Information Retrieval Offline Evaluation

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### Course overview



### This lecture



### How would you evaluate a search system?

Yandex						
		information retrieval X 🛫	Search			
Web	W	Information retrieval - Wikipedia, the free encyclopedia en.wikipedia.org > Information retrieval *				
Images		Information retrieval (IR) is the activity of obtaining information resources relevant to an information need from a collection of information resources. Searches can be based on full-				
Video		text or other content-based indexing.				
Translate	n.	Information retrieval				
More		en.academic.ru > dic.nst/enwiki/9176 + 1950: The term " <b>information retrieval</b> " appears to have been coined by Calvin Mooers. 1951: Philip Bagley conducted the earliest experiment in computerized document <b>retrieval</b> in a				
	Ψ	Information retrieval - Psychology Wiki - Wikia psychology.wikia.com > wiki/Information_retrieval >				
		Assessment   Biopsychology   Comparative   Cognitive   Developmental   Language   Individual differences   Personality   Philosophy   Social   Methods   Statistics   Clinical   Educational   Industrial   Professional items   World psychology  .				
	ŃP	Introduction to Information Retrieval				
		Infl. Station Leader 1 - 100000 - 1 - 100000 - 1 - 100000 - 10000000 - 100000000				
	•1)	Information retrieval - Wikiquote en.wikiquote.org > wikinformation_retrieval + Information retrieval is the activity of obtaining information resources relevant to an Information need from a collection of Information resources, and the part of Information science, which suites of these activity.				

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### Taxonomy of evaluation approaches



Diane Kelly, "Methods for Evaluating Interactive Information Retrieval Systems with Users"





### 2 Metrics



### Outline

### 1 Test collections

- Components of test collections
- Evaluation campaigns

### 2 Metrics

### 3 Summary





### **1** Test collections

Components of test collections

Metrics

Summary

### What components should a test collection comprise?



Picture taken from http://nationwidepathology.co.uk

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### What components should a test collection comprise?

- Test documents
- Test queries
- Ground truth



## Use a document collection that is representative for the application in terms of the number, size, and type.



- Where can we get test queries?
  - Query log
  - Example queries from potential users
- How many queries should we get?
  - The more the better
  - At least 50

### Ground truth

wars in netherlands in 17th century

#### Anglo-Dutch Wars - Wikipedia, the free encyclopedia https://en.wikipedia.org/wiki/Anglo-Dutch Wars -

The Anglo-Dutch wars (Dutch: Engels-Nederlandse Oorlogen or Engelse Zeeoorlogen) were ... A view of the Dutch factory at Ambon, early to mid-17th century.

#### First Anglo-Dutch War - Wikipedia, the free encyclopedia https://en.wikipedia.org/wiki/First\_Anglo-Dutch\_War -

1654, depicts the final battle of the First Anglo-Dutch War. ... By the middle of the 17th century the Dutch had built by far the largest mercantile fleet in Europe, ...

#### 1652-1674 Anglo-Dutch Wars - Rijksmuseum

https://www.rijksmuseum.n/en/...dutch.../1652-1674-anglo-dutch-wars 
In the 17th century, England fought three wars with the Republic in a little over twenty
years. Rivalry between the two mercantile nations and European power ...

#### The Anglo-Dutch wars - Het Geheugen van Nederland www.geheugenvannederland.nl/?/en/collecties/nederland\_engeland/... •

(Dutch-English (Naval) Wars). Three of them were fought in the seventeenth century, one in the eighteenth. Trade conflicts and naval supremacy were at stake in ...

#### Anglo-Dutch Wars | European history | Britannica.com www.britannica.com/topic/Anglo-Dutch-Wars -

Jul 4, 2014 - Anglo-Dutch Wars, also called Dutch Wars, Dutch Engelse Oorlogen, (English Wars), the four 17th- and 18th-century naval conflicts between ...





X







### Relevance judgements

- Where can we get relevance judgements?
  - Users
  - Independent judges
  - Crowdsourcing
- How many relevance judgements should we get?
  - The more the better
  - More judged queries, fewer judgements per query
  - Multiple judges
- Graded relevance
  - 4 perfect
  - 3 excellent
  - 2 good
  - 1 fair
  - 0 bad

## Pooling

- Impossible to obtain judgments for all documents
- Depth-k pooling
  - consider multiple search systems (by participants)
  - 2 consider top-k results from each system
  - ③ remove duplicates
  - ④ present documents to judges in a random order
- Produces a large number of judgments for each query
- Still incomplete
- Other methods
  - Automatic evaluation
  - Minimum test collection

### Multiple assessors

Inter-assessor agreement, Cohen's kappa coefficient

$$\kappa = \frac{P(A) - P(E)}{1 - P(E)}$$

- Expected chance agreement P(E)
- Values

- 0.67 0.8 acceptable
- $\bullet\ < 0.67 low$

• For more than two assessors, average pair-wise coefficients

### Components of test collections

### Documents

### Queries

### Judgements



search engine evaluation Amsterdam web search University of Amsterdam information studies







#### **1** Test collections

- Evaluation campaigns

### Evaluation campaigns

- Text REtrieval Conference (TREC)
  - US National Institute of Standard and Technology (NIST)
  - http://trec.nist.gov
- Cross-Language Education and Function (CLEF)
  - Mainly European
  - http://www.clef-campaign.org
- NII Test Collections for IR (NTCIR)
  - National Institute of Informatics of Japan (NII)
  - http://research.nii.ac.jp/ntcir/index-en.html
- Российский семинар по Оценке Методов Информационного Поиска (РОМИП)
  - Russian version of TREC
  - http://romip.ru

## Text REtrieval Conference (TREC)



http://trec.nist.gov

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## TREC greatest hits

Track	Dataset	Year	Documents	Queries
Ad hoc track	TREC 1–8	1994–1999	1,89 million	450
Web track	WT10G	2000-2001	1,692,096	100
	ClueWeb09	2009–2012	1,040,809,705	200
	ClueWeb12	2013–2014	733,019,372	100
Terabyte track	GOV2	2004–2006	25,205,179	150

### NIST assessors



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### Outline

### 1 Test collections

### 2

### Metrics

- Unranked evaluation
- Ranked evaluation
- User-oriented evaluation
- Evaluating other aspects of search
- Evaluating metrics

### 3 Summary

### Outline



#### Metrics

#### • Unranked evaluation

- Ranked evaluation
- User-oriented evaluation
- Evaluating other aspects of search
- Evaluating metrics

### Precision and recall

• Precision is the fraction of retrieved items that are relevant

 $Precision = \frac{\#(relevant items retrieved)}{\#(retrieved items)} = P(relevant | retrieved)$ 

• Recall is the fraction of relevant items that are retrieved

$$\mathsf{Recall} = rac{\#(\mathsf{relevant items retrieved})}{\#(\mathsf{relevant items})} = P(\mathsf{retrieved} \mid \mathsf{relevant})$$

Manning et al., "Introduction to Information Retrieval"

### Precision and recall

	Relevant	Non-relevant
Retrieved	true positives (TP)	false positives (FP)
Not retrieved	false negatives (FN)	true negatives (TN)

$$P = \frac{TP}{TP + FP}$$
$$R = \frac{TP}{TP + FN}$$

Manning et al., "Introduction to Information Retrieval"

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### Precision-recall curve



Manning et al., "Introduction to Information Retrieval"

### F-measure

#### • F-measure

$$F = \frac{1}{\alpha \frac{1}{P} + (1 - \alpha) \frac{1}{R}} = \frac{(\beta^2 + 1)PR}{\beta^2 P + R},$$

where 
$$\beta^2 = \frac{1-\alpha}{\alpha}$$
  
• F1-measure ( $\alpha = 0.5, \beta^2 = 1$ )  
 $F_1 = \frac{2PR}{P+R}$ 

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### Any problems with the metrics so far?

## The ranking of items is not taken into account

### Outline



#### Metrics

Unranked evaluation

#### Ranked evaluation

- User-oriented evaluation
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### Precision and recall

• Precision at rank k

$$P@k = \frac{\#(\text{relevant items at } k)}{k}$$

• Recall at rank k

$$R@k = rac{\#(\text{relevant items at }k)}{\#(\text{relevant items})}$$

### Other common metrics

• Reciprocal rank

$$RR = \frac{1}{\text{rank of first relevant item}}$$

• Average precision (AP)

$$AP = \frac{\sum_{d \in rel} P@k_d}{\#(\text{relevant items})}$$

- Average over multiple queries
  - mean P@k
  - In mean R@k
  - MRR
  - MAP

### Any problems with the metrics so far?

### User search behavior is not taken into account

### Outline



#### Metrics

- Unranked evaluation
- Ranked evaluation

#### User-oriented evaluation

- Evaluating other aspects of search
- Evaluating metrics

### User search behavior



Picture taken from Google-Studie-Betrachtungsverlauf-einer-SERP.pdf

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### Discounted cumulative gain (DCG)

- Graded relevance  $R_k \in \{0, 1, 2, 3, 4\}$
- Cumulative gain

$$CG = \sum_{k=1}^{N} (2^{R_k} - 1)$$

• Gain is **discounted** by rank

$$D(k) = \frac{1}{\log(k+1)}$$

Discounted cumulative gain

$$DCG = \sum_{k=1}^{N} \frac{2^{R_k} - 1}{\log(k+1)}$$

Normalized DCG

$$NDCG = \frac{DCG}{DCG_{ideal}}$$

### Rank-biased precision (RBP)

- View next item with probability  $\boldsymbol{\theta}$
- Stop with probability  $1 \theta$
- Probability of looking at rank k

$$P(\text{look at } k) = \theta^{k-1}$$

• Average number of examined items

Avg. exam 
$$= \sum_{k=1}^{\infty} k \cdot P(\text{look at } k) \cdot P(\text{stop at } k)$$
$$= \sum_{k=1}^{\infty} k \cdot \theta^{k-1} \cdot (1-\theta)$$
$$= \frac{1}{1-\theta}$$

### Rank-biased precision (RBP)

• Utility at rank k

$$U@k = P(\text{look at } k) \cdot R_k = \theta^{k-1} \cdot R_k$$

• Average utility of all results

$$RBP = \frac{\sum_{k=1}^{N} U@k}{Avg. \text{ exam}} = (1-\theta) \cdot \sum_{k=1}^{N} \theta^{k-1} \cdot R_k$$

•  $\theta$  is usually close to 1

### Expected reciprocal rank (ERR)

• Reciprocal rank

$$RR = rac{1}{ ext{rank of first relevant item}}$$

- If an item is relevant  $(R_k)$  then stop
- Otherwise  $(1 R_k)$ , continue with probability  $\theta$

Probability of looking at rank k

$$P( ext{look at } k) = \prod_{i=1}^{k-1} (1-R_i) \cdot heta$$

• Probability of reciprocal rank = k

$$P(RR = \frac{1}{k}) = R_k \cdot \prod_{i=1}^{k-1} (1 - R_i) \cdot \theta$$

### Expected reciprocal rank (ERR)

• Expected reciprocal rank

$$\begin{aligned} \mathsf{ERR} &= \sum_{k=1}^{N} \frac{1}{k} \cdot \mathsf{P}(\mathsf{RR} = \frac{1}{k}) \\ &= \sum_{k=1}^{N} \frac{1}{k} \cdot \theta^{k-1} \cdot \mathsf{R}_{k} \cdot \prod_{i=1}^{k-1} (1 - \mathsf{R}_{i}) \end{aligned}$$

 ${\ensuremath{\, \bullet \,}}\ \theta$  is usually close to 1

### Outline



#### Metrics

- Unranked evaluation
- Ranked evaluation
- User-oriented evaluation
- Evaluating other aspects of search
- Evaluating metrics

### Evaluating other aspects of search

Intent-aware measures

$$M(q) = \sum_i M_i(q) P(i \mid q)$$

- Novelty and diversity:  $\alpha$ -nDCG
- Session-based evaluation

### Outline



#### Metrics

- Unranked evaluation
- Ranked evaluation
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### Evaluating metrics

- No established methodology
- Metrics for metrics
  - Discriminative power
  - Intuitiveness/concordance
- Correlation with online metrics and experiments

### Discriminative power

Metric	Discriminative Power
Precision	50.1~%
Precision2	30.8~%
DCG	48.6~%
ERR	39.3~%
uSDBN	$51.1 \ \%$
EBU	35.1~%
rrDBN	21.1~%
uDCM	34.7 %
rrDCM	26.0~%
uUBM	33.3~%

A. Chuklin et al. "Click model-based information retrieval metrics"

### Correlation with online metrics

		-RR			
	Max-	Min-	Mean-	UCTR	PLC
Precision	-0.117	-0.163	-0.155	0.042	-0.027
Precision2	0.026	0.093	0.075	0.092	0.094
DCG	0.178	0.243	0.237	0.163	0.245
ERR	0.378	0.471	0.469	0.199	0.399
EBU	0.374	0.467	0.464	0.198	0.397
rrDBN	0.384**	0.475	0.473	0.194**	$0.399^{-1}$
rrDCM	0.387**	0.478	0.476	0.194**	$0.400^{-1}$
uSDBN	0.322**	0.412**	0.407**	0.206**	0.370**
uDCM	0.374**	0.466**	0.463**	$0.198^{}$	0.396**
uUBM	0.377-*	0.469**	0.467**	$0.198^{}$	0.398-*

A. Chuklin et al. "Click model-based information retrieval metrics"





#### 2 Metrics



## Offline evaluation summary

- Test collection
  - Test documents
  - Test queries
  - Ground truth
- Metrics
  - Unranked
  - Ranked
  - User-oriented
  - Evaluating metrics

### What are the advantages of offline evaluation?

If we have a test collection then...

- Cheap
- Fast
- Reusable

### What are the disadvantages of offline evaluation?

- Creating a test collection is expensive
- Actual users are not considered

### Materials

- Croft et al., Chapter 8
- Manning et al., Chapter 8
- Evangelos Kanoulas

A Short Survey on Online and Offline Methods for Search Quality Evaluation Proceedings of RuSSIR, 2015

### Materials

DCG

Kalervo Järvelin, Jaana Kekäläinen **Cumulated gain-based evaluation of IR techniques** ACM Transactions on Information Systems, 2002

#### RBP

Alistair Moffat, Justin Zobel

Rank-biased precision for measurement of retrieval effectiveness ACM Transactions on Information Systems , 2008

ERR

Olivier Chapelle, Donald Metlzer, Ya Zhang, Pierre Grinspan **Expected reciprocal rank for graded relevance** Proceedings of CIKM, 2009

Evaluation of metrics

Aleksandr Chuklin, Pavel Serdyukov, Maarten de Rijke Click model-based information retrieval metrics Proceedings of SIGIR, 2013

# Hands-on

### Course overview



### Next lecture

