S_r$ , set of relation symbols
            $S_x$ , set of object symbols that appear as first arguments
            $S_y$ , set of object symbols that appear as second arguments
            $R$ , ground  $r(x, y)$  atoms formed from the symbols in  $S_r, S_x$ , and  $S_y$ 
  output: a semantic network,  $\{(\gamma_r, \gamma_x, \gamma_y) \in \Gamma_r \times \Gamma_x \times \Gamma_y : (\gamma_r, \gamma_x, \gamma_y) \text{ contains at least one true ground atom}\}$ 
  for each  $i \in \{r, x, y\}$ 
     $\Gamma_i \leftarrow \text{unitClusters}(S_i)$ 
   $\text{mergeOccurred} \leftarrow \text{true}$ 
  while  $\text{mergeOccurred}$ 
     $\text{mergeOccurred} \leftarrow \text{false}$ 
    for each  $i \in \{r, x, y\}$ 
       $\text{CandidateMerges} \leftarrow \emptyset$ 
      for each  $(\gamma, \gamma') \in \Gamma_i \times \Gamma_i$ 
         $\Delta P \leftarrow \text{change in } P(\{\Gamma_r, \Gamma_x, \Gamma_y\} | R) \text{ if } \gamma, \gamma' \text{ are merged}$ 
        if  $\Delta P > 0$ ,  $\text{CandidateMerges} \leftarrow \text{CandidateMerges} \cup \{(\gamma, \gamma')\}$ 
      sort  $\text{CandidateMerges}$  in descending order of  $\Delta P$ 
       $\text{MergedClusters} \leftarrow \emptyset$ 
      for each  $(\gamma, \gamma') \in \text{CandidateMerges}$ 
        if  $\gamma \notin \text{MergedClusters}$  and  $\gamma' \notin \text{MergedClusters}$ 
           $\Gamma_i \leftarrow (\Gamma_i \setminus \{\gamma, \gamma'\}) \cup \{\gamma \cup \gamma'\}$ 
           $\text{MergedClusters} \leftarrow \text{MergedClusters} \cup \{\gamma\} \cup \{\gamma'\}$ 
           $\text{mergedOccurred} \leftarrow \text{true}$ 
  return  $\{(\gamma_r, \gamma_x, \gamma_y) \in \Gamma_r \times \Gamma_x \times \Gamma_y : (\gamma_r, \gamma_x, \gamma_y) \text{ contains at least one true ground atom}\}$ 

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Thank You