

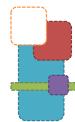
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(<u>nongaussian@hanyang.ac.kr</u>)

X Slides by Chris Manning and Pandu Nayak

Situation

- Thanks to your stellar performance in IR class, you quickly rise to VP of Search at internet retail giant 11st.com. Your boss brings in her/his nephew Sergey, who claims to have built a better search engine for 11st. Do you
 - Laugh derisively and send him to rival Gmarket?
 - Counsel Sergey to go to Stanford and take IR?
 - Try a few queries on his engine and say "Not bad"?
 - **-** ... ?

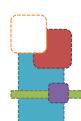


What could you ask Sergey?

- How fast does it index?
 - Number of documents/hour
 - Incremental indexing 11st adds 10K products/day
- How fast does it search?
 - Latency and CPU needs for 11st's 5 million products
- Does it recommend related products?
- This is all good, but it says nothing about the quality of Sergey's search
 - You want 11st's users to be happy with the search experience

How do you tell if users are happy?

- Search returns products relevant to users
 - How do you assess this at scale?
- Search results get clicked a lot
 - Problem: misleading titles/summaries can cause users to click
- Users buy after using the search engine
 - Or, users spend a lot of \$ after using the search engine
- Repeat visitors/buyers
 - Do users leave soon after searching?
 - Do they come back within a week/month/...?

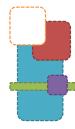


Happiness: elusive to measure

- Most common proxy: relevance of search results
 - But how do you measure relevance?
- Pioneered by Cyril Cleverdon in the Cranfield Experiments
 - http://en.wikipedia.org/wiki/Cranfield_Experiments





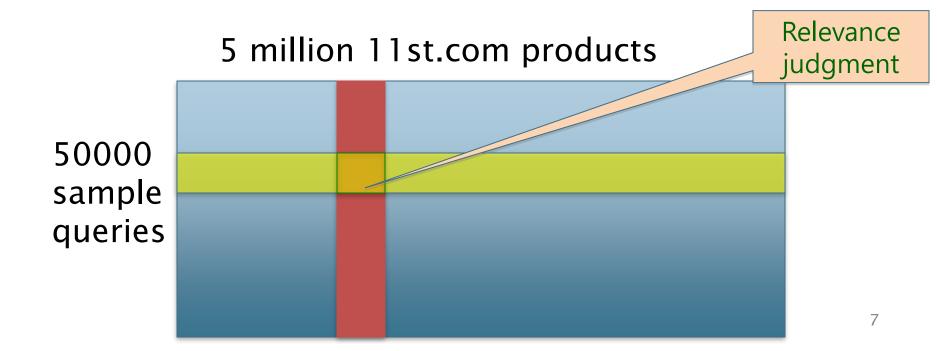


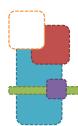
Measuring relevance

- Three elements:
 - A benchmark document collection
 - A benchmark suite of queries
 - An assessment of either <u>Relevant</u> or <u>Non-relevant</u> for each query and each document



- Benchmark documents 11st.com products
- Benchmark query suite more on this
- Judgments of document relevance for each query (e.g., rating score)





Relevance judgments

- Binary (relevant vs. non-relevant) in the simplest case, more nuanced (0, 1, 2, 3 ...) in others
- What are some issues already?
- 5 million times 50K takes us into the range of a quarter trillion judgments
 - If each judgment took a human 2.5 seconds, we 'd still need 10¹¹ seconds, or nearly \$300 million if you pay people \$10 per hour to assess
 - 10K new products per day

SMALL TALK: CROWDSOURCING

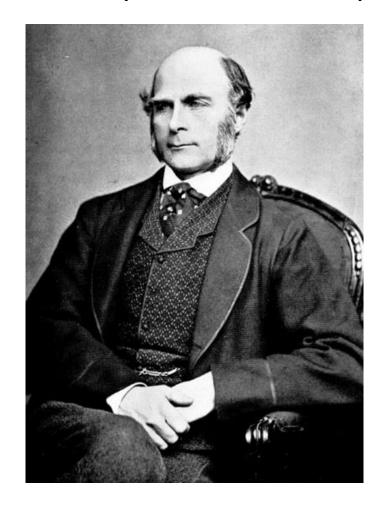


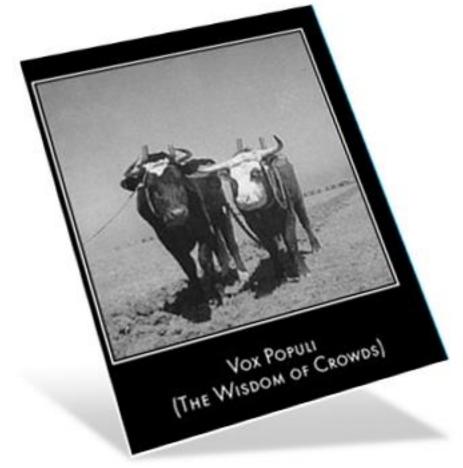
Crowd source relevance judgments?

- Present query-document pairs to lowcost labor on online crowd-sourcing platforms
 - Hope that this is cheaper than hiring qualified assessors
- Main takeaway you get some signal, but the variance in the resulting judgments is very high

Traditional Crowdsourcing

Weight-judging competition (Francis Galton, 1906): 1,197 (mean of 787 crowds) vs. 1,198 pounds (actual measurement)

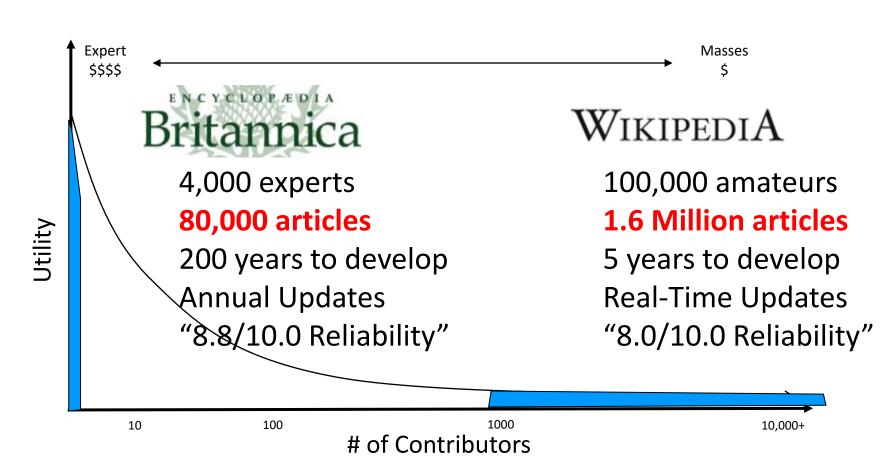


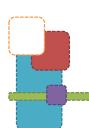


[Slide by Dongwon Lee, PennState Univ



Economics & Wikinomics





What is Crowdsourcing?

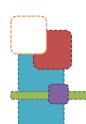
Online

- Crowd typically form into online communities based on the Web site
- The crowd submits solutions to the site or produce its contents

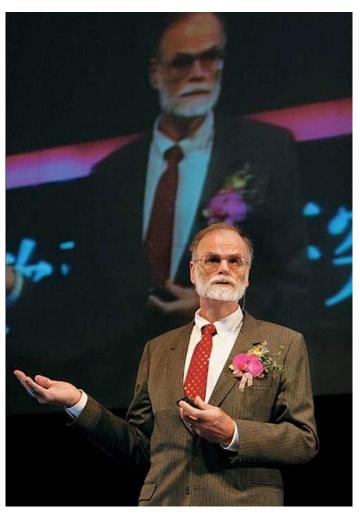
Distributed problem solving and production

A problem is often divided into many micro tasks

Answers from crowd are collected and merged together to derive the final solution



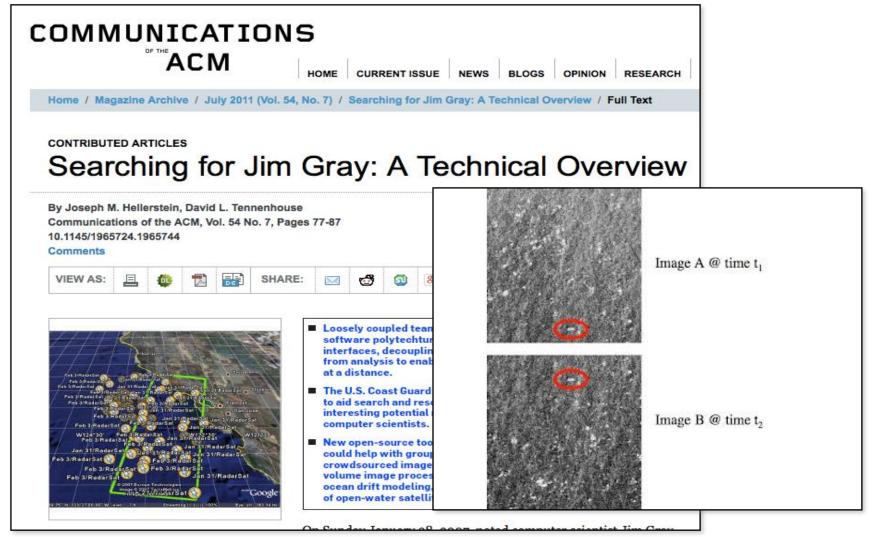
Eg, Finding "Jim Gray", 2007



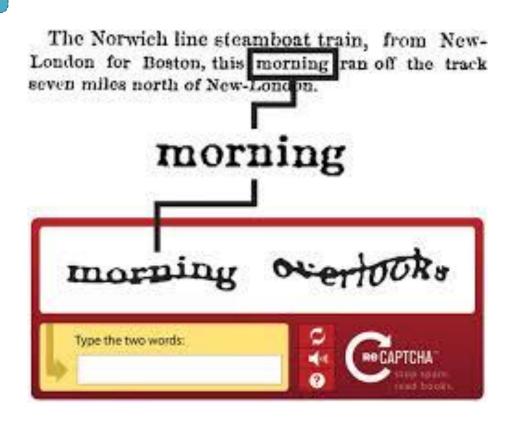
An American computer scientist who received the **Turing Award** in 1998 for seminal contributions to database and transaction processing research

On Sunday, January 28, 2007, during a short solo sailing trip near San Francisco, Gray and his 40-foot yacht, Tenacious, were reported **missing** by his wife

Eg, Finding "Jim Gray", 2007







As of 2012

Captcha: 200M every day

ReCaptcha: 750M to date



Eg, reCAPCHA

OCR Transcription

The Hreckinridge and Lane Democrats, having taken courage at the recent eastern advises, are boxxxxxxxxx energetically for the campaign: Several prominent Democrats who at first favored DonoLea, are coming out, for the other aide, apparently under Democracy of "lifornia", urging the party to support HaeeslipsIDas, has recently been published, which manifestly bss strengthened that aide of the [xxxxxxxxxx]: It is signed by 65 Democrats, many of whom occupy respectable and prominent positions in the party, 22 of them are Federal office-holders, [xxxxx] more are recipients of Federal patronage, and the others. represent a mass of politicians giving the document [xxxx] xxxxxxx mTheDcu8las Democrats are also active The Irish and German vote will mostly go with the branch of the party, but it is 17 IT newspapers have declared for DonGres, 13 for BaseSlaalDGS and 9 remain non-committal, with even chances of going either way. Under these circumstances the Republicans entertain not unjustifiable hopes that the Democratic divisions may be so equal,- ly balanced as to give the State [xx] LlaCOLV. Same very (xxxxxxxx) Bell and Everett meetings have been held in different parts of the State, bat thus far that party does not exhibit much rank sad ale air en.

reCAPTCHA Transcription

The Breckinridge and Lane Democrats, having taken courage at the recent eastern advices, are organizing energetically for the campaign. Several prominent Democrats who at first favored Douglas, are coming out for the other side, apparently under the pressure of Federal influence. An address to the National Democracy of California, urging the party to support Breckinridge has recently been published, which manifestly has strengthened that side of the question. It is signed by 65 Democrats, many of whom occupy respectable and prominent positions in the party, 22 of them are Federal office-holders, eight more are recipients of Federal patronage, and the others represent a mass of politicians giving the document most weight. The Douglas Democrats are also active The Irish and German vote will mostly go with that branch of the party, but it is difficult to estimate which wing is the stronger. Thus far 17 Democratic newspapers have declared for Douglas, 13 for Breckinridge and 9 remain non-committal, with even chances of going either way. Under these circumstances the Republicans entertain not unjustifiable hopes that the Democratic divisions may be so equally balanced as to give the State to Lincoln. Some very respectable Bell and Everett meetings have been held in different parts of the State, but thus far that party does not exhibit much rank and file strength.



Duolingo: Crowdsourcing new languages



utilizes the power of crowds to make learning a language free

GERMAN TEXT:

Falls Pakistans Geschichte ein Indikator ist, so könnte Musharrafs Entscheidung, das Kriegsrecht zu verhängen, jener sprichwörtliche Tropfen sein, der das Fass zum Überlaufen bringt.

PROFESSIONAL HUMAN

TRANSLATION (20 cents per word):

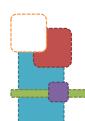
If Pakistan's history is any indicator, Musharraf's decision to impose martial law may prove to be the proverbial straw that breaks the camel's back.

GOOGLE TRANSLATE:

If Pakistan's history is any indicator, it could Musharraf's decision to impose martial law, be that proverbial straw that breaks the camel's back.

DUOLINGO:

If Pakistan's history is an indicator, Musharraf's decision to impose martial law could be the straw that breaks the camel's back.



Characteristics of Crowdsourcing

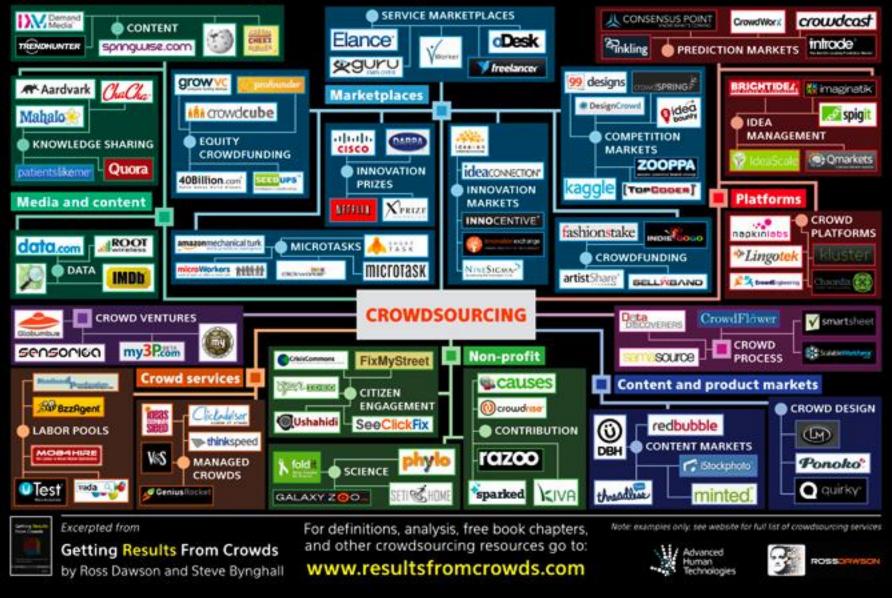
Benefits of Crowdsourcing

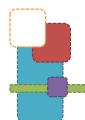
- Problems can be explored at comparatively little cost
- Payment is by results
- The organization can tap a wider range of talent than might be present in its own organization
- Turn customers into designers
- Turn customers into marketers

Problems with Crowdsourcing

- Quality
- Intellectual property leakage
- No time constraint
- Not much control over development or ultimate product
- III-will with own employees
- Choosing what to crowdsource
 & what to keep in-house

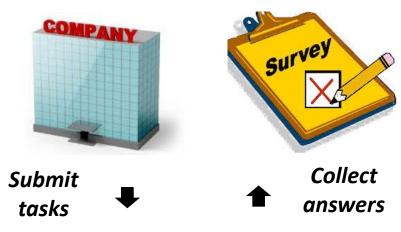
Crowdsourcing landscape Beta v2





Three Participants

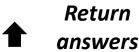
- Requesters
 - People submit some tasks
 - Pay rewards to workers
- Marketplaces
 - Provide crowds with tasks
- Crowds
 - Workers perform tasks



amazon mechanical turk™ Artificial Artificial Intelligence

Find tasks





CloudCrowd



AMT: mturk.com

Workers

- Register w. credit account (only US workers can register as of 2013)
- Bid to do tasks for earning money

Requesters

Status | Account S

Qualificatio

- First deposit money to account
- Post tasks
- Gather results
- Pay to workers if results are satisfactory

give businesses and developers acress to an on-demand, Sanable workforce.

Workers

from thousands of tasks and work wh

lechanical Turk is a marketplace t

Your Account

Introduction | Dashboar

200,645 HITs available. View the

Requesters

Make Money

by working on HITs

HITs - Human Intelligence Tasks - are individual tasks that you work on. Find HITs now.

As a Mechanical Turk Worker you:

- · Can work from home
- Choose your own work hours
- Get paid for doing good work



Get Results

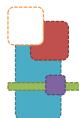
from Mechanical Turk Workers

Ask workers to complete HITs - Human Intelligence Tasks - and get results using Mechanical Turk. Register Now

As a Mechanical Turk Requester you:

- Have access to a global, on-demand, 24 x 7 workforce
- · Get thousands of HITs completed in minutes
- Pay only when you're satisfied with the results





AMT: HIT

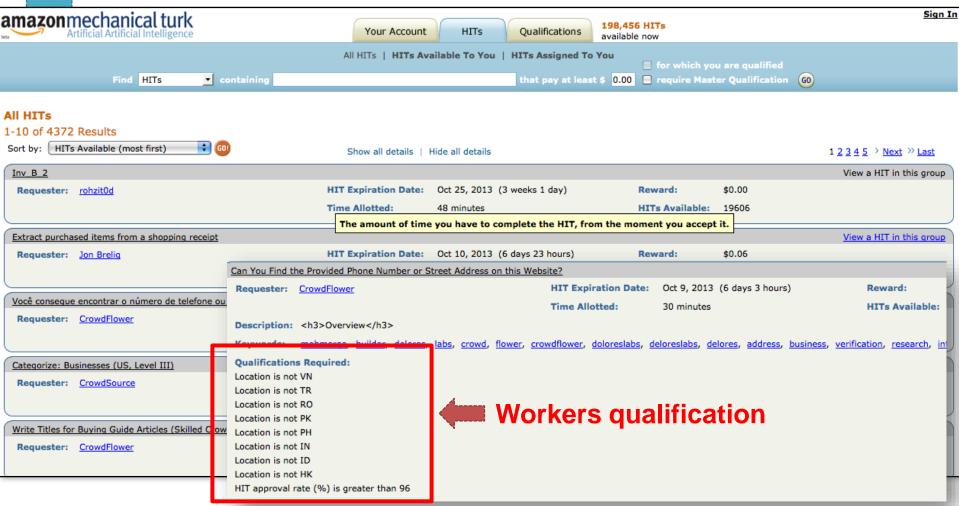
- Tasks
 - Called HIT (Human Intelligence Task)
 - Micro-task
- Eg
 - Data cleaning
 - Tagging/labeling
 - Sentiment analysis
 - Categorization
 - Surveying
 - Photo moderation
 - Transcription

Requester: Sergey Vasilyev	Reward: \$0.05 per HIT	HITs Available: 1	Duration: 15 minutes
Qualifications Required: HIT appr	oval rate (%) is not less than 75		
Franslate a text betwe	en the markers belo	ow from Englis	sh to Russian.
Human translation onl	vi Machine tranlatio	ne will be reid	acted
numan translation om	y: Maciline traillatio	ilis Will be reje	cteu.
====== FROM HER	E ======		
Hello!			
am test text message to be translat	ed from English to Russian.		
f you ask me, I was born in a mind o			
vho tests the MTurk API to start a ve	ry promising service later.		
TILL HERE			
		~~	
Any notes? Advices? E	motions? (Ontional)		
Ally Hotes: Advices: E	motions: (Optional)	,	
		//	

Translation task



AMT: HIT List



AMT: HIT Example



AMT: HIT Example

You will be asked to answer a series of questions based on identifying visual features from the bird image on the left. Closely follow the specific instructions for each question. Holding the mouse over each selectable option for 1 second will provide additional instructions or examples.

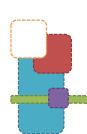


RETURN TO OUR TOPIC: QUALITY MEASURES



What else?

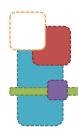
- Still need test queries (how to select them?)
 - Must be *germane* to docs available
 - Must be representative of actual user needs
 - Random query terms from the documents generally not a good idea
 - Sample from query logs if available
- Classically (non-Web)
 - Low query rates not enough query logs
 - Experts hand-craft "user needs"



Some public test Collections

TABLE 4.3 Common Test Corpora

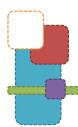
Collection	NDocs	NQrys	Size (MB)	Term/Doc	Q-D RelAss
ADI	82	35			
AIT	2109	14	2	400	>10,000
CACM	3204	64	2	24.5	
CISI	1460	112	2	46.5	
Cranfield	1400	225	2	53.1	
LISA	5872	35	3		
Medline	1033	30	1		
NPL	11,429	93	3		
OSHMED	34,8566	106	400	250	16,140
Reuters	21,578	672	28	131	
TREC	740,000	200	2000	89-3543	» 100,000



Now we have the basics of a benchmark

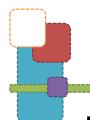
- Let's review some evaluation measures
 - Precision
 - Recall
 - DCG

— ...



Evaluating an IR system

- Note: user need is translated into a query
- Relevance is assessed relative to the user need, not the query
 - E.g., <u>Information need</u>: My swimming pool bottom is becoming black and needs to be cleaned.
 - Query: pool cleaner
- Assess whether the doc addresses the underlying need, not whether it has these words



Unranked retrieval evaluation: Precision and Recall

Binary assessments

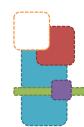
Precision: fraction of retrieved docs that are relevant = P(relevant|retrieved)

Recall: fraction of relevant docs that are retrieved

= P(retrieved|relevant)

	Relevant	Nonrelevant
Retrieved	tp	fp
Not Retrieved	fn	tn

• Recall
$$R = tp/(tp + fn)$$



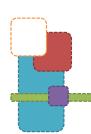
Rank-Based Measures

- Binary relevance
 - Precision@K (P@K)
 - Mean Average Precision (MAP)
 - Mean Reciprocal Rank (MRR)
- Multiple levels of relevance
 - Normalized Discounted Cumulative Gain (NDCG)

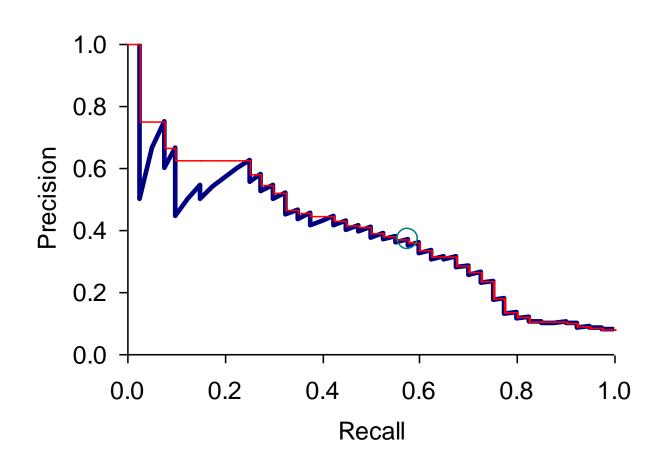
Precision@K

- Set a rank threshold K
- Compute % relevant in top K
- Ignores documents ranked lower than K
- Ex:
 - Prec@3 of 2/3
 - Prec@4 of 2/4
 - Prec@5 of 3/5
- In similar fashion we have Recall@K





A precision-recall curve



Mean Average Precision

- Consider rank position of each *relevant* doc
 - K₁, K₂, ... K_R
- Compute Precision@K for each K₁, K₂, ... K_R
- Average precision = average of P@K

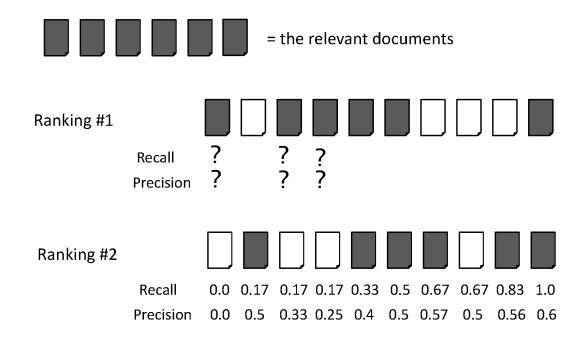
– Ex:

has AvgPrec of
$$\frac{1}{3} \cdot \left(\frac{1}{1} + \frac{2}{3} + \frac{3}{5}\right) \approx 0.76$$

MAP is Average Precision across multiple queries/rankings

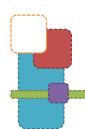


Average Precision



Ranking #1: (1.0 + 0.67 + 0.75 + 0.8 + 0.83 + 0.6)/6 = 0.78

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



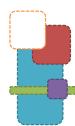
MAP

Recall 0.2 0.2 0.4 0.4 0.4 0.6 0.6 0.6 0.8 1.0 Precision 1.0 0.5 0.67 0.5 0.4 0.5 0.43 0.38 0.44 0.5

Recall 0.0 0.33 0.33 0.67 0.67 1.0 1.0 1.0 1.0 Precision 0.0 0.5 0.33 0.25 0.4 0.33 0.43 0.38 0.33 0.3

average precision query 1 = (1.0 + 0.67 + 0.5 + 0.44 + 0.5)/5 = 0.62average precision query 2 = (0.5 + 0.4 + 0.43)/3 = 0.44

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



Mean Average Precision

- If a relevant document never gets retrieved, we assume the precision corresponding to that relevant doc to be zero
- MAP is macro-averaging: each query counts equally
- Now perhaps most commonly used measure in research papers
- Good for web search?
- MAP assumes user is interested in finding many relevant documents for each query
- MAP requires many relevance judgments in text collection

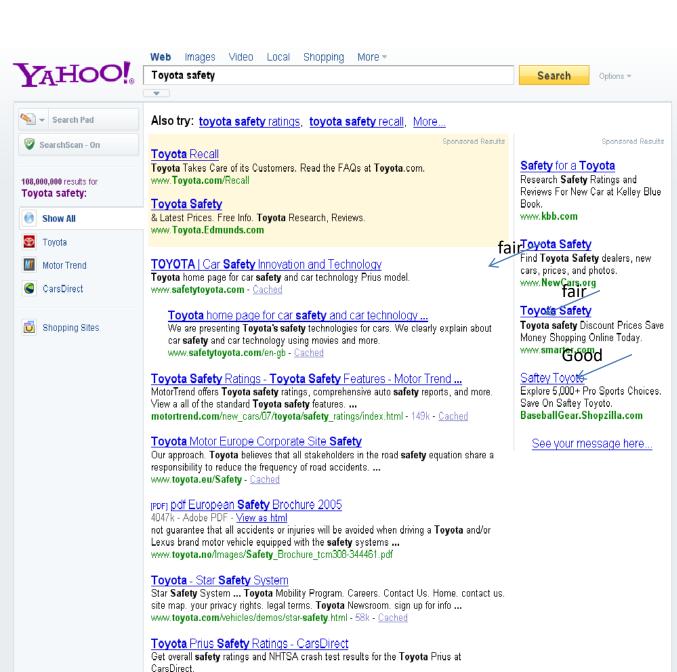


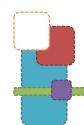
Mean Reciprocal Rank

- Consider rank position, K, of the first relevant doc
 - Could be only clicked doc
- Reciprocal Rank score = $\frac{1}{K}$
- MRR is the mean RR across multiple queries

$$MRR = ?$$

BEYOND BINARY RELEVANCE

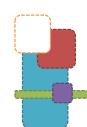




Discounted Cumulative Gain

 Popular measure for evaluating web search and related tasks

- 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



Discounted Cumulative Gain

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

Summarize a Ranking: DCG

- What if relevance judgments are in a scale of [0,r]?
 r>2
- Cumulative Gain (CG) at rank n
- The relevance value of the doc ranked at the 1st place
- Let the ratings of the n documents be r₁, r₂, ...r_n (in ranked order)
- $CG = r_1 + r_2 + ... r_n$
- Discounted Cumulative Gain (DCG) at rank n
 - $-DCG = r_1 + r_2/log_2 + r_3/log_2 + ... r_n/log_2 n$
 - We may use any base for the logarithm



Discounted Cumulative Gain

DCG is the total gain accumulated at a particular rank p:

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

• Alternative formulation:

$$DCG_p = \sum_{i=1}^{p} \frac{2^{rel_i} - 1}{\log(1+i)}$$

- used by some web search companies
- emphasis on retrieving highly relevant documents

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

10 ranked documents judged on 0-3 relevance scale:

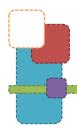
Discounted gain:

$$3, 2/1, 3/1.59, 0, 0, 1/2.59, 2/2.81, 2/3, 3/3.17, 0$$

= $3, 2, 1.89, 0, 0, 0.39, 0.71, 0.67, 0.95, 0$

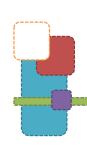
DCG:

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



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
 - The ideal ranking would first return the documents with the highest relevance level, then the next highest relevance level, etc
- Normalization useful for contrasting queries with varying numbers of relevant results
- NDCG is now quite popular in evaluating Web search



NDCG - Example

4 documents: d₁, d₂, d₃, d₄

i	Ground Truth		Ranking Function ₁		Ranking Function ₂	
	Document Order	r _i	Document Order	r _i	Document Order	r _i
1	d4	2	d3	2	d3	2
2	d3	2	d4	2	d2	1
3	d2	1	d2	1	d4	2
4	d1	0	d1	0	d1	0
	NDCG _{GT} =1.00		NDCG _{RF1} =1.00		NDCG _{RF2} =0.9203	

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

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

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

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



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 R N R N N N N N R R
System 2 N R N N R R R N N N

- What is the MAP of each system? Which has a higher MAP?
- Compute NDCG of each system
- Does this result intuitively make sense? What does it say about what is important in getting a good MAP score?



Human judgments are

- Expensive
- Inconsistent
 - Between raters
 - Over time
- Decay in value as documents/query mix evolves
- Not always representative of "real users"
 - Rating vis-à-vis query, vs underlying need
- So what alternatives do we have?

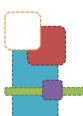
USING USER CLICKS



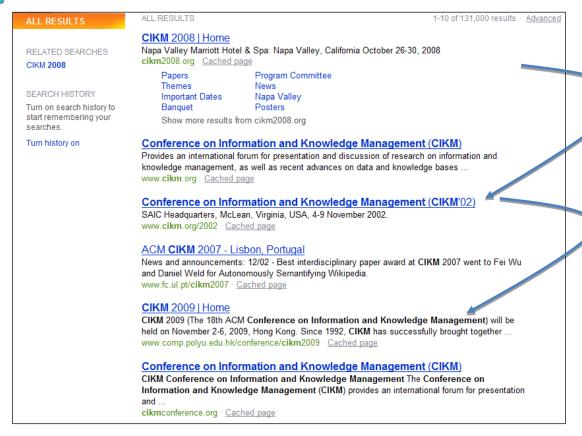
What do clicks tell us?



Strong position bias, so absolute click rates unreliable



Relative vs absolute ratings



User's click sequence

Hard to conclude <u>Result1 > Result3</u> Probably can conclude <u>Result3 > Result2</u>



Comparing two rankings via clicks (Joachims 2002) 두 Ranking을 섞어 놓고 어느 결과를 사람들이

클릭하는지 관찰!

Query: [support vector machines]

Ranking A

Kernel	machines
1/611161	111001111103

SVM-light

Lucent SVM demo

Royal Holl. SVM

SVM software

SVM tutorial

Ranking B

Kernel machines

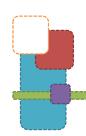
SVMs

Intro to SVMs

Archives of SVM

SVM-light

SVM software



Interleave the two rankings

This interleaving starts with B

Kernel machines

Kernel machines

SVMs

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Lucent SVM demo

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Royal Holl. SVM

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Remove duplicate results

Kernel machines

Kernel machines

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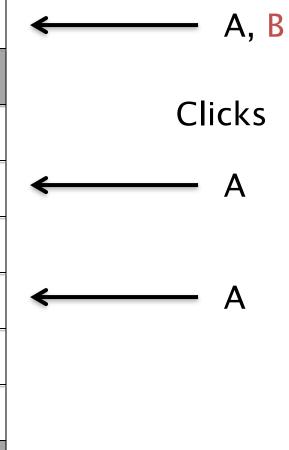
. .



Count user clicks

Ranking A: 3 Ranking B: 1

Kernel machines Kernel machines **SVMs SVM-light** Intro to SVMs Lucent SVM demo Archives of SVM Royal Holl. SVM **SVM-light**





Interleaved ranking

- Present interleaved ranking to users
 - Start randomly with ranking A or ranking B to evens out presentation bias
- Count clicks on results from A versus results from B
 Better ranking will (on average) get more clicks





A/B testing at web search engines

- Purpose: Test a single innovation
- Prerequisite: You have a large search engine up and running.
- Have most users use old system
- Divert a small proportion of traffic (e.g., 1%)
 to an experiment to evaluate an innovation
 - Interleaved experiment
 - Full page experiment



- Given a set of pairwise preferences (baseline) P
- We want to measure two rankings A and B
- Define a proximity measure between A and P
 - And likewise, between B and P
- Want to declare the ranking with better proximity to be the winner
- Proximity measure should reward agreements with P and penalize disagreements

Kendall tau distance

- Let X be the number of agreements between a ranking (say A) and P
- Let Y be the number of disagreements
- Then the Kendall tau distance between A and P is (X-Y)/(X+Y)
- Say P = $\{(1,2), (1,3), (1,4), (2,3), (2,4), (3,4)\}$ and A= $\{(1,3,2,4)\}$
- Then X=5, Y=1 ...
- (What are the minimum and maximum possible values of the Kendall tau distance?)

Example

Suppose we rank a group of five people by height and by weight:

Person	А	В	С	D	Е
Rank by Height	1	2	3	4	5
Rank by Weight	3	4	1	2	5

Assume that people answers that they prefer to:

$$(A > B), (B > D), (B > E), (C > D), (A > C), (A > E)$$

Do people like a tall one? Or heavy one?