

Introduction to Text Mining and Web Search



Some slides are borrowed from Prof. Marti Hearst, Christopher Manning, Louis Eisenberg, Bing Liu, and Prabhakar Raghavan

Objectives

- To have a rough idea of text mining and web search
 - Not about deep techniques
- Connections with StreamSpin
 - Mobile services as component of SS

Corporate Knowledge "Ore"

Stuff not very accessible via standard data-mining

- Email
- Insurance claims
- News articles
- Web pages
- Patent portfolios
- IRC
- Scientific articles

- Customer complaint letters
- Contracts
- Transcripts of phone calls with customers
- Technical documents

Definitions of Text Mining

- Text mining mainly is about somehow extracting the information and knowledge from text;
- 2 definitions:
 - Any operation related to gathering and analyzing text from external sources for business intelligence purposes;
 - Discovery of knowledge previously unknown to the user in text;
- Text mining is the process of compiling, organizing, and analyzing large document collections to support the delivery of targeted types of information to analysts and decision makers and to discover relationships between related facts that span wide domains of inquiry.

Text Mining

- Text classification
- Text clustering
- Named entity recognition
- Information extraction
- Information retrieval engine
- Web spider/search
- Question answering
- Opinion mining
- Summarization
- Topic detection and tracking
- ...

Text Classification (Categorization) Clustering

Text Classification and its application Clustering and its application

Is this spam?

From: "" <takworlld@hotmail.com> Subject: real estate is the only way... gem oalvgkay

Anyone can buy real estate with no money down

Stop paying rent TODAY !

There is no need to spend hundreds or even thousands for similar courses

I am 22 years old and I have already purchased 6 properties using the methods outlined in this truly INCREDIBLE ebook.

Change your life NOW !

Click Below to order: http://www.wholesaledaily.com/sales/nmd.htm

More Examples

Assign labels to each document or web-page:

- Labels are most often topics such as Yahoo-categories e.g., "finance," "sports," "news>world>asia>business"
- Labels may be opinion
 e.g., "like", "hate", "neutral"
- Labels may be domain-specific binary

 e.g., "interesting-to-me": "not-interesting-to-me"
 e.g., "spam": "not-spam"
 e.g., "contains adult language": "doesn't"
- Labels may be genres

e.g., "editorials" "movie-reviews" "news"

Classification

- Training data
 - A description of an instance by a set of features.
 - Issue: how to represent text documents.
 - Feature selection
 - A fixed set of categories
 - Build classification model
- Apply classification model to new data instance described by a set of features

Classification model

- K nearest neighbor
- Naïve Bayesian model
- Rule based model
 - Decision tree
 - Association rule based model
- Support Vector Machine
- Neural Network

Yiming Yang & Xin Liu, A re-examination of text categorization methods. *Proceedings of SIGIR*, 1999

[•]

Clustering

- Clustering: the process of grouping a set of objects into classes of similar objects
 - unsupervised learning: no training data
 - A common and important task that finds many applications in IR and other places
 - Whole corpus analysis/navigation
 - Better user interface
 - For improving recall in search applications
 - Better search results

For better navigation of search results

- For grouping search results thematically
 - clusty.com / Vivisimo

Clusty Web+ News	Images Shopping Encyclopedia Gossip Customize!	
Beta text clustering	Cluster	
Advanced Search Help Tell a Friend Tell us what you think! Download the Clusty Toolbar		
Cluster by: Topics 💌	Top 187 results of at least 251,221 retrieved for the query <u>text clustering</u> (<u>Details</u>)	
text clustering (187) Mining (30) Gene (19) Clustering algorithms (13) Text Document Clustering (8) Receptor (13)	 Apple 64-bit Xserve G5 - Clustering A High-density, affordable server with unparalleled functionality and usability. Featuring massive processing power, high-speed I/O and remote management tools that make it easy to deploy. www.apple.com Custom Configured Clustering Solutions A spen Systems custom designs, manufactures and supports Beowulf Clusters. www.aspsys.com 	
 Lab (6) Model, Self-Organising Hybrid (9) Ontology-based Text Clustering (6) Text Categorization (7) 	Search Res 1. Delphion: Text Clustering Image: Search Res Citation Link » Clustering « » PatentLab-II » PDF Express » Data Extract » Other Services Text Clustering Overview Delphion's Text Clustering transforms obscure, textual information into usefi www.delphion.com/products/research/products-cluster - Wisenut, GigaBlast, Lycos, MSN 2. Vivisimo Clustering - automatic categorization and meta-search software - Image: Search Clustering and meta-search software - Vivisimo's document clustering and meta-search softwarePublisherâ.g: Cluster larger text collections Â. Â. Â. Â. fv organization with document clustering	

Applications and Models of Clustering

- Examples of Clustering Applications
 - Marketing: Help marketers discover distinct groups in their customer bases, and then use this knowledge to develop targeted marketing programs
 - Insurance: Identifying groups of motor insurance policy holders with a high average claim cost
 - City-planning: Identifying groups of houses according to their house type, value, and geographical location
- Clustering models
 - Partitioning Methods
 - Hierarchical Methods
 - Density-Based Methods
 - Grid-Based Methods
 - Model-Based Clustering Methods

WEB SEARCH

Main components of search engines Advertisement and search engineer



The Web

Content aggregators

Content consumers

Brief (non-technical) history

- Early keyword-based engines
 - Altavista, Excite, Infoseek, Inktomi, Lycos, ca.
 1995-1997
- <u>Paid placement</u> ranking: Goto.com (morphed into Overture.com \rightarrow Yahoo!)
 - Your search ranking depended on how much you paid
 - Auction for keywords: <u>casino</u> was expensive!

Brief (non-technical) history

- 1998+: Link-based ranking pioneered by Google
 - Blew away all early engines save Inktomi
 - Great user experience in search of a business model
 - Meanwhile Goto/Overture's annual revenues were nearing \$1 billion
- Result: Google added paid-placement "ads" to the side, independent of search results
 - Internet providers to create content-rich <u>broadband</u> services ; 2003: Yahoo follows suit, acquiring Overture (for paid placement) and Inktomi (for search);

Ads vs. search results

 Google has maintained that ads (based on vendors bidding for keywords) do not affect other rankings in search results

Sponsored Links

Fly to Aalborg

Aalborg Flights with a flight site for the right flight. www.cheapflightsite.co.uk

Studying In Denmark

The Official Guide to studying in Denmark. www.Denmark.dk/Study

Search = study in aalborg

Web Aalborg University: Study Administration - Aalborg University Aalborg University The Study Administration (Studieforvaltningen) Fredrik Bajers Vej 5 DK -9220 Aalborg East Denmark. Please note that The International ... en.aau.dk/About+Aalborg+University/University+Structure/Administration/Study+Administration - 8k - Cached - Similar pages Study Abroad - International Students - Aalborg Universitet (AAU) f you are enrolled at Aalborg University as a student and are interested in a study period abroad, Ms. Mariann Simonsen at the International Office is ... studyguide.aau.dk/enrolledstudents/studyabroad - 7k - Cached - Similar pages PAU Home: New in DK? - Study Cards - Aalborg University f not already received, you can get a PhD study card from the International Doctoral School. This will prove that you are a PhD student at Aalborg ... www.pau.aau.dk/newindk/1041744 - 7k - Cached - Similar pages [PDF] Aalborg brochure #2 File Format: PDF/Adobe Acrobat - View as HTML ar program of study at Aalborg. Foreign language skills, are only required for students who wish to study in a. foreign language. ...

studyabroad.uoregon.edu/brochure/aalborg.pdf - Similar pages

Ads vs. search results

- Other vendors (Yahoo!, Live Search) have made similar statements from time to time
 Any of them can change anytime
- We will focus primarily on search results independent of paid placement ads
 - Although the latter is a fascinating technical subject in itself
 - So, we'll look at it briefly later

Web search basics



How Search Engines Work

Three main parts:

Crawler: Gather the contents of all web pages (using a program called a crawler or spider)
Indexer: Organize the contents of the pages in a way that allows efficient retrieval (indexing)
Ranker: Take in a query, determine which pages match, and show the results (ranking and display of results)

Crawler

Crawling Issues

How to crawl?

- Quality: "Best" pages first
- *Efficiency*: Avoid duplication (or near duplication)
- Etiquette: Robots.txt, Server load concerns
- How much to crawl? What to crawl?
 - *Coverage*: How big is the Web? How much do we cover?
 - Relative Coverage: How much do competitors have?

• How often to crawl?

- Freshness: How much has changed?
- Pages change (25%,7% large changes)
 - At different frequencies

Basic crawler algorithm

- Begin with known "seed" pages (on a queue)
- Fetch and parse them
 - Extract URLs they point to
 - Place the extracted URLs on a queue
- Fetch each URL on the queue and repeat

Four Laws of Crawling

- A Crawler must show identification
- A Crawler must obey the robots exclusion standard
 - <u>http://www.robotstxt.org/wc/norobots.html</u>
 - For a URL, create a file URL/robots.txt
- A Crawler must not hog resources
 Politeness don't hit a server too often
- A Crawler must report errors

Example robots.txt file

www.whitehouse.gov/robots.txt (just the first few lines)

User-agent:	*
Disallow:	/cgi-bin
Disallow:	/search
Disallow:	/query.html
Disallow:	/help
Disallow:	/360pics/text
Disallow:	/911/911day/text
Disallow:	/911/heroes/text
Disallow:	/911/messages/text
Disallow:	/911/patriotism/text
Disallow:	/911/patriotism2/text
Disallow:	/911/progress/text
Disallow:	/911/remembrance/text
Disallow:	/911/response/text
Disallow:	/911/sept112002/text
Disallow:	/911/text
Disallow:	/ConferenceAmericas/text
Disallow:	/GOVERNMENT/text
Disallow:	/QA-test/text
Disallow:	/aci/text
Disallow:	/afac/text
Disallow:	/africanamerican/text
Disallow:	/africanamericanhistory/text
Disallow:	/agencycontact/text
Disallow:	/americancompetitiveness/text
Disallow:	/apec/2003/text
Disallow:	/apec/2004-summit/text
Disallow:	/apec/2004/text
- · · ·	

What really gets crawled?

- A small fraction of the Web that search engines know about; no search engine is exhaustive
- Not the "live" Web, but the search engine's index
- Not the "Deep Web", but start to do this
 - E.g., buy some stuff in Amazon, password protected
 - In 2000, a research project by Berkeley: 91,000
 <u>terabytes</u>. By contrast, the surface Web (which is easily reached by search engines) is only about 167
 terabytes
- Mostly HTML pages but other file types too: PDF, Word, PPT, etc.

Lots of tricky aspects

- Servers are often down or slow
- Hyperlinks can get the crawler into cycles
- Some websites have spam in the web pages
- Now many pages have dynamic content
 - E.g. javascript, the deep web

Lots of tricky aspects

- The web is **HUGE**
 - Distributed crawling
 - Crawl order
 - Most sites, stay at \leq 5 levels of URL hierarchy
 - Which URLs are most promising for building a high-quality corpus
 - Filtering duplicates, and Mirror detection
 - Fetch from the fastest, inlink and outlink
 - Malicious pages, Spider traps

Indexer

Index (the database)

Record information about each page

- List of words
 - In the title?
 - How far down in the page?
 - Was the word in boldface?
- URLs of pages pointing to this one
- Anchor text on pages pointing to this one

– <u>AAU DB</u>

Inverted Index

- How to store the words for fast lookup
- Basic steps:
 - Make a "dictionary" of all the words in all of the web pages
 - For each word, list all the documents it occurs in.
 - Often omit very common words
 - "stop words"
 - Sometimes **stem** the words
 - (also called morphological analysis)
 - cats -> cat
 - running -> run

Inverted Index Example



Query processing: AND

• Consider processing the query:

Computer AND science

- Locate *computer* in the Dictionary;
 - Retrieve its postings.
- Locate science in the Dictionary;
 - Retrieve its postings.



The merge

 Walk through the two postings simultaneously, in time linear in the total number of postings entries



If the list lengths are x and y, the merge takes O(x+y) operations. Crucial: postings sorted by docID.

Inverted Index

- In reality, this index is HUGE
- Need to store the contents across many machines
- Need to do optimization tricks to make lookup fast.
Discussion: what information is missing from the simple index?

- Frequency
 - Party in Aalborg
- Words are the same important?
 Introduction to XML
- Location, in title, in anchor text
- Proximity
- •

term frequency: tf

- tf the frequency of a tem in a page
- Weighting tf is the relative importance of
 - 0 vs. 1 occurrence of a term in a doc
 - 1 vs. 2 occurrences
 - 2 vs. 3 occurrences ...
- Unclear: while it seems that more is better, a lot isn't proportionally better than a few
 - Can just use raw tf
 - Another option commonly used in practice:

$$wf_{t,d} = 0$$
 if $tf_{t,d} = 0$, $1 + \log tf_{t,d}$ otherwise

Score computation

• Score for a query q = sum over terms t in q:

$$= \sum_{t \in q} t f_{t,d}$$

- [Note: 0 if no query terms in document]
- This score can be zone-combined
- Can use *wf* instead of *tf* in the above
- Still doesn't consider term scarcity in collection (*ides* is rarer than *of*)

Weighting should depend on the term overall

- Which of these tells you more about a doc?
 - 10 occurrences of *house*?
 - 10 occurrences of *the*?
- Would like to attenuate the weight of a common term
 - But what is "common"?

Document frequency

• *df* = number of docs in the corpus containing the term, cf = number of occurrences in a collection

Word	cf	df
ferrari	10422	17
insurance	10440	3997

- Document/collection frequency weighting is only possible in known (static) collection.
- So how do we make use of *df* ?

tf x idf term weights

- tf x idf measure combines:
 - term frequency (tf)
 - or *wf*, some measure of term density in a doc
 - inverse document frequency (*idf*)
 - measure of informativeness of a term: its rarity across the whole corpus
 - could just be raw count of number of documents the term occurs in (*idf_i* = 1/*df_i*)
 - but by far the most commonly used version is:

$$idf_i = \log\left(\frac{n}{df_i}\right)$$

• See Kishore Papineni, NAACL 2, 2002 for theoretical justification

Ranker

Results ranking

- Search engine receives a query, then
- Looks up the words in the index, retrieves many documents, then
- Rank orders the pages and extracts "snippets" or summaries containing query words.
 - Most web search engines assume the user wants all of the words (Boolean AND, not OR).
- These are complex and highly guarded algorithms unique to each search engine.

Some ranking criteria

- For a given candidate result page, use:
 - Number of matching query words in the page
 - Proximity of matching words to one another
 - Location of terms within the page
 - Location of terms within tags e.g. <title>, <h1>, link text, body text
 - Anchor text on pages pointing to this one
 - Frequency of terms on the page and in general
 - Link analysis of which pages point to this one
 - (Sometimes) Click-through analysis: how often the page is clicked on
 - How "fresh" is the page
- Complex formulae combine these together. How ?

One possible ranking

- First retrieve all pages meeting the text query (say *venture capital*).
- Order these by their link popularity (either variant on the previous page).

Link analysis

- First generation: using link counts as simple measures of popularity.
- Two basic suggestions:
 - <u>Undirected popularity:</u>
 - Each page gets a score = the number of in-links plus the number of out-links (3+2=5).
 - <u>Directed popularity:</u>
 - Score of a page = number of its in-links (3).



Spamming simple popularity

- *Discussion*: How do you spam each of the following heuristics so your page gets a high score?
- Each page gets a score = the number of inlinks plus the number of out-links.
- Score of a page = number of its in-links.

Measuring Importance of Linking

- PageRank Algorithm
 - Idea: important pages are pointed to by other important pages
 - Method:
 - Each link from one page to another is counted as a "vote" for the destination page
 - But the importance of the starting page also influences the importance of the destination page.
 - And those pages scores, in turn, depend on those linking to them.

Pagerank scoring

 Imagine a browser doing a random walk on web pages:

– Start at a random page



- At each step, go out of the current page along one of the links on that page, equiprobably
- "In the steady state" each page has a longterm visit rate - use this as the page's score.

Not quite enough

- The web is full of dead-ends.
 - Random walk can get stuck in dead-ends.
 - Makes no sense to talk about long-term visit rates.



Teleporting

- At a dead end, jump to a random web page.
- At any non-dead end, with probability 10%, jump to a random web page.
 - With remaining probability (90%), go out on a random link.
 - 10% a parameter.

Pagerank



Machine Learned Ranking

- Goal: Automatically construct a ranking function
 - Input:
 - Large number training examples
 - Features that predict relevance
 - Relevance metrics
 - Output:
 - Ranking function
- Enables rapid experimental cycle
 - Scientific investigation of
 - Modifications to existing features
 - New feature

Ranking Features

- A0 A4 anchor text score per term
- W0 W4 term weights
- L0 L4 first occurrence location
 - (encodes hostname and title match)
- SP spam index: logistic regression of 85 spam filter variables (against relevance scores)
- F0 F4 term occurrence frequency within document
- DCLN document length (tokens)
- ER Eigenrank
- HB Extra-host unique inlink count
- ERHB ER*HB
- A0W0 etc. A0*W0
- QA Site factor –
 logistic regression of 5 site link and url count ratios
- SPN Proximity
- FF family friendly rating
- UD url depth

(from Jan Pedersen's lecture)

Ranking Decision Tree



Advertisement and Manipulating ranking

advertisement

- How does SE sell keyword to advertisers?
 - One keyword can be sold to multiple advertisers
 - The more you bid, the more chance that it appears
 - It is not true. It is secret to rank ads.
 - An optimization problem to maximize its revenue
 - Bid, relevance to query, click rate

advertisement

- Where to place advertisement?
 - Adword: Appear in the sidebar
 - Adsense: Appear in the other web pages
 - Apply to google; a good website, e.g. blog; is related to a keyword?
 - Google then sends a script to insert ad to the webpages
 - Google shares revenue with host providers
 - How to decide if a page is related to a search keyword?
 - How to detect spam click to advertisement?
 - The quality

Trademarks and paid placement

- Consider searching Google for *geico*
 - Geico is a large insurance company that offers car insurance
- Sponsored Links

Car Insurance Quotes

Compare rates and get quotes from top car insurance providers. www.dmv.org

It's Only Me, Dave Pell

I'm taking advantage of a popular case instead of earning my traffic. www.davenetics.com

Fast Car Insurance Quote

21st covers you immediately. Get fast online quote now! <u>www.21st.com</u>

Who has the rights to your name?

- Geico sued Google, contending that it owned the trademark "Geico" – thus ads for the keyword *geico* couldn't be sold to others
 - Unlikely the writers of the constitution contemplated this issue
- Courts recently ruled: search engines can sell keywords including trademarks
 - <u>Personal names</u>, too
- No court ruling yet: whether the ad itself can use the trademarked word(s) e.g., *geico*

The trouble with paid placement

- It costs money. What's the alternative?
- <u>Search Engine Optimization:</u>
 - "Tuning" your web page to rank highly in the search results for select keywords
 - Alternative to paying for placement
 - Thus, intrinsically a marketing function
 - Also known as Search Engine Marketing
- Performed by companies, webmasters and consultants ("Search engine optimizers") for their clients

Simplest forms

- Early engines relied on the density of terms
 - The top-ranked pages for the query *aalborg resort* were the ones containing the most *aalborg*'s and *resort*'s
- SEOs responded with dense repetitions of chosen terms
 - -e.g., aalborg resort aalborg resort aalborg resort
 - Often, the repetitions would be in the same color as the background of the web page
 - Repeated terms got indexed by crawlers
 - But not visible to humans on browsers

Can't trust the words on a web page, for ranking.

Variants of keyword stuffing

• Misleading meta-tags, excessive repetition

Meta-Tags =

"... London hotels, hotel, holiday inn, hilton, discount, booking, reservation, sex, mp3, britney spears, viagra, ..."

Search engine optimization (Spam)

- Motives
 - Commercial, political, religious, lobbies
 - Promotion funded by advertising budget
- Operators
 - Contractors (Search Engine Optimizers) for lobbies, companies
 - Web masters
 - Hosting services
- Forum
 - Web master world (<u>www.webmasterworld.com</u>)
 - Search engine specific tricks
 - Discussions about academic papers ^(C)
 - More pointers in the Resources

More spam techniques

Cloaking

-Serve fake content to search engine spider

-DNS cloaking: Switch IP address. Impersonate



FAQ: Cloaking & Stealth Technology

Tutorial: Cloaking and Stealth Technology

Featured as an ongoing multi part section newsletter, we are offering you all the stuff you to know, straight from the horse's mou Learn the secrets of the pros – subscriptic terminated anytime you wish.

"Stealth, Cloaking, Phantom Tech

FAQ		
 <u>What are Ghost</u> Pages? 	 What are the mechanics or 	
<u>What are Doorway</u> <u>Pages, then?</u>	cloaking? • <u>What's a key</u>	
And Hallway Pages?	switch? switch?	
How are cloaked pages submitted?	simple redire technique?	
How about changing stealth pages?	 <u>VVhat about</u> penalization? 	

fanto spide

Tutorial on Cloaking & Stealth Technology

Don't risk nasty surprises from spiders sneaking on your site under wraps!

Sure, they tend to add and switch engines, IPs and User Agents almost all the time, and keeping up with their antics is a grueling task at best.

But it's also a fact that professional traffic evaluation, stealthing technology and even page submission management depend on reliable search engine reference data, if you don't want to waste your valuable resources on inventing the wheel over and over

And consider the risks: one single unrecognized spider crawling your doorways or debunkig your stealth pages, and your top ranking with that engine may be gone for keeps! If they don't har you from page submissions

More spam techniques

• Doorway pages

 Pages optimized for a single keyword that re-direct to the real target page

• Link spamming

- Mutual admiration societies, hidden links, awards
 more on these later
- Domain flooding: numerous domains that point or re-direct to a target page

•

Acid test

- Which SEO's rank highly on the query *seo*?
- Web search engines have policies on SEO practices they tolerate/block

– See pointers in Resources

- Adversarial IR: the unending (technical) battle between SEO's and web search engines
- See for instance <u>http://airweb.cse.lehigh.edu/</u>

Question Answering

Transitional QA Community based QA

Can google find answers?

- Question: *How much money did IBM spend on advertising in 2002?*
- Answer: I dunno, but I'd like to ... 🛞




ELECTION 2008 & Yahoo! Answers



How can we engage more people in the democratic process?

I will be asking questions to help create dialogue around this and many other important topics so please add me to

Network so that we can begin exchanging befully make changes that will benefit the

etails)

ers Staff note: Yahoo! Answers is a forum for II over the world to engage with one another prmation on topics that interest them. This is ement. We are not siding with any candidate eneral or for the 2008 US elections. We're eople from all perspectives will realize the that the Answers community can have, and for future discussions.

ously impressed by the thoughtfulness of the my question. There is, of course, no single gage more people in the democratic eve we all need to do more work on nore people to participate in the process and ery community is included in the discussion. Indreds of worthy and thought-provoking fact, some were so intriguing that I arranged onally with their creators. To hear my with one user, who wrote about the parallels y and democracy, click below.

no.com/video/play?vid=74... 7382 answers - Report Abuse





Hillary Clinton

OFFICIAL

John McCain

What would you do to stop wasteful government spending in Washington?

Additional Details

6 months ago

Yahoo! Answers Staff Note: Watch a video of John McCain discussing government spending.

http://video.yahoo.com/video/play?vid=f3...

6 months ago

Yahoo! Answers Staff note: Yahoo! Answers is a forum for people from all over the world to engage with one another and to find information on topics that interest them. This is not an endorsement. We are not siding with any candidate or party -- in general or for the 2008 US elections. We're hopeful that people from all perspectives will realize the great insights that the Answers community can have, and will turn to us for future discussions.

5 months ago

Yahoo! Answers staff note: Read Sen. John McCain's own response to the Yahoo! Answers community.

http://blog.360_vaboo_com/blog-d8ph0dcor... 6 months ago_16713 answers_Report Abuse Enter a search term and click on the links under "Search Feature" in the table below to find specific information about those features, or click on the links in the "Sample Query" column to view sample search results.

help q&a	Send
----------	------

Personalization	
Change personalization setting	save on / save off / clear
Set / view / clear your location	set location 10012 / view location / clear location
Search Queries Personalized*	pizza / weather / movies

Search Feature	Sample Query
Local	<u>sushi 94040</u>
Weather	weather boston
Glossary	define zenith
Sports **	score red sox
Movies	movies 94110
Stocks	stock tot
Zip Codes	zip code 72202
Directions	directions pasadena ca to 94043
Maps	map 5th avenue new york
Flights ***	flight aa 2111
Area Codes	area code 650
Products	price ipod player 40qb
<u>Q&A</u>	abraham lincoln birthday
Airlines ***	united airlines
Translation	translate hello in french
Web Snippets	web hubble telescope
Calculator	<u>1 us pint in liters</u>
Currency Conversion	8 usd in yen



Typical TREC QA Pipeline

"A simple factoid question"



Process limited types of questions

CURRENT Community QA ARCHITECTURE



PROPOSED CQA ARCHITECTURE





Problem: find questions and answers in a forum thread

PROBLEM AND MOTIVATIONS

- Problem: Find questions, context and answers in forums
- Extract knowledge in QA form from forums
 - Large amount of existing QA pairs in forums
 - TripAdvisor: ~3M QA pairs (Jan 2008)
 - Yahoo! Answer Travel: ~700K (Jan 2008)
- Provide effective and efficient search & browsing UI
- Organize and manage forum information
 - Highlight the main content of a thread
 - Index question and answer pairs

MAIN CHALLENGES: QUESTION DETECTION

- Simple rules are not adequate.
 - 30% questions have no '?' (out of 1,000).
 - 9% sentences with '?' are not questions.
- Questions can be expressed in imperative form.
 - "I am wondering where I can buy cheap and good clothing in beijing."
- Question mark is often omitted in forums.
- Short informal expressions, such as "really?", are not questions.

QUESTION DETECTION



LABELED SEQUENTIAL PATTERN(LSP)

- LSP: sequence \rightarrow label
 - E.g. question: *i* want to buy an office software and wonder which software company is best.
 - wonder which...is \rightarrow question
- Two measures of LSP: a labeled sequence database *D* and a LSP *p*
 - Support of p:the percentage of labeled sequences in D that contain p.

$$\sup(p) = \frac{|\{s \mid s \in D, p \subseteq s\}|}{|D|}$$

- Confidence of p: the probability of the sequence of p being true $conf(p) = \frac{sup(p)}{sup(p.sequence)}$
- Mine LSP given minsup and minconf threshold

Labeled sequence database:
S1(Q): a d e f
S2(Q): afef
S3(NQ): d a f

LSP1: < a, e, f > \rightarrow Q S1,S2 contain <a,e,f> \rightarrow Q S1,S2 contain <a,e,f> \rightarrow Q S1,S2 contain <a,e,f> Sup(LSP1)=2/3 Conf(LSP1)=2/2

LSP2: < a, $f > \rightarrow Q$ S1,S2 contain <a, $f > \rightarrow Q$ S1,S2,S3 contain <a,f >Sup(LSP2)=2/3 Conf(LSP2)=2/3

QUESTION DETECTION

Data	Method	$\operatorname{Prec}(\%)$	$\operatorname{Rec}(\%)$	$F_1(\%)$
Q-TUnion	5W-1H words	69.0	14.8	24.4
	Question Mark	96.8	78.4	86.6
	SM [18]	81.9	87.8	84.6
	Our	96.5	98.5	97.5
Q-TInter	5W-1H words	69.0	15.3	25.0
	Question Mark	98.7	77.6	86.9
	SM [18]	92.7	86.8	89.7
	Our	97.8	97.0	97.4

- Ours and SM use Ripper classifier but different set of features
- 2,316 LSPs for questions (1,074 contains "?") and 2,789 for nonquestions

MAIN CHALLENGES

- Answer detection
 - Multiple QA threads overlap and exist in parallel.
 - Reply relationship between posts is not explicit.
 - A post may answer many questions.
 - A question may have multiple answers.

ANSWER DETECTION

- Problem: Given a question and a set of candidate answers, rank
- Benchmark methods
 - Cosine similarity
 - KL divergence language model
 - Query likelihood language model
 - Classification based re-ranking

UNSUPERVISED APPROACH

- Our improvements: Make use of the dependency relationships between answers to improve the benchmark methods
 - if a candidate answer is related to an authoritative candidate answer with high score, the candidate answer is likely to be an answer
 - Build a graph on candidate answers of a question, rerank based on the graph
 - Build graph: using asymmetric generation probability
 - Assign weight:
 - Compute reranking score:

ANSWER DETECTION ON A-TUNION DATA

Method	Question with answer		
	P@1	MRR	MAP
NA	0.644	0.718	0.618
Lex	0.649	0.756	0.721
Cla	0.722	0.818	0.774
CS	0.686	0.789	0.737
QL	0.697	0.791	0.719
KL	0.709	0.809	0.762
G+CS	0.739	0.830	0.784
G+QL	0.761	0.843	0.775
G+KL	0.816	0.882	0.842

• G+KL performs 15% better than KL

OPINION MINING

- Problems & Application
- Approaches

Examples

- Product review mining: What features of the ThinkPad T43 do customers like and which do they dislike?
- **Review classification:** Is a review positive or negative toward the movie?
- Tracking sentiments toward topics over time: Is anger ratcheting up or cooling down?
- Search engines do not search for opinions
 - Opinions are hard to express with a few keywords
 - How do people think of Motorola Cell phones?
 - Current search ranking strategy is not appropriate for
 - opinion retrieval/search

Application

- Businesses and organizations: product and service benchmarking. Market intelligence.
 - Business spends a huge amount of money to find consumer sentiments and opinions.
 - Consultants, surveys and focused groups, etc
- Individuals: interested in other's opinions when
 - Purchasing a product or using a service,
 - Finding opinions on political topics,
- Ads placements: Placing ads in the user-generated content
 - Place an ad when one praises a product.
 - Place an ad from a competitor if one criticizes a product.
- Opinion retrieval/search: providing general search for opinions.



Products

See also: Web, Images, Video, News, Maps, More V

Apple iPhone Smart Phone



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Q

User reviews Product details Exp	orders over \$25 ship free	
<u>All user reviews</u>	General Comments View by: Positive comments (85%) Negative comments (15%)	Amazon.com <u>iphone</u> Find Low Prices and Multip
General Comments (497 comments) 85% positive	The lphone is the best cell phone I had so far and i think that if	On iphone shopping.yahoo.com
Ease Of Use (217 comments)	you buy it you will love it. <u>More</u> ruben_93 catalog.ebay.com 8/23/2008	By Iphone Explore 5,000+ Cell Phone: Save On By Iphone!
Features (132 comments) 93% positive	I have decided to buy it because I like it very much and good product the apple iphone 16GB, has very good quality <u>More</u> davidxerach catalog.ebay.com 5/30/2008	CellPhones.Shopzilla.com See your message here
Affordability (126 comments) 52% positive	We are very happy with the item, it is really a great product. <u>More</u>	
Screen (70 comments) 80% positive	Overall this is great item and I highly recommend this device to travelers. More	
Appearance (64 comments) 95% positive	clink1381 catalog.ebay.com 4/26/2008	

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General approach

- Identify features
 - Useful to identify opinion (positive, negative, neutral)
- Use classification approach or unsupervised approach to identify polarity (sentiment)
 - document level
 - Sentence level
 - Feature level

Opinion Words or Phrases

- Find relevant words, phrases, patterns that can be used to express subjectivity
- **Determine** the polarity of subjective expressions
 - also called polar words, opinion bearing words, etc, e.g.,
 - Positive: beautiful, wonderful, good, amazing, Ron
 Paul is the only honest man in Washington
 - Negative: bad, poor, terrible.
 - curious, peculiar, odd, likely, probable, The two species are likely to flower at different times

Three main ways to compile a list of opinion words

- Manual approach: not a bad idea, only an one-time effort
- Corpus-based approaches
 - This car is *beautiful* **and** *spacious*
- Dictionary-based approaches

– WordNet, e.g. synonyms and antonyms

Domain and context dependent

- Domain dependent
 - "go read the book" most likely indicates positive sentiment; for book reviews, but negative sentiment for movie reviews
 - Unpredictable for movie plot, for car's steering abilities
- Contextual Polarity :
 - Some opinion words are context independent (e.g., good).
 - Some are context dependent (e.g. long)

Example

Philip Clap, President of the National Environment Trust, sums up well the general thrust of the reaction of environmental movements: there is no reason at all to believe that the polluters are suddenly going to become reasonable.

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Philip Clap, President of the National Environment **Trust**, sums up **well** the general thrust of the reaction of environmental movements: there is no **reason** at all to believe that the **polluters** are suddenly going to become **reasonable**.

Example

Philip Clap, President of the National Environment **Trust**, sums up well the general thrust of the reaction of environmental movements: there is no reason at all to believe that the pollute are suddenly going to become reasonable. Contextual

polarity

prior polarity

SUMMARY

- Background on classification and clustering
- Basic knowledge on search engines

Crawler, indexer, ranker

- Question Answering and QA extraction
 QA detection in forum threads
- Opinion Mining