On obtaining effort based judgments for Information retrieval

Manisha Verma, Emine Yilmaz, Nick Craswell UCL UCL Microsoft

Motivation

1. Log based evaluation

- Relevance determined implicitly or by clicks
- Large scale collection of user data
- Suffers from several (click, presentation etc) biases.

2. Batch evaluation

- Relevance labels assigned by trained judges
- Smaller test collections

Judging Interfaces

Instructions

Suppose you submitted the following query to a search engine and document below was shown as result.

Search query: what are clouds If page does not load please visit: http://www.weatherwizkids.com/weather-clouds.ht



Instructions

Suppose you submitted the following query to a search engine and two documents are shown as results. Please mark the document that you would prefer to see for the query

Search query: abraham lincoln



Structure Oriented Features

Relevance Prediction

- Simple assumptions about real information needs

These two forms of evaluation often do not completely agree with each other ([1] and [2]) They agree with each other only when there is a significant **gap** in quality of the systems compared ([3] and [4]).





- At present, *relevance is primary factor* for judging documents. It does not consider 'User effort' (Yilmaz et al. [5])
- A judge can spend a lot of time evaluating correctness of document for a given query. An **impatient user may not** spend as much time studying the document!.

Couds Rain & Floods Thunderstorms Lightning Wind Would you be satisf • Yes • No • Somewhat • Can not judge (skip How difficult was it f • Very easy • Easy • Somewhat difficult • Very difficult • Very difficult • Somewhat difficult	The droplets are so small and li How are cl All eir contains water, but near the invisible gas called water vapor. Wh Cool air can't hold as much water w condenses onto tiny pieces of dust ied (happy) with this search result? to understand the document?	ght that they can float in the air. ouds formed? a ground it is usually in the form of an an warm air rises, it expands and cools. apor as warm air, so some of the vapor that are floating in the air and forms a Is this document relevant to the query? Non Relevant Somewhat Relevant Relevant Highly Relevant Is the language easy to read? Very easy Easy Somewhat difficult Very difficult 2	 Would you be satisfied (happy) with this search result? Prefer left (I would like to see the left document in search results) Prefer right (I would like to see the right document in search results) Prefer none (I would not like to see these documents) Skip these documents (I cannot judge which document I would prefer to see) 			
 Very easy Easy Somewhat difficult Very difficult 						
		<u>Resu</u>	<u>ilts</u>			
Factors important for User Satisfaction			Effort and Preference Agreement			
FACTOR		p-Value	FACTOR	p-Value		
Fi	ndability ⁺	0.003	Findability	0.60*		
Re	Readability ⁻ 0.3		Readability	0.51		
Unde	Understandability ⁺ 0.0		Understandability	0.51		
Re	Relevance ⁺ 0		Relevance 0.72*			
		Featu	res	1		

User Model

1. When users first access the page, they *quickly scan it* to determine portions relevant to the query.

FINDABILITY

- 2. This is *followed by reading* these paragraphs/snippets. READABILITY
- 3. Finally, user *focusses on understanding* these nuggets of information.

UNDERSTANDABILITY

Contributions

- 1. Identify *factors that characterize* user effort.
- 2. Conduct experiments to *obtain explicit judgments* for these factors.

avgSumChar	Avg #chars in summary	fTable	Fraction of Tables
docCLI	CLI Index of document	maxWinPos	Max window pos with all query terms
docWords	#words in document	qWinO	Fraction of outlinks with query terms
qTermsInTitle	#query terms in title	fBoldItalics	Fraction of bold, italics and strong
sumWords	#words in summary	flmg	Fraction of images
tRatio	Fraction of #words and	minWinPos	Min window pos with all query terms
	#tags in html	countH	#Headings with query terms
<u>L</u>			

Findability Prediction

Text Features

FEATURE	p-val	FEATURE	p-val	FEATURE	p-val	FEATURE	p-val
fTable⁻	0.00	minWinPos+	0.00	tRatio⁻	0.01	qTermsInTitle ⁺	0.01
avoSumChar ⁻	0.01	1 meanPosOut	0.01				
argeanienai eie				countH⁻	0.02	aWinO ⁺	0.01
docCLI ⁻	0.02	sumWords ⁺	0.02				
	0.04	flmg ⁺	0.02	maxWinPos⁻	0.04	fBoldItalics ⁺	0.02
maxvvinPos				doo\\/ordo ⁻	0 01	flma ⁺	0.04
				uuuuvulus	0.04	IIIIg	0.04



3. Finally analyse the *effect of incorporating effort* into retrieval evaluation.

Methodology

- Collect effort based (explicit) judgments for each document for above parameters.
- Study user preferences
- Control for relevance: Collect user preferences with sideby-side comparison for documents of same relevance grade.



- 1. W. Hersh, A. Turpin, S. Price, B. Chan, D. Kramer, L. Sacherek, and D. Olson. Do batch and user evaluations give the same results? In Proc. SIGIR, Athens, Greece, 2000.
- 2. A. H. Turpin and W. Hersh. Why batch and user evaluations do not give the same results? In Proc. SIGIR, New Orleans, USA, 2001.
- 3. J. Allan, B. Carterette, and J. Lewis. When will information retrieval be "good enough"? In Proc. SIGIR, Salvador, Brazil, 2005.
- 4. A. Al-Maskari, M. Sanderson, P. Clough, and E. Airio. *The good and the bad system:* Does the test collection predict users' effectiveness? In Proc. SIGIR, 2008.
- 5. E. Yilmaz, M. Verma, N. Craswell, F. Radlinski, and P. Bailey. *Relevance and effort: An* analysis of document utility. In Proc. CIKM, 2014.