Cluster Hypothesis in Low-Cost IR Evaluation with **Different Document Representations**

Kai Hui, Klaus Berberich

Max Planck Institute for Informatics

Motivation

Cluster hypothesis: documents that are relevant to the same query should be more similar with each other

Low-cost evaluation with cluster

Document Representations

Bag-of-words (*BOW*): with *tf-idf* weight

Expanded BOW with word embedding (*EBOW*): use the similarity matrix among word embedding to expand the document

hypothesis: if the cluster hypothesis is satisfied, the manual assessments of document relevance can be done partially by automatic method, e.g., text classification or clustering

N

Two influential factors: similarity measures and document representations

Low-Cost Evaluation and Cluster Hypothesis

N

N

N

representation

Latent Dirichlet Allocation (LDA): generative topical modeling Latent semantic analysis (LSA): topical modeling based on matrix decomposition Neural network based vectorization (*Para2Vec*): co-train word embedding together with paragraph N embedding as memory

Benchmarks

- Triple Test: comparison of similarity between (Rel, Rel) against similarity between (Rel, Non-Rel)
- Knn Test: the precision of relevance in **k** most similar documents of a given relevant document



Conclusions

- Agreement to the cluster hypothesis is not good enough for low-cost evaluation
- Improvement with word embedding is non-trivial
- **EBOW perform best:** red bold number indicates significance

Task	Benchmark	BOW	EBOW	LDA	LSA	Para2Vec
	Triple Test	0.61	0.62 (1.1%)	0.51 (-17%)	0.47 (-24%)	0.53 (-14%)
Adhoc	Knn Test @5	0.62	0.62 (0.4%)	0.53 (-15%)	0.58 (-6%)	0.57 (-8%)

