Here or There Preference Judgments for Relevance Carterette, Bennett, Chickering and Dumais

CS 6784 Advanced Machine Learning

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Motivation

How much would you pay for each of these cars?





Α.	\$30,000	C. \$90,000
Β.	\$45,000	D. \$100,000

Motivation

Now: For which car would you pay more?

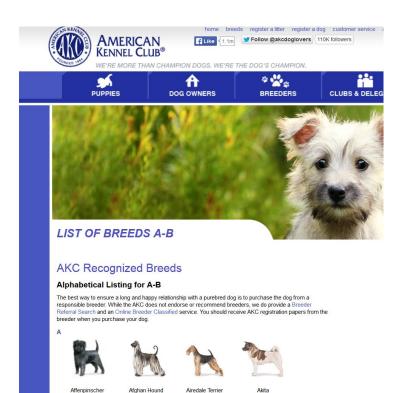




Here or There Preference Judgments for Relevance

Motivation

Query: Dog breeds



10



News

SD Legislature OKs bill against dog breed policies

| March 4, 2014 | Updated: March 4, 2014 3:54pm



PIERRE, S.D. (AP) – The South Dakota House has passed a measure to keep local governments from setting policies targeting specific dog breeds.



Motivation

- How are search results evaluated?
 - Binary (relevant/non-relevant)
 - On a numeric scale (highly relevant less relevant)
- Problems:



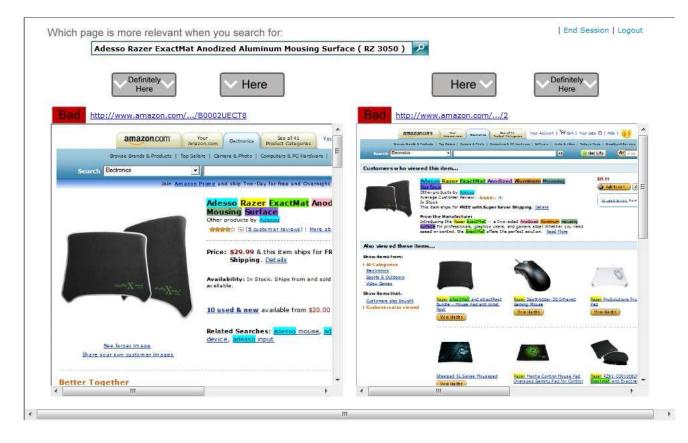
- Compared three types of judgments:
 - Absolute



Absolute

	hen you search for:		
Adesso Razer Exac	tMat Anodized Aluminum Mousing Sur	face (RZ 3050) 💋	
	Fair Good	Excellent	
	Bad http://www.amazon.com	n//B0002UECT8	
	amazon.com	Your Amazon.com Bectronics Product Categories You	
	Search Bectronics	Top Sellers Camera & Photo Computers & PC Hardware	
	Join Ar	mazon Prime and ship Two-Day for free and Overnight	
		Adesso Razer ExactMat Anod Mousing Surface Other products by Adesso	
		Carter and the second s	
		Price: \$29.99 & this item ships for FR Shipping. <u>Details</u>	
	in Xan	Availability: In Stock. Ships from and sold available.	
		10 used & new available from \$20.00	
	See jarger im age	Related Searches: adesso mouse, ad device, adesso input	
	<u>See larger image</u> Share vour own ouatomer images		
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Graded preferences

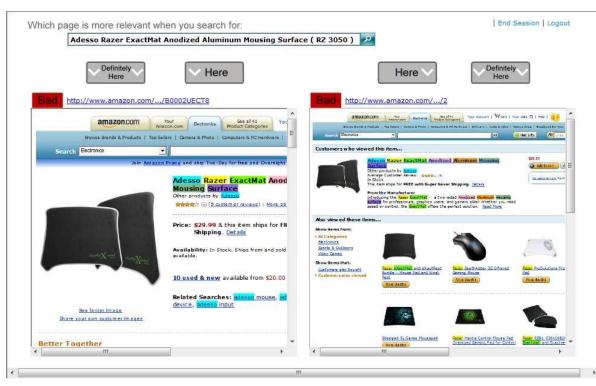


- Experimental set-up:
 - 51 queries, each ~12 results
 - Each <result, result> pair was rated for preferences

Judgments

- 6 assessors, all Microsoft employees
- Guidelines for absolute five-point scale
- Pages judged as "bad" removed from results

- Transitivity:
 - Holds 99 % on average
 - But ...



Low agreement for absolute judgements

	Bad	Fair	Good	Excellent	Perfect
Bad	0.579	0.29	0.118	0.014	0.000
Fair	0.208	0.332	0.309	0.147	0.003
Good	0.095	0.348	0.286	0.26	0.011
Excellent	0.011	0.167	0.264	0.535	0.022
Perfect	0.000	0.042	0.125	0.25	0.583

Explicit preferences better than inferred from absolute statements

	A < B	A,B bad	A > B	Total		
A < B	0.752 0.033		0.215	2580		
A,B bad	0.208	0.208 0.567 0		413		
A > B 0.201 C		0.034	0.765	2757		
(a) explicit preferences						
	A < B	A,B bad	A > B	Total		
A < B	0.657	0.051	0.292	2530		
A,B bad	0.297	0.38	0.323	437		
A > B	0.278	0.053	0.669	2654		

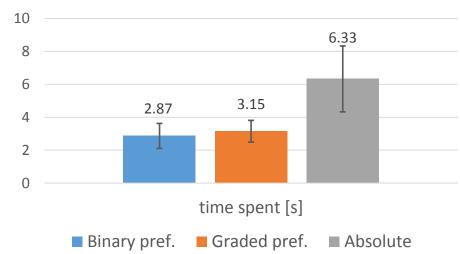
(b) inferred preferences

Graded preferences vs. binary preferences

		A < B	A,B bad	А	> B	Total		
A < B		0.752	0.033	0.2	215	2580		
A,B bad		0.208	0.567	0.2	225	413		
A > B		0.201	0.034	0.765		2757		
(a) preferences								
	A << B	A < B	A, B bad	A > B	A >> B	Total		
A << B	0.247	0.621	0.000	0.132	0.000	219		
A < B	0.059	0.661	0.043	0.221	0.015	2288		
A, B bad	0.000	0.244	0.453	0.300	0.002	406		
A > B	0.012	0.212	0.051	0.670	0.055	2389		
A >> B	0.000	0.180	0.005	0.680	0.134	194		

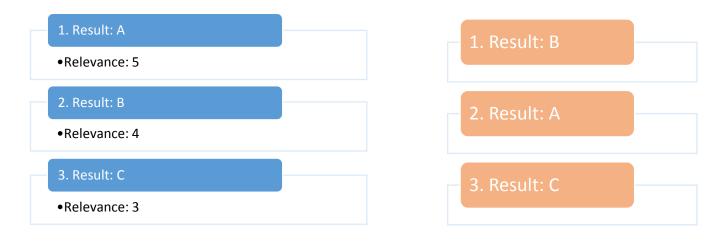
(b) graded preferences

Making preference statements is faster than making absolute statements



Median seconds per judgment

- Want to evaluate entire queries
- Traditionally: graded relevance

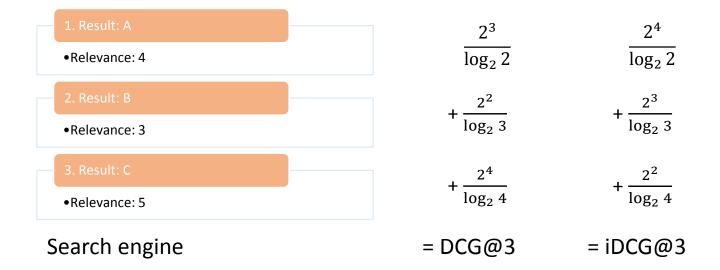


Gold standard from assessors

DCG (discounted cumulative gain)

• DCG@k =
$$\sum_{i=1}^{k} \frac{2^{rel_i-1}}{\log_2 i+1}$$

• nDCG@k = $\frac{DCG@k}{iDCG@k}$

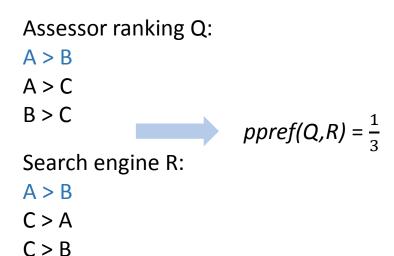


Activity: How would you evaluate preferences?

Assessor preferences: A > B	1. Result: A
A > C B > C	2. Result: B
	3. Result: C

Search engine: A > B C > A C > B

- Simple idea: proportion of correctly ranked pairs
 - Named *ppref*



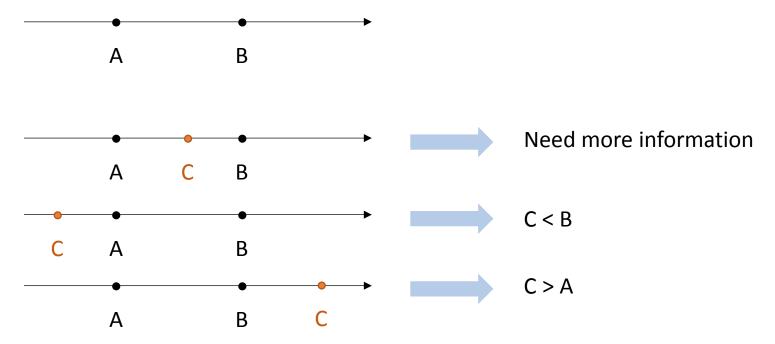
They show similar behaviour

	NDCG	ppref	wpref
DCG	1.00	0.873	0.866
NDCG		0.873	0.866
ppref			0.940

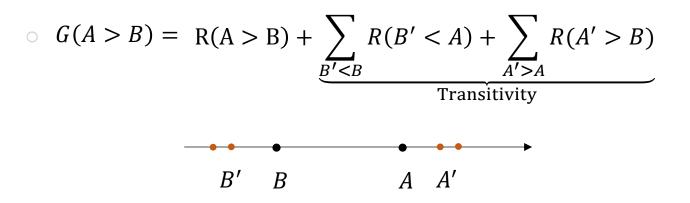
Agreement on system differences.

- Task: Compare two search engines E₁ and E₂
- Problem: O(n²) pairs for n results
- Solutions:
 - Use transitivity (given > 98 % of the time)
 - Eliminate "bad" judgments
 - Evaluate pairs (i, j) with high utility first

- Estimating utility:
 - If A > B in both rankings, value for *ppref* is the same
 - Make use of transitivity:

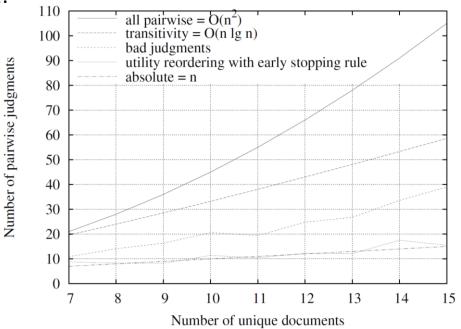


- Gain of a preference pair:
 - Let R(A > B) the subgain we get by just knowing A > B



- Expected utility of a pair (A,B):
 - U(A,B) = p(A > B)G(A > B) + p(B > A)G(B > A)
 - $\circ p(.)$ is assumed to be uniform

- Then:
 - Choose pair (A,B) with maximum utility at each step
 - Stop when remaining pairs cannot change overall result
- Performance:



Conclusions

Preferences

- Are faster and easier for humans to state
- Cause lower disagreement rates
- Graded preferences don't add value
- Suitable algorithms (e.g., RankSVM)

Issues

- No guarantees for judging heuristics
- Full evaluation if we just want to compare two search engines?