https://www.youtube.com/watch?v=bScsFi6DaoM

## CORPUS ANALYSIS

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NLP - Spring 2019

## Administrivia

Assignment 0

Assignment 1 out

- due Monday at 11 am (don't wait until the weekend!)
$\square$ no code submitted, but will require coding
$\square$ will require some command-line work

Reading

## NLP models

How do people learn/acquire language?

## NLP models

A lot of debate about how human's learn language
$\square$ Rationalist (e.g. Chomsky)
$\square$ Empiricist

From my perspective (and many people who study NLP)...
$\square$ I don't care :)

Strong AI vs. weak AI: don't need to accomplish the task the same way people do, just the same task
$\square$ Machine learning
$\square$ Statistical NLP

## Vocabulary

## Word

$\square$ a unit of language that native speakers can identify
$\square$ words are the blocks from which sentences are made
Sentence
$\square$ a string of words satisfying the grammatical rules of a language
Document
$\square$ A collection of sentences
Corpus
$\square$ A collection of related texts

## Corpus examples

Any you've seen or played with before?

## Corpus characteristics

What are some defining characteristics of corpora?

## Corpus characteristics

monolingual vs. parallel
language
annotated (e.g. parts of speech, classifications, etc.)
source (where it came from)
size

## Corpus examples

## Linguistic Data Consortium

- http://www.Idc.upenn.edu/Catalog/byType.jsp

Dictionaries
$\square$ WordNet - 206K English words

- CELEX2 - 365K German words

Monolingual text
$\square$ Gigaword corpus

- 4M documents (mostly news articles)
- 1.7 trillion words
- 11 GB of data (4GB compressed)
$\square$ Enron e-mails
- 517 K e-mails


## Corpus examples

Monolingual text continued
$\square$ Twitter

- Chatroom
- Many non-English resources

Parallel data
$\square \sim 10 \mathrm{M}$ sentences of Chinese-English and Arabic-English

- Europarl
- $\sim 25 \mathrm{M}$ sentence pairs with English with 21 different languages
$\square$ 260K sentences of English Wikipedia—Simple English Wikipedia


## Corpus examples

## Annotated

- Brown Corpus
- $1 M$ words with part of speech tag
$\square$ Penn Treebank
- 1 M words with full parse trees annotated
- Other treebanks
- Treebank refers to a corpus annotated with trees (usually syntactic)
- Chinese: 51 K sentences
- Arabic: 145K words
- many other languages...
- BLIPP: 300M words (automatically annotated)


## Corpora examples

Many others...
$\square$ Spam and other text classification
$\square$ Google n-grams

- 2006 (24GB compressed!)
- 13M unigrams
- 300M bigrams
- 1B 3,4 and 5-grams
$\square$ Speech
- Video (with transcripts)


## Corpus analysis

Corpora are important resources

Often give examples of an NLP task we'd like to accomplish

Much of NLP is data-driven!

A common and important first step to tackling many problems is analyzing the data you'll be processing

## Corpus analysis

## What types of questions might we want to ask?

How many...

- documents, sentences, words

On average, how long are the:

- documents, sentences, words

What are the most frequent words? pairs of words?

How many different words are used?

Data set specifics, e.g. proportion of different classes?

## Corpora issues

Somebody gives you a file and says there's text in it

Issues with obtaining the text?
$\square$ text encoding
$\square$ language recognition
$\square$ formatting (e.g. web, xml, ...)
$\square$ misc. information to be removed
■ header information

- tables, figures
- footnotes


## A rose by any other name...

## Word

$\square$ a unit of language that native speakers can identify
$\square$ words are the blocks from which sentences are made

Concretely:
$\square$ We have a stream of characters
$\square$ We need to break into words

- What is a word?
- Issues/problem cases?
$\square$ Word segmentation/tokenization?


## Tokenization issues: ‘

Finland's capital...


## Tokenization issues: ‘

## Finland's capital...

Finland<br>Finland 's<br>Finland 's<br>Finlands<br>Finland s<br>Finland's

What are the benefits/drawbacks?

## Tokenization issues: ‘

## Aren't we



## Tokenization issues: ‘

## Aren't we

Aren't
Arent

Are n't
Aren $\dagger$

Are not

## Tokenization issues: hyphens

Hewlett-Packard
co-education
take-it-or-leave-it
state-of-the-art
lower-case

26-year-old

## Tokenization issues: hyphens

## Hewlett-Packard

co-education

Keep as is
merge together

- HewlettPackard
- stateoftheart

Split on hyphen

- lower case
- co education

What are the
benefits/drawbacks?
state-of-the-art
lower-case

## More tokenization issues

## Compound nouns: San Francisco, Los Angelos, ...

$\square$ One token or two?

Numbers
$\square$ Examples

- Dates: 3/12/91
- Model numbers: B-52
- Domain specific numbers: PGP key - 324a3df234cb23e
- Phone numbers: (800) 234-2333
- Scientific notation: 1.456 e-10


## Tokenization: language issues

Lebensversicherungsgesellschaftsangestellter

## 'life insurance company employee'

Opposite problem we saw with English (San Francisco)

German compound nouns are not segmented

German retrieval systems frequently use a compound splitter module

## Tokenization：language issues

## 莎拉波娃现在居住在美国东南部的佛罗里达。

Where are the words？

## thisissue

Many character based languages（e．g．Chinese）have no spaces between words
－A word can be made up of one or more characters
$\square$ There is ambiguity about the tokenization，i．e．more than one way to break the characters into words
$\square$ Word segmentation problem
－can also come up in speech recognition

## Word counts: Tom Sawyer

How many words?

- 71,370 total
$\square 8,018$ unique

Is this a lot or a little? How might we find this out?
$\square$ Random sample of news articles: 11 K unique words

What does this say about Tom Sawyer?
$\square$ Simpler vocabulary (colloquial, audience target, etc.)

## Word counts

What are the most frequent words?

What types of words are most frequent?

| Word | Frequency |
| :--- | :--- |
| the | 3332 |
| and | 2972 |
| a | 1775 |
| to | 1725 |
| of | 1440 |
| was | 1161 |
| it | 1027 |
| in | 906 |
| that | 877 |
| he | 877 |
| l | 783 |
| his | 772 |
| you | 686 |
| Tom | 679 |
| with | 642 |

## Word counts

8 K words in vocab
71 K total
occurrences
how many occur once? †wice?

| Word <br> Frequency | Frequency of frequency |
| :--- | :--- |
| 1 | 3993 |
| 2 | 1292 |
| 3 | 664 |
| 4 | 410 |
| 5 | 243 |
| 6 | 199 |
| 7 | 172 |
| 8 | 131 |
| 9 | 82 |
| 10 | 91 |
| $11-50$ | 540 |
| $51-100$ | 99 |
| $>100$ | 102 |

## Zipf's "Law"



The frequency of the occurrence of a word is inversely proportional to its frequency of occurrence ranking

Their relationship is log-linear, i.e. when both are plotted on a log scale, the graph is a straight line

## Zipf's law

## At a high level:

- a few words occur very frequently
- a medium number of elements have medium frequency
$\square$ many words occur very infrequently


## Zipf's law

$$
f=C \frac{1}{r}
$$

The product of the frequency of words ( $f$ ) and their rank ( $r$ ) is approximately constant

Constant is corpus dependent, but generally grows roughly linearly with the amount of
 data

## Zipf Distribution




Illustration by Jacob Nielsen

## Zipf's law: Brown corpus



## Zipf's law: Tom Sawyer

| Word | Frequency | Rank |
| :--- | :--- | :--- |
| the | 3332 | 1 |
| and |  | 2 |
|  |  |  |
| $\qquad$ | $f=C \frac{1}{r}$ |  |

$$
\begin{aligned}
C & =f * r \\
& =3332
\end{aligned}
$$

$$
\begin{gathered}
f=3332 * \frac{1}{2} \\
=1666
\end{gathered}
$$

## Zipf's law: Tom Sawyer

| Word | Frequency | Renk |
| :--- | :--- | :--- |
| the | 3332 | 2 |
| and | 2972 | 2 |

$$
\begin{aligned}
C & =f * r \\
& =3332
\end{aligned}
$$

$$
\begin{gathered}
f=3332 * \frac{1}{2} \\
=1666
\end{gathered}
$$

## Zipf's law: Tom Sawyer

| Word | Frequency | Rank |
| :--- | :--- | :--- |
| the | ***** | 1 |
| and | 2972 | 2 |
| a | $?$ | 3 |
|  |  |  |
|  | $f=C \frac{1}{r}$ |  |
|  |  |  |

$$
\begin{aligned}
C & =f * r \\
& =2972 * 2 \\
& =5944
\end{aligned}
$$

$$
\begin{gathered}
f=5944 * \frac{1}{3} \\
=1981
\end{gathered}
$$

## Zipf's law: Tom Sawyer

| Word | Frequency | Rank |
| :--- | :--- | :--- |
| the | $* * * * *$ | 1 |
| and | 2972 | 2 |
| a | 1775 | 3 |
|  |  |  |
|  | $f=C l l$ |  |
|  |  |  |
|  |  |  |

$$
\begin{aligned}
C & =f * r \\
& =2972 * 2 \\
& =5944
\end{aligned}
$$

$$
\begin{gathered}
f=5944 * \frac{1}{3} \\
=1981
\end{gathered}
$$

## Zipf's law: Tom Sawyer

| Word | Frequency | Rank |
| :--- | :--- | :--- |
| he | 877 | 10 |
| friends | $?$ | 800 |
|  |  |  |
| $\qquad$ | $f=C \frac{1}{r}$ |  |

$$
\begin{aligned}
C & =f * r \\
& =877 * 10 \\
& =8770
\end{aligned}
$$

$$
f=8770 * \frac{1}{800}
$$

$$
=10.96
$$

## Zipf's law: Tom Sawyer

| Word | Frequency | Rank |
| :--- | :--- | :--- |
| he <br> friends | 877 <br> 10 | 10 |
|  | $f=C \frac{1}{r}$ |  |
|  | $f 00$ |  |

$$
\begin{aligned}
C & =f * r \\
& =877 * 10 \\
& =8770
\end{aligned}
$$

$$
f=8770 * \frac{1}{800}
$$

$$
=10.96
$$

## Zipf's law: Tom Sawyer

| Word | Frequency | Rank | C=f*r |
| :--- | :--- | :--- | :--- |
| the | 3332 | 1 | 3332 |
| and | 2972 | 2 | 5944 |
| a | 1775 | 3 | 5235 |
| he | 877 | 10 | 8770 |
| but | 410 | 20 | 8400 |
| be | 294 | 30 | 8820 |
| Oh | 116 | 90 | 10440 |
| two | 104 | 100 | 10400 |
| name | 21 | 400 | 8400 |
| group | 13 | 600 | 7800 |
| friends | 10 | 800 | 8000 |
| family | 8 | 1000 | 8000 |
| sins | 2 | 3000 | 6000 |
| Applausive | 1 | 8000 | 8000 |

What does this imply about C/zipf's law? How would you pick C?

## Sentences

## Sentence

$\square$ a string of words satisfying the grammatical rules of a language

Sentence segmentation

- How do we identify a sentence?
- Issues/problem cases?
$\square$ Approach?


## Sentence segmentation: issues

A first answer:
$\square$ something ending in a: . ? !
$\square$ gets $90 \%$ accuracy

Dr. Dave gives us just the right amount of homework.

Abbreviations can cause problems

## Sentence segmentation: issues

A first answer:
$\square$ something ending in $\mathrm{a}:$. ? !
$\square$ gets $90 \%$ accuracy

The scene is written with a combination of unbridled passion and sure-handed control: In the exchanges of the three characters and the rise and fall of emotions, Mr. Weller has captured the heartbreaking inexorability of separation.
sometimes: : ; and - might also denote a sentence split

## Sentence segmentation: issues

A first answer:
$\square$ something ending in a: . ?!
$\square$ gets $90 \%$ accuracy
"You remind me," she remarked, "of your mother."

Quotes often appear outside the ending marks

## Sentence segmentation

Place initial boundaries after: . ? !

Move the boundaries after the quotation marks, if they follow a break

Remove a boundary following a period if:
$\square$ it is a known abbreviation that doesn't tend to occur at the end of a sentence (Prof., vs.)
$\square$ it is preceded by a known abbreviation and not followed by an uppercase word

## Sentence length

What is the average sentence length, say for news text?

| Length | percent | cumul. percent |
| :--- | :--- | :--- |
| $1-5$ | 3 | 3 |
| $6-10$ | 8 | 11 |
| $11-15$ | 14 | 25 |
| $16-20$ | 17 | 42 |
| $21-25$ | 17 | 59 |
| $26-30$ | 15 | 74 |
| $31-35$ | 11 | 86 |
| $36-40$ | 7 | 92 |
| $41-45$ | 4 | 96 |
| $46-50$ | 2 | 98 |
| $51-100$ | 1 | 99.99 |
| $101+$ | 0.01 | 100 |

