

REWRITING LITERATURE HSTORY WITH BIG DATA

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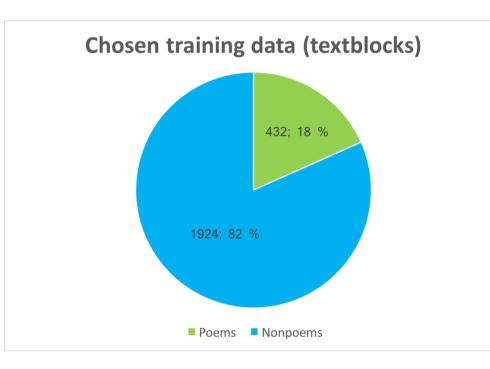


Data: National Library's digitalized newspaper archives - approx. 1,9M pages from years 1820-1910.

Canataanu m 1957	(Bähetetty.)	MG amanan
Rewät-aamu w. 1857. (Lähetetty.) Rä ferran fewähällä Käwelin laaffosfa, lamulla heliällä, Kuffaisten tuoffusfa. Baan ajatuffet juoffit Kipoista fyytiä, Re maita myöten wiilfit, Tawoitit tähtiä. Isa polfu minut juohti Eräälle pirtille, Ee filmihini hohti Ja näytti fiistille. Rä luulin yffinääni Olewain liiffeellä; Baan fuuluipa jo ääni	Koirain werosta. Koirille on konsti tullut, Peräti jo pentuloille, Kun on saatettu sakolle, Seitettynä hengen uhka Onnettomien osakii, Run on wero wiskattuna, Rahan makso määrättynä Salliloitten haukunnasta, Billahännän winkunasta, Billahännän winkunasta, Bustunnasta loppafuitten Raksin markoin kaupungeissa, Soka koiran kohtaloksi, Run jo alettiin anoa, Mustiloille muistutella, Baatia jo wahtiloilta, Seristellä hekkoloilta, Kaiketikin Rastoreilta, Sopa raiskoilta rahoja.	Warpunen. Jos ruikutella woifin Ma kielin fataisin Tai kiurusena visin Kohoowa pilwihin, — Ma aina kiitteleisin Sulvista Suomea, Ja Jumalalle weisin Ma huokauksensa. Han silmät ehkä loi Myös meihin raukkoih Ja päiwän paistaa so Saloihin synkkihin. Ei ääni mulle suotu Do satakielisen;
Pirttifen wierellä.	Mistäs pennut pennit faiwat, Lahi markat makjoihinfa? Wunna alkaiwat anad	Seitoffi fiipi luotu,

- Are there differences in poem topics between newspapers or geographical areas?
- How did the topics develop through

FINDING POEMS WITH MACHINE LEARNING



A Support Vector Machine classifier was used to classify textblocks into poem or non-poem classes. The word frequencies were used as features for the training classifier. The initial cross-validation error of 6 % was improved to 93 % by

TOPIC MODELING

Structural topic modeling (stm) is a method to find topics in a set of documents. The user defines the number of topics (k) and the algorithm statistically analyses the frequencies of the words in the texts to find the groups of words which "go together".

With our training data of approximately 100 handpicked poems, we used the stm R package, which yielded promising results – the topics seemed like plausible poem themes.

decades?

• Are there sub-topics for common themes, such as nationalistic poems?

MORPHOLOGY OF POEMS

Could some morphological features highlight regional differences, changes in time, or allow us to differentiate types of poetry within the corpus? The features we wanted to look for were:

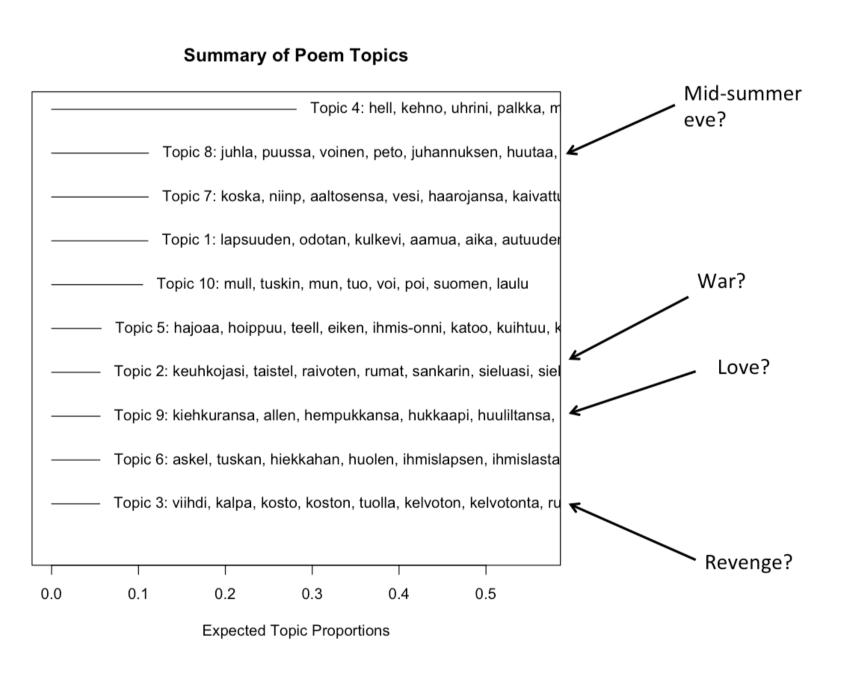
- Adjectives
- Verbs (the hypothesis being that poems containing large numbers of verbs would be more active in meaning)
- Interjections and exclamations such as

using an exhaustive search over the parameter values, and carefully choosing the training data.



The accuracy of the predictions on the early 19th century was over 50 %, however it decreased in the late 19th century. It would be interesting to take a closer look to investigate the reasons behind this observation.

Identified poems (whole corpus) poems.



We also performed the stm to lemmatised texts, which caused changes in the topics, but did not necessarily improve the model. There are several different types of stm algorithms, and it would be useful to compare their performance to determine which type would be best for analysing

"Oi!", which we know are typical for poetry

Steps of the analysis:

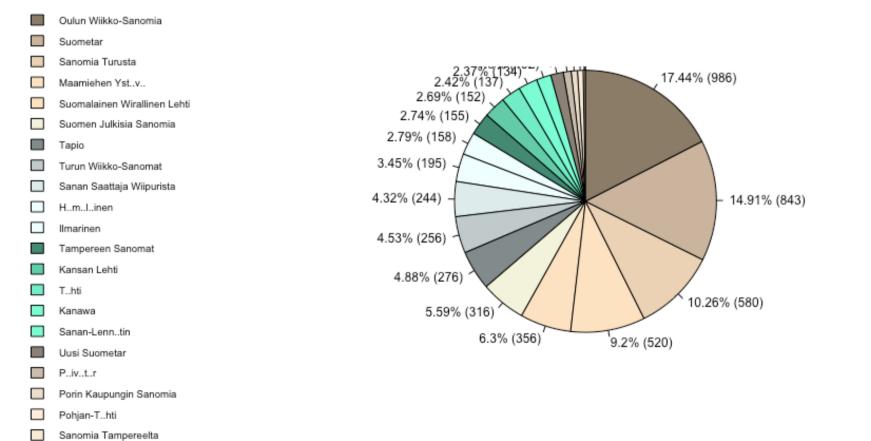
- Selecting the training data,
- Uploading the data to the server and running a morphological analysis tool (LAS) on the data (POS-tagging the data)
- Processing the JSON files in R
- Analysis of results

What we managed to do:

- POS-tagging on the corpus 1820-1870
- A small-scale test (manual)

Issues:

• Processing JSON files with parsed trees has proven difficult, and we have not been able to come up with a working script



With the complete poem corpus and improved stm analysis, we could shed light to e.g. following questions:

- What kind of topics are there and what lacksquareare their prevalences?
- Problems within the data (OCR errors resulting in wrong tagging of words)









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