

Clickstream Analysis for Internet Marketing

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Outline

- Principles of Internet Marketing
- Defining Clickstream Data
 - What does it tell us about an anonymous user?
- Analysis of Clickstream Data
 - User Profiling
 - Measuring Brand Equity: Choice Modeling
 - Cognitive Lock-In: Power Law of Practice
 - Forecasting Purchases using Clickstream Data

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Are there any principles behind Internet Marketing?

Interactive Marketing/
1-to-1 Marketing/
Relationship Marketing/
Personalization/
Customer Relationship Management

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Interactive Marketing

Blattberg and Deighton (Sloan Management Review 1991) discuss the differences between traditional mass marketing and interactive marketing:

- Use of actual behavior to identify customers/prospects
- Customized prices/product offers to the individual
- Customizing the advertising message via selected binding (split cable).
- Distribution: Direct links with the customer
- Sales force: Improved monitoring, less discretion, better data for the sales person

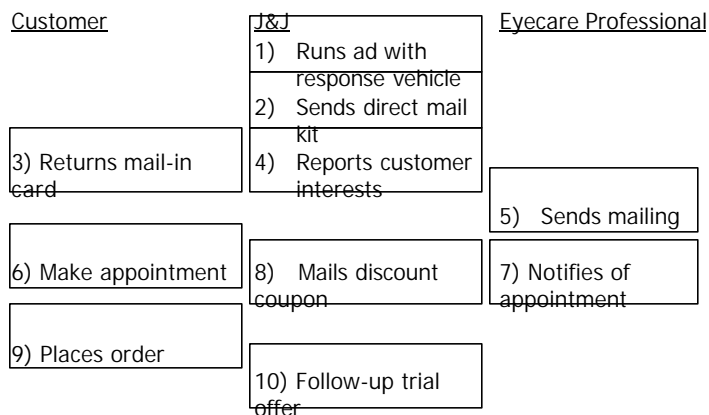
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Comparing Mass and 1to1 Marketing

Product manager sells one product at a time to as many customers as possible	➔	Sell as many products as possible to one customer at a time
Differentiate his products	➔	Differentiate his customer
Acquire a constant stream of new customers	➔	Tries to get a constant stream of business from current customers
Concentrates on economies of scales	➔	Concentrates on economies of scope

Peppers and Rogers, **The One to One Future** 5

Acuvue Integrated Communications Program



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The Web's Contribution

- Addressable marketing is not new
 - Mail and telephone have been around
 - For a long time the most valuable assets of companies like LL Bean, Fidelity Investments, American Express are their electronic customer transaction histories
- The difference between the web and these other technologies is that addressability and the collection of transaction histories is almost automatic

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Interactive Marketing Requires...

- Ability to *identify* end-users
- Ability to *differentiate* customers based on their value and their needs
- Ability to *interact* with your customers
- Ability to *customize* your products and services based on knowledge about your customers

Peppers, Rogers, and Dorf (1999)

Information is key!

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Example: American Airlines

- Updated in late 1998, has the capability to build custom pages for each of the airline's 2 million registered users
- Prior to the web, there was no cost-efficient way to tell millions of customers about a special fare available only this weekend, to a destination you personally will find attractive

[Source](#)



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Definition of Clickstream Data

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What is clickstream data?

- A record of an individual's movement through time at a web site
- Contains information about:
 - Time
 - URL content
 - User's machine
 - Previous URL viewed
 - Browser type

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A clickstream example

Household ID: Female born 12Jul42

Demographics: Philadelphia Area, Male and Female Married (husband born: 27Sep46), 3 members in household, income: \$75,000-\$99,999, Graduated College, employed 35 or more hours, 1 child age 13 to 17 (daughter born: 5Jul80), own single family home, white collar, own car & truck, microwave, three dogs, five cats

18JUL97:18:55:57	47	www.voicenet.com/
18JUL97:18:56:44	37	www.weather.com/
18JUL97:18:57:25	105	www.weather.com/weather/us/cities/MI_Traverse_City.html
18JUL97:19:03:00	7	www.weather.com/weather/us/cities/MI_Traverse_City.html
18JUL97:19:03:56	2	www.weather.com/weather/us/cities/MI_Traverse_City.html
18JUL97:19:03:58	6	www.weather.com/weather/us/cities/HI_Lahaina.html
18JUL97:19:04:58	2	www.weather.com/weather/us/cities/HI_Lahaina.html
18JUL97:19:05:00	1	www.weather.com/weather/us/cities/MI_Traverse_City.html
18JUL97:19:15:24	39	www.weather.com/weather/us/cities/MI_Traverse_City.html
18JUL97:19:17:00	7	www.weather.com/weather/us/cities/MI_Traverse_City.html
18JUL97:19:17:07	13	www.realastrology.com/
18JUL97:19:17:20	44	www.realastrology.com/libra.html

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What does the ISP see?

www.voicenet.com would see the following requests:

18JUL97:18:55:57	www.voicenet.com/
18JUL97:18:56:44	www.weather.com/
18JUL97:18:57:25	www.weather.com/weather/us/cities/MI_Traverse_City.html
18JUL97:19:03:58	www.weather.com/weather/us/cities/HI_Lahaina.html
18JUL97:19:17:07	www.realastrology.com/
18JUL97:19:17:20	www.realastrology.com/libra.html

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What does the server see?

www.weather.com would see the following requests:

18JUL97:18:56:44	www.weather.com/
18JUL97:18:56:44	www.weather.com/weather/us/cities/HI_Lahaina.html
18JUL97:18:57:25	www.weather.com/weather/us/cities/MI_Traverse_City.html

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Sources of clickstream data

- Web Servers
 - Each hit is recorded in the web server log
- Media Service Providers
 - DoubleClick, Flycast
- ISP/Hosting Services
 - AOL, Juno, Bluelight.com
- Marketing Research Companies
 - Media Metrix, NetRatings, PCData

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Processing clickstream data

- IS Departments
- Site Activity Tracking Suppliers
- Auditing and Verification Suppliers
- Marketing Research Suppliers

What do you do with your data?

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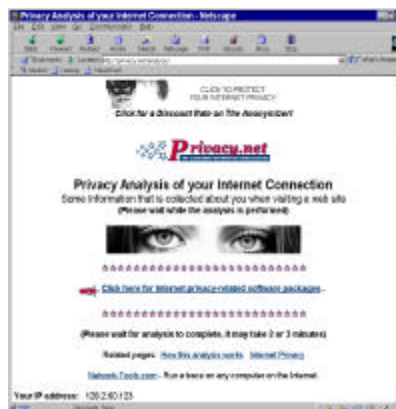
What can we say about an 'anonymous' user?

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The Raw Clickstream Data

All information sent by my web browser when requesting <http://www.privacy.net/analyze>:

Accept: image/gif, image/x-xbitmap, image/jpeg, image/pjpeg, image/png, */*
Accept-Language: en Connection: Keep-Alive
Host: www.privacy.net
Referer: http://www.privacy.net/
User-Agent:
 Mozilla/4.75 [en] (WinNT; U)
Pragma: no-cache
Cookie: Date=10/18/00;
 Privacy.net=Privacy+Analysis
Accept-Encoding: gzip
Accept-Charset: iso-8859-1,*,utf-8



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Using Domain lookups

- If we match my domain, cmu.edu, with its registered zip code, "15213", we can think about geodemographic marketing
- The most likely visitor from the "15213" zip is the University USA segment
- What can we do with a ZIP Code?

[ENDS/MicroVision Lifestyle Game](#)

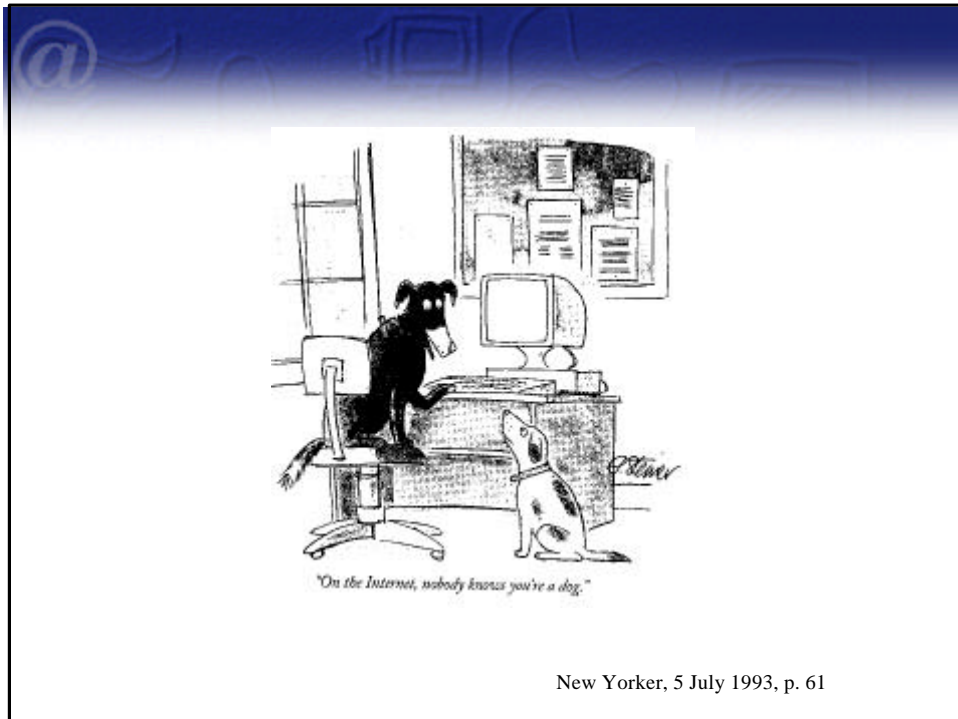


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User Profiling

What does 'where you go' say about 'who you are'?

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Clickstream Example: Who is this web person?


Example

Session Number	Start Time	Seconds Actively Viewing	Seconds Elapsed between start and stop	URL
1	06.Jul.1995:20:49:31	4	100	http://www.merrysbeanies.com/merrysbeanies/pokadach.htm
2	06.Jul.1995:20:51:11	52	132	http://www.merrysbeanies.com/merrysbeanies/pokadach.htm
3	06.Jul.1995:20:53:23	51	85	http://www.aol.com/inetfind/kids/home.html
4	06.Jul.1995:20:54:49	51	219	http://www.merrysbeanies.com/merrysbeanies/pokadach.htm
5	06.Jul.1995:20:58:20	50	1162	http://www.blueplanet.com/fun/packman/
6	07.Jul.1995:00:14:28	51	74	http://find.web.aol.com/channel/find/mainchannel/find?query=eric+mc
7	07.Jul.1995:00:15:42	50	162635	http://www.lycos.com/network/
8	08.Jul.1995:21:26:17	4	113	http://www.zvooorder.com/st/Mogon.asp
9	08.Jul.1995:21:28:10	19	0	http://www.zvooorder.com/st/Verification.asp
10	08.Jul.1995:21:28:29	4	43	http://find.web.aol.com/channel/find/mainchannel/find?query=stargate
11	08.Jul.1995:21:29:12	1	601	http://members.aol.com/aneof
12	08.Jul.1995:21:29:13	2	65	http://members.aol.com/aneof
13	08.Jul.1995:21:40:10	36	0	http://ids.simplenet.com/stargate/links.htm
14	08.Jul.1995:21:40:54	132	172	http://www.e-net.or.jp/usetecol/1701/rdaf
15	08.Jul.1995:21:43:46	1	20	http://www.e-net.or.jp/usetecol/1701/rdaf06/99sg1.html
16	08.Jul.1995:21:44:06	1	5	http://www.e-net.or.jp/usetecol/1701/rdaf
17	08.Jul.1995:21:44:11	1	15	http://www.e-net.or.jp/usetecol/1701/rdaf06/99.html
18	08.Jul.1995:21:44:26	6	6	http://www.e-net.or.jp/usetecol/1701/rdaf06/99.html
19	08.Jul.1995:21:44:32	1	15	http://www.e-net.or.jp/usetecol/1701/rdaf
20	08.Jul.1995:21:44:47	1	16	http://www.e-net.or.jp/usetecol/1701/wase/05main.html
21	08.Jul.1995:21:45:00	57	85	http://www.e-net.or.jp/usetecol/1701/wase/05main.html

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How much have you learned about this person?

- Gender
- Age
- Race
- Marital Status
- Geographic Location
- City Size
- Household Size
- Household Composition
- Household Income
- Rent or Own
- Education
- Age and presence of children
- Car or truck ownership
- Dog or cat ownership
- Female
- 34 years old
- White
- Single
- East South Central
- 250,000-499,999
- 2 household members
- Female head living with others related
- \$25,000-\$29,999
- Own
- Graduated High School
- No Children Under 18
- Two cars, no trucks
- No dogs or cats

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What is this user's gender?

Web sites visited during one month:

48%	aol.com	63%	libertynet.org
64%	astronet.com	39%	lycos.com
75%	avon.com	27%	netradio.net
52%	blue-planet.com	57%	nick.com
56%	cartoonnetwork.com	59%	onhealth.com
54%	cbs.com	83%	onlinepsych.com
76%	country-lane.com	44%	simplenet.com
47%	eplay.com	76%	thriveonline.com
41%	halcyon.com	59%	valupage.com
70%	homearts.com	71%	virtualgarden.com
66%	ivillage.com	66%	womenswire.com

Percentage of female viewers using PC Meter data

Bayesian Updating Formula

Application of Bayesian hypothesis updating.

$$\bar{p} = \frac{p \cdot \bar{p}}{p \cdot \bar{p} + (1 - p)(1 - \bar{p})}$$

New probability \bar{p} = $\frac{p \cdot \bar{p}}{p \cdot \bar{p} + (1 - p)(1 - \bar{p})}$
 New information p Old probability \bar{p}

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Statistical Analysis

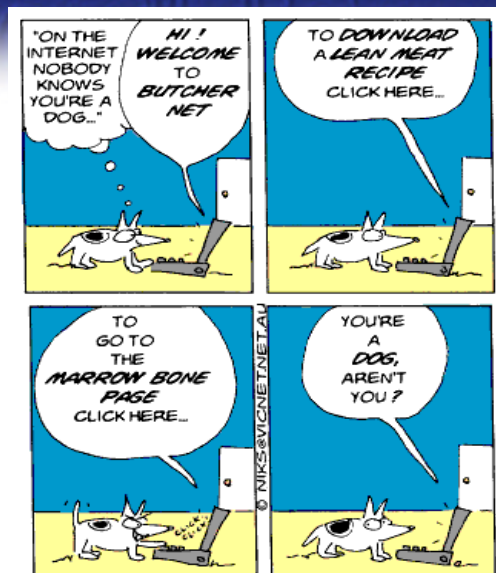
	A	B	C	D	E	F	G
		Ad Network	Probability a Female Visits the site	Probability visitor is Female given visits	Probability visitor is female for Ad Network		
1	(Email)						
2	OverseasInterview		45%	45.00%	45.00%		
3	zol.com	AOL	46%	43.03%	45.00%		Ad Network
4	astranet.com	X	64%	57.31%	45.00%		DoubleClick
5	zvtv.com	None	75%	60.13%	45.00%		
6	likeplanet.com	?	52%	61.36%	45.00%		
7	carlsonnetwork.com	X	56%	64.74%	45.00%		
8	cbs.com	DoubleClick	54%	66.70%	48.59%		
9	country-sites.com	None	76%	65.30%	48.59%		
10	epny.com	X	47%	64.62%	48.59%		
11	halcyon.com	?	41%	62.71%	48.59%		
12	hormearts.com	X	70%	66.74%	48.59%		
13	ivillage.com	DoubleClick	66%	68.29%	65.09%		
14	iberlynet.org	DoubleClick	63%	68.99%	76.05%		
15	lycos.com	X	39%	68.43%	76.05%		
16	netradio.net	X	27%	65.07%	76.05%		
17	nick.com	DoubleClick	57%	66.65%	80.88%		
18	onhealth.com	X	59%	67.79%	80.88%		
19	onlinepsych.com	DoubleClick	83%	69.54%	95.36%		
20	simplest.com	?	44%	68.43%	95.36%		
21	threesonline.com	AOL	76%	69.03%	95.36%		
22	valupage.com	None	59%	69.07%	95.36%		
23	vg.com	X	71%	69.63%	95.36%		
24	womenwire.com	X	66%	69.97%	95.36%		

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Results

- Analysis shows that there is a $>99\%$ probability this user is female.
 - Using only DoubleClick sites the probability is 95%.
- Using all user data for one month:
 - 90% of men are predicted with $>80\%$ confidence (81% accuracy)
 - 25% of women are predicted with $>80\%$ confidence (96% accuracy)

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[Http://www.moreinfo.com/au.cranlerma/fo12.htm](http://www.moreinfo.com/au.cranlerma/fo12.htm)

What is the value of profiling



No targeting

		Truth	
		Male	Female
Prediction	Male	40%	10%
	Female	25%	25%

		Truth	
		Male	Female
Prediction	Male	\$ 0.50	\$ 0.50
	Female	\$ 0.50	\$ 0.50

Expected Value \$ 0.50

Targeting

		Truth	
		Male	Female
Prediction	Male	40%	10%
	Female	25%	25%

		Truth	
		Male	Female
Prediction	Male	\$ 1.00	\$ 0.20
	Female	\$ 0.20	\$ 0.50

Expected Value \$ 0.60

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Choice Modeling

“The Great Equalizer? The Role of Shopbots in Electronic Markets” by Erik Brynjolfsson and Michael D. Smith (2000)

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Which would you choose?

A Amazon
\$43.90
5-10 days

B Kingsbooks.com
\$32.06
16 days

C 1Bookstreet.com
\$35.96
6-21 days

D Barnesandnoble.com
\$37.19
5-9 days

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Modeling the Choice Decision with a Multinomial Logit

$$\exp\{\text{Value}_A\} = .5 - .2 \text{ Price} - .02 \text{ Delivery} = .0002$$

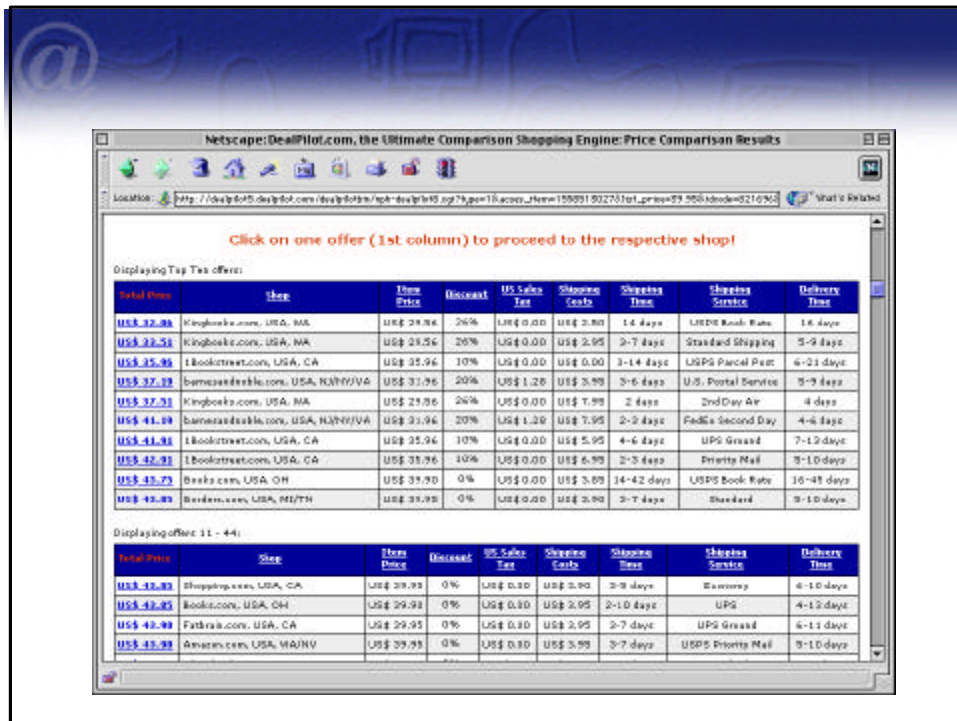
$$\exp\{\text{Value}_B\} = 0 - .2 \text{ Price} - .02 \text{ Delivery} = .0010$$

$$\exp\{\text{Value}_C\} = 0 - .2 \text{ Price} - .02 \text{ Delivery} = .0008$$

$$\exp\{\text{Value}_D\} = .2 - .2 \text{ Price} - .02 \text{ Delivery} = .0003$$

$$\text{Probability of Choosing A} = \frac{e^{.0002}}{e^{.0002} + e^{.001} + e^{.0008} + e^{.0003}} = 7\%$$

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Data

Offer Data: Total Price, Item Price, Shipping Cost, US Sales Tax, Shipping Time, Delivery Time, Delivery "n/a"

Session Data: Date/Time, ISBN, IP Number, Sort Key

Customer Data: Cookie Number, Cookies On (97.1%), *Prior Visits, Prior Choices*

Choice Data: Click-Through, Last Click-Through

Data Statistics

- 69 Days (August 25 - November 1, 1999)
- 1,513,439 Offers
- 39,654 Sessions
- 20,227 Unique Visitors (7,478 Repeat Visitors)

Shopbot Choice Model

<u>Parameter</u>	<u>Estimate</u>	
<i>Total Price</i>		
Item Price	-.19	(\$1.00)
Shipping Price	-.37	(\$1.95)
U.S. Tax	-.43	(\$2.26)
<i>Delivery Average</i>	-.02	(\$.10)
<i>Delivery "n/a"</i>	-.37	(\$1.94)
<i>"Big 3"</i>		
Amazon	.48	(\$2.52)
BarnesandNoble	.17	(\$.89)
Borders	.27	(\$1.42)

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Implications

- Customers approximately twice as sensitive to changes in shipping charges and sales tax as item price
- Price and delivery time are important
 - "Big 3" holds \$1.13 (.284/.252) advantage over unbranded offers
 - Amazon.com holds \$1.85 advantage over unbranded offers
 - BN/Borders hold \$0.72 advantage over unbranded offers
- Shopbot customers highly price sensitive, how generalizable are these results?

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The Power Law of Practice

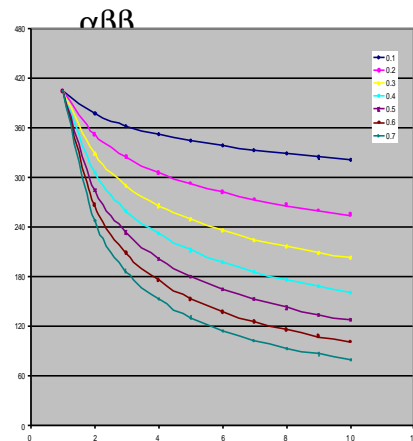
“Cognitive Lock In” by Eric Johnson,
Jerry Lohse, and Steve Bellman (2000)

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Power Law of Practice:

$$T = \beta N^\alpha$$

- Macro ‘empirical regularity’
- Describes increases in performance in many domains.
 - Text editing and other HCI tasks. (Newell and Rosenbloom, 1981)
 - Learning curve for firms(Zangwill & Cantor 1998)



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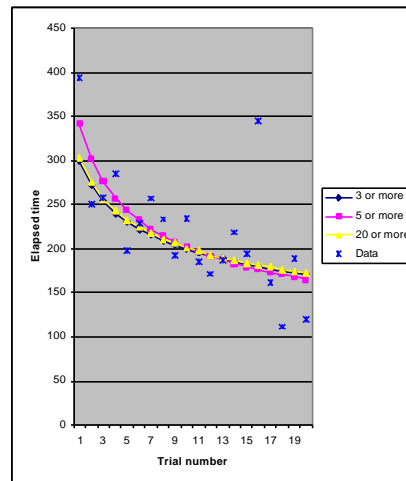
The argument

- By using a site, I learn, lowering the cost of further use.
 - Analogy to learning a supermarket, software,
 - Raises relative switching costs.
- The site (designer) and I coproduce solutions.
- Lower costs → More learning → More loyalty, purchases, etc.

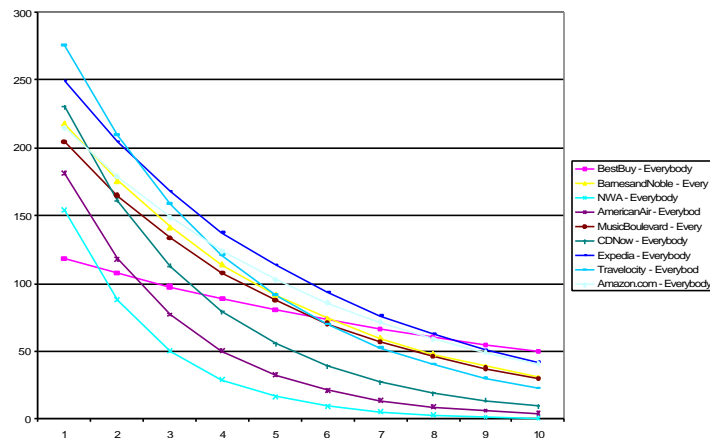
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Illustrative Example: Visits to Amazon

- MediaMetrix: 6 months, all visits.
- Power law provides a good fit $r^2 = .45$
- Note decrease between 1st and 5th visit is about 200 seconds.
- Time savings = \$1.44



Power Law Learning Curves



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Implications

- Rapid site redesign is bad.
 - Refresh content, not navigation.
 - Simpler navigation is asset.
- Raising the cost of price shopping
- Brand (interface) extension may be a viable strategy.

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Using the clickstream to predict purchases

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Problem

- How do we predict (or influence) a purchase decision?
- How much information is there in the path that an individual takes about purchase? Price sensitivity?
- What is an appropriate form of the model?

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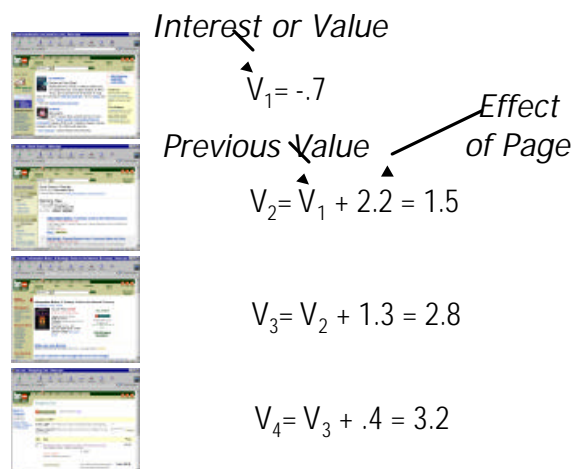
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Motivation

- Assume that our independent variable is a latent measure of interest (e.g., utility) instead of time web used
- Again relate current interest to past values of interest and unpredictable changes
- Also incorporate measures of information and content about a web page
- Finally, assume the coefficients follow a switching model for 'surfing' behavior and 'goal' oriented behavior

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Illustrating the Model



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Review

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Summary

- A foundation for Internet Marketing can be found in One-to-One Marketing
- Information is critical, the most valuable information is purchase history
- New and better measures of the information content of clickstream data and customer acquisition costs are needed

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Interactive Marketing Requires...

- Identify
Clickstream data can be used to identify your visitors
- Differentiate
User Profiling can be used to differentiate them
- Interact
Choice modeling can predict what they want
- Customize
The next step is responding to this information, will explore this in the area of web advertising

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