Testing the Satisficing Model Using Web Browsing Data

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Introduction

- Choice alone cannot be used to test the Satisficing model
- We need other data
- Choice process data is one option
- Another would be to use data directly on search
- Assume we observe exactly what alternatives have been looked at, and in what order

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Satisficing and Search Data

- Satisficing make 2 predictions
- 1. Object chosen should be last searched (unless they have search all available alternatives)
 - Search stops when an above reservation alternative is found
 - That alternative is then chosen
- 2. Value of the best option currently seen should predict probability of continuing to search
 - Higher value alternatives more likely to be above reservation level
 - More likely that search will stop

An Alternative Model

- Fixed search set size
 - Before starting to search decide how many alternatives to look at
 - Search that number of alternatives regardless of what is seen along the way
 - Note that such behavior is not optimal if one can 'dynamically optimize'
- Fixed search set implies
 - Last object seen not necessarily the one purchased
 - Value of object seen not predictive of whether search will continue

Data

- Web browsing data allows us to approximate search data
 - We can record what websites a subject has looked at
 - (Note this is not the same thing as a subject understanding what is on the website)
- Dataset: 152,000 users from ComScore
 - Company that records web browsing activity (!)
 - Date
 - Time
 - Duration
 - Purchase description, price and quantity

Data

• Concentrate on purchase of books

Table 1—Transactions and Visits by Bookstore							
		pactions	Visits				
Bookstore	Number	Percentage	Number	Percentage			
Amazon	10,197	65.5	249,593	76.3			
Barnes and Noble	3,042	19.6	25,758	7.9			
Book Clube							
christianbook.com	615	3.9	3,968	1.2			
doubledaybookclub.com	465	3.0	4,001	1.2			
eharlequin.com	61	0.4	3,647	1.1			
literaryguild.com	322	2.1	3,500	1.1			
mysteryguild.com	187	1.2	2,095	0.6			
Other Bookstore							
1bookstreet.com	10	0.1	120	0.0			
allbookstless.com	5	0.0	199	0.1			
alldirect.com	27	0.2	490	0.1			
ecampus.com	114	0.7	1,206	0.4			
powells.com	68	0.4	1,320	0.4			
vagsitybooks.com	16	0.1	218	0.1			
walmart.com	183	1.2	25,663	5.5			
booloamillion.com	246	1.6	2,290	0.7			
Total	15,561	100.0	327,074	100.0			

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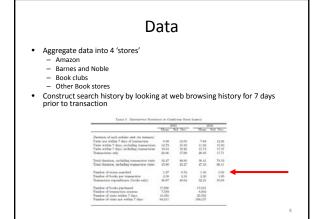
Data

- Aggregate data into 4 'stores'

 - Amazon Barnes and Noble

 - Book clubs Other Book stores
- Construct search history by looking at web browsing history for 7 days prior to transaction

		2007	2011	
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Darwins of such unbain real (to minute)				
Visits not within 7 days of transaction	0.00	1100	7.69	12.3
Vaits within 7 dury, excluding transactions	12.72	15.65	31.02	15.0
Vasts within 7 days, including transactions	19.74	19.26	13.74	17.35
Transactions only	25.36	17.69	26.08	17.71
Total duration, excluding transaction visits	32.47	49.50	39.41	79.5
Deal duration, including transaction visits	43.66	63.27	47.43	96.1
Number of stores marched	1:27	9.54	1.30	4.5
Number of books per transaction	2.38	2.10	2.20	1.00
Transaction expenditures (broke only)	SMAT		30.21	35.69
Number of books purchased	17.656		17,631	
Number of transaction senten-	7.559		8.000	
Number of yorks within 7 days	19.350		25.536	
Number of visits not within 7 date	94,011		149,127	



Data

- Assume that the product the consumer wants is homogenous
 - They really want a copy of 'Inferno' by Dan Brown
- Search is over prices
 - Price of book in purchased store observed directly
 - Price of books in other store imputed from most recent purchase

Results

- 1. Consumers do not maximize on price
 - Buy from lowest priced store in 63% of observations
 - Average loss \$2.60 compared to lowest available
 - BUT this difference is not due entirely to unawareness
 - Average loss relative to lowest of stores searched is \$1.99

Results

2. Some consumers do NOT buy the last searched product

Search window	No. of stores visited	Percentage	If 2 or more stores, bought from:	Percentage	Precentage exhausted search?
7 Days	One	76			
	2 or more	24	Last store sampled Recalled	65	55
& Dana	One	77			
	2 or more	23	Last store sampled Recalled	64 36	55
5 Dans	One	79	(NOCHARIO)	,00	99
2 Dalia	2 or more	21	Last store sampled Recalled	63 37	55
4 Days	Oter	85			
	2 or more	20	Last store sampled Recalled	61	55
3 Days	One	102			-
	2 or more	18	Last store sampled Recalled	61	56
2 Dans	One	54			
	2 or more	16	Last store sampled Recalled	61	56
1 Day	One	Mi.			
	2 or more	14	Last store sampled Recalled	61	56
Same day	One	90	SUCCESSIVE.		
	2 or more	10	Last store sampled Recalled	62 38	58

Results

2. Some consumers do NOT buy the last searched product

Table 5—Test of "No Recall" Hypothesis					
Search window	No. of stores visited	Parastage	If 2 or more stores,	Demontary	Precentage exhausted search?
7 Days	One	76			
	2 or more	24	Last store sampled Socialist	65	55
6 Days		- 77			
	2 or more	23	Last store sampled Recalled	64 36	55
5 Days	One	79			
	2 or more	21	Last store sampled Recalled	63 37	55
4 Days	One	85			
	2 or more	20	Last store sampled Recalled	61 29	55
3 Days	One	82			
	2 or more	18	Last store sampled Recalled	61 39	56
2 Dans	One	54			
	2 or more	16	Last store sampled Recalled	61 39	56
1 Day	One	86			
	2 or more	14	Last store sampled Recalled	61 39	56
Same day	One	90			
	2 or more	10	Last store sampled Recalled	62 38	58

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Results

3. Observed price does NOT affect the decision to continue searching

Table 5—Price of the First Store by Number of Searches

Price of the first store	Once	Twice	Total
Lower or equal	63.55%	61.89%	63.32%
Higher	36.45%	38.11%	36.68%
Number of observations	2,244	349	2,593

Summary

- We have introduced the 'Satisficing' model of incomplete attention
- Shown that satisficing can be optimal in the face of per-item search
- Shown that it is difficult to test satisficing with standard choice data
- Introduced two data sets which can be used to test satisficing
 - Choice processSearch data
- In the lab, satisficing seems to do a reasonable job of explaining behavior
- But in web search, behavior seems better described by a 'fixed search' algorithm

 But data set does not have a lot of power