

Presentation of :

Knowledge discovery by automated identification and ranking of implicit relationships

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What is knowledge discovery?

- ◆ Knowledge Discovery is
"the nontrivial process of identifying valid, novel, potentially useful, and ultimately understandable patterns in data"
(Fayyad, Piatetsky-Shapiro, and Smyth 1996).

Background: Literature Mining for knowledge discovery

- ◆ Information overload
- ◆ Millions of journal articles recording scientific findings
- ◆ No one can read them all: need automated approaches

Contribution of Swanson

- ◆ “undiscovered public knowledge”
- ◆ “non-interactive literatures”
- ◆ A-B-C model

A - - - - - B - - - - - C



Past work: Swanson and others who use co-occurrence

- ◆ Swanson
 - based on keywords in titles
- ◆ Others
 - MeSH terms
 - Mapping text to UMLS concepts

In both cases the size of the domain is a problem.

Approach in this paper

- ◆ Use the A – B – C model as a basis
 - Choose the A terms
 - Literature mining to find associations to A terms (B entities)
 - Query B objects to find relationships to other objects – C
 - Look at implicit (not explicit) A-C relationships
 - Rank relationships to find the statistically exceptional ones

The A Set

- ◆ Data Sources:
 - OMIM - diseases and clinical phenotypes
 - HGNC - genes
 - LocusLink - genes
 - MeSH - chemical compounds and drugs
- ◆ 33,539 unique objects (85,234 including synonyms)

Identify Relationships

- ◆ Co-occurrence in MEDLINE record
 - Abstract
 - sentence
- ◆ Caveat – co-occurrence may not always the existence of a biologically meaningful relationship
- ◆ Need a way to estimate the importance of co-occurrence

Importance of Co-occurrence

- ◆ Fuzzy logic – not 0 or 1, somewhere in between
- ◆ Score based on frequency
- ◆ Calculate expected value based on relative connectivity
 - Assume a random network
 - How far does this relationship deviate

Implementation: estimate of precision and recall

- ◆ (Why are they putting this section here?)
- ◆ In general, precision and recall measurements are difficult in text-mining
 - Gold standard
 - Test corpus

Precision

- ◆ Manual estimation based on sample
- ◆ Looked at 25 randomly selected MEDLINE records
 - Found that 2 objects co-mentioned within the same sentence were more likely related (83%) than objects mentioned in abstract (53%)
 - Sentence co-mentions alone misses relationships (43%)

Trivial vs. non-trivial relationships

- ◆ Found non-persistent relationships
 - In first half of MEDLINE but not in second half
 - Assumed false or not interesting relationships
- ◆ Rates similar to power decay function
- ◆ Decided OK to use as error probability

Recall

- ◆ Studied recall rates using abstracts vs. full text articles
 - Chose 4 objects, one of each type, had to have 2 review articles in last 3 yrs
 - Compared objects
- ◆ Results
 - 30 objects in the literature not in database, for various reasons
 - 141 of 181 objects in database (78%)
 - 98% could have been because terms were in abstracts (spelling errors)

Processing MEDLINE records

- ◆ 12,037,763 MEDLINE records
- ◆ Created a network of 3,482,204 unique relationships between objects
- ◆ Many objects had a high number of connections – unwieldy number
- ◆ Needed to rank potential significance
- ◆ Obs/Exp calculated

Cardiac Hypertrophy

- ◆ An example – why?
- ◆ Looked at compounds with implicit relationships to cardiac hypertrophy
- ◆ Cholopromazine ranked high
- ◆ Mouse trials showed CPZ lowers hypertrophy
- ◆ A relationship between cardiac hypertrophy/CPZ not mentioned previously in literature

Discussion

- ◆ A new relationship was found
- ◆ Shortcomings of method
 - Finding uninteresting relationships
 - Time consuming to find nature of relationship
 - Comparison to random network model assumes text is non-random. Is it?
 - Method has utility as information increases

Comments

- ◆ Confusing paper
 - Structure
 - Formulas
 - GDB and Genome Ontology – were they used?
- ◆ Why cardiac hypertrophy? Did it come to the top in results or was it originally a disease of interest?

Text Mining Issues

- ◆ Evaluation of methods – precision/recall
- ◆ The human component – someone must decide whether connection is interesting and potentially useful
- ◆ Collaboration