605.744: Information Retrieval Problem Set (Module 5)

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1. (10%) Briefly describe the three key assumptions of the Cranfield paradigm for information retrieval evaluation.

Answer:

2. (10%) What is pooling and why is it used in large-scale text retrieval evaluations?

Answer:

3. (50%) Consider a query with 10 relevant documents whose docids are: D3, D27, D30, D39, D51, D54, D69, D72, D81, and D96. Assume that all other documents are not relevant. On this query two retrieval systems FastSearch and Telescope produce the following ranked lists. (Note: D17 is the 1st ranked doc by FastSearch; D4 is its 2nd ranked doc, etc ...)

FastSearch: D17, D4, D69, D54, D37, D41, D89, D85, D3, D5, D91, D39 Telescope: D3, D1, D94, D27, D50, D54, D16, D7, D72, D39, D95, D62

(a) How many relevant documents are found by each system?

Answer:

(b) For both systems what is P@10 (precision at 10 documents) for this query?

Answer:

(c) For FastSearch what is the uninterpolated precision at 30% Recall?

Answer:

(d) Assuming that FastSearch returns no other documents other than this top-12 ranked list, what is FastSearch's Recall for this query?

Answer

(e) For both systems what is average precision on this query?

Answer:

4. (15%) Given two retrieval systems (called A and B), is it possible for System A to be better than System B in average precision, but for System B to have higher P@10 than System A? Briefly justify your response.

Answer:

5. (15%) Consider the contingency tables below for the word pairs (bicycle, helmet) and (bicycle, repairs). Suppose we are looking to expand a query containing the word bicycle by adding some potentially useful search terms. Using pointwise mutual information (PMI) to score

candidate terms, calculate scores for both helmet and repairs, and indicate which of the two would be the better expansion term. $N=15{,}000$ documents. Use base 2 logs.

$$PMI(x,y) = log_2\left(\frac{P(x,y)}{P(x)P(y)}\right) = log_2\left(\frac{N \times a}{(a+b)(a+c)}\right)$$

A: docs	B: docs
with both	with first
terms	term, but
together	not second
C: docs	D: docs
with	that
second	contain
term, but	neither
not first	term

	has helmet	missing helmet
has bicycle	22	54
missing bicycle	87	14837

	$_{ m has}$ $_{ m helmet}$	missing helmet
has bicycle	31	45
missing bicycle	164	14760

Answer: