

Computational methods for complexity and burstiness in natural language

Jefrey Lijffijt

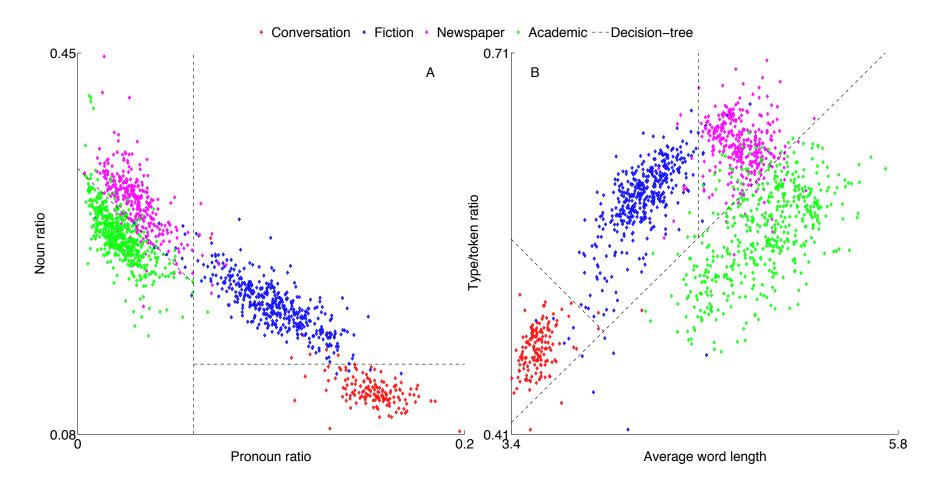
Joint work:

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Early work: Classification and clustering of genres





Paper under rotation.

Early work: Word sense disambiguation

Query words that are difficult to query directly

- Premodifying –ing participles
- Use automatically tagged + parsed corpus
 - Noisy tags and poor parsing
- Goal: generate rules
 - [POS = AJ0 and SUBST_NEXT = True \rightarrow Prem. -ing part.]
- 0-1 Classification problem
- Turo Vartiainen, Jefrey Lijffijt. Premodifying -ing participles in the parsed BNC. To appear in Corpus Linguistics and Variation in English: Theory and Description. Amsterdam/New York: Rodopi.



Recent work: Inter-arrival time distributions

Count space between consecutive occurrences: and

Finnair believes that it will be able to resume its scheduled service to **and** from New York on Monday, after two days of cancellations caused by hurricane Irene. All three airports serving New York City have been closed because of the hurricane **and** Finnair was forced to cancel flights on Saturday **and** Sunday. The airline is not certain when its scheduled service can be resumed, but the assumption is that Monday afternoon's flight from Helsinki will depart. Some Finnair passengers whose final destination is not New York have been rerouted **and** some have delayed travel plans. The company has also offered ticket holders a refund. *YLE*

 $\blacksquare IA_{and} = \{29, 9, 39, 29\}$

Hypothesis: this captures the behavior pattern of words



Given two corpora (collections of texts) S and T

- Corpus with fiction prose
- \blacksquare S = male authored, T = female authored

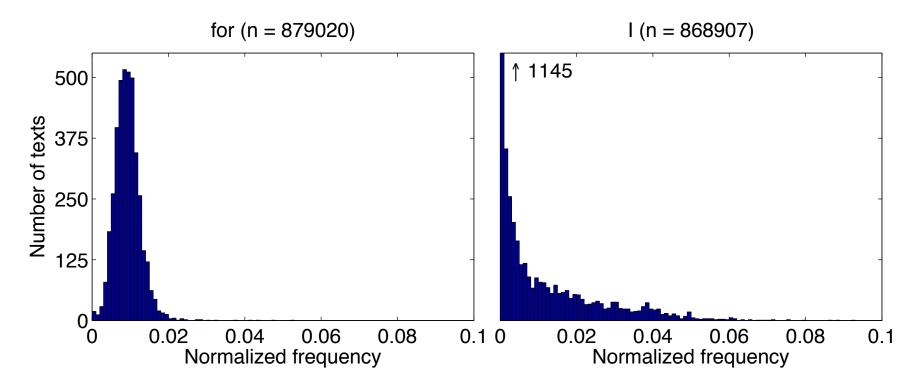
Find all words that are significantly more frequent in S than in T, or vice versa

Word	Freq in S	Freq in T
sergeant	57	32
Total	400.000	400.000

Is this statistically significant?

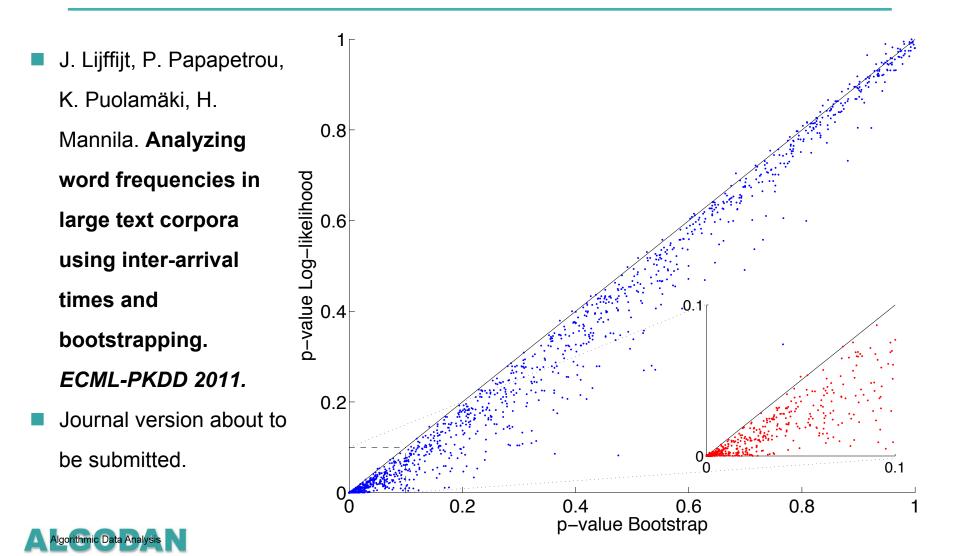


■ Words are bursty → do not use bag-of-words model



Data: British National Corpus, 4049 texts





$\alpha \leq 0.01$ in a text of 2000 words

Word	Freq in BNC (x10 ⁶)	Weibull β	Binomial test	Bootstrap test
а	2.2	1.01	61	72
for	0.9	0.93	29	37
1	0.9	0.57	29	110

β is the shape parameter of the Weibull fit
 Smaller β gives larger differences



ECML-PKDD 2011 08/09/2011 8 Analyzing word frequencies in large text corpora Jefrey Lijffijt

Upcoming/now: Burstiness and inter-arrival times

Segmentation of text

Find change-points in sequence of inter-arrival times

Clustering of words

- We know burstiness is related to parts-of-speech
- Can we group words based on inter-arrival times?

Stochastic model for inter-arrival times

- Current best (Weibull) does not fit that well
- How to take into account correlations



Upcoming/now: Burstiness and inter-arrival times

Relating diseases, genes and tissues through text mining of scientific articles (PubMed abstracts)

- Disease \rightarrow key-words \rightarrow genes
- Collaboration with Hautaniemi Lab



Computational methods for complexity and burstiness in natural language

Early work

Classification and clustering of text

- Word sense disambiguation
- Later work
 - Burstiness and inter-arrival times of words
 - Finding key-words: statistical testing for linguistics
- Upcoming/now
 - Clustering words based on inter-arrival times
 - Segmentation of text based on inter-arrival times
 - Stochastic models for inter-arrival times
 - Relating genes and diseases through key-word analysis

