## **Modeling User Behavior and Interactions** Lecture 1: Modeling Searcher Behavior **Eugene Agichtein Emory University Eugene Agichtein RuSSIR 2009: Modeling User Behavior and Interactions Emory University**

#### **Overview of the Course**

- Lecture 1: Modeling searcher behavior
- Lecture 2: Interpreting behavior  $\rightarrow$  relevance
- Lecture 3: Using behavior data  $\rightarrow$  ranking
- Lecture 4: Personalizing search with behavior
- Lecture 5: Search user interfaces

#### **Lecture 1: Models of Search Behavior**

- Understanding user behavior at micro-, meso-, and macro- levels
- Theoretical models of information seeking
- Web search behavior:
  - Levels of detail
  - Search Intent
  - Variations in web searcher behavior
  - Click models



#### Levels of Understanding User Behavior [Daniel M. Russell, 2007]

 Micro (eye tracking): lowest level of detail, milliseconds

 Meso (field studies): mid-level, minutes to days Web lunn David Nova Freeds COORD Set David of Control 2005 500 2005 500 2005 Web Set David of Control 2005 500 2005 500 2005 Web Set David of Control 2005 500 2005 500 2005 Participation of Control 2005



 Macro (session analysis): millions of observations, days to months

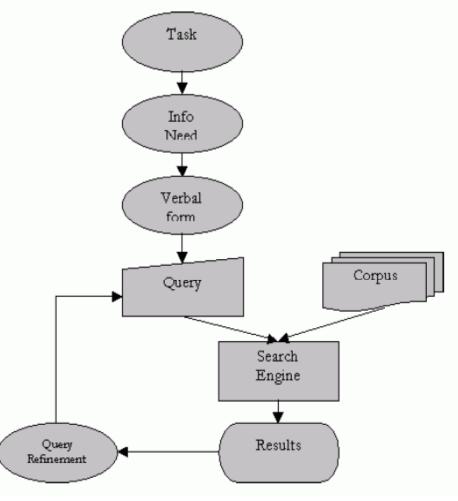




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#### **Models of Information Seeking**

- "Information-seeking ... includes recognizing ... the information problem, establishing a plan of search, conducting the search, evaluating the results, and ... iterating through the process."-Marchionini, 1989
  - Query formulation
  - Action (query)
  - Review results
  - Refine query



Adapted from: M. Hearst, SUI, 2009



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#### **Key Concept: Relevance**

- Intuitively well understood
  - same perception globally "y'know"
  - a "to" and context always present
- Relevance:
  - a relation between objects P & Q along property R
  - may also include a measure S of the strength of connection
- Example: topical relevance (document on the correct topic)



#### **Relevance clues**

- What makes information or information objects relevant? What do people look for in order to infer relevance?
  - Topicality (subject relevance)
  - Extrinsic (task-, goal- specific)
- Information Science "clues research":
  - uncover and classify attributes or criteria used for making relevance inferences



#### **IR Relevance Models**

- All IR and information seeking models have relevance at their base
- Traditional IR model has most simplified (topic) version of relevance (topical)

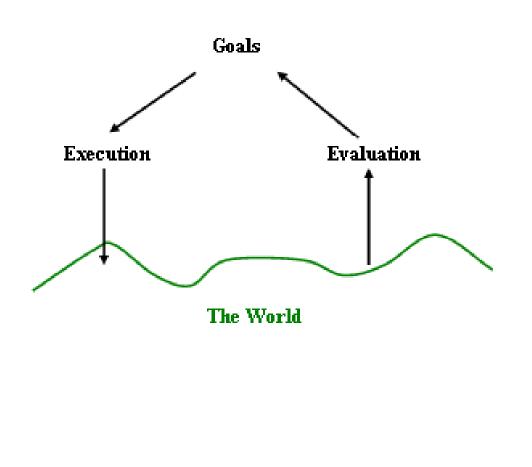
- Enough to make progress

- Variety of integrative models have been proposed
  - more complex models = increased challenge to evaluation and implementation in practice



#### **Cognitive Model of Information Seeking**

- Static Info Need
  - Goal
  - Execution
  - Evaluation





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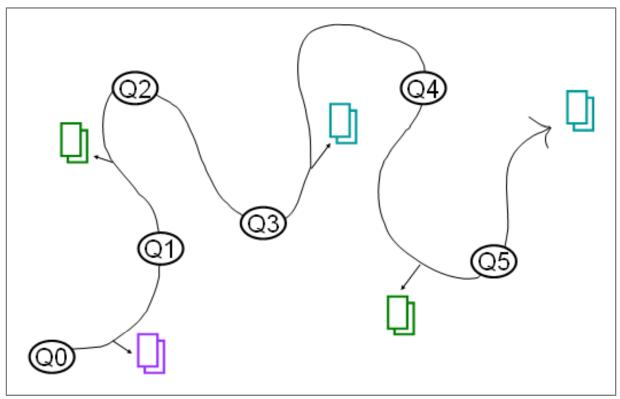
#### **Relevance dynamics**

- Do relevance inferences and criteria change over time for the same user and task, and if so, how?
- As user progresses through stages of a task:
  - the user's cognitive state changes
  - the task changes as well



## **Dynamic "Berry Picking" Model**

#### Information needs change during interactions



**[Bates, 1989]** M.J. Bates. The design of browsing and berrypicking techniques for the online search interface. *Online Review*, 13(5):407–431, 1989.



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#### **Information Foraging Theory**

# **Goal:** maximize rate of information gain.

Patches of information →websites Basic Problem: should I

#### continue in the current patch or look for another patch?

**Expected gain** from **continuing** in current patch, **how long** to continue searching in that patch







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01	Paris La Defense / La Defense Ce Hotel Info: 800-246-8357	entre ** * * * ± Show Amenities & Rates	From \$113.00 LOWEST AVG. MONTLY MATE LOW DOICY SELECT
2	Hotel Lutetia ***** Left Bank, St-Germain-Des-Pr +s / Paris Hotel Info: 800-246-8357	± Show Amenilies & Rates	from \$225.00 Louiss ava. mainty mare Louis avar
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-4	Hotel De Vendome **** ConcordeMadeleine / Paris Hotel Info: 800-246-8357	2 Show Amenities & Rates	From \$438.00 LOWEST AVE. MUMILE MATE COMPARE BUILDING
-5	Le Meurice ***** City Center / Paris Hotel Info: 800-246-8357	± Show Amenilies & Rates	From \$623.00 LOWEST AVG. MIGHTLY RATE LOW DESCE
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70	L'Hotel De Sers **** Next To The Champs Elysees / Paris Hotel Info: 800-246-8357	± Show Amenilies & Rates	From \$368.00 LOWEST AVS. MIGHTLY RATE SOURCENTER SELECT
8 IMAGE COMING SOON	Jolly Hotel Lotti ***** On A Right Bank Street / Paris Hotel Info: 800-246-8357	* Show Amenities & Rates	From \$229.00 LOWEST AVS. HOMELY PART COM PRECE COMPARTS
19	Hotel De Crillon **** Tuileries Gardens / Paris Hotel Info: 600-246-8357	± Show Amenities & Rates	COMEST AUG. HORITLY RATE COMEST AUG. HORITLY RATE COMESTIC

#### otel Search

bal: Find eapest 4-star tel in Paris.

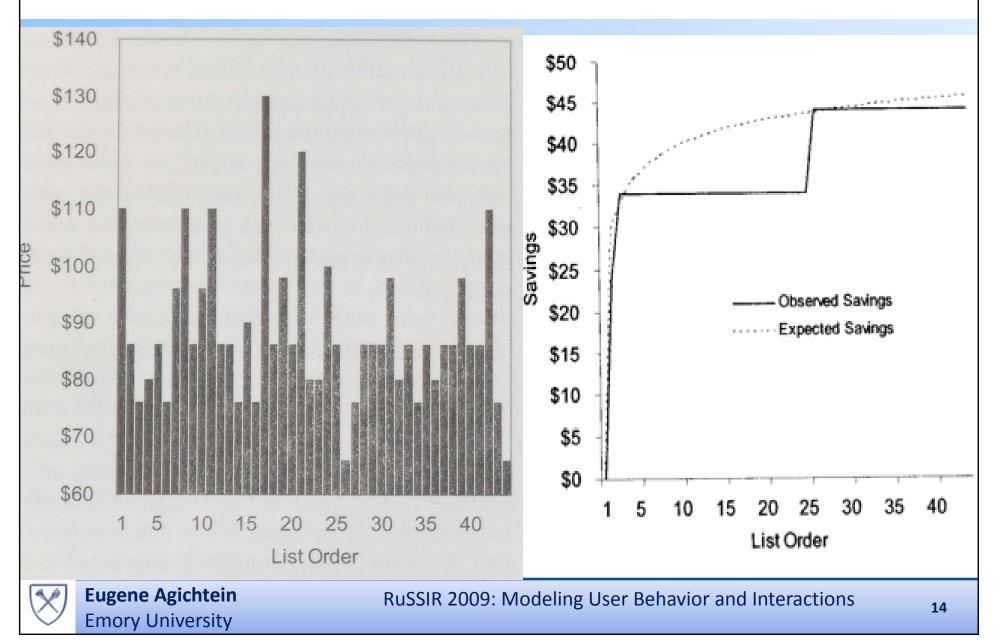
ep 1: pick hotel arch site

ep 2: scan list

#### ep 3: goto 1

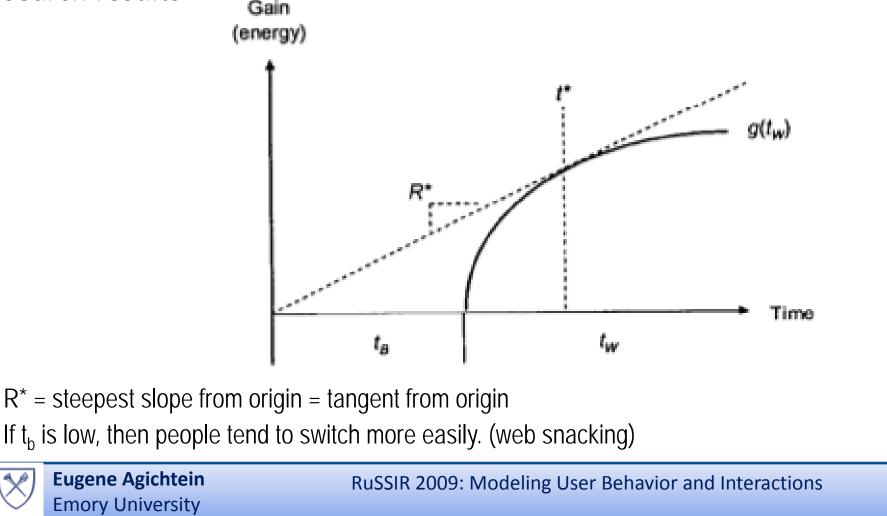
vior and Interactions

#### Example: Hotel Search (cont'd)



### -Charnov's Marginal Value Theorem

Diminishing Returns Curve; 80% of users don't scan past the 3<sup>rd</sup> page of search results



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#### **Browsing vs. Search**

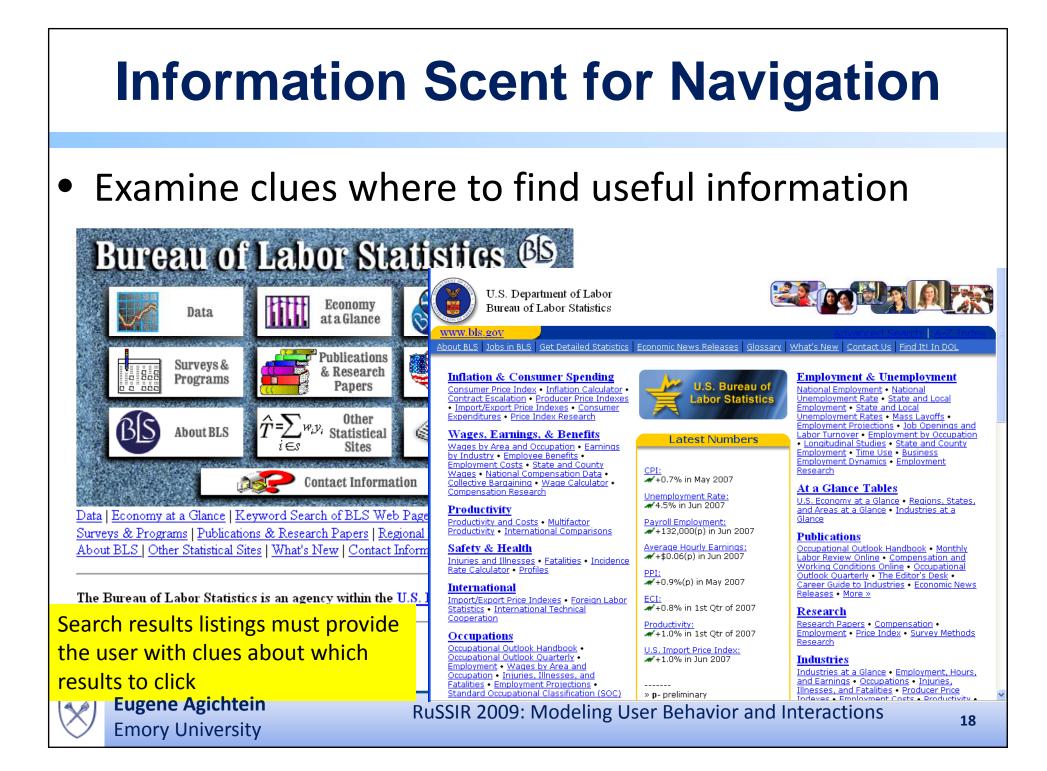
- Recognition over recall (I know it when I see it)
- Browsing hierarchies/facets more effective than querying



#### Orienteering

- Searcher issues a quick, imprecise to get to approximately the right information space region
- Searchers follow known paths that require small steps that move them closer to their goal
- Expert searchers starting to issue longer queries





#### **Summary of Models**

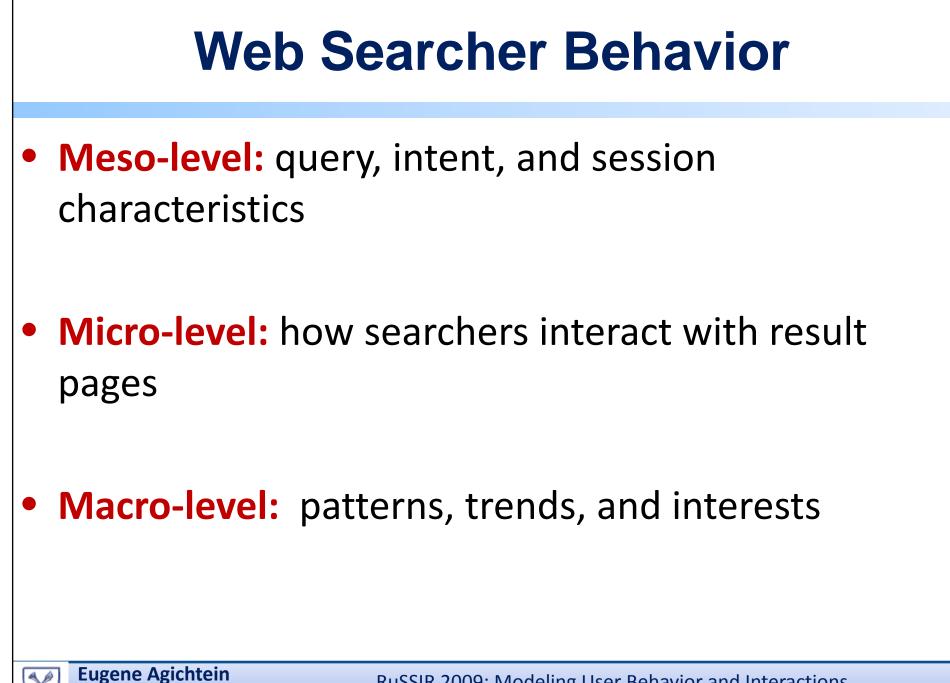
- Many cognitive models proposed
- Classical IR Systems research mainly uses the simplest form of relevance (topicality)
- Open questions:
  - How people recognize other kinds of relevance
  - How to incorporating other forms of relevance (e.g., user goals/needs/tasks) into IR systems

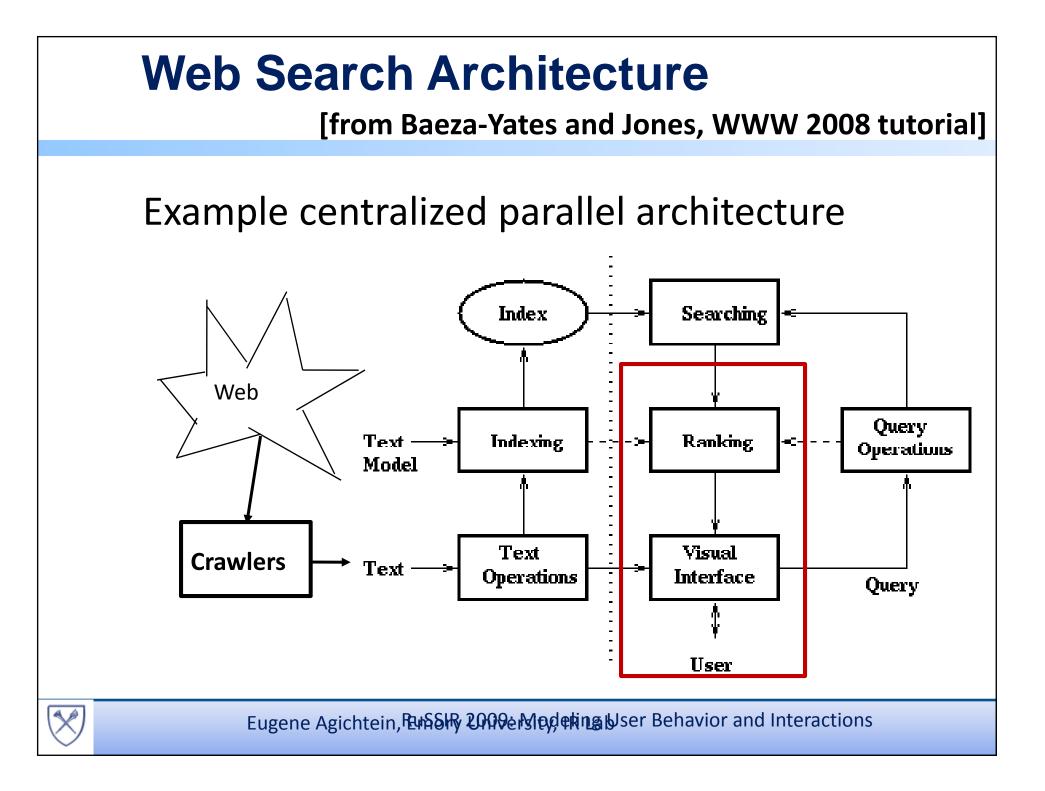


#### **Lecture 1: Models of Search Behavior**

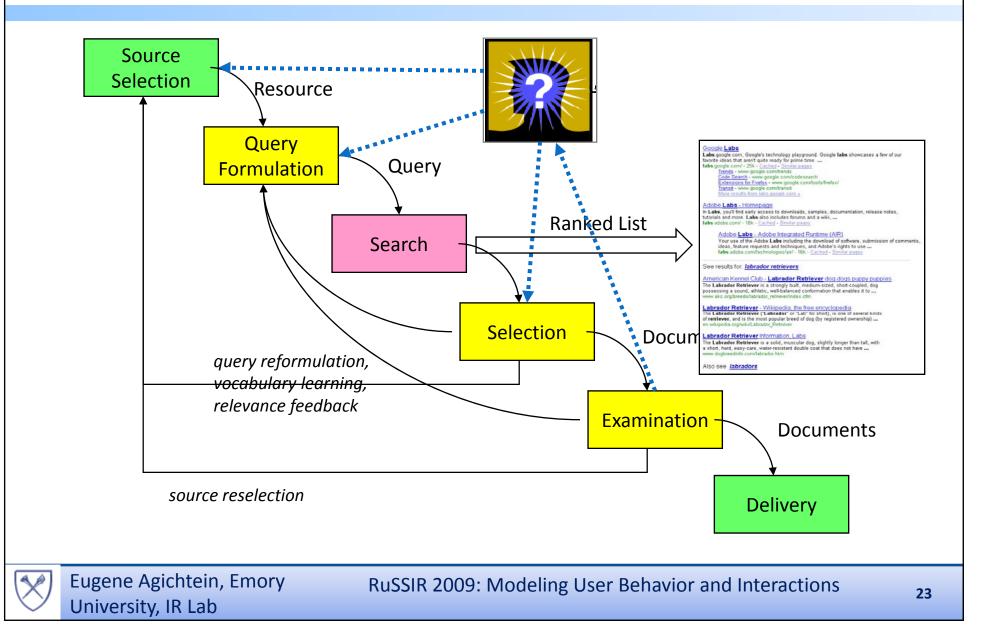
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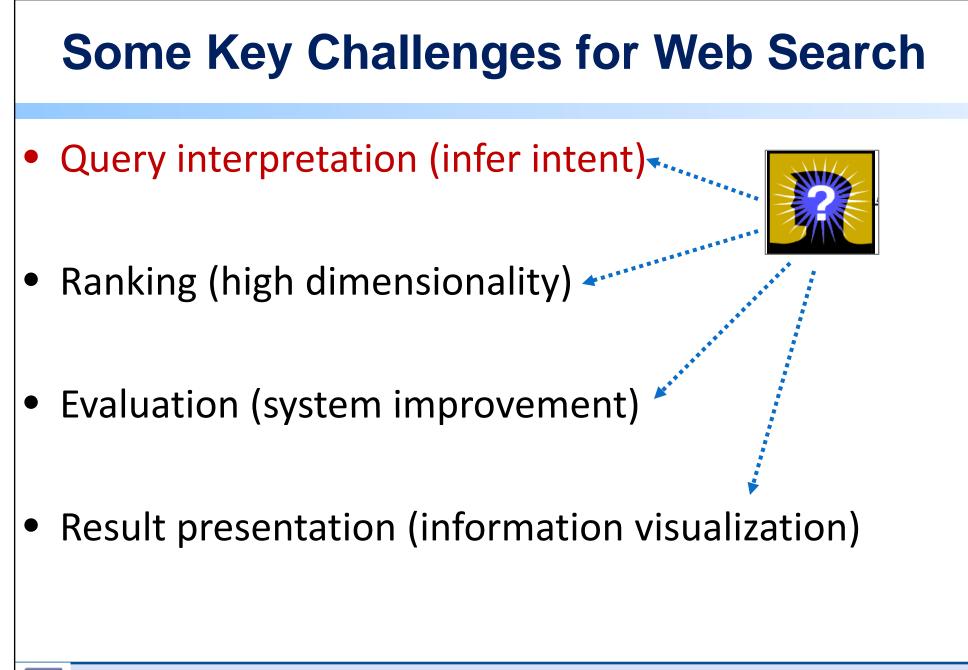


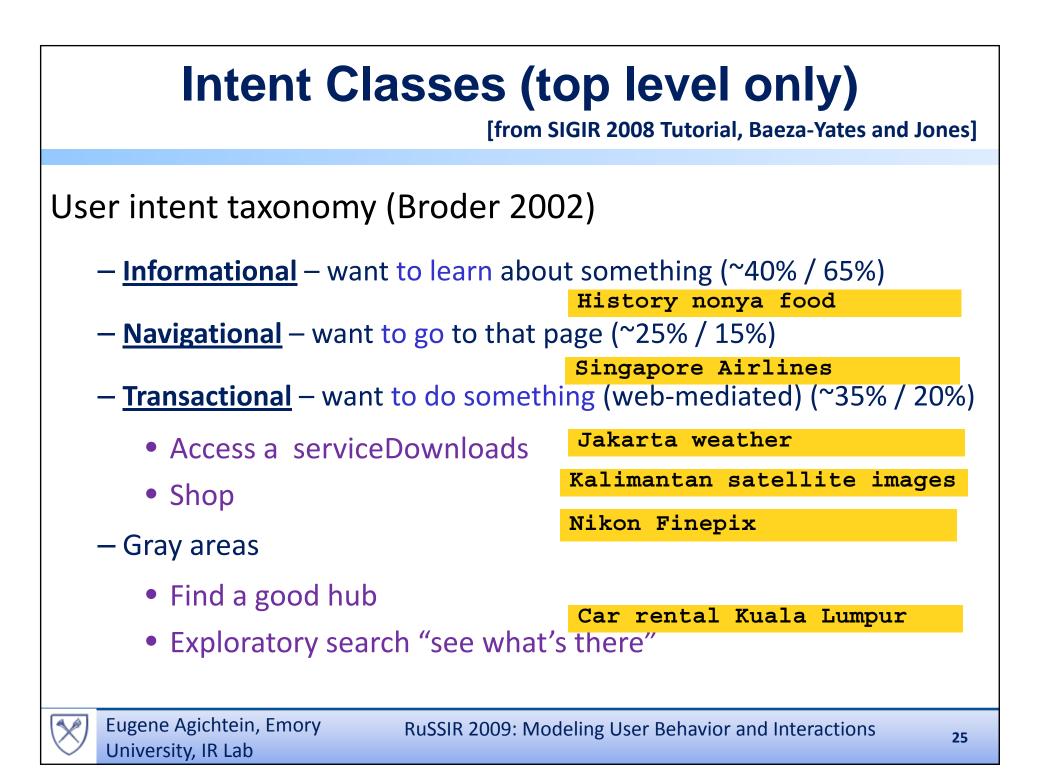




#### Information Retrieval Process (User view)



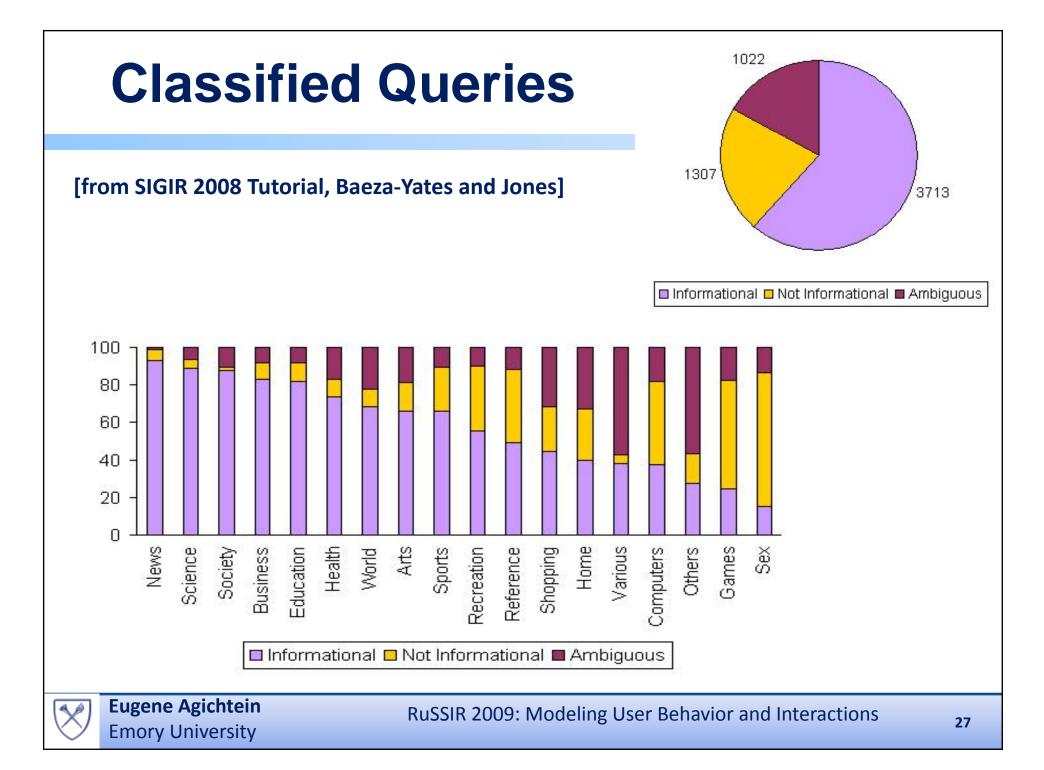




#### **Web Search Queries**

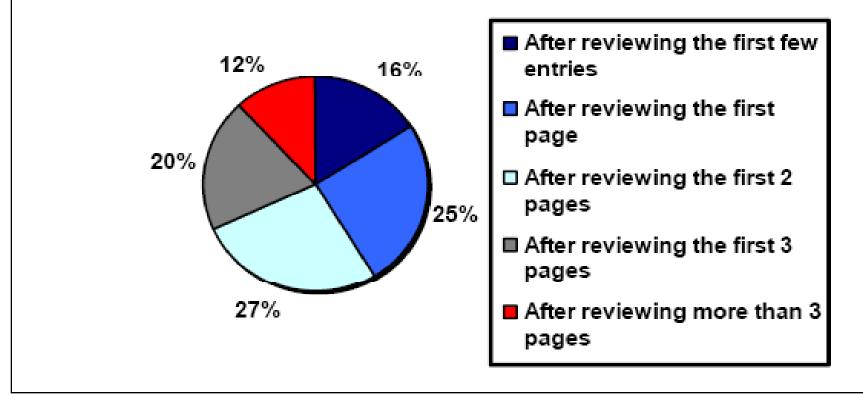
- Cultural and educational diversity
- Short queries and impatient interaction
  - Few queries posed and few answers seen (first page)
  - Reformulation common
- Smaller and different vocabulary
  - Not "expert" searchers!
  - "Which box do I type in?"





#### **People Look at Only a Few Results**

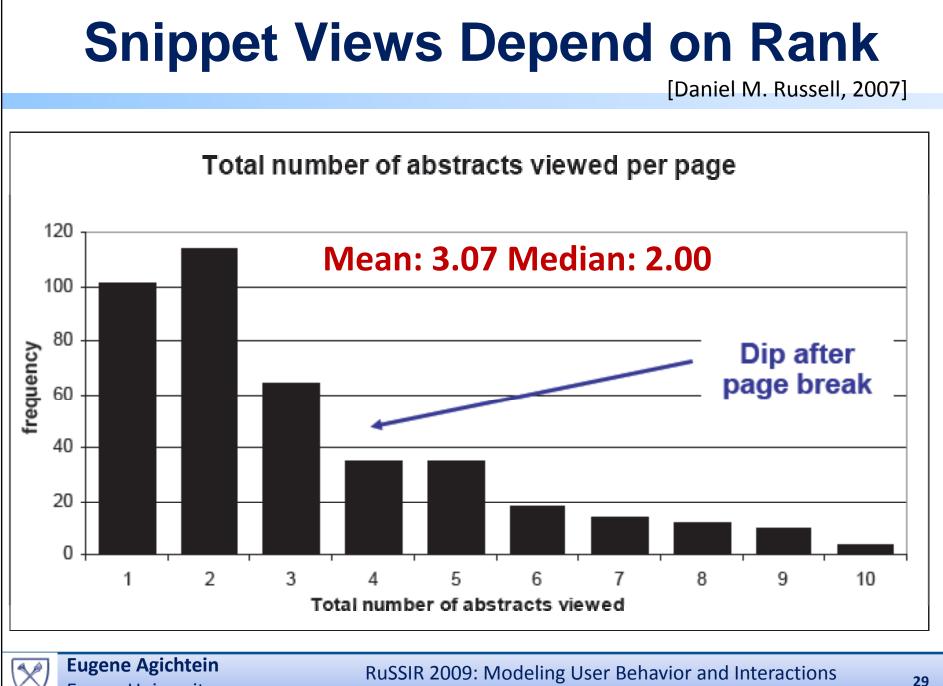
"When you perform a search on a search engine and don't find what you are looking for, at what point do you typically either revise your search, or move on to another search engine? (Select one)"



(Source: <a href="mailto:iprospect.com">iprospect.com</a> WhitePaper\_2006\_SearchEngineUserBehavior.pdf)

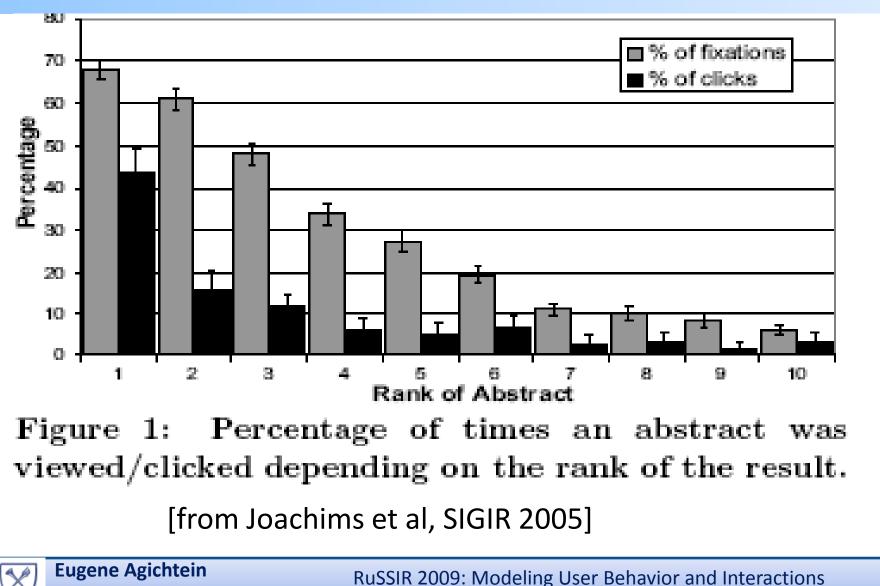


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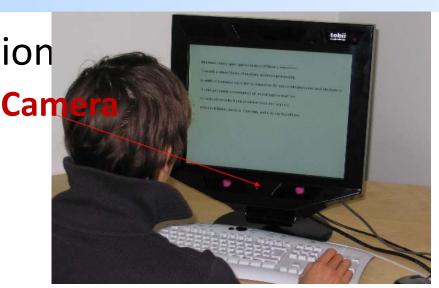
#### **Snippet Views and Clicks Depend on Rank**



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#### "Eyes are a Window to the Soul"

- Eye tracking gives information about search interests: Can
  - Eye position
  - Pupil diameter
  - Seekads and fixations





Eugene Agichtein, Emory University, IR Lab

#### **Micro-level: Examining Results**

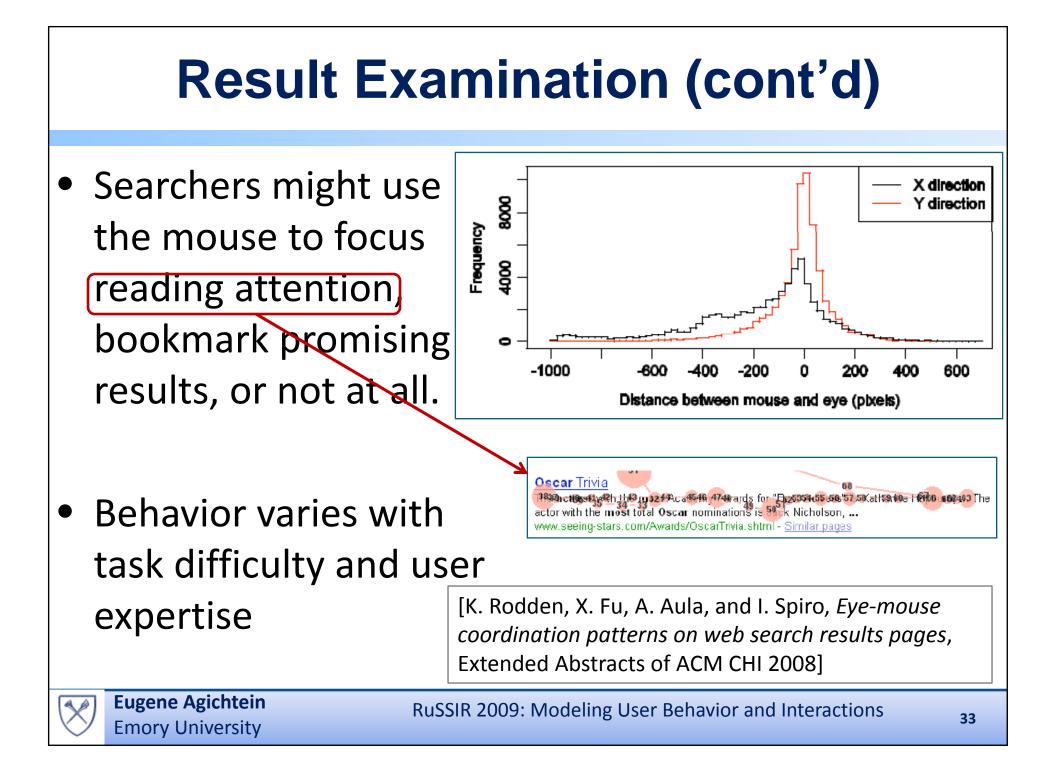
[Daniel M. Russell, 2007]

- Users rapidly scan the sea
- What they see in lower si judgment of higher result
- Spend most time scrutini
  - Trust the ranking





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## Macro-Level (Session) Analysis

- Can examine theoretical user models in light of empirical data:
  - Orienteering?
  - Foraging?
  - Multi-tasking?
- Search is often a multi-step process:
  - Find or navigate to a good site ("orienteering")
  - Browse for the answer there: [actor most oscars] vs. [oscars]
- Teleporting
  - "I wouldn't use Google for this, I would just go to..."
- Triangulation
  - Draw information from multiple sources and interpolate
  - Example: "how long can you last without food?"



#### **Users (sometimes) Multi-task**

[Daniel M. Russell, 2007]

100: Google Search [free roulette] (4s) (DUPE) (p=78)

- 102: Google Result 7 www.getlyrical.com/general/free\_casino\_games/free\_online\_roulette.html (3s)
- 103: Google Result 7 www.getlyrical.com/general/free\_casino\_games/free\_online\_roulette.html (19s) (DUPE) (p=100)

106: Google Result 8 www.saliu.com/Roulette.htm (56s) (p=100)

112: Google Search [shockwave] (4s)

114: Google Result 3 www.shockwave.com/sw/home/ (10s)

117: Google Result 5 sdc.shockwave.com/shockwave/download/download.cgi (16s) (p=112)

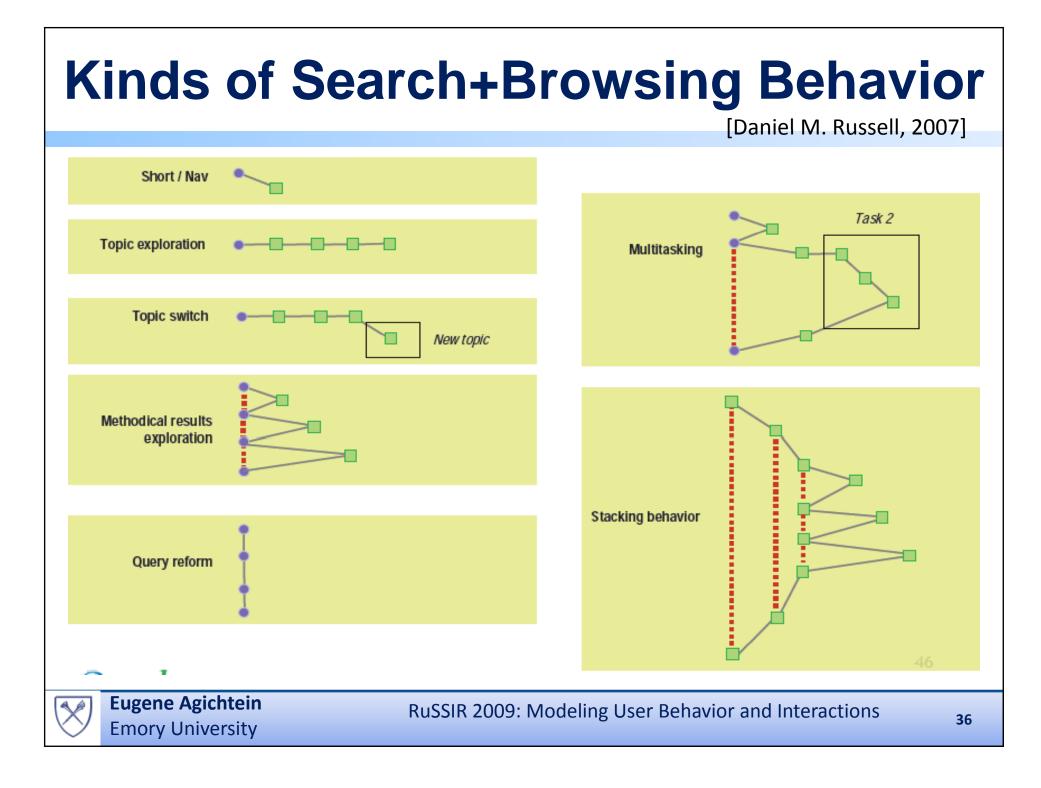
120: Google Search [free roulette] (3s) (DUPE) (p=78)

- 122: Google Result 1 www.ildado.com/free\_roulette.html (15s) (DUPE)
- 124: Google Search [free proffessional roulette] (2s)
- 126: Google Search (spell correct) [free professional roulette] (10s)
- 128: Google Result 3 imagesculptor.com/Roulette/free-roulette-professional-system.php (5s)
- 129: Google Result 3 imagesculptor.com/Roulette/free-roulette-professional-system.php (8s) (DUPE) (p=126)

133: Google Result 7 www.amazon.com/exec/obidos/tg/detail/-/B0007XRSQ4?v=glance (2s) (p=126)



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#### Variance in Behavior between Novice and Expert Searchers [White & Morris, 2007]

- Some people are more expert at searching than others
  - Search expertise, not domain expertise
- Find characteristics of these "advanced search engine users" in an effort to better understand how these users search
- If we can better understand what advanced searchers are doing maybe we can improve the search experience for everyone



#### **Characterizing Advanced Searchers**

[White & Morris, 2007]

- Four advanced operators used: +, -, "", and "site:"
  - ~1% of submitted queries contained at least one operator
  - 51K users (9%) of users used query operators at least once
- padvanced used to denote the percentage of a user's queries that contain advanced operators
  - Non-advanced users (padvanced = 0%)
  - Advanced users (padvanced > 0%)
- Included users who issued > 50 queries
  - ~38K (20%) advanced users
  - ~151K (80%) non-advanced users



## **Findings: Query/Result-click**

[White & Morris, 2007]

- Factor analysis to study the relationships among the dependent variables
- Factor analysis revealed two factors that could account for ~84% of the variance:
  - Factor A = Querying
    - Query properties associated with position of clicks in result list
  - Factor B = Result-click
    - Querying frequency associated with the likelihood that user will click on a search result and click latency



#### **Search Sessions**

[White & Morris, 2007]

digital dpreview.com pmai.org cameras S1 **S**3 S4 S2 Session - Query  $\rightarrow$  Timeout **S**3 digitalcamera-hq.com Query trail S5 S2 **S7 S6** - Query  $\rightarrow$  End trail event **S**5 **S6 S8** digital Another query camera S6 S9 • Type URL canon canon lenses • Visit homepage S1 S10 S11 amazon Check Web-based S13 S10 S12 S14 online service howstuffworks.com amazon.com Close browser Session timeout



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## Findings – Post-query browsing

[White & Morris, 2007]

#### Advanced users:

- Traverse trails faster
- Spend less time viewing each Web page
- Follow query trails with fewer steps
- Revisit pages less often
- "Branch" less often

	Feature	Padvanced					
		о%	> 0%	≥ 25%	≥ 50%	≥ 75%	
σ	Session Secs	701.10	706.21	792.65	903.01	1114.71	
g	Trail Secs	205.39	159.56	156.45	147.91	136.79	
	Display Secs	36.95	<b>32.9</b> 4	34.91	33.11	30.67	
า	Num. Steps	4.88	4.72	4.40	4.40	4.39	
•	Num. Revisits	1.20	1.02	1.03	1.03	1.02	
	Num.	1.55	1.51	1.50	1.47	1.44	
1	Branches						
	% <sub>Trails</sub>	72.14%	27.86%	.83%	.23%	.05%	
	% <sub>Users</sub>	79.90%	20.10%	.79%	.18%	.04%	

Non-advanced Advanced

More advanced  $\rightarrow$ 



#### Findings – Post-query browsing [White & Morris, 2007]

 Greater the proportion of queries with advanced syntax the more focused their search interactions become

- Shorter query trails
- Less "branchy" query trails
- Session time increases but search time drops with increases in padvanced
  - Perhaps more advanced users are multitasking between search and other activities



#### **Lecture Plan**

- Understanding user behavior at micro-, meso-, and macro- levels
- Theoretical models of information seeking
- ✓ Web search behavior:
  - ✓ Levels of detail
  - ✓ Search Intent
  - $\checkmark$  Variations in web searcher behavior
  - Keeping found things found
  - Click models



### **ReFinding Behavior**

[From Teevan et al, 2007]

- 40% of the queries led to a same user had clicked on in a past search session.
   Teevan et al., 2007
- What's the URL for this year's RuSSIR?
  - Does not really matter,
     it is faster to re-find it

```
Google russir 2009
                                                                                                                                                                                    Search Advanced Search
      Web Show options.
                                                                                                                                                                                                                                                                        Results 1 - 1
      Did vou mean: russia 2009
       Results include your SearchWiki notes for russir 2009. 
Share these notes
       RuSSIR'2009: 3rd Russian Summer School in Information Retrieval
       The 3rd Russian Summer School in Information Retrieval will be held September 11-16, 2009 in
      Petrozavodsk, Russia. The school is co-organized by the Russian ...
      romip.ru/russir2009/eng/index.html - Cached - Similar - 💬 \Lambda 👽
       RuSSIR 2009: call for participation | PASCAL 2
       RuSSIR 2009 is co-located with the yearly ROMIP meeting (http://romip.ru/) and Russian
       Conference on Digital Libraries 2009 ...
       www.pascal-network.org/?q=node/106 - Cached - Similar - P T
                       3rd Russian Summer School in Information Retrieval (RuSSIR 2009 ...
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                       Conference on Digital Libraries 2009 ...
                       www.pascal-network.org/?q=node/78 - Cached - Similar - 💬 🗟 🗵
       RuSSIR'2009: 3rd Russian Summer School in Information Retrieval
      The Republic of Karelia is located in the North-West of Russia. Karelia is often called "stony
       lake-forest land" and "the lungs of Europe", highlighting the ...
       romip.ru/russir2009/eng/venue.html - Cached - Similar - Cached - Similar - Revealed - Similar
       LINGUIST List 20.171: Applied Ling, Computational Ling, Text ...
       Jan 19, 2009 ... 3rd Russian Summer School in Information Retrieval (RuSSIR 2009). ...
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       3rd Russian Summer School in Information Retrieval (RuSSIR 2009)
       FIRST CALL FOR COURSE PROPOSALS ... The 3rd Russian Summer School in Information
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       sci.tech-archive.net/Archive/sci.image...11/msg00046.html - Cached - Similar - Similar - Cached - Similar - Similar - Cached - Similar - Simi
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       Третья Российская летняя школа по информационному поиску (RuSSIR 2009) пройдет с
      11 по 16 сентября в Петрозаводске, непосредственно перед ежегодным ...
       community.livejournal.com/ru ir/74129.html - Similar - 💬 🗟 🔀
       SEMICON Russia - Home
       Dear SEMICON Russia 2009 Participants! Thank you very much for taking part in our porgram.
      We are looking forward to see you at SEMICON Russia 2010 ...
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      SELF-EVALUATION FORMS - CORDIS: FP7: Find a Call
Done
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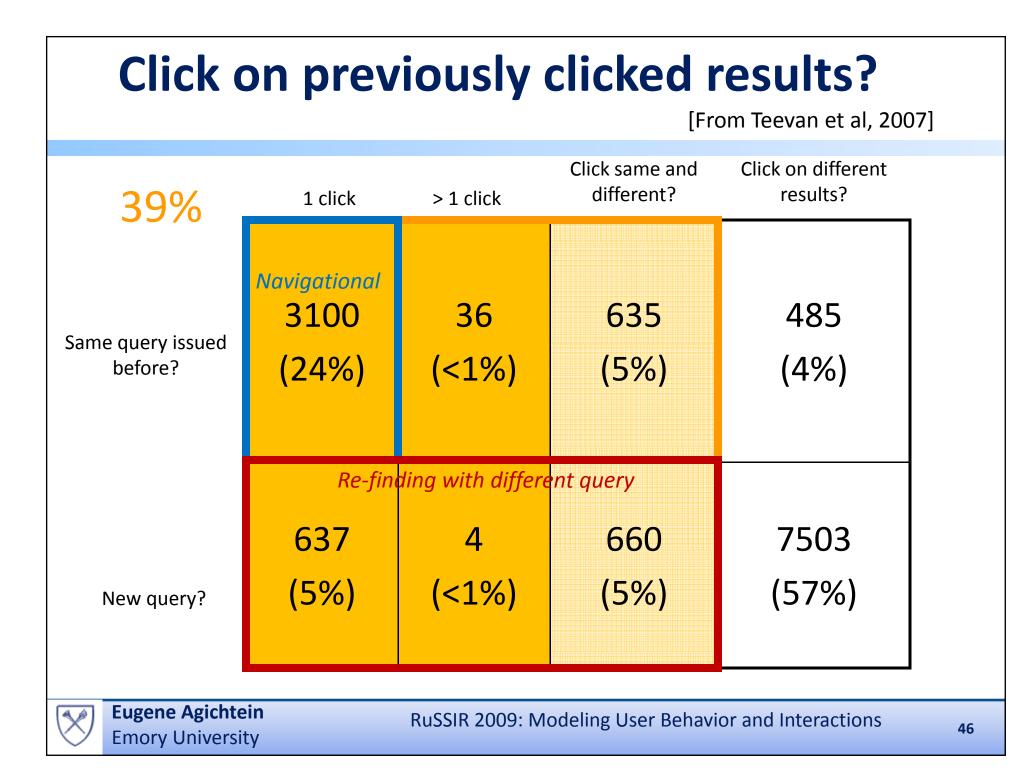
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#### What Is Known About Re-Finding

[From Teevan et al, 2007]

- Re-finding recent topic of interest
- Web re-visitation common [Tauscher & Greenberg]
- People follow known paths for re-finding
  - Search engines likely to be used for re-finding
- Query log analysis of re-finding
  - Query sessions [Jones & Fain]
  - Temporal aspects [Sanderson & Dumais]





#### How Queries Change [From Teevan et al. 2007]

- Many ways queries can change
  - Capitalization ("new york" and "New York")
  - Word swap ("britney spears" and "spears britney")
  - Word merge ("walmart" and "wal mart")
  - Word removal ("orange county venues" and "orange county music venues")
- 17 types of change identified
  - 2049 combinations explored
  - Log data and supplemental study
  - Most normalizations require only one type of change



## Rank Change Reduces Re-Finding

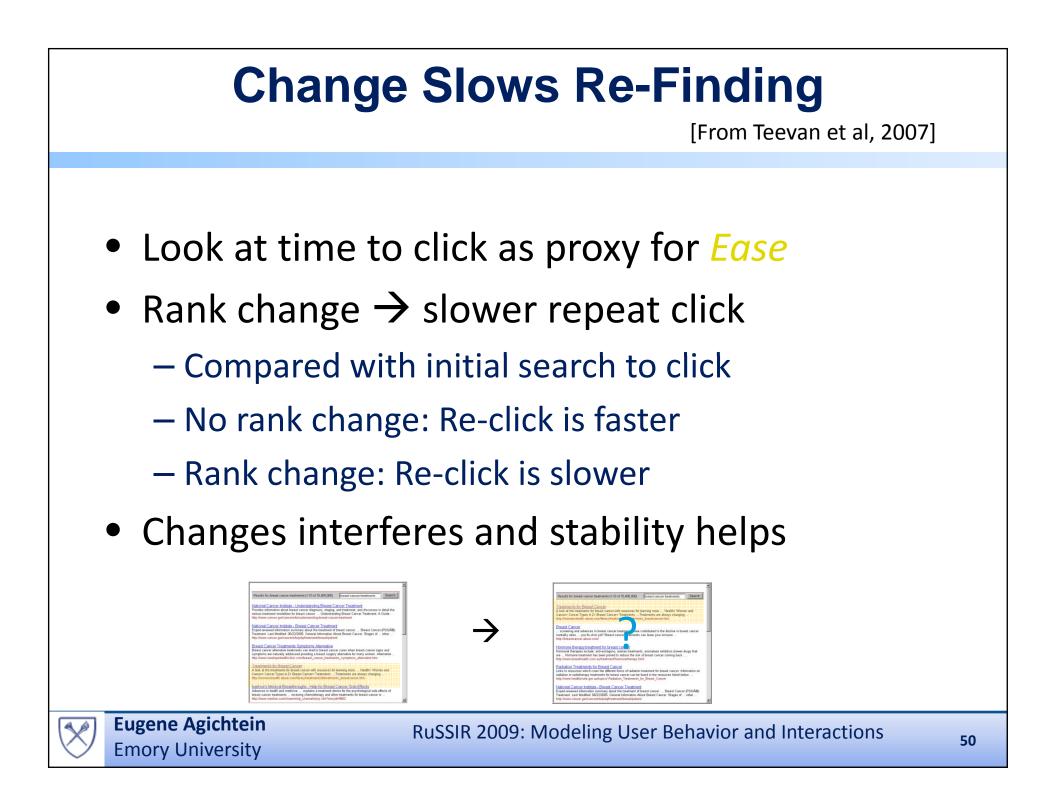
- Results change rank
- Change reduces probability of repeat click
  - No rank change: 88% chance
  - Rank change: 53% chance
- Why?
  - Gone?
  - Not seen?
  - New results are better?



#### Gone? Not Seen? Better?

[From Teevan et al, 2007]

Results for breast cancer treatm				
National Cancer Institute - U Provides information about breast various treatment modalities for bi http://www.cancer.gov/cancerinfo/	Results for breast cancer treatments (1-10 of 15,400,000) breast cancer treatments Search Treatments for Breast Cancer			
National Cancer Institute - B Expert reviewed information sumn Treatment: bast Modified: 06/22/2 http://www.cancer.gov/cancerinfo/	A look at the treatments for breast cancer with resources for learning more Health> Women and Cancer> Cancer Types A Z> Breast Cancer> Treatments Treatments are always changing http://womenshealth.about.com/library/treatments/bltreatments_breastcancer.htm			
Breast Cancer Treatments's Breast cancer alternative treatments symptoms are naturally advresse http://www.newhopehealt/clinic.c				
<u>Treatments for D east Canc</u> A look at the treatments for breas Cancer> Cancer Types A Z> Brea http://womenshealth.about.com/li	Hormone therapy treatment for breast cancer Hormonal therapies include: anti-estrogens, ovarian treatments, aromatase inhibitors (newer drugs that are Hormone treatment has been proved to reduce the risk of breast cancer coming back http://www.breasthealth.com.au/treatment/hormonetherapy.html			
Ivanhoe's Medical Breakthro Advances in health and medicine. breast cancer treatments recei http://www.ivanhoe.com/channels         Radiation Treatments for Breast Cancer           Links to resources which cover the different forms of radiation treatment for breast cancer. Informat radiation or radiotherapy treatments for breast cancer can be found in the resources listed below http://www.healthinsite.gov.au/topics/ Radiation_Treatments_for_Breast_Cancer				
Eugene Agichtein	National Cancer Institute - Breast Cancer Treatment Expert-reviewed information summary about the treatment of breast cancer Breast Cancer (PDQ®): Treatment. Last Modified: 06/22/2005. General Information About Breast Cancer. Stages of other http://www.cancer.gov/cancerinfo/pdq/treatment/breast/patient			
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## **Helping People Re-Find**

- Potential way to take advantage of stability
  - Automatically determine if the task is re-finding
  - Keep results consistent with expectation
  - Simple form of personalization
- Can we automatically predict if a query is intended for re-finding?



## **Predicting the Query Target**

- For simple navigational queries, predict what URL will be clicked
- For complex repeat queries, two binary classification tasks:
  - Will a new (never visited) result be clicked?
  - Will an old (previously visited) result be clicked?



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### **Predicting Navigational Queries**

- Predict navigational query clicks using
  - Query issued twice before
  - Queries with the same one result clicked
- Very effective prediction
  - 96% accuracy: Predict one of the results clicked
  - 95% accuracy: Predict first result clicked
  - 94% accuracy: Predict only result clicked



#### **Predicting More Complex Queries**

- Trained an SVM to identify
  - If a new result will be clicked
  - If an old result will be clicked
- Effective features:
  - Number of previous searches for the same thing
  - Whether any or the results were clicked >1 time
  - Number of clicks each time the query was issued
- Accuracy around 80% for both prediction tasks



## **Re-Finding Summary**

- Log analysis supplemented by a user study
- Re-finding is very common
  - Navigational queries are particularly common
  - Categorized potential re-finding behavior
  - Explored ways query strings are modified
- Stability of result rank impacts re-finding tasks
- Can identify refinding queries with 80-90% accuracy

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## Automatic Click models

- Clickthrough and subsequent browsing behavior of individual users influenced by many factors
  - Relevance of a result to a query
  - Visual appearance and layout
  - Result presentation order
  - Context, history, etc.



#### **Hypothesis 1: No Bias**

[Craswell et al., 2008]

Our baseline

$$c_{di} = r_d = c_{dj}$$

- cdi is P( Click=True | Document=d, Position=i )
- rd is P( Click=True | Document=d )
- Why this baseline?
  - We know that rd is part of the explanation
  - Perhaps, for ranks 9 vs 10, it's the main explanation
  - It is a bad explanation at rank 1 e.g. Eve tracking

Attractiveness of summary ~= Relevance of result

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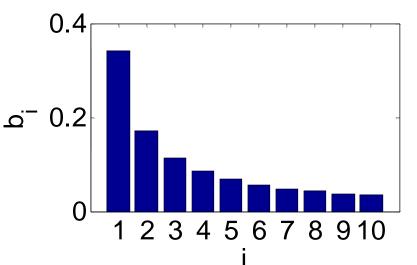


[Craswell et al., 2008]

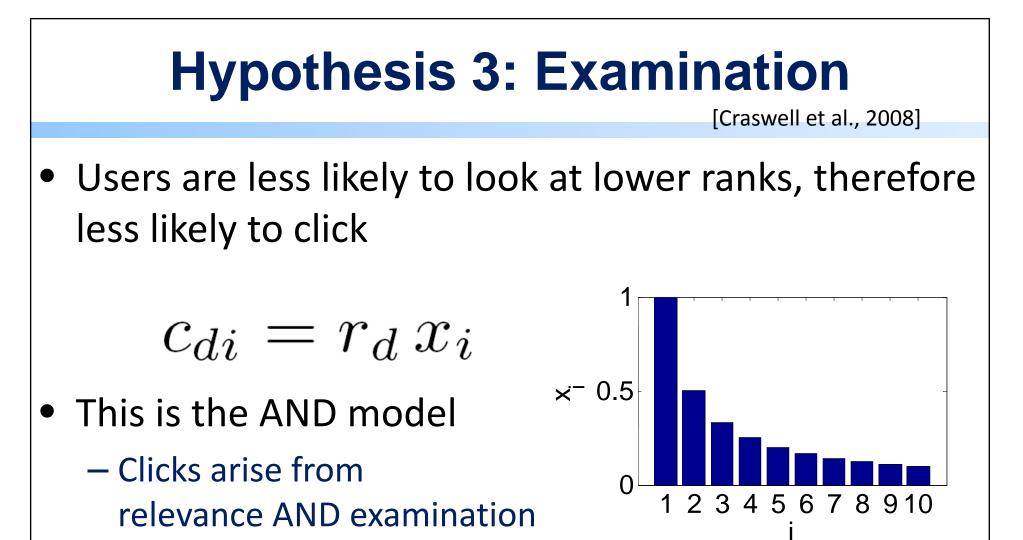
- There are two types of user/interaction
  - Click based on relevance
  - Click based on rank (blindly)

$$c_{di} = \lambda r_d + (1 - \lambda) b_i$$

- A.k.a. the OR model:
  - Clicks arise from relevance OR position

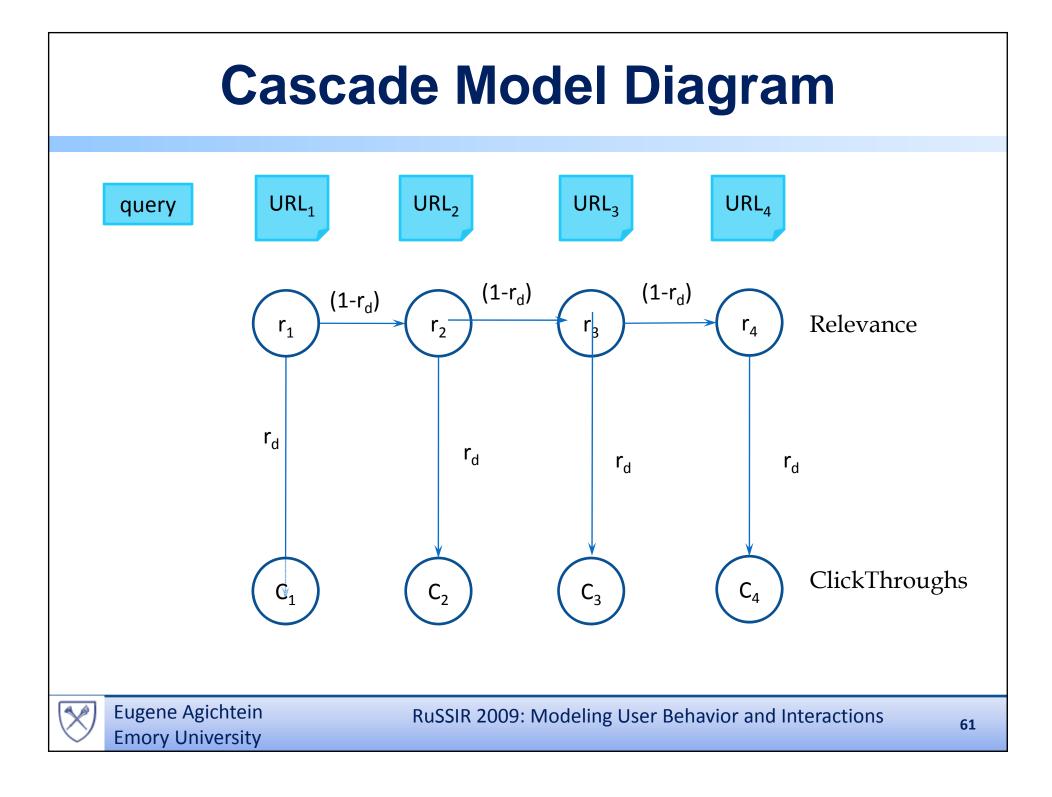






– Probability of examination  $x_i$  does not depend on what else is in the list





#### Hypothesis 4: Cascade

[Taylor et al., 2008]

- Users examine the results in rank order
- At each document d
  - Click with probability  $r_d$
  - Or continue with probability (1-r<sub>d</sub>)

$$c_{di} = r_d \prod_{j=1}^{i-1} (1 - r_{docinrank:j})$$



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## Cascade Model Example

[Craswell et al., 2008]

- 500 users typed a query
- 0 click on result A in rank 1
- 100 click on result B in rank 2
- 100 click on result C in rank 3

This may seem different from the formulation on the previous slide, but is precisely equivalent

- Cascade (with no smoothing) says:
- 0 of 500 clicked A  $\rightarrow$  rA = 0
- 100 of 500 clicked B → rB = 0.2
- 100 of remaining 400 clicked C → rC = 0.25

#### **Cascade Model Seems Closest to Reality**

[Craswell et al., 2008]

Model Cross Entropy  $0.141 \pm 0.0055$ Best Possible  $0.225 \pm 0.0052$ Cascade  $0.236 \pm 0.0063$ Logistic Examination  $0.247 \pm 0.0072$  $0.250 \pm 0.0073$ Baseline

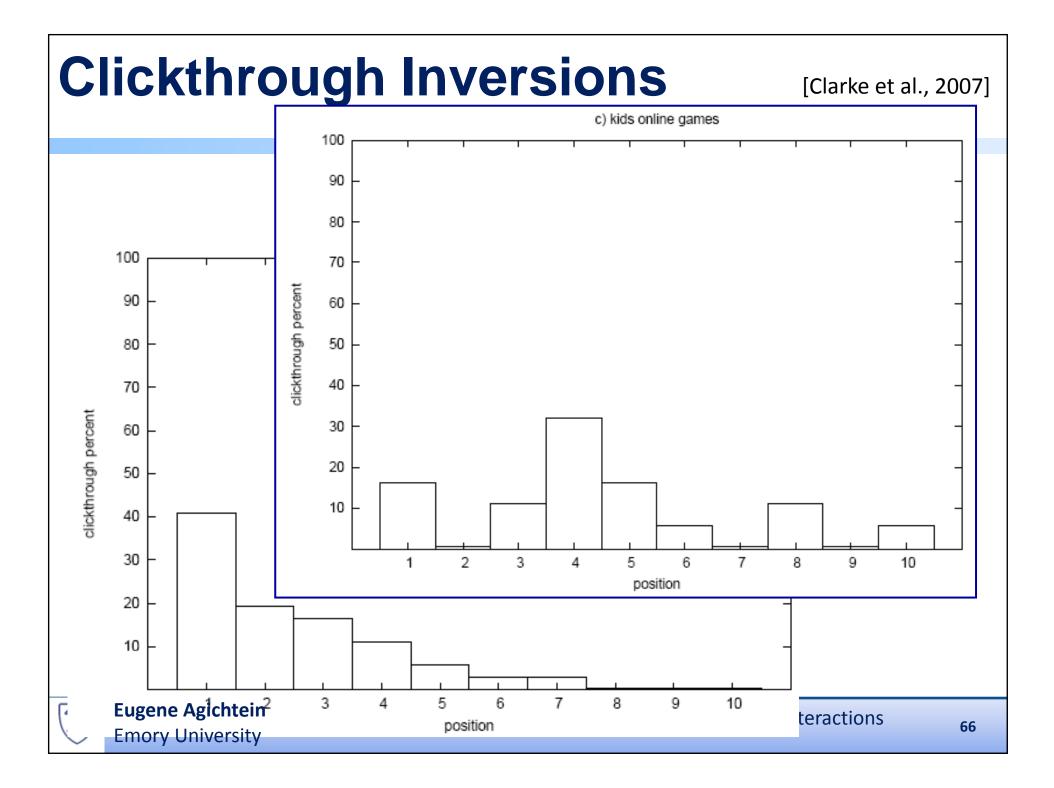
Best possible: Given the true click counts for ordering BA



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#### Problem: Users click based on result "Snippets" [Clarke et al., 2007] title 10 kids online games Yahool Kids: Games, music, movies, videos and fun for kids Featuring comprehensive safe surfing, games; homework help, and many kid friendly activities. vahooligans.vahoo.com · 12/17/2006 · Cached page FEMA For Kids: Games www.fema.gov/kids/games1.htm - 12/16/2006 - Cathed page Boston Children's Museum : Kids : Online Games KIDS' QUEST. Boston Children's Museum gratefully acknowledges The Center for Disease Control's National Center on Birth Defects and Development Disabilities for its .... www.bostonkids.org/kids/online games.html + 12/17/2006 + Cached page snippet URL **Eugene Agichtein RuSSIR 2009: Modeling User Behavior and Interactions** 65 **Emory University**



#### **Relevance is Not the Dominant Factor!**

[Clarke et al., 2007]

Relationship	Number	Percent
rel(A) < rel(B)	119	33.5%
rel(A) = rel(B)	134	37.7%
rel(A) > rel(B)	102	28.7%

Figure 3: Relevance relationships at clickthrough inversions. Compares relevance between the higher ranking member of a caption pair (rel(A)) to the relevance of the lower ranking member (rel(B)), where caption A received fewer clicks than caption B.



#### **Snippet Features Studied**

[Clarke et al., 2007]

Feature Tag	Description
MissingSnippet	snippet missing in caption A and present in caption B
SnippetShort	short snippet in caption A ( $< 25$ characters) with long snippet ( $> 100$ characters) in caption B
TermMatchTitle	title of caption A contains matches to fewer query terms than the title of caption B
TermMatchTS	title+snippet of caption A contains matches to fewer query terms than the title+snippet of caption B
TermMatchTSU	title+snippet+URL of caption A contains matches to fewer query terms than caption B
TitleStartQuery	title of caption B (but not A) starts with a phrase match to the query
QueryPhraseMatch	title+snippet+url contains the query as a phrase match
MatchAll	caption B contains one match to each term; caption A contains more matches with missing terms
URLQuery	caption B URL is of the form www.query.com where the query matches exactly with spaces removed
URLSlashes	caption A URL contains more slashes (i.e. a longer path length) than the caption B URL
URLLenDIff	caption A URL is longer than the caption B URL
Official	title or snippet of caption B (but not A) contains the term "official" (with stemming)
Home	title or snippet of caption B (but not A) contains the phrase "home page"
Image	title or snippet of caption B (but not A) contains a term suggesting the presence of an image gallery
Readable	caption B (but not A) passes a simple readability test



# Feature Importance [Clarke et al., 2007]

Feature Tag	INV+	INV-	%+	CON+	CON-	%+	$\chi^2$	p-value
MissingSnippet	185	121	60.4	144	133	51.9	4.2443	0.0393
SnippetShort	20	6	76.9	12	16	42.8	6.4803	0.0109
TermMatchTitle	800	559	58.8	660	700	48.5	29.2154	<.0001
TermMatchTS	310	213	59.2	269	216	55.4	1.4938	0.2216
TermMatchTSU	236	138	63.1	189	149	55.9	3.8088	0.0509
TitleStartQuery	1058	933	53.1	916	1096	45.5	23.1999	< .0001
QueryPhraseMatch	465	346	57.3	427	422	50.2	8.2741	0.0040
MatchAll	8	2	80.0	1	4	20.0		0.0470
URLQuery	277	188	59.5	159	315	33.5	63.9210	<.0001
URLSlashes	1715	1388	55.2	1380	1758	43.9	79.5819	< .0001
URLLenDiff	2288	2233	50.6	2062	2649	43.7	43.2974	< .0001
Official	215	142	60.2	133	215	38.2	34.1397	<.0001
Home	62	49	55.8	64	82	43.8	3.6458	0.0562
Image	391	270	59.1	315	335	48.4	15.0735	<.0001
Readable	52	43	54.7	31	48	39.2	4.1518	0.0415



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#### **Important Words in Snippet**

[Clarke et al., 2007]

$\operatorname{Rank}$	Term	$\chi^2$	influence
1	encyclopedia	114.6891	$\rightarrow$
2	wikipedia	94.0033	$\downarrow$
3	official	36.5566	
4	and	28.3349	$\uparrow$
5	tourism	25.2003	$\uparrow$
6	$\operatorname{attractions}$	24.7283	
7	free	23.6529	Ļ
8	sexy	21.9773	$\uparrow$
9	medlineplus	19.9726	$\downarrow$
10	information	19.9115	Ŷ

Figure 6: Words exhibiting the greatest positive  $(\uparrow)$  and negative  $(\downarrow)$  influence on clickthrough patterns.



#### **Click Models Summary**

Models proposed to simulate searcher click process

- Increasingly sophisticated and theories
- Assume searcher is rational and consistent

But, searchers are not rational or careful:

Attracted/repelled by simple features of summaries

Will incorporate summary and browsing info to extract relevance information from clicks (next lecture)



### Lecture 1 Summary. Questions?

- Understanding user behavior at micro-, meso-, and macro- levels
- Theoretical models of information seeking
- Web search behavior:
  - ✓ Levels of detail
  - ✓ Search Intent
  - $\checkmark$  Variations in web searcher behavior
  - ✓ Keeping found things found
  - ✓ Click models



#### **References and Further Reading**

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