### Multimedia Information Retrieval and Technology

Lecture 9 Evaluation II

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Room: SD555



### Evaluation of unranked retrieval results Evaluation of ranked retrieval results

- ? Eleven-point interpolated average precision
- MAP (Mean Average Precision)
- ?Precision@K
- ? Normalized Discounted Cumulative Gain

For any information need  $q_j \in Q$ , we denote the set of relevant documents as  $\{d_1, \dots, d_{m_j}\}$ 

Precision( $d_k$ ) is the precision of retrieved results when we reach the document  $d_k$ .

For a single information need, Average Precision is the average of the precision value obtained for the set of top *k* documents existing after **each relevant document** is retrieved.

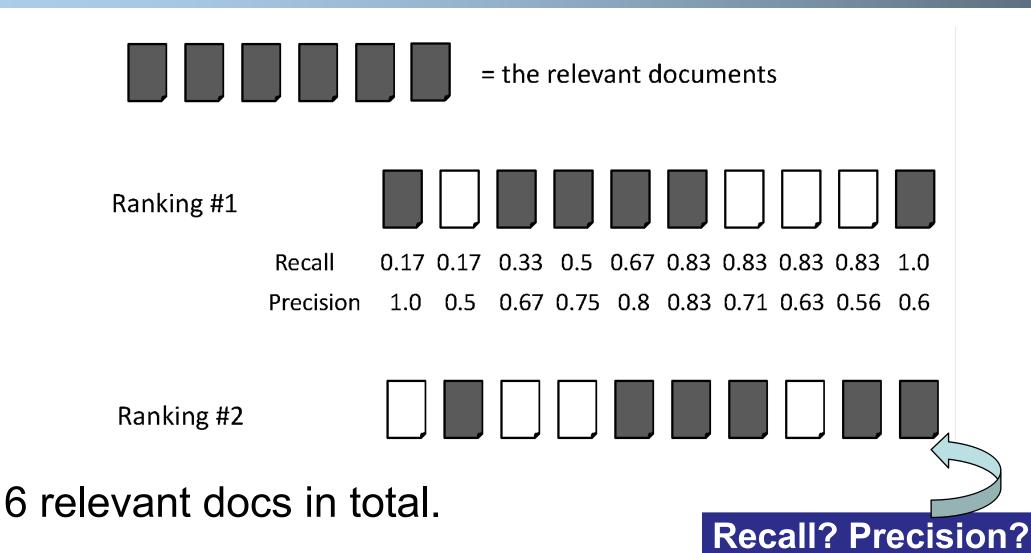
$$AP = \frac{1}{m_j} \sum_{k=1}^{m_j} Precision(d_k)$$

For an information need  $q_j \in Q$ , the IR system to be evaluated has returned the following documents:

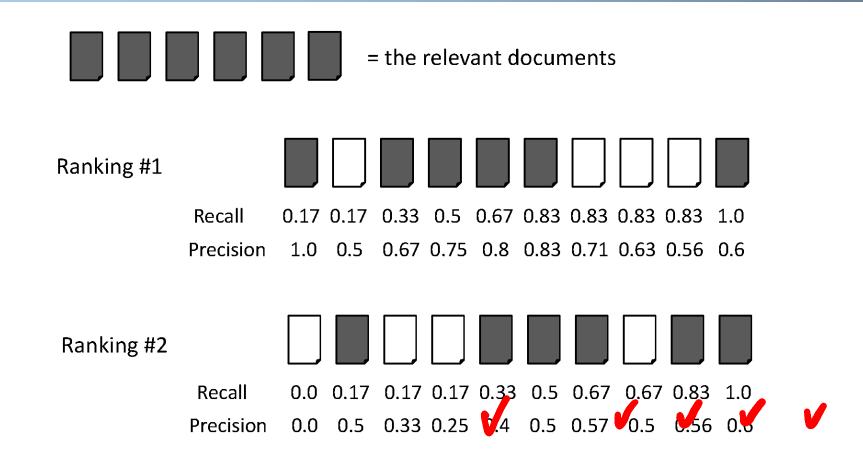
			•	
		Р	R	
1	R	1	0.01	$\simeq 0.87$
2	R		0.02	
3	N	2/3	0.02	
4	N	2/4	0.02	
5	R	3/5	0.03	
6	N	3/6	0.03	

Here,  $m_j$ =3, AP=?



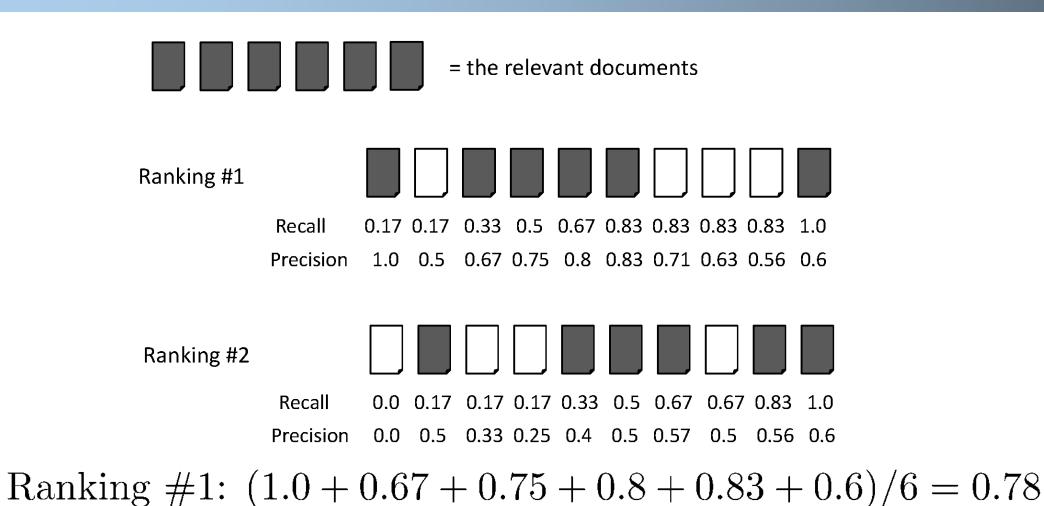


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AP=?





Ranking #2: (0.5 + 0.4 + 0.5 + 0.57 + 0.56 + 0.6)/6 = 0.52



#### Exercise I

Consider an information need for which there are 4 relevant documents in the collection. Contrast two systems run on this collection. Their top 10 results are judged for relevance as follows (the leftmost item is the top ranked search result):

System 1 RNRNN NNNRR System 2 NRNNR RRNNN

What is the AP of each system? Which has a better performance?



#### Solution

AP(System 1) = 
$$(1/4)*(1+(2/3)+(3/9)+(4/10)) = 0.6$$
  
AP(System 2) = $(1/4)*(1/2 + 2/5 + 3/6 + 4/7) = 0.493$ 

System1 has a higher average precision

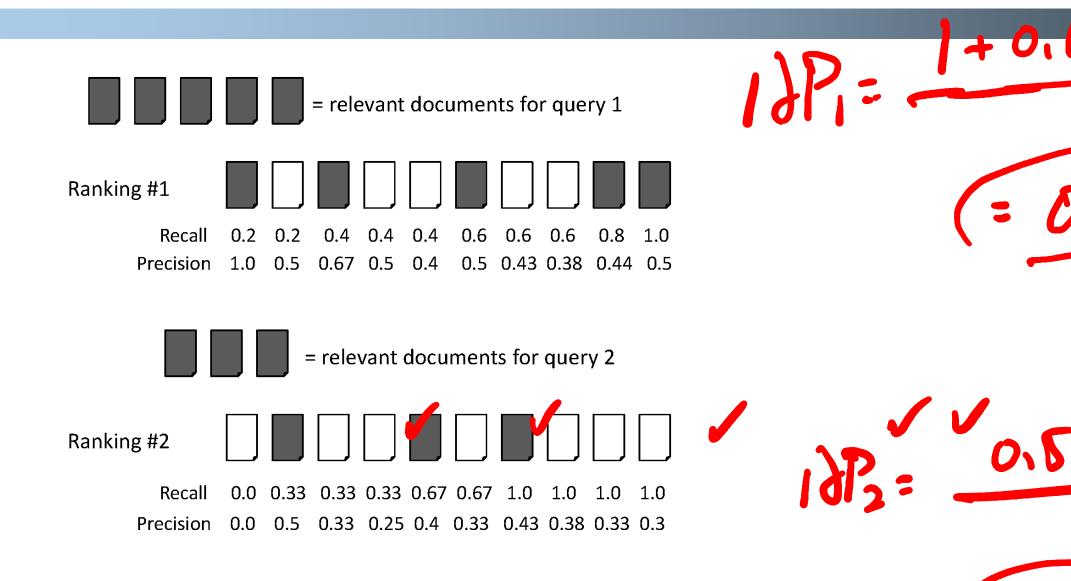
### Mean Average Precision

Then:

$$MAP(Q) = \frac{1}{|Q|} \sum_{j=1}^{|Q|} \frac{1}{m_j} \sum_{k=1}^{m_j} Precision(d_k)$$

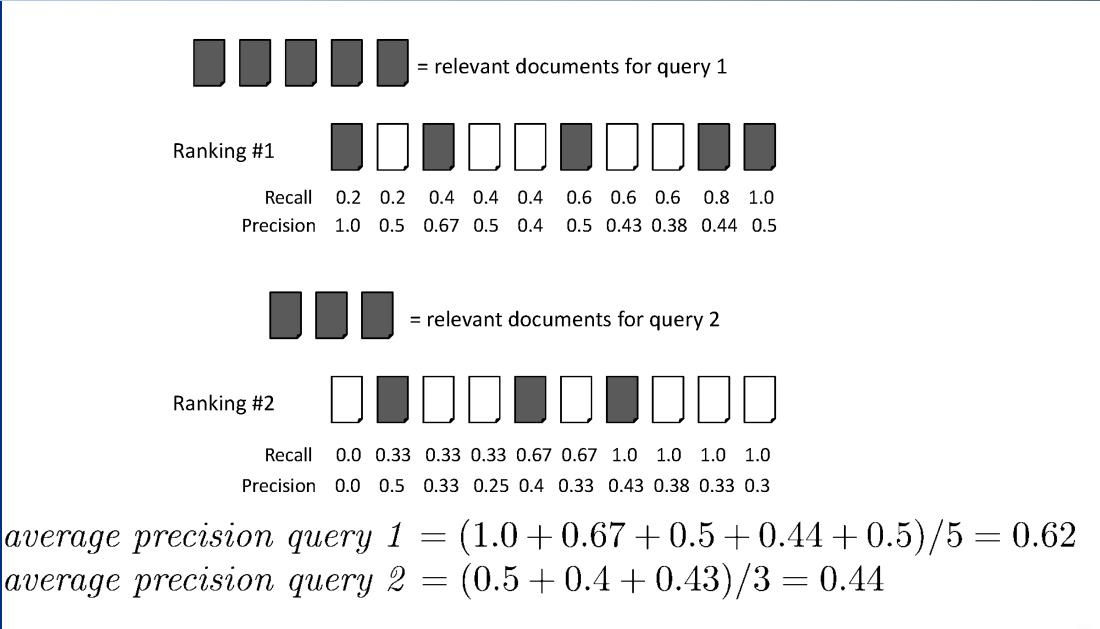
MAP is Average Precision across multiple queries and rankings

### Exercise II: MAP





#### Solution



mean average precision = (0.62 + 0.44)/2 = 0.53

### Mean average precision

MAP is macro-averaging: each query counts equally

Now perhaps most commonly used measure in research papers. Among evaluation measures, MAP has been shown to have especially good discrimination and stability.

Using MAP, fixed recall levels are not chosen, and there is no interpolation.

### Mean average precision

MAP assumes user is interested in finding many relevant documents for each query

MAP requires many relevance judgments in text collection

Good for web search?

What matters is rather how many good results there are on the first page or the first three pages.



### Evaluation of unranked retrieval results Evaluation of ranked retrieval results

- ? Eleven-point interpolated average precision
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### Precision@K

Measuring precision at fixed low levels of retrieved results, such as 10 or 30 documents.

Ignores documents ranked lower than K



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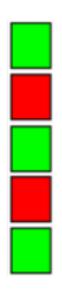
### Precision@K

### Example:

Prec@3 of 2/3

Prec@4 of 2/4

Prec@5 of 3/5



In similar fashion we have Recall@K

#### Exercise II

The following list of Rs and Ns represents relevant (R) and nonrelevant (N) returned documents in a ranked list of 20 documents retrieved in response to a query from a collection of 10,000 documents. The top of the ranked list (the document the system thinks is most likely to be relevant) is on the left of the list. This list shows 6 relevant documents. Assume that there are 8 relevant documents in total in the collection.

#### RRNNN NNNRN RNNNR NNNNR

e. Assume that these 20 documents are the complete result set of the system. What is the AP of the system based on the result above?



### Solution

- a) 0.3
- b) 3/7
- c) 0.33 or 0.3
- d) 0.364
- e) 0.555

## Evaluation of unranked retrieval results

Evaluation of ranked retrieval results

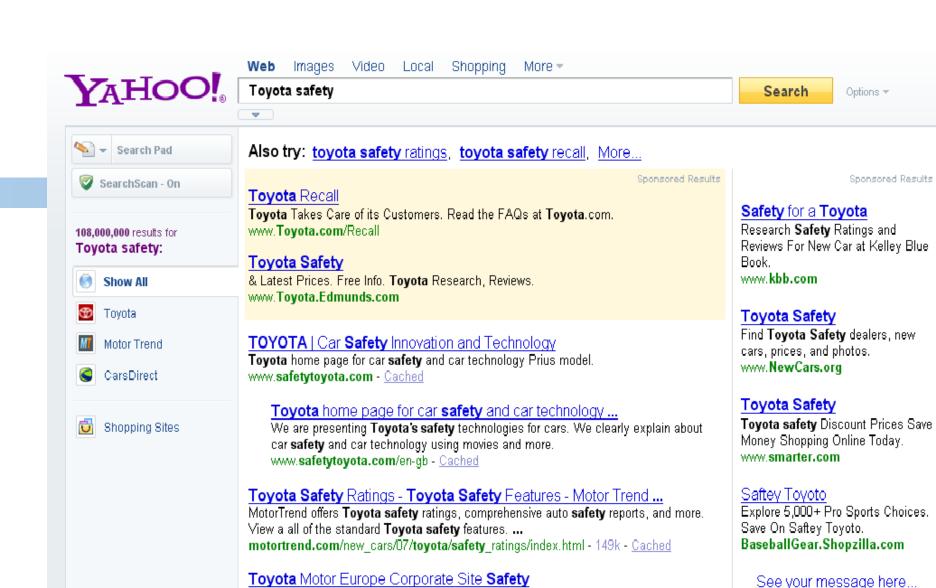
- ? Eleven-point interpolated average precision
- ?MAP
- ?Precision@K
- Normalized Discounted Cumulative Gain

#### Discounted Cumulative Gain

A Popular measure for evaluating web search and related tasks. A measure of ranking quality.

#### Two assumptions:

- Highly relevant documents are more useful than marginally relevant documents;
- the lower the ranked position of a relevant document, the less useful it is for the user, since it is less likely to be examined.



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responsibility to reduce the frequency of road accidents. ...

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not guarantee that all accidents or injuries will be avoided when driving a **Toyota** and/or Lexus brand motor vehicle equipped with the **safety** systems ... www.toyota.no/lmages/Safety Brochure tcm308-344461.pdf

Our approach. Toyota believes that all stakeholders in the road safety equation share a

#### Toyota - Star Safety System

Star **Safety** System ... **Toyota** Mobility Program. Careers. Contact Us. Home. contact us. site map. your privacy rights. legal terms. **Toyota** Newsroom. sign up for info ... www.toyota.com/vehicles/demos/star-safety.html - 58k - Cached

#### Toyota Prius Safety Ratings - CarsDirect

Get overall **safety** ratings and NHTSA crash test results for the **Toyota** Prius at CarsDirect.

#### Discounted Cumulative Gain

Uses *graded relevance* as a measure of usefulness, or *gain*, from examining a document

Eg: 5 ranked documents judged on 0-3 relevance scale

D1	3
D2	2
D3	3
D4	0
D5	1

D1	3
D3	3
D4	0
D2	2
D5	1 ;

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### Summarize a Ranking: DCG

Let the relevance ratings of the top n documents be  $rel_1, rel_2, ..., rel_n$  (in ranked order)

#### Cumulative Gain (CG) at rank n

$$CG = rel_1 + rel_2 + ... + rel_n$$

total gain accumulated at a particular rank n

#### Discounted Cumulative Gain

Gain is accumulated starting at the top of the ranking and may be reduced, or *discounted*, at lower ranks.

Typical discount is:

1/log (rank)

With base 2, the discount at rank 4 is 1/2, and at rank 8 it is 1/3



#### Discounted Cumulative Gain

#### Discounted Gain:

$$DG = \frac{rel_i}{\log_2 i}$$

DCG is the total discounted gain accumulated at a particular rank n:

$$DCG_p = rel_1 + \sum_{i=2}^{p} \frac{rel_i}{\log_2 i}$$



## DCG Example

Docs	D1	D2	D3	D4	D5	D6	D 7	D8	D9	D10
Gain	3	2	3	0	0	1	2	2	3	0
Discounted gain	3									
DCG	3									

## DCG Example

Docs	D1	D2	D3	D4	D5	D6	D 7	D8	D9	D10
Gain	3	2	3	0	0	1	2	2	3	0
Discounted gain	3	2/1 =2	3/1.59 =1.89	0	0	1/2.59 =0.39				
DCG	3	5	6.89	6.89	6.89	7.28				

**DCG** Example

Perfecto DI D3 D9 D2 D7 D8 D6 D4 B3 D10

Docs	D1	D2	D3	D4	D5	D6	D 7	D8	D9	D10
Gain	3	2	3	0	0	1	2	2	3	0
Discounted gain	3	2/1 =2	3/1.59 =1.89	0	0	1/2.59 =0.39	0. 71	0.67	0.95	0
DCG	3	5	6.89	6.89	6.89	7.28	7. 99	8.66	9.61	9.61

### Recap: Relevance judgments

Relevance can reasonably be thought of as a scale, with some documents highly relevant and others marginally so..

This decision is referred to as the *gold standard* or *ground truth* judgment of relevance.

### NDCG (Normalized DCG)

The ideal (perfect) ranking would first return the documents with the highest relevance level, then the next highest relevance level, etc

The ideal (perfect) ranking is based on the *gold* standard or *ground truth* judgment of relevance

### Summarize a Ranking: NDCG

# Normalized Discounted Cumulative Gain (NDCG) at rank *n*:

Normalize DCG at rank *n* by the DCG value at rank *n* of the ideal ranking

$$NDCG_{rank n} = \frac{Actual DCG_{rank n}}{Ideal DCG_{rank n}}$$

- Normalization is useful for contrasting queries with varying numbers of relevant results
- Popular in evaluating web search.

### Example

Docs	D1	D2	D3	D4	D5	D6	D 7	D8	D9	D10
Gain	3	2	3	0	0	1	2	2	3	0
Discounted gain	3									
DCG	3									

Perfect (ideal) ranking:

3, 3, 3, 2, 2, 2, 1, 0, 0, 0

ideal DCG values:



#### Perfect (ideal) ranking:

3, 3, 3, 2, 2, 2, 1, 0, 0, 0

ideal DCG values:

3, 6, 7.89, 8.89, 9.75, 10.52, 10.88, 10.88, 10.88, 10.88

Actual DCG(3, 2, 3, 0, 0, 1, 2, 2, 3, 0):

3, 5, 6.89, 6.89, 6.89, 7.28, 7.99, 8.66, 9.61, 9.61



Docs	D1	D2	D3	D4	D5	D6	D 7	D8	D9	D10
Gain	3	2	3	0	0	1	2	2	3	0
Discounted gain	3	2/1 =2	3/1.59 =1.89	0	0	1/2.59 =0.39	0. 71	0.67	0.95	0
DCG	3	5	6.89	6.89	6.89	7.28	7. 99	8.66	9.61	9.61

### Example

NDCG at rank 8?

$$NDCG_{rank n} = \frac{Actual DCG_{rank n}}{Ideal DCG_{rank n}}$$

NDCG values (divide actual by ideal): 1, 0.83, 0.87, 0.76, 0.71, 0.69, 0.73, 0.8, 0.88, 0.88

#### NDCG - Exercise

4 documents: d<sub>1</sub>, d<sub>2</sub>, d<sub>3</sub>, d<sub>4</sub>

	Ground Tr	ruth	Ranking F	unction <sub>1</sub>	Ranking Function <sub>2</sub>		
i	Document Order	r <sub>i</sub>	Document Order	$\mathbf{r_i}$	Document Order	r <sub>i</sub>	
1	d4	2	d3	2	d3	2	
2	<b>d3</b>	2	<b>d4</b>	2	d2	1	
3	d2	1	d2	1	<b>d4</b>	2	
4	<b>d</b> 1	0	d1	0	d1	0	

Please compare the NDCG at rank 4 for ranking function 1 and 2.



#### Solution

$$DCG_{GT} = 2 + \left(\frac{2}{\log_2 2} + \frac{1}{\log_2 3} + \frac{0}{\log_2 4}\right) = 4.6309$$

$$DCG_{RF1} = 2 + \left(\frac{2}{\log_2 2} + \frac{1}{\log_2 3} + \frac{0}{\log_2 4}\right) = 4.6309$$

$$DCG_{RF2} = 2 + \left(\frac{1}{\log_2 2} + \frac{2}{\log_2 3} + \frac{0}{\log_2 4}\right) = 4.2619$$

$$MaxDCG = DCG_{GT} = 4.6309$$

### Solution