Texts Words Meaning

Elisabeth Bommes

mbr targeting Ladislaus von Bortkiewicz Chair of Statistics Humboldt-Universität zu Berlin http://lvb.wiwi.hu-berlin.de



Motivation — 1-1

One fits All?

Depends on the domain and the question.

Outline

- 1. Motivation ✓
- 2. Obtaining Data
- 3. Processing Data
- 4. Sentiment Classification
- 5. Sentiment Classification

Obtaining

- □ Ready to use data sets
 - Gold standard corpora
 - ► Classification: Reuters Newswire, Twenty Newsgroups, Youtube Spam, ...
 - ► Sentiment: reviews from Amazon, TripAdvisor, Rotten Tomatoes, ...
- APIs
 - ► Twitter, New York Times, Guardian
 - Only in real time?
- □ Crawling / Scraping
 - Data for (almost) any domain of interest
 - ► Legal? Terms of Service, robots.txt



Gold Standard Corpora

Wissler et al. (2014)

- Standard collections for training and evaluation of algorithms
- Manual annotations and more structure thain plain text
 - Syntactical information, lexical knowledge, semantic associations
 - E.g. entities, grammatical structures
- Multiple experts view the data independently
 - Inter-annotator agreement
 - Creation is time-consuming and expensive
- - ▶ 4.5 million English words, GSC for syntactical tagging



Legality of Web Scraping

- - ► Added value by search engines
 - ▶ What about log in systems, paywalls, ...?
- Highly context specific
 - Commercial v non-commercial
 - Internal v third party use
- Technicalities
 - ▶ Humans don't access thousands of pages, bandwidth usage
 - Denial-of-service (DoS) attack



European Union

- □ Ryanair Ltd v PR Aviation BV (2015)
 - ► PR Aviation: price comparison of flights
 - Copyright and database right infringement?
 - ► ToS prohibited data extraction for commercial purposes
- Decision by Court of Justice of the European Union
 - ▶ No infringement of intellectual property, no creative input
 - ► ToS still apply, liability in terms of breach of contract
- - Scraping of news headlines and links to articles
 - ▶ Intellectual property is infringed because of creative input



United States

Pro

- □ First Amendment protects information gathering
- Unfair market power of Facebook, Google, LinkedIn, ...

Contra

- Copyright infringement

- Trespass to chattels



LinkedIn v hiQ and vice versa

If you exclude someone from sites like LinkedIn, Facebook and Twitter, you are excluding them from the modern version of the town square.

Laurence Tribe, Harvard law professor

- ightharpoonup hiQ startup predicting who is when quitting their job
- □ LinkedIn: CFAA violation, hiQ: blocked
- LinkedIn ordered to give access to public profiles



Academia is save, right?



Aaron Swartz

- Harvard research fellow
- Automatic download of articles from JSTOR
- Download via laptop in a closet at MIT

- Possible penalty of \$1 million and 35 years in prison

Unclear outcome, suicide on January 11, 2013



Good Manners and Ethics

- □ Data as a product
- Cost of bandwidth
- □ Rapid requests slow services for real humans
- □ robots.txt

Google's robots.txt

```
User-agent: *
Disallow: /search
Allow: /search/about
Allow: /search/howsearchworks
Disallow: /sdch
Disallow: /groups
Disallow: /index.html?
Disallow: /?
Allow: /?hl=
Disallow: /?hl=*&
Allow: /?hl=*&gws rd=ssl$
Disallow: /?hl=*&*&gws rd=ssl
Allow: /?gws rd=ssl$
Allow: /?pt1=true$
Disallow: /imgres
Disallow: /u/
Disallow: /preferences
Disallow: /setprefs
Disallow: /default
Disallow: /m?
Disallow: /m/
Allow: /m/finance
```



Ethical Scraping for Academia

- Technical
 - Use API if provided
 - Appear as a bot, not as a human
 - Provide user agent string with contact data
 - Decreased rate of requests
- Usage
 - Strictly non-commercial
 - ▶ Restrict further access to academia



Scraping and Crawling



Scraping How To



Obtaining Data — 2-13

Beer Advocate



Individual Review

Sort by: Recent | High | Low | Top Raters | Alström Bros

first ← prev | 1-25 | 26-50 | 51-75 | next → last

✓ Ratings: 5,928 |

Reviews: 1,915

4.13/5 rDev -2.1%

look: 4.5 | smell: 4 | taste: 4 | feel: 4 | overall: 4.5

A Great One, a very good IPA with added thickness from chocolate malt. Pours dark and foamy, hoppy, ahh! the taste of citrus C-hop, (there is also Amarillo and Simcoe) and the c-hop aroma. Background roastiness. Full mouthfeel. Delightful as a meal in itself.

259 characters

Sammy, Dec 23, 2007

Source



xPath Inspector

div#rating_fullview | 494.8×5179 1,915

4.13/5 rDev -2.1%

look: 4.5 | smell: 4 | taste: 4 | feel: 4 | overall: 4.5

A Great One, a very good IPA with added thickness from chocolate malt. Pours dark and foamy, hoppy, ahh! the taste of citrus C-hop, (there is also Amarillo and Simcoe) and the c-hop aroma. Background roastiness. Full mouthfeel. Delightful as a meal in itself.

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Sammy, Dec 23, 2007



Example: Scraper

Processing

- Often depends on theoretical model
 - ► Time series models for stock returns
 - Example 2
- Just a side note in scientific articles
 - ▶ Well established "gold standard" processing for many domains
 - No scientific news value



Example: Beer Review

An OK lager. Light, crisp, but nothing special. Stacked against the great pilsners of the world or similar offerings, this beer is mediocre. But still solid enough to put it head and shoulders above any macro.

Nouns

An OK lager. Light, crisp, but nothing special. Stacked against the great pilsners of the world or similar offerings, this beer is mediocre. But still solid enough to put it head and shoulders above any macro.

NOUN

POS Tagging is not perfect

An OK lager. Light, crisp, but nothing special. Stacked against the great pilsners of the world or similar offerings, this beer is mediocre. But still solid enough to put it head and shoulders above any macro.

NOUN

- Which topic? Beer

Verbs

An OK lager. Light, crisp, but nothing special. Stacked against the great pilsners of the world or similar offerings, this beer is mediocre. But still solid enough to put it head and shoulders above any macro.

NOUN, VERB

Dependency Tree

Adjectives

An OK lager. Light, crisp, but nothing special. Stacked against the great pilsners of the world or similar offerings, this beer is mediocre. But still solid enough to put it head and shoulders above any macro.

NOUN, VERB, ADJ

Text Normalization: Basics

[(an, ok lager), (light, crisp, but, nothing, special), (stacked, against, the, great, pilsners, of, the, world, or, similar, offerings, this, beer, is, mediocre), (but, still, solid, enough, to, put, it, head, and, shoulders, above, any, macro)]

- Lowercase
- Removal of non alphabetic characters
- Word and sentence lemmatization

Text Normalization: Lemmatization

```
(stacked, stack), (pilsners, pilsner)
(offerings, offering), (is, be)
(it, -PRON-), (shoulders, shoulder)
```

Text Normalization: Special Cases

- Dates
 - ► Canonical form: $22.11.2017 \rightarrow 11/22/2017$
 - ightharpoonup Relative to absolute: Yesterday ightarrow 11/22/2017
- Abbreviations
 - ▶ Common: United States of America → USA
 - Specific: ordinary least squares (OLS)
- Numbers and units
 - ▶ Hundred \rightarrow 100
 - ▶ $$100 \rightarrow 100_dollar \rightarrow 10000_cent$

What is the most important aspect of the domain?



Text Filtering: Noun Chunks

```
(An, OK, lager), (Light), (the, great pilsners)(the, world), (similar, offering), (this, beer)(it), (any, macro)
```

- □ Parse dependencies of nouns
- Additional filtering for noun, adjective, verb combinations

Evaluation Processing

Pred	Unprocessed		Processed	
True	-1	1	-1	1
-1	677	723	725	675
1	498	902	307	1,093
Accuracy	0.56		0.65	
Recall	0.58	0.56	0.70	0.62
Precision	0.48	0.64	0.52	0.78

Evaluation Pipeline

Pred	Unprocessed		Processed	
True	-1	1	-1	1
-1	1,304	96	1,274	126
1	70	1,330	94	1,306
Accuracy	0.94		0.92	
Precision	0.93	0.95	0.91	0.93
Recall	0.95	0.93	0.93	0.91

Sentiment Distribution

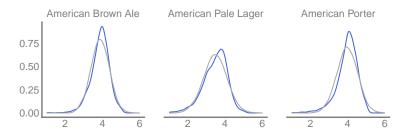


Figure: Estimated Densities of Beer Ratings and Simulated Normal Distribution



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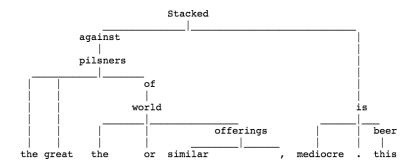
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Appendix

Appendix — 7-1

Grammar based Dependency Tree



Back

