START OF QUIZ Student ID: 90412503, Yin, Ting

Topic: Lecture 7 Source: Lecture 7

Explain why the cosine similarity between a document and query vector is roughly equivalent to adding up the TF-IDF scores of each word in the document that occurs in the query. (2)

Topic: Lecture 8 Source: Lecture 8

In class, I mentioned that high k value for BM25 TF weighting rewards documents with many, many instances of a term in them. Explain why that's the case. (2)

Topic: Lecture 8 Source: Lecture 8

What is the reasonining behind substituting TF-IDF with Okapi BM25? (1)

Topic: Lecture 7 Source: Lecture 7

Define P@R. (1)

Topic: Lecture 6 Source: Lecture 6

Why do we need a "human in the loop" for topic modeling? (1)

Topic: Lecture 6 Source: Lecture 6

In some ways, we could consider Beta distributions themselves to be an embedding of a topic. Explain, and explain how we might be able to leverage that. (2)

Topic: Lecture 5 Source: Lecture 5

The Frobenius norm looks very similar to a distance metric we've already observed. Explain which one. (1)

Topic: Lecture 5 Source: Lecture 5

Why can we be confident that a low-rank approximation of a matrix contains the most important information in a document? (1)

Topic: Long

Source: Lecture 7

Imagine that we have 2 information retrieval systems, and we are evaluating on the same test set, which has 10 relevant documents. The first system returns them in positions [1, 5, 7, 15, 25, 50, 60, 70, 71, 90]. The second returns the documents at positions [2, 3, 6, 8, 10, 62, 80, 83, 91, 95]. Make an argument for each system being better, and provide support for both. Explain which system you would rather use, and why. If there are any other considerations, list them. (3)

END OF QUIZ