# Language Analytics

Session 2: String Processing with Python

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#### Course outline

- 1. Introductions
- 2. String Processing with Python
- 3. NLP for linguistic analysis
- 4. Text Classification 1
- 5. Text Classification 2
- 6. Word embeddings

- 7. Language modelling 1
- 8. Language modelling 2
- 9. BERT
- 10. More BERT
- 11. Project pitches
- 12. Generative models
- 13. Social impact

# Break

• Following Hunston (2012: 2):

[...] a collection of naturally occurring examples of language, consisting of anything from a few sentences to a set of written texts [...]

A corpus is planned, though chance may play a part in the text collection, and it is designed for some linguistic purpose. The specific purpose of the design determines the selection of texts [...]

If a corpus represents, very roughly and partially, a speaker's experience of language, the access software re-orders that experience so that it can be **examined in ways that are usually impossible** 

Hunston (2012: 3)

#### Specialised

- Texts of a particular type
- 'Representative'

#### General

- Texts of many diverse types
- Unlikely to be 'representative'

#### Comparative

- Two or more designed along the same lines
- Enables direct comparison

#### Parallel

- •Two or more in different languages
- Texts from one translated into different languages

#### Historical

 Texts from different periods of time

• Why use corpora?

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      - What about 10? 100? 10,000? More?

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  - Exploring the great unread (Cohen 2009: 59)
    - We can manually study a handful of texts in detail
      - What about 10? 100? 10,000? More?
- Text corpora allow us to make empirical, quantifiable observations of language in use that can easily scale and move between 'close' and 'distant' readings

#### Discussion

 Are there possible limitations of our working definition of a text corpus or text corpora?

How might text corpora be relevant in your discipline(s)?

- Are there any specific corpora that interest you?
  - What type of corpora are they?

# 2. Exploring corpora

Word frequencies

• Concordances, KWIC

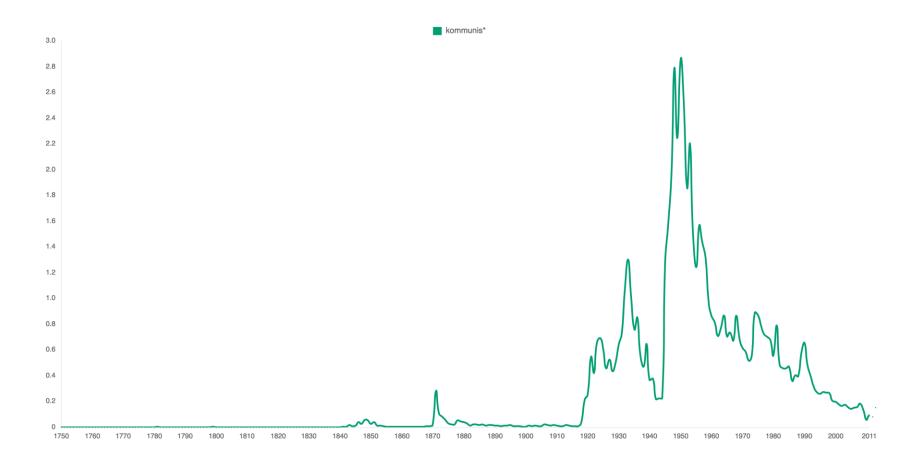
Collocation

### 2. Exploring corpora – word frequency

- Word frequency tells us...
  - how often an individual word appears
  - how often it appears relative to corpus size
  - if there are different distributions in different corpora
  - if distributions change over time
- Online examples
  - Google n-gram viewer
    - https://books.google.com/ngrams
  - Smurf
    - http://labs.statsbiblioteket.dk/smurf/
  - English corpora online

https://www.english-corpora.org/

### 2. Exploring corpora – word frequency



*kommunistisk\** in Danish newspapers over time

# 2. Exploring corpora – word frequency

SECTION	ALL	1800	1810	1820	1830	1840	1850	1860	1870	1880	1890	1900	1910	1920	1930	1940	1950	1960	1970	1980	1990
FREQ	41637	0	0	4	2	20	51	41	158	149	37	21	40	2727	2716	6326	12265	7251	4528	2813	2040
WORDS (M)	1589	5.0	7.1	11.6	28.1	30.4	33.0	34.2	37.1	60.0	51.2	64.7	79.8	71.7	95.2	94.8	121.0	152.0	163.3	183.7	177.1
PER MIL	26.19	0.00	0.00	0.34	0.07	0.66	1.55	1.20	4.26	2.48	0.72	0.32	0.50	38.04	28.53	66.70	101.39	47.69	27.72	15.31	11.52
SEE ALL SUB- SECTIONS AT ONCE																					

#### 2. Exploring corpora - concordances

• Concordances (key words in context, KWIC)...

[...] bring together many instances of use of a word or phrase, allowing the user to observe regularities in use that tend to remain unobserved when the same words or phrases are met in their normal contexts. (Hunston 2012: 9)

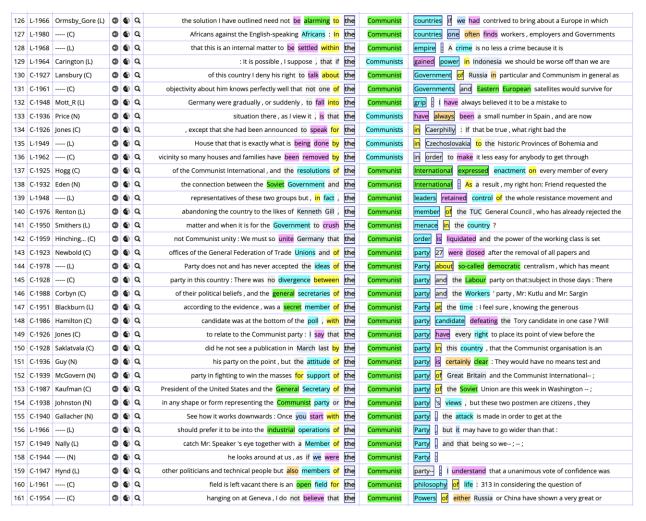
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- The so-called distributional hypothesis
  - 'You shall know a word by the company it keeps' (Firth, J. R. 1957:11)
  - In short, the context in which a word is used determines meaning
  - Words used in similar contexts tend to have similar meanings

### 2. Exploring corpora – concordances



#### Collocation is...

[...] the statistical tendency of words to co-occur. A list of the collocates of a given word can yield similar information to that provided by concordance lines, with the difference that more information can be processed more accurately by the statistical operations of the computer than can be dealt with by the human observer. (Hunston 2012: 12)

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- How likely are words to co-occur within a given window size?
  - E.g. ±5 words from our target word

- Collocation essentially models the following question:
  - Given the occurrence of a word u, how likely is it to also see word v within a given distance from u?
- One way to do this is to calculate the *strength of association* between two words
- We do this by comparing the expected frequencies with the observed frequencies
- For a more comprehensive overview of available metrics, see Evert (2010)
  - http://collocations.de

- One useful and easy to understand association measure is (pointwise) mutual information (Church & Hanks 1990)
- Do two words co-occur more than expected if they were independent?

$$PMI(word_1, word_2) = \log_2\left(\frac{p(word_1 \& word_2)}{p(word_1) p(word_2)}\right)$$

- In ordinary language...
  - P(word₁) = probability of word₁ appearing in corpus
  - P(word<sub>2</sub>) = probability of word<sub>2</sub> appearing in corpus
  - P(word<sub>1</sub> & word<sub>2</sub>) = probability of word<sub>1</sub>, word<sub>2</sub> appearing together in given window size

# An example: $w_1$ = hat; $w_2$ =rolled; window =±5

A vast deal of coolness and a peculiar degree of judgement, are requisite in catching a hat.

A man must not be precipitate, or he runs over it; he must not rush into the opposite extreme, or he loses it altogether. [...]

There was a fine gentle wind, and Mr. Pickwick's hat rolled sportively before it.

The wind puffed, and Mr. Pickwick puffed, and the hat rolled over and over as merrily as a lively porpoise in a strong tide; and on it might have rolled, far beyond Mr. Pickwick's reach, had not its course been providentially stopped, just as that gentleman was on the point of resigning it to its fate.

(adapted from Evert & Krenn (2003))

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HELP	?		EDEO -	ALL .	%	ML	
		DARTY	FREQ	ALL 702610		MI	
1		PARTY COUNTRIES	4185 1789	702610 441806	0.60	4.83	
2					0.40	4.27	
3	0	WORLD	1311	542904	0.24	3.53	
4		CHINA	1214	65215	1.86	6.47	
5	0	CHINESE	834	44795	1.86	6.47	
6		SOVIET	702	86598	0.81	5.27	
7		PROPAGANDA	685	30353	2.26	6.75	
8	0	INTERNATIONAL	552	224102	0.25	3.56	
9		RUSSIA	546	105374	0.52	4.63	
10		COMMUNIST	478	23725	2.01	6.59	
11		RUSSIAN	417	55264	0.75	5.17	
12		BLOC	416	7406	5.62	8.07	
13		PARTIES	415	238199	0.17	3.06	
14		SPREAD	325	49652	0.65	4.97	
15		AGGRESSION	317	18447	1.72	6.36	
16		INFLUENCE	298	132912	0.22	3.42	
17		FASCIST	265	4612	5.75	8.10	
18		SOCIALIST	263	44898	0.59	4.80	
19		COMMUNISM	253	8618	2.94	7.13	
20		THREAT	252	63602	0.40	4.24	
21		EASTERN	250	45746	0.55	4.70	
22		SOCIALISM	249	15755	1.58	6.24	
23		MOVEMENT	243	98345	0.25	3.56	_
24		LEADERS	240	65545	0.37	4.13	_
25		Ré	230	12304	1.87	6.48	_
26		COMMUNISTS	217	8042	2.70	7.01	
27		INFILTRATION	209	1382	15.12	9.50	_
28		SOCIALISTS	197	8860	2.22	6.73	-
29		ACTIVITIES	194	105498	0.18	3.13	-
30		FASCISM	192	2260	8.50	8.66	_

#### Discussion

- Given the methods we've so discussed so far:
  - How might each of these methods be useful for you?
    - Word frequency
    - Concordances
    - Collocation

# Move over to UCloud now!

https://cloud.sdu.dk

#### References

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- \* Hunston, S. (2002). Corpora in Applied Linguistics. Cambridge: Cambridge University Press