3D Semantic knowldege retrieval

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I. Introduction

- The method used to capture the essential semantic information is dimension reduction:
 - selecting the most important dimensions from a co-occurrence matrix decomposed using Singular Value Decomposition. (Deerwester, Dumais, Furnas, Landauer, & Harshman, 1991)

I. Introduction

- LSA is a machine-learning model:
 - that induces representations of the meaning of words
 - by analyzing the relation between words and passages in large bodies of text (Corpus)

I. Introduction

 As a result, LSA offers a way of assessing semantic similarity between any two samples of text in an automatic, unsupervised way. (Landauer & Dumais, 1997).

I. Introduction

- LSA has been used in applied settings with a high degree of success in areas like:
 - automatic essay grading
 - (Foltz, Laham, & Landauer, 1999)
 - automatic tutoring to improve summarization skills in children
 - (E. Kintsch, Steinhart, Stahl, Matthews, Lamb, & the LSA Research Group, 2000).

I. Introduction

- As a model LSA's most impressive achievements have been:
 - in human language acquisition simulations (Landauer & Dumais, 1997)
 - and modeling of high-level comprehension phenomena like metaphor understanding, causal inferences and judgments of similarity (Kintsch, 2001).

I. Introduction

- Due to multidimensional nature of semantic space, LSA results are hard to visualize.
- Our goal was to develop a tool for visualizing this semantic relationship.

II. LSA implementation in R

Algorithm developed in R:

- 1. scan text:
 - Corpus with 6000 diferent words over 1000 documents
 - From "The catcher in the rye"

II. LSA implementation in R



- 2. normalize text:
 - remove puntuation
 - word stemming using Porter's algortihm

II. LSA implementation in R

• 3. Build Word by Document matrix



II. LSA implementation in R



• 4. Singular Value Decomposition

II. LSA implementation in R

• 5. Build new low dimensional matrix



- II. LSA implementation in R
- 6.a Build term by term distances matrices:



- Cosin, length & 1/eucl

III. Results

- 3D plots for word "Lago" (Lake)
 - Dimension reduction:
 - 100
 - 200

III. Results



- 3d plot for word Lake in a 100 dimensional space





- 3d plot for word Lake in a 200 dimensional space

III. Results

- 3D plots for word "Soda"
 - Dimension reduction:
 - 100

• 200

III. Results



3d plot for word "Soda" in a 100 dimensional space

III. Results



3d plot for word "Soda" in a 200 dimensional space

IV. Conclusion

- R offers an enviroment for Isa visualization.
- Applications:
 - Psychology
 - Linguistics
 - Cognitive Science
 - User modelling
 - Etc...
- Future work:
 - Dimension reduction using cmdscale has offer ad-hoc results in some cases.

Literature cited

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