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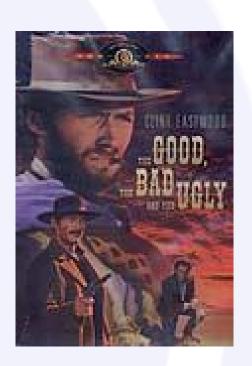
Mining E-commerce Data The Good, the Bad, and the Ugly

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- The Good
 E-commerce is the killer domain for data mining
- The Bad You need more than web logs and you must conflate many data sources
- The Ugly
 Pre-processing and post-processing are hard
- Stories from mining real data "Peeling the onion" on observations to yield insight
- Summary

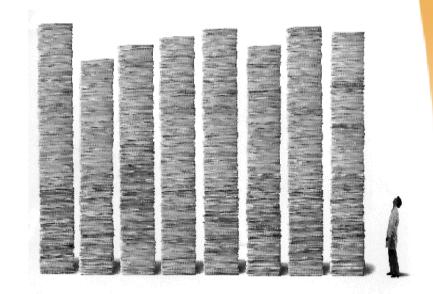
Overview



Successful data mining benefits from:

- Large amount of data (many records)
- Rich data with many attributes (wide records)
- Clean data collection (avoid GIGO)
- Actionable domain (have real-world impact)
- Measurable return-on-investment (did the recipe help)

E-commerce has all the right ingredients





- Clickstreams generate huge amounts of data
- Yahoo! has 1 billion page views a day.
 Web log data for page views is 10GB per hour!
- New e-commerce sites, even if small, generate sufficient data for effective mining quickly If you sell five items an hour on average, that's

5 items * 24 hours * 30 days / 2% conversion * 8 clicks-in-session >

1.4 million page views



Effective site design can log many attributes about what was shown or purchased:

- Product and product attributes
- Assortment attributes (when multiple products are shown)
- Promotions shown

Rich Records

- Visit attributes (e.g., visit count)
- Customer attributes (when known through login/registration)

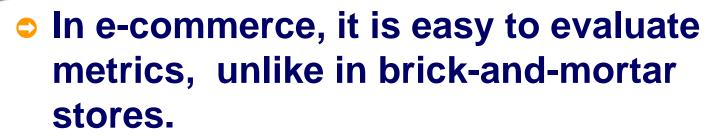
- Collect data directly at webstore No legacy systems
- Collect what is needed by design Not as an afterthought
- Collect electronically reliable data
 No humans typing survey data from forms
- Sample at the right granularity level Architecture design principle: sample at the customer or session level, never at page view level



A bank discovered that almost 5% of their customers were born on the exact same date

Can you explain?

- Few data mining discoveries had a real impact on businesses.
 - Taking action requires changing complex systems, procedures, and human habits HARD in general Easier in the electronics world
- In e-commerce, many discoveries can be made actionable by
 - Changing web sites (e.g., personalization)
 - Targeted campaigns
 - Changing advertising strategies based on ROI
- Easy to offer cross-sells or up-sells Contrast with changing actual store layouts

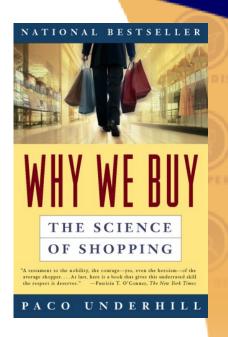


See Why We Buy: the Science of Shopping by Paco Underhill

In e-commerce it is easy to measure the *effect* of changes.

One can easily set control groups on a web site

- Response to e-mails and surveys is days, not weeks and months
- The web is an experimental laboratory
 It is easy to change and measure the effect



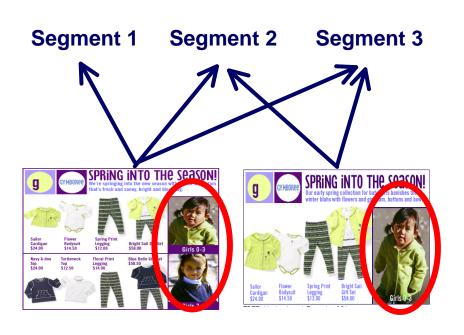
Measure ROI

E-mail campaigns: Immediate ROI



GYMBOREE

One of our customers, Gymboree, sent e-mail campaign based on analysis of website data of registered users: 7 email designs to 4 segments



Results:

- Very high clickthrough rate of 22% (normal is 10%)
- Average order size was 36% higher than normal
- Email with two age groups of the same gender outperformed that with single age (medium targeting)
- Lifestyle images better than products

Firms need web intelligence, not log analysis
-- Forrester Report, Nov 1999

Web logs provide little data, even in the Extended Common Log Format (ECLF)

Host

The Bad

- Time
- Request, e.g., an html page
- Referrer
- User agent (browser identifier)
- IP address
- Cookie
- Bytes, status, ...

What is on the Web Page?

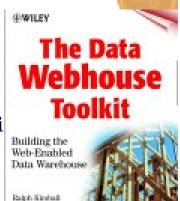


- Weblogs designed for analyzing web servers, not for mining e-commerce transactions and clickstreams
- Given a URL, what was displayed?
 - Reverse URL mapping. Very brittle.

http://www.amazon.com/exec/obi dos/ASIN 41580/105-9856660-9155942

is The Data WebHouse Toolkit

N 47 3768 9/qi



Blichward More

Hard to derive attributes of the product, such as soft cover, author, edition, year?

Dynamic content, which is becoming more common makes web log analysis harder

- **The same URL will display different items**
- URLs are amazingly long in dynamic sites and information is in the application server session:

http://www.im.aa.com/American?BV_EngineID=dealikcjfekgbfdmcflmcfkhdgfh.7 &BV_Operation=Dyn_RawSmartLink&BV_SessionID=%40%40%40%40%4008226 17159.0968100982%40%40%40%40%40%form%25destination=index-member.tmpl&BV_ServiceName=American

 Personalized content (e.g., recommended cross-sell) is practically impossible to reconstruct from web logs

- HTTP is stateless
- Sessionizing is still a research topic

 Measuring the Accuracy of Sessionizers for Web Usage Analysis Berent,

 Mobasher, Spiliopoulou, and Wiltshire, in Proceedings of the Web Mining

 Workshop at the First SIAM International Conference on Data Mining, 2001
- Recreating user sessions is heuristic based:
 - IP addresses
 - Cookies
 - Browser type

Some events cannot be determined from weblogs:

- Add to shopping cart needed to compute value of abandoned shopping carts
- Change quantity of item in cart
- Promotion offered on page
- Out of stock shown on the page
- Dynamically constructed media (e.g., Flash)
- Search common keywords or keywords that were not found

 (an important warning to an e-commerce site)

- On one of our sport-related sites, the top searched keywords were:
 - Baseball
 - Video
 - Softball
 - Volleyball
 - Pins
 - Equestrian
 - Videos
 - Posters
 - Music
 - Poster

What is common to the words in red?

- On one of our sports-related sites, the top searched keywords (in order) were:
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 - Video
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 - Equestrian
 - Videos
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 - Music
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Searches for red words yielded zero results!

- Some words just need a synonym
- Some words should send a strong message about items the store should carry!

Weblogs do not typically contain sufficient information to extract failed searches.

This isn't fancy analytics, but it's crucial.

About 11% of searches fail

Matching Web Logs to DB

- Given a request, how do you
 - Match it to the customer in your database that filled a registration form?
 - Determine if this is the customer's second visit or the 100th visit?
 - Determine if the customer previously purchased?
- These common requests are very hard to implement as an afterthought
- They are even harder when you try to find "scenarios" that match multiple events

Using hits and page views to judge site success is like evaluating a musical performance by its volume -- Forrester Report, 1999

- Most often-requested measures relate to conversion rates (buyers to browsers)
- Especially useful by referrer (e.g., ad)
- Given an HTTP request that has one of your ads as the referrer field, how can you tell if it resulted in a sale?

 On one of our sites, we saw the following in their initial rampup period

Referrer	# Sessions	% of traffic	# Sales	Conv rate
ShopNow	16,178	6.9%	6	0.04%
FashionMall	19,685	8.4%	17	0.09%
MyCoupons	2,052	0.9%	170	8.28%

- Conversion rates differ by a factor of over 200!
- Knowing the likelihood of purchase dramatically changes the message to present

- RYICE Bad" IS Notal Stone Bad

 STEMS O- MARKETING WEB STORE
 - Ignore web logs They are at the wrong granularity level to be useful
 - Log the information yourself at the application layer
 - The application knows what is on the page
 - The app controls sessions
 - The app can log business events
 - The app can tie a visitor to their customer information upon login
 - Also see Structure and Content Preprocessing by Rob Cooley for more information

- There are several hard problems:
 - Crawlers
 - Handling large amounts of data (previously mentioned)
 - Data transformations for analysis
 - Marketing-level insight
- These are excellent research topics



- Search crawlers
- Shopping bots

Crawlers

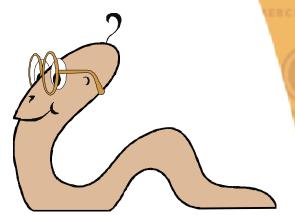
- **○** IE5 offline viewer
- E-mail harvesters Evil
- Students learning Perl scripts



30% of sessions come from bots/crawlers (most are measure of service bots such as Keynote)

Good

Fairly hard problem
 Some try to hide themselves



 80% of the time spent in data analysis is typically spent transforming data

- What can be done today:
 - Automate transfer of data from webstore environment to data warehouse
 - Provide data transformation UI
 - Provided "canned" transformations for common business problems
- Business users without "data" or "analyst" in their title cannot spend the time to learn how to transform data

- Everything is a GO!
 - You collected data correctly
 - You built a data warehouse
 - You transformed the data
 - You ran a simple Perceptron
 (1-layer) neural network that
 predicts the target well



What does the 237-dimensional hyperplane represent?

Insight must be comprehensible to biz users
 Sometimes required for legal reasons
 (e.g., no discrimination)

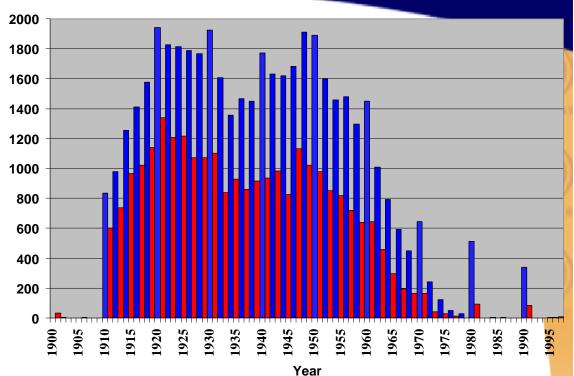




- **Teasers are all real-world example**
- Data miners have to face surprising observations
- **Example from this conference, PAKDD 2001**
- The Kowloon Shangri-La employees change eight carpets every midnight
- Which carpets and why?



The KDD CUP 98 data contained anomalies for date of birth [Georges and Milley, SIGKDD Explorations 2000]



- Spikes on years ending in zero (white dots on blue)
- **Solution** Few individuals born prior to 1910
- Many more individuals who were born on even years (blue) as on odd years (red)

Why?



- A site has gender on the registration form
- Acxiom, a syndicated data provider, also provides gender
- A very large discrepancy found between
 - Males according to registration form and
 - Acxiom provided data

Why?

Hint: Acxiom only conflicted with females, claiming some females are males.

Never in the other direction





Recall that Conversion Rate is the ratio of buyers to browsers.

- High conversion rates are desired
- Reports showed some products have really low conversion rates?

Why?



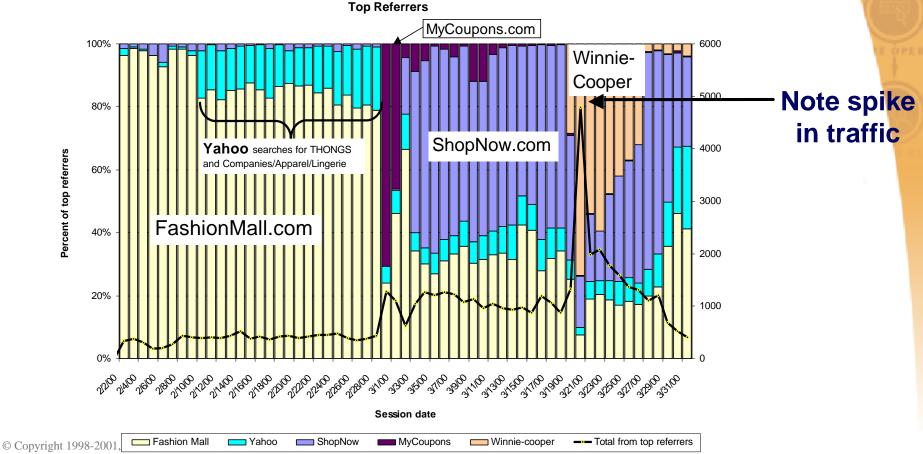


- Product Conversion Rate is the ratio of product purchases to product views
- High can conversion rates be over 100%

Teaser - Who is Winnie?



Referring site traffic for Gazelle.com, a leg-wear and leg-care web retailer. From KDD Cup 2000 Who is Winnie Cooper? What can you do about it?



- Winnie-cooper is a 31 year old guy who wears pantyhose
- He has a pantyhose site
- 8,700 visitors came from his site in a few days (!)
- Actions:
 - Make him a celebrity and interview him about how hard it is for a men to buy pantyhose in stores
 - Personalize for XL sizes



- KDNuggets, Software for Web Mining http://www.kdnuggets.com/software/web.html
- WEBKDD Workshops in Web Mining http://robotics.Stanford.EDU/~ronnyk/WEBKDD2000/index.html http://robotics.Stanford.EDU/~ronnyk/WEBKDD2001/index.html
- WEB Mining Tutorials

Resources (I)

- ► E-commerce and Clickstream Mining, Jon Becher and Ron Kohavi, First SIAM International Conference on Data Mining, 2001 http://robotics.Stanford.EDU/~ronnyk/miningTutorialSlides.pdf
- Web Mining for E-Commerce, Jaideep Srivastava,
 The Fifth Pacific Asia Conference on Knowledge Discovery and Data Mining, 2001

Mastering

- The Data Webhouse Toolkit: Building the Web-Enabled Data Warehouse by Ralph Kimball, Richard Merz. ISBN: 0471376809 (Jan 2000)
- Mastering Data Mining: The Art and Science of Customer Relationship Management by Michael J. A. Berry, Gordon Linoff. ISBN: 0471331236
- The Data Mining and Knowledge Discovery special issue on Application of Data Mining to Electronic Commerce (volume 5, 1/2) January/April 2001. Special issue:

http://www.wkap.nl/issuetoc.htm/1384-5810+5+1/2+2001

Book ISBN: 0792373030

http://www.amazon.com/exec/obidos/ASIN/0792373030



- Web Mining Research: A Survey
 http://www.acm.org/sigs/sigkdd/explorations/issue2-1/contents.htm#Kosala
- Web Data Mining course at DePaul University by Bamshad Mobasher http://maya.cs.depaul.edu/~classes/cs589/lecture.html
- Integrating E-commerce and Data Mining: Architecture and Challenges, WEBKDD'2000 http://robotics.Stanford.EDU/~ronnyk/ronnyk-bib.html
- Drinking from the Firehose: Converting Raw Web Traffic and E-Commerce Data Streams for Data Mining and Marketing Analysis by Rob Cooley http://www.webusagemining.com/sys-tmpl/webdataminingworkshop/



- Analyzing Web Site Traffic, Sane Solutions http://www.sane.com/products/NetTracker/whitepaper.pdf
- Web Mining, Accrue Software http://www.accrue.com/forms/webmining.html

Resources (IV)

Direct effect of web on established retailers may not be large, but lessons learned will affect other channels, so additional ROI comes from improvements to other channels

The webstore provides an experimental laboratory

and a trend-discovery system

Which cross-sells work?

Which ads are effective?

What are people looking for (failed searches for pokédex)

B2C E-Commerce 1999

Amazon 1999



\$20.2 B

Wal-Mart
1999 revenues:
\$162.8 B

- Good: E-commerce is the killer-domain for data mining with all the right ingredients
- Bad: Good data collection is hard
 - Web logs are information poor
 - New sites should log clickstream and events in the app
 - Existing sites should extract data from HTML traffic (e.g., sniffer packages). Plan to upgrade to a better architecture

Ugly:

- Data transformations take longer than you expect.
- You must "peel the onion" for interesting insight (see KDD CUP 2000 http://www.ecn.purdue.edu/KDD

Take Home Messages (III)



Always involve the business user

Many "interesting" discoveries turn out to be a result of

some intentional activity. "Peel the onion."

- Business users want simple, comprehensible results
 - Reports are not glamorous but most often needed
 - Simple algorithms are most useful especially if coupled with good visualizations
- The web is a measurement and experiments lab
 - Half the discoveries will carry over to the "real world"