Top2Vec as a tool for topic identification in fiction. A study on contemporary Czech prose.

8. 02. 2024 Charlotte Panušková

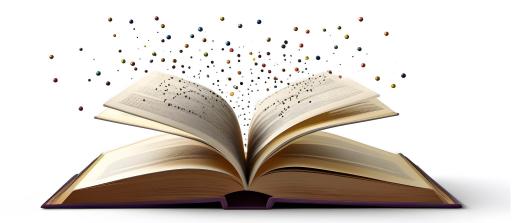
Ústav pro českou literaturu AV ČR Institute of Czech Literature of the CAS

Outline

- 1. Introduction
- 2. Background
- 3. Research Methodology
- 4. Topic Modelling Techniques
- 5. Results and Observations
- 6. Conclusion
- 7. Q&A

Research question

- Trends in contemporary Czech literature -> topics?
- How does topic modelling relate to traditional literary studies?



Trends in Czech literature

1990s - Two streams emerged

- fantasy, imaginative prose associated with the metropolis (mostly Prague)
- authenticity, connected by the author's "self"

2000s - Works set in specific times and places

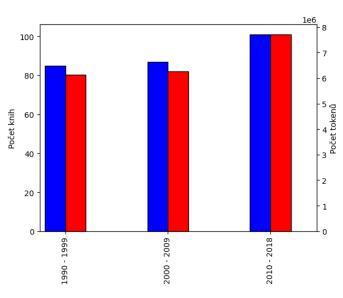
- historical traumas (holocaust, expulsion of Germans ...)
- the atmosphere of life in (post)communist society
- **2010s** Continuation of the trend from the previous decade
 - Relationship crisis
 - (Auto)biographies

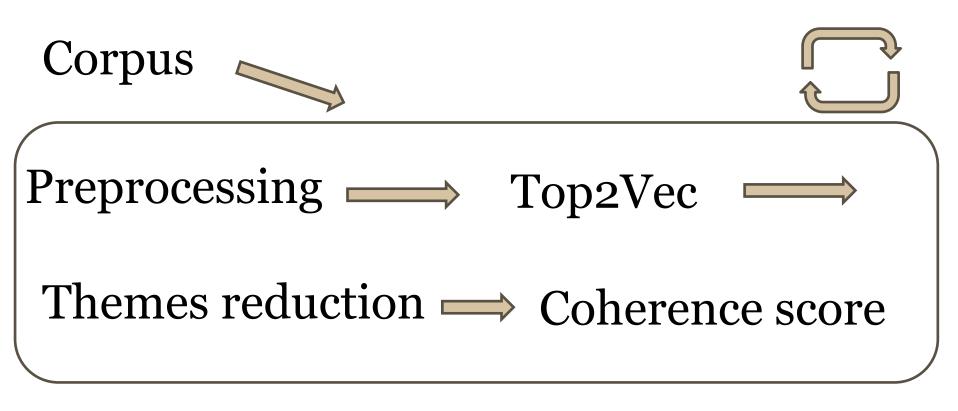
Topic Modelling in DLS

- Jockers, Matthew L., and David Mimno. 'Significant Themes in 19th-Century Literature', 2013.
- Schöch, Christof. 'Topic Modeling Genre: An Exploration of French Classical and Enlightenment Drama', 2021.
- Van Zundert, Joris J et. al. 'What Do We Talk About When We Talk About Topic?', 2022.

Corpus

- Publicly available corpus from the Czech National Corpus, SYN v9
- Only fiction -> prose
- Not representative, however diverse
- documents split into sections and mixed
- 273 books
- Divided into 3 decades based on the date of first publication







Preprocessing

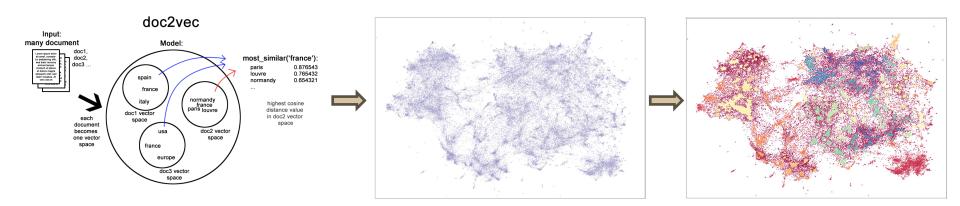
- The corpus is already lemmatized.
- Removal of Named Entity Recognition (NER) NameTag
- Removal of stopwords.
- Splitting documents into parts.

Top2Vec

doc2vec

UMAP

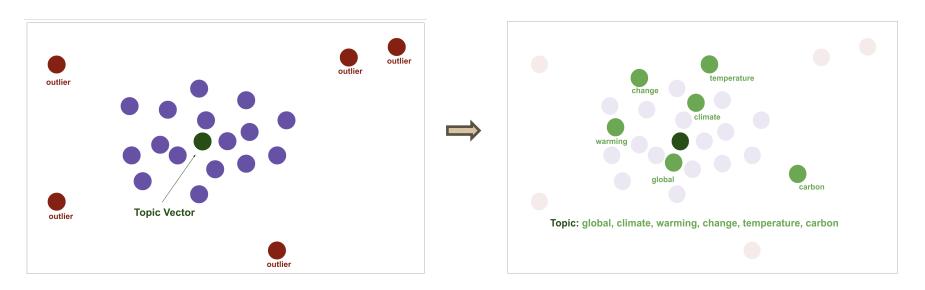
HDBSCAN



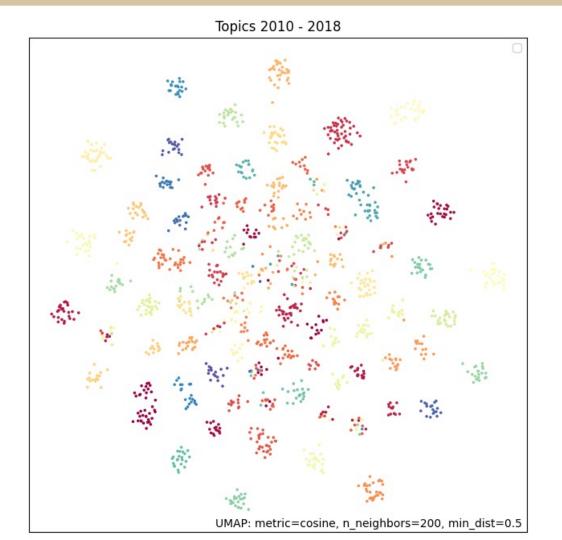
Pham, Hieu Huu Quang. (2017). An Empirical Approach to Sentiment Analysis with Doc2Vec. Retrieved from the University of Minnesota Digital Conservancy, https://hdl.handle.net/11299/190914. Angelov, D. (2020). Top2vec: Distributed representations of topics. *arXiv preprint arXiv:2008.09470*.

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Top2Vec

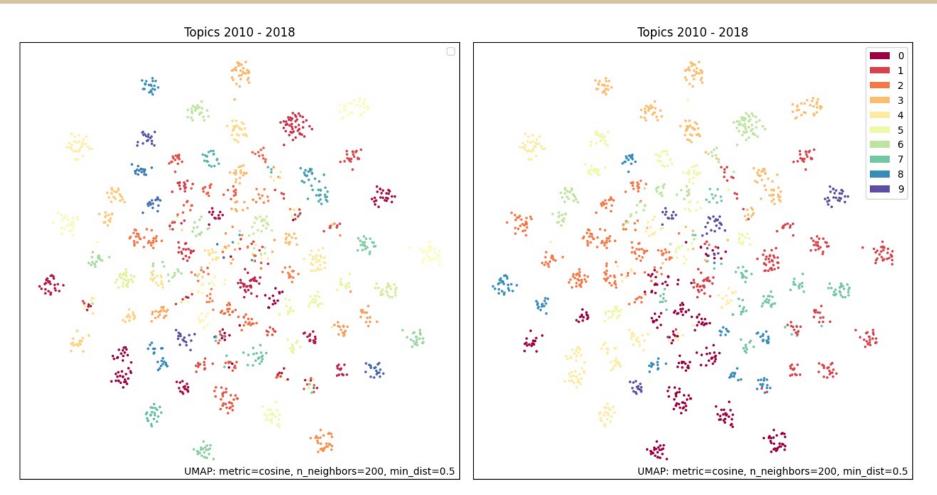


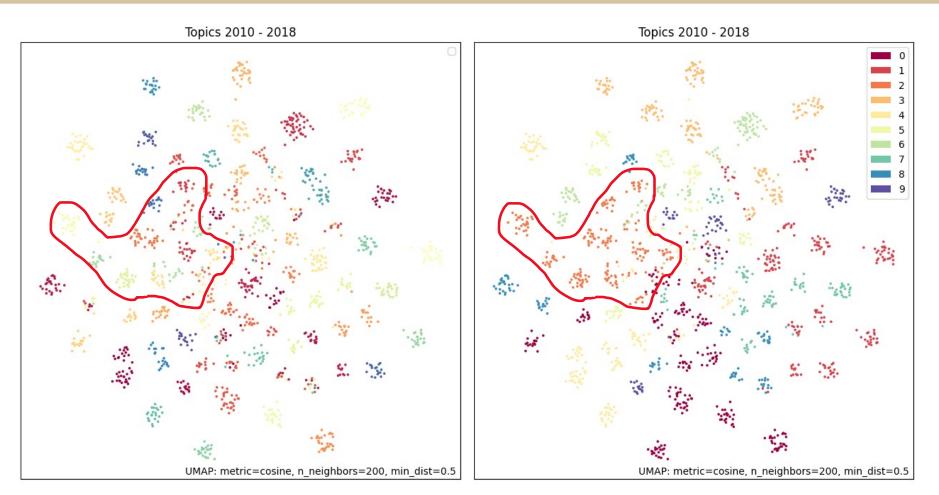
Angelov, D. (2020). Top2vec: Distributed representations of topics. *arXiv preprint arXiv:2008.09470*.

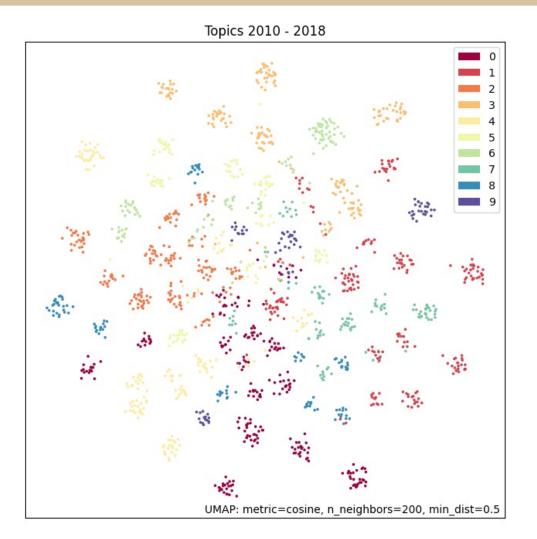


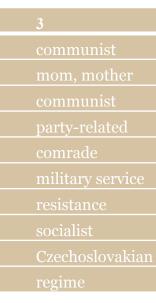
Coherence score

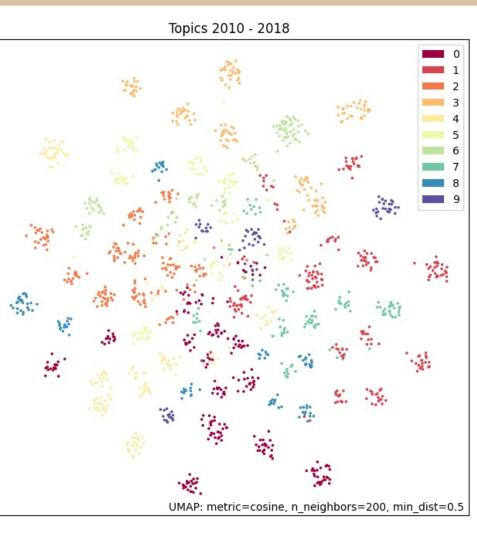
Tokens	Topics	C_v
1000	10	0,447
1000	20	0,465
1000	30	0,471
2000	10	0,479
2000	20	0,470
2000	30	0,468

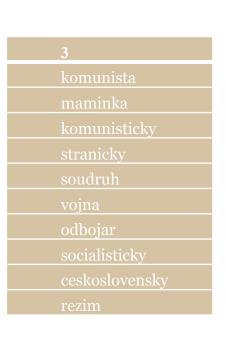






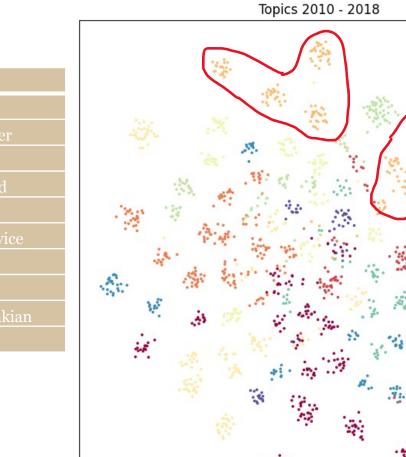


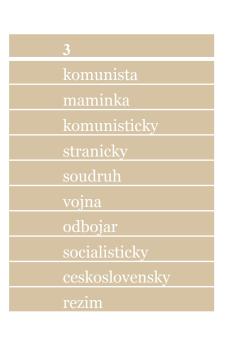




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UMAP: metric=cosine, n_neighbors=200, min_dist=0.5

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mom, mothe

communist

party-related

comrade

military service

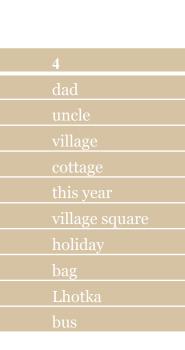
resistance

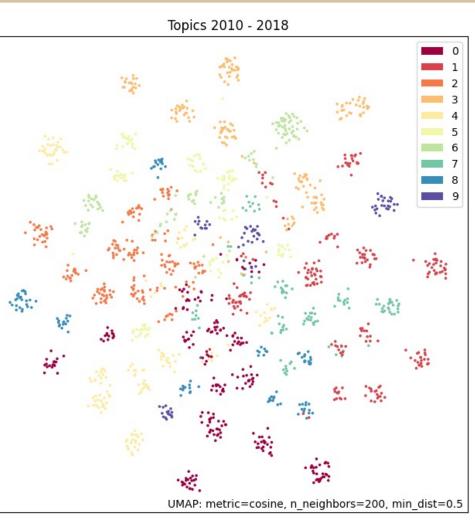
socialist

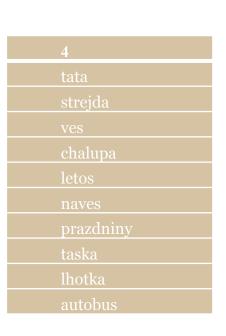
Czechoslovakian

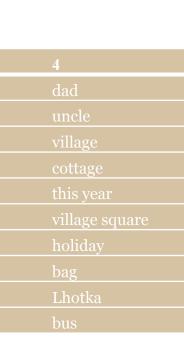
egime

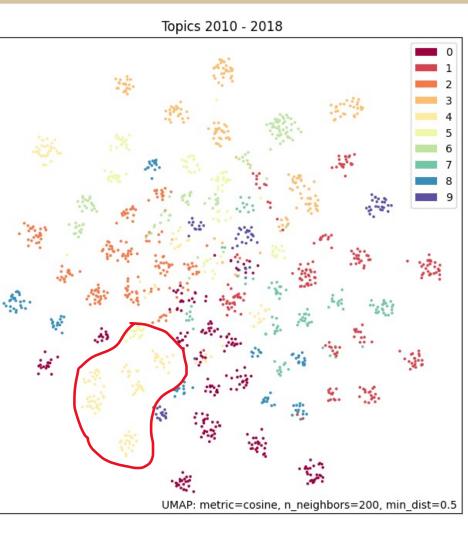
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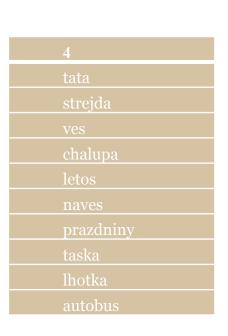


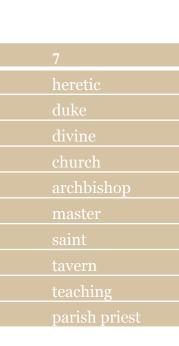


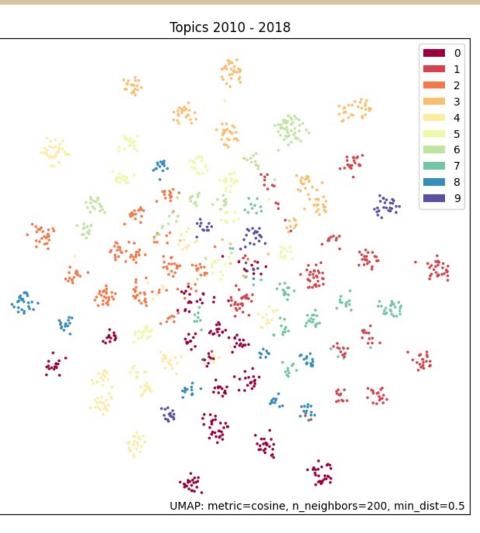


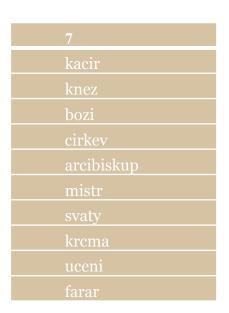


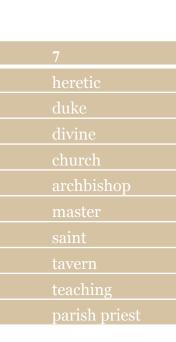


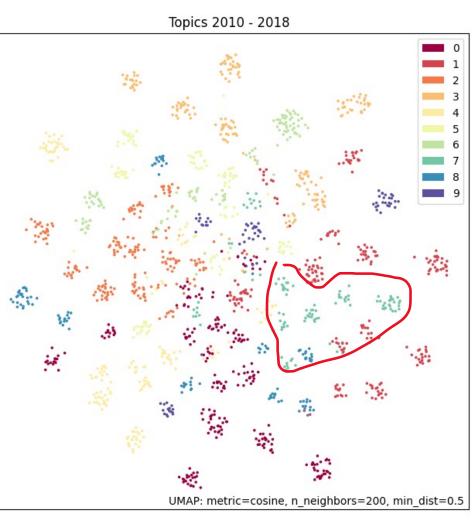




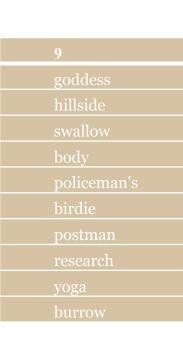


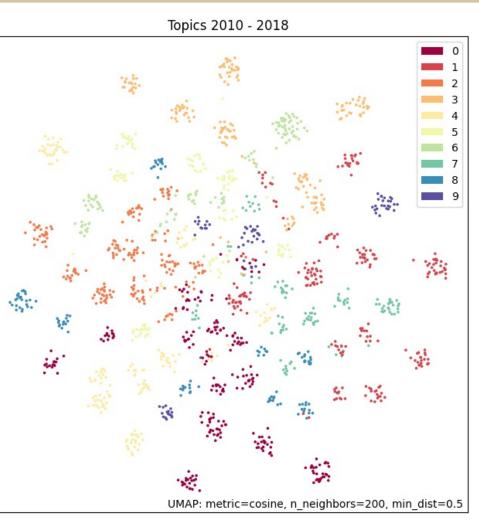


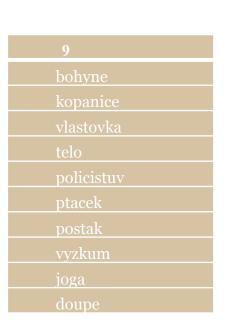


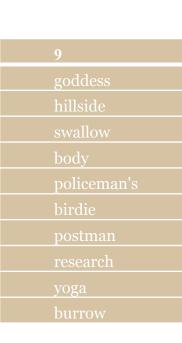


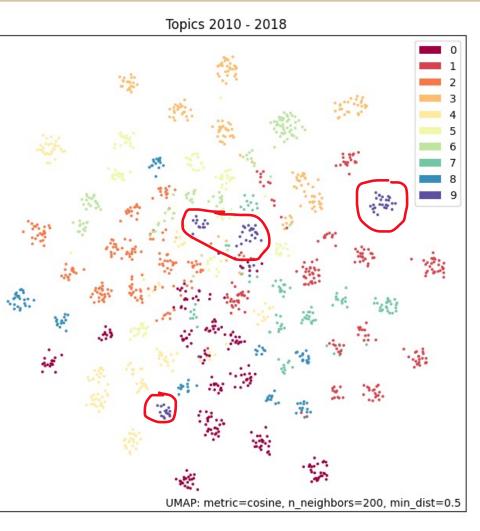
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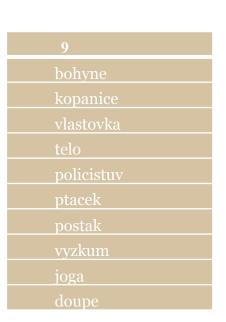






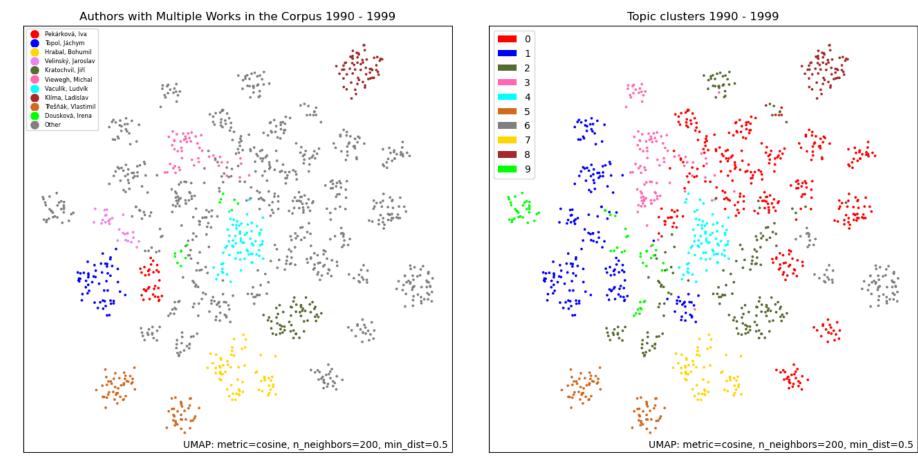






Observation

- Themes are fairly coherent
- Words are not always representative; it is necessary to look at specific works.
- Problem of dividing books into parts the longer the book, the greater the weight.
- Authors stay within one topic cluster



Results

- What is a 'topic' according to Top2Vec?
 - Motif, Theme
 - Genre
 - Author's style
- Literary system?
- The words of the topics corresponded to literary trends only if the trend was clearly defined (eg. Communism)

Future work

- Representative corpus
- More robust model
- Comparison of multiple models

Thank you



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