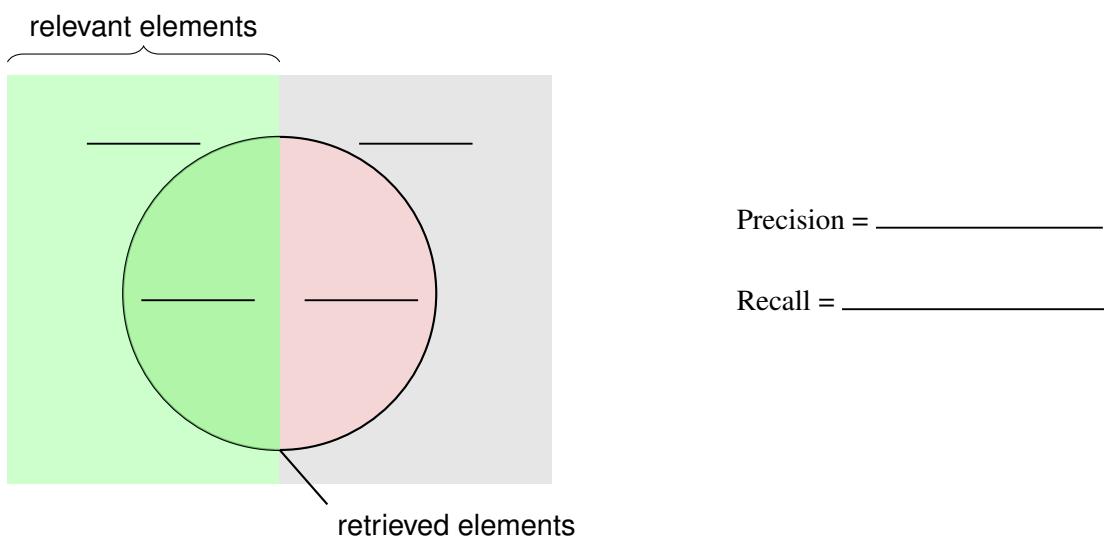


Lab Class IR:II**Exercise 1 : Effectiveness measures in Set Retrieval**

The diagram below illustrates the relationship between “relevant” and “retrieved” elements and True Positives (TPs), True Negatives (TNs), False Positives (FPs), and False Negatives (FNs) in an information retrieval task.

- (a) Label the four regions in the diagram with TP, FP, FN, and TN.
- (b) Complete the formulas for precision and recall in the fields provided.

**Exercise 2 : Measure Intuition (Set & Ranked Retrieval)**

Describe the following measures, each in 1–3 sentences:

- (a) Recall
- (b) Query throughput
- (c) Indexing space overhead
- (d) Normalized Discounted Cumulative Gain (nDCG)

Exercise 3 : Measure Suitability (Set Retrieval)

Name two search tasks where the Precision measure (not Precision@ k !) is good or bad at modelling the user behavior, respectively. Give a short explanation (1 sentence).:

- (a) For which search task is it a good fit?
- (b) For which search task is it a bad fit?

Exercise 4 : Effectiveness vs. Efficiency (Set & Ranked Retrieval)

Which of the following measures describe the effectiveness, and which describe the efficiency of a retrieval system? Give a short reason (up to 10 words)

- (a) Query latency
- (b) nDCG
- (c) Recall
- (d) Index size

Exercise 5 : Practical Limitations of Measuring Recall (Set Retrieval)

Recall measures how many of the relevant documents were successfully retrieved.

- (a) Explain in your own words what happens when recall is very high.
- (b) Is there a trivial way to maximize recall? What would be the consequence for precision?
- (c) In practice, information retrieval systems often work with very large document collections. What difficulties arise when trying to measure recall in such settings?

Exercise 6 : Precision, Recall, and F-Measure (Set Retrieval)

Two information retrieval systems A and B , have been tested on the same topics. The number of relevant documents retrieved by each system is shown below:

System	Relevant Retrieved	Total Retrieved
A	8	10
B	5	6

Assume the total number of relevant documents in the collection is 12.

- (a) Compute the precision p and recall r for each system.
- (b) Determine which system is better in terms of precision p and which is better in terms of recall r .
- (c) Compute the F-measure (i.e., harmonic mean $\frac{2 \cdot p \cdot r}{p + r}$ of precision p and recall r) for each system.
- (d) Based on the F-score, determine which system performs better overall.
- (e) Explain why the F-measure is a better single measure in this scenario.

Exercise 7 : Harmonic vs. Arithmetic and Geometric Means (Set Retrieval)

The F-measure induces a total order and is computed as the harmonic mean of precision p and recall r . Compare the harmonic, geometric, and arithmetic means of precision and recall. Explain why the harmonic mean ([Figure 1](#)) is often preferred over the geometric mean ([Figure 2](#)) or the arithmetic mean ([Figure 3](#)) in evaluating performance metrics. Discuss how each mean behaves when precision and recall differ significantly, and justify why the harmonic mean provides a more balanced evaluation in such cases.

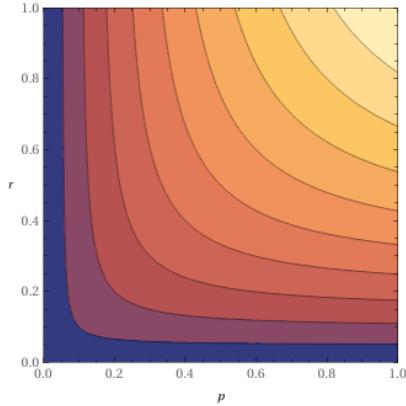


Figure 1: Harmonic mean $\frac{2 \cdot p \cdot r}{p+r}$ [\[plot\]](#)

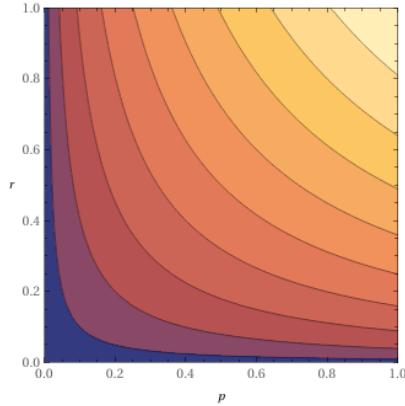


Figure 2: Geometric mean $\sqrt{p \cdot r}$ [\[plot\]](#)

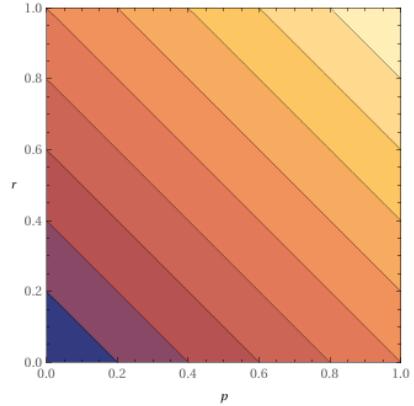


Figure 3: Arithmetic mean $\frac{p+r}{2}$ [\[plot\]](#)

Exercise 8 : Precision and Recall (Set Retrieval)

Consider a retrieval system that achieves a recall of r_1 and a precision of p_1 when retrieving k_1 documents. Suppose we retrieve more documents, $k_2 > k_1$, and compute the corresponding recall r_2 and precision p_2 .

Determine whether the following statements are true or false:

- (a) $r_2 \geq r_1$
- (b) $p_2 \geq p_1$

Exercise 9 : Interactive Retrieval

Name three main differences between ad hoc retrieval and interactive retrieval.

Exercise 10 : Cranfield Paradigm

Which items are required for a laboratory experiment for ad hoc retrieval? This setup is also referred to as an experiment under the Cranfield Paradigm.

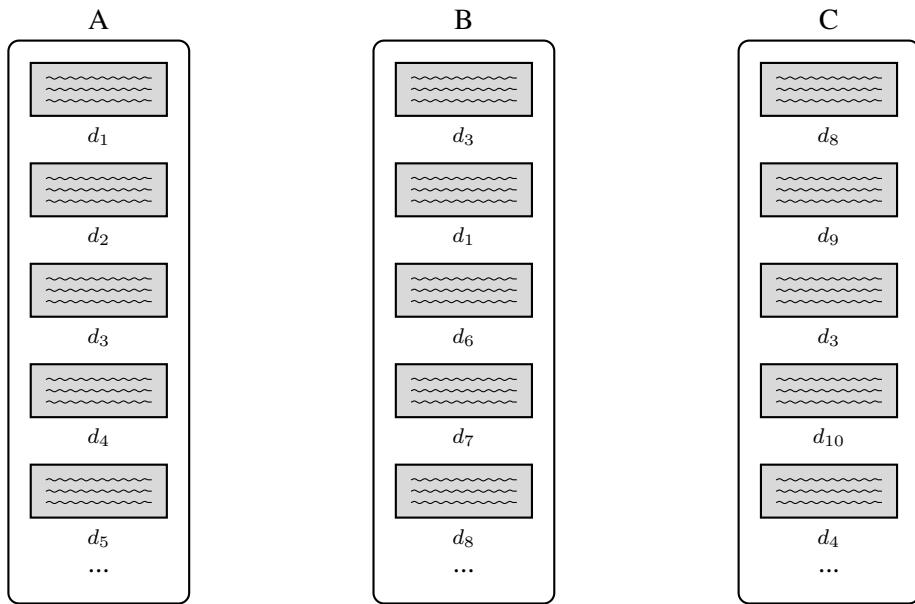
Exercise 11 : Experimental setup: Relevance judgements

Which dimensions should be considered for carrying out manual assessment for relevance judgement.

Exercise 12 : Pooling

Given a corpus with documents denoted as d_i indexed by three retrieval systems A , B , and C , each producing a ranked list of documents for a query q (where the top is considered most relevant), answer the following:

- (a) For $k = 3$ pooling depth, which documents are included in the resulting document “pool”?
- (b) Explain the limitations of this pooling approach in the context of this example.



Exercise 13 : Kappa statistics

Given the judgements of two annotators A, B on a given topic, Scott's κ measures their agreement as follows:

$$\kappa = \frac{p_o - p_e}{1 - p_e}$$

where p_o denotes the proportion of agreement observed, and p_e the expected proportion of agreement by chance.

- (a) Compute the Scott's κ measure for the binary relevance judgements given in Table 1.
- (b) How would you interpret the results (using Table 2)?

Table 1: Annotators A and B have made the following $n = 600$ binary relevance judgements.

		B		\sum
		yes	no	
A	yes	456	44	500
	no	13	87	100
\sum		469	131	600

Table 2: Interpretations to measured kappa values (disputed).

κ	Agreement
< 0	poor
0.01 - 0.20	slight
0.21 - 0.40	fair
0.41 - 0.60	moderate
0.61 - 0.80	substantial
0.81 - 1.00	almost perfect