University of Twente

Information Retrieval Modeling

Russian Summer School in Information Retrieval







PART 5 Evaluation

Course material

 Djoerd Hiemstra and Wessel Kraaij, Evaluation of Multimedia Retrieval Systems, In Multimedia Retrieval, Springer, pages 347-365, 2007

Goal

- An introduction to doing real (measurable, repeatable) research
- Getting acquainted with the TREC paradigm

Science?

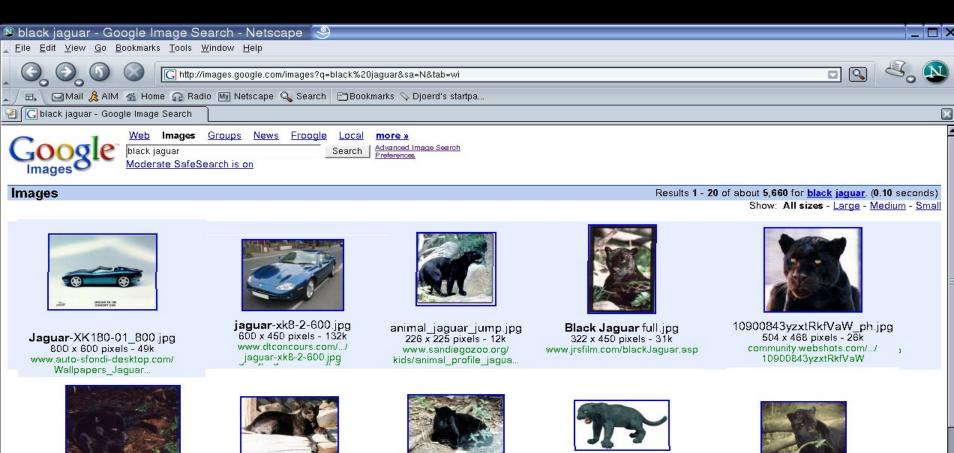
- At some point during your studies (or in your professional life), you might have to do a scientific evaluation...
 - choose data
 - choose suitable evaluation measures: assign values to results of your system
 - choose a statistical methodology: determine whether observed differences are significant

- To start with you need
 - An IR system (or two)
 - A collection of documents
 - A collection of requests
- Then you run your experiment
 - Input (index) the documents
 - Put each request to the system
 - Collect the output

- Then you need to
 - Evaluate the output, document by document
 - Discover (??) the good documents your system has missed
 - Analyse the results

- What is a document?
 - traditionally: a package of information structured by an author
- What is a request?
 - a description of a topic of interest
 - a partial representation of an underlying information need
- What is a system?
 - a device that accepts a request and delivers of identifies documents
 - "device" may be an organisation: involve people(!)

- Assuming that documents are either relevant or not, the objective is:
 - To retrieve relevant documents
 - Not to retrieve non-relevant documents





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- Evaluation measures
 - precision = r/n: fraction of retrieved documents that is relevant
 - $-\text{recall} = {}^r/_R$: fraction of relevant documents that is retrieved

r: number of relevant documents retrieved

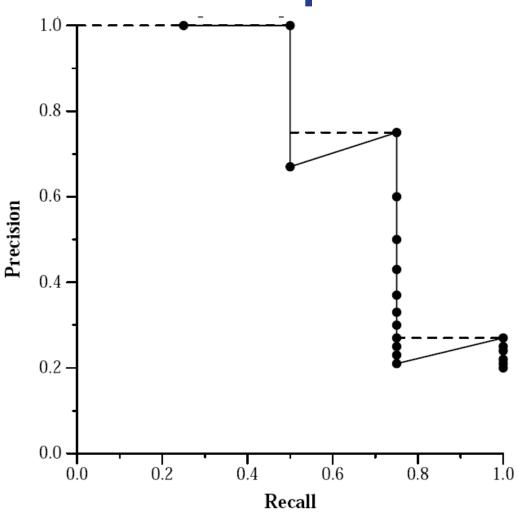
n: number of documents retrieved

R: number of relevant documents

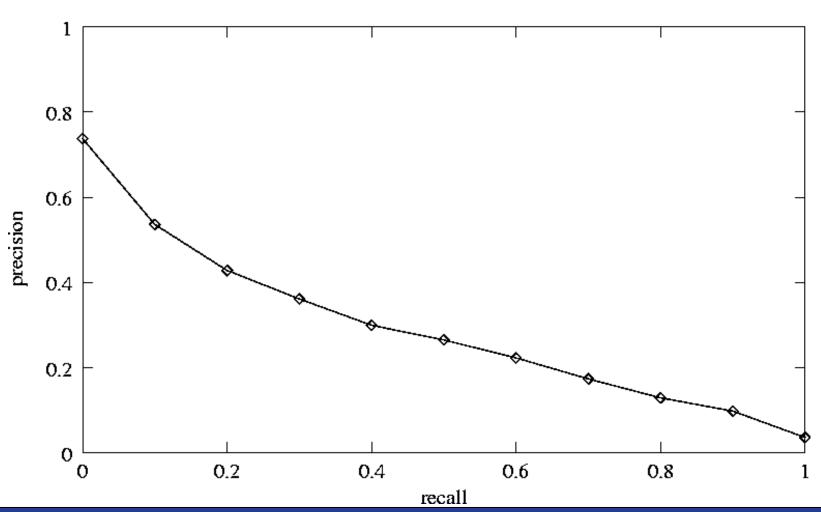
What about ranked output?

- Report precision for positions in the ranked list
 - -5, 10, 20 document retrieved
- Report precision for some recall levels
 - precision at 0.1, 0.2, etc.

Recall-precision plot



Recall-precision plot



- Problems with IR system evaluation
 - costly (involves users)
 - which documents did the system miss?
 - hard to repeat in same settings (learning / fatigue effects)
 - we need a complete system(!) we do not in general know how to evaluate components



The TREC paradigm

doing laboratory tests

Benchmark collections

- Consists of three parts:
 - documents (realistic contents and size)
 - requests (textual description of information need; realistic, "real" application)
 - relevance assessments: how useful is the retrieved document?
- How to design?
 - Cranfield →TREC →CLEF, INEX, ROMIP



What is TREC?

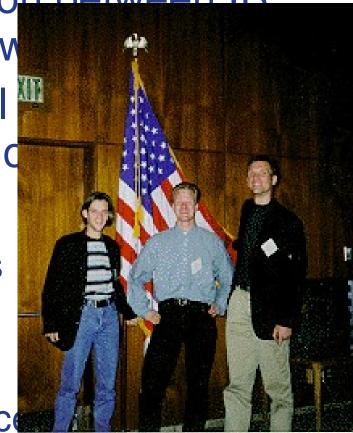
 Competition/collaboration between IR research groups world-w

 Run by the US National Standards and Technological

• TREC provides:

- common test collections
- common tasks
- common measures

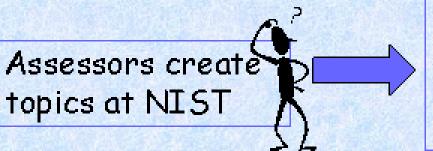
common evaluation proce



What is TREC?

- A workshop series that provides the infrastructure for large-scale testing of text retrieval technology
 - realistic test collections
 - uniform, appropriate scoring procedures
 - a forum for the exchange of research ideas and for the discussion of research methodology

TREC approach



Topics are sent to participants, who return ranking of best 1000 documents per topic

Systems are evaluated using relevance judgments

NIST forms pools of unique documents from all submissions which the assessors judge for relevance

An example TREC topic

```
<top>
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<num> 405

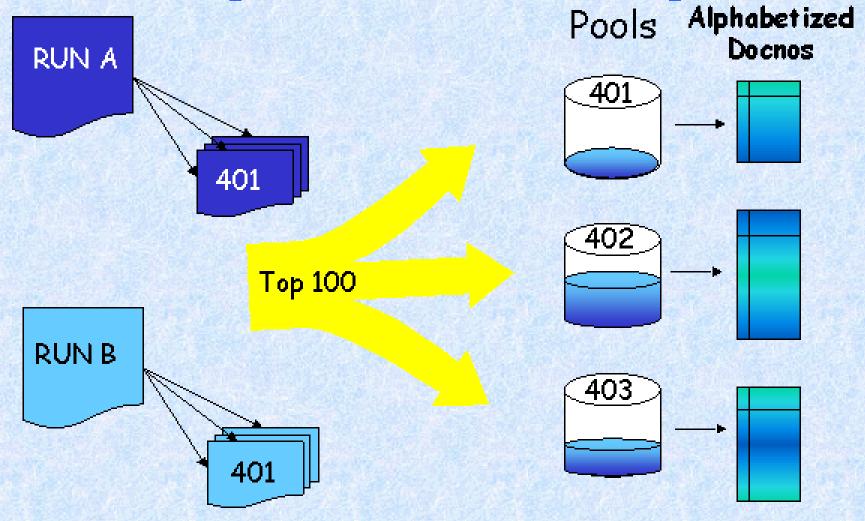
<title> cosmic events

<desc> What unexpected or unexplained cosmic
 events or celestial phenomena, such as
 radiation and supernova outbursts or new
 comets, have been detected?

<narr> New theories or new interpretations
 concerning known celestial objects made as a
 result of new technology are not relevant.

</top>

Creating Relevance Judgments







TREC assumptions about relevance

- Relevance of one element does not affect the relevance of another element
- Relevance is a binary decision, i.e., a document is either relevant or not
- A document is relevant if it would help in writing an article about the subject
 - relevant? topicality? clarity? recency? accuracy? trustworthiness?

TREC assumptions about systems

- A system is a programme
 - the user is outside the system
- A system is an input-output device
 - query in, documents out
 - most real searches involve interaction

Quality of a test collection?

- Two concerns:
 - Consistency of the judgements: do the results of the experiments critically depend on the particular choices of human judges?
 - Completeness of the judgements: do the results critically depend on the pool construction process, i.e. on the systems that participated in INEX?

Consistency of the judgements

- Experiment: 10 topics assessed twice by two different assessors
- Dutch CLEF collection, overlap: 0.465
- TREC: overlap between: 0.421 and 0.494
 (Overlap = size of intersection of the relevant document sets divided by the size of the union of the relevant document sets.)
- (Overall agreement 93.4 %)

Completeness of judgements

- Can we use the collection for future experiments?
- What if my run is not judged?
- Experiment: recompute for each official run the average precision as if it was not in the pool, i.e. ignoring the relevant documents uniquely found by that run

Completeness: What if my run is not judged?

run name	unjudged /	<u>judged</u>	avg.prec.	difference	unique rel.
ut1	0.4222	0.4230	0.0008	0.2 %	55
aplmonla	0.3943	0.4002	0.0059	1.5 %	29
tnonn3	0.3914	0.3917	0.0003	0.1 %	2
humNL01x	0.3825	0.3831	0.0006	0.2 %	5
tlrnltd	0.3760	0.3775	0.0015	0.4 %	10
tnoen1	0.3246	0.3336	0.0090	2.8 %	32
AmsNIM	0.2770	0.2833	0.0063	2.3 %	32
aplbiennl	0.2692	0.2707	0.0015	0.6 %	7
oce2	0.2363	0.2405	0.0042	1.8 %	21
glaenl	0.2113	0.2123	0.0010	0.5 %	8
oce1	0.2024	0.2066	0.0042	2.1 %	23
medialab	0.1600	0.1640	0.0040	2.5 %	23
EidNL2001A	0.1339	0.1352	0.0013	<u>1.0 %</u>	8
		mean	: 0.0031	1.2 %	20
	standard	deviation	0.0027	1.0 %	15

Significance testing

- When is one system better than another?
 - Maybe the average difference can be contributed to chance?
 - Need a reasonable amount of queries (e.g. 50), which should be a random sample of all possible queries for a given task

Significance testing

- Two hypotheses
 - null-hypothesis H_0 : there is no difference between system A and system B
 - alternative hypothesis H_1 : either system A consistently outperforms system B, or system B consistently outperforms system A
- Show that, given the evaluation results,
 H₀ is indefensible

Significance testing

- Test statistics should behave differently under H₀ than under H₁:
 - Paired tests: for each query the performance difference between system A and B consist of a mean difference μ and some error.
 - H_0 : $\mu = 0$; H_1 : $\mu \neq 0$;
 - Paired t-test: assumes that errors are normally distributed. Under H₀ the distribution is Student's t
 - Paired sign test: assumes equal probability of positive and negative error. Under H₀ the distribution is binomial

Conclusion

- To evaluate your system, use a benchmark collection.
- Choose appropriate evaluation measures
- Base your conclusions on statistical tests

References

- Stephen Robertson. Evaluation in information retrieval.
 Lecture Notes in Computer Science, Springer, pp. 81-92. 2000.
- Ellen Voorhees and Lori Buckland, The Sixteenth Text REtrieval Conference Proceedings (TREC 2007). NIST, 2007
- Donna Harman and/or Ellen Voorhees. Appendix:
 Common evaluation measures. In "Proceedings of most of the TREC proceedings".

Acknowledgements

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 - Stephen Robertson (Microsoft Research)
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