

From topic modeling to mapping discursive areas

Topic modeling:

- based on the assumption of semantic clustering (thematic organization) of textual data
- complicated calculation of topics
- · producing discontinuities rather than continuities.

Sometimes, we are not interested in distinguishing individual topics but in a more general discursive structure of textual data. For example, we know what "topics" are in migration stories: leaving the home country (- traveling - refugee camps) - arrival to the host country - adaptation (- return). We can wonder if there are more macrostructures than topics

--> analysis of frequent word co-occurrences.

Data & Analysis

Data: the corpus of migrant stories from iamamigrant.org; cleaned (paratextuals, stop-words) and divided in male and female subsets

Analytical procedure:

- 1. selection of 50 most frequent semantic words; excluding highly contextual words such as *people*, *time*, *lot*, *month*, *day*, *dont*, *im* etc.
- 2. splitting the corpus into context units (stories=paragraphs)
- 3. counting co-occurrences of selected words within the context units
- 4. creation of a standardized matrix of distances between words using
- Jaccard's similarity coefficient $J(A,B) = \frac{|A \cap B|}{|A \cup B|} = \frac{|A \cap B|}{|A| + |B| |A \cap B|}$. 5. computation of a two-dimensional graph using multidimensional scaling technique (MDS)

Software

Word frequencies: R



Counting word co-occurrences and calculation of distances:

Co-occurrence Analysis

Software (COOA)

Multidimensional scaling: R

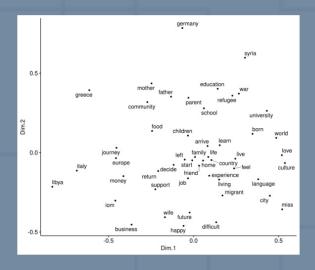




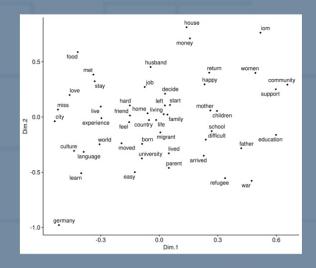
Results I - gender differences

Interpretation guide:

- the most frequent words are displayed closer to the centre of the graph as they have many co-occurrences with all other words; less frequent words are, on the other hand, found at the periphery;
- it is essential to look at the structure as a whole, not at the exact position of a particular word;
- dimensions (axes) do not have substantive meaning, and the graph can be arbitrarily rotated and turned.

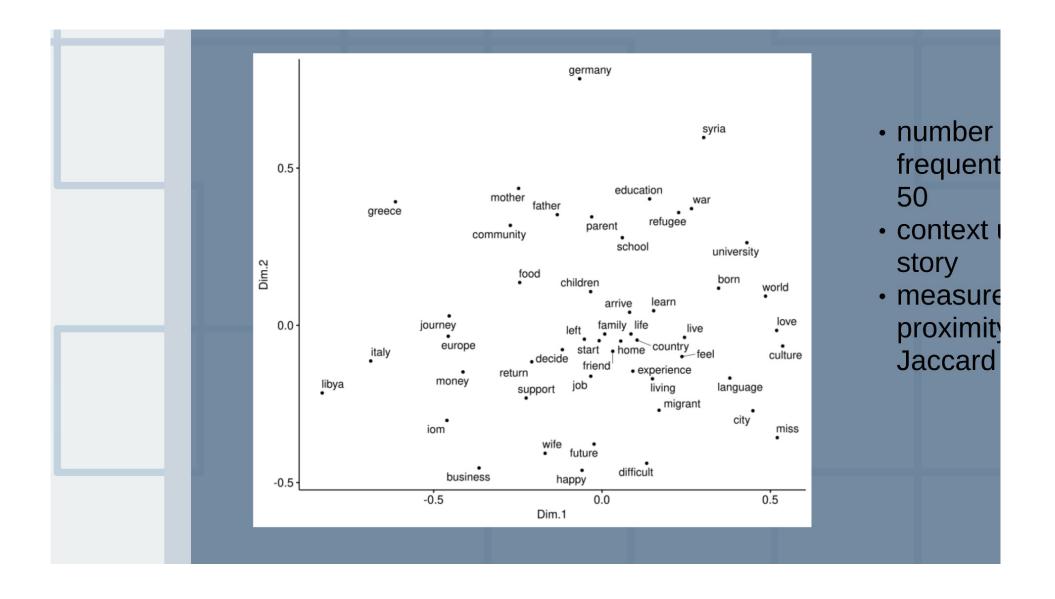


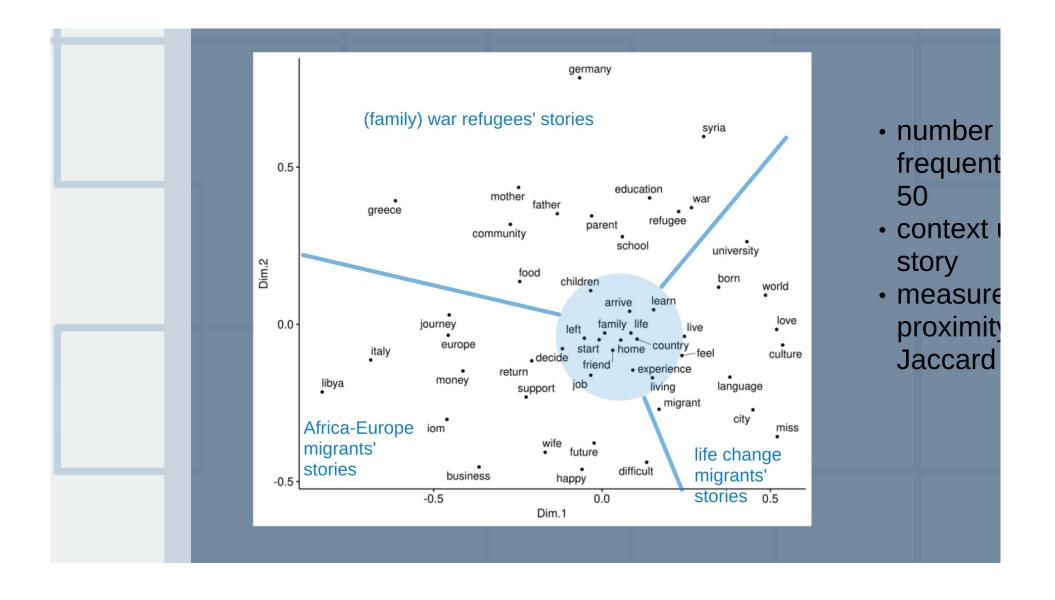
- number of frequent words:50
- context unit: story
- measure of proximity: Jaccard coeff.

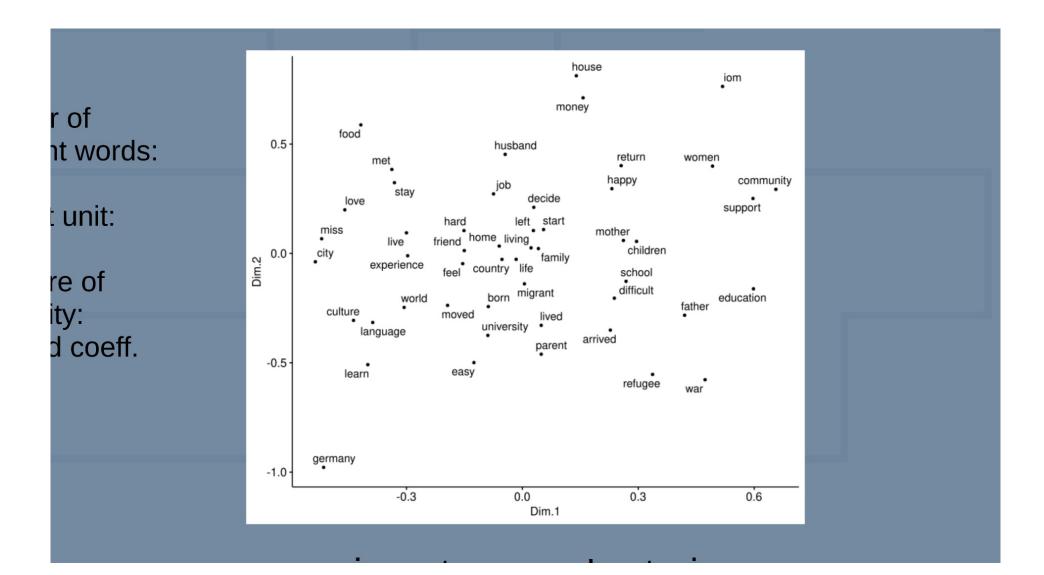


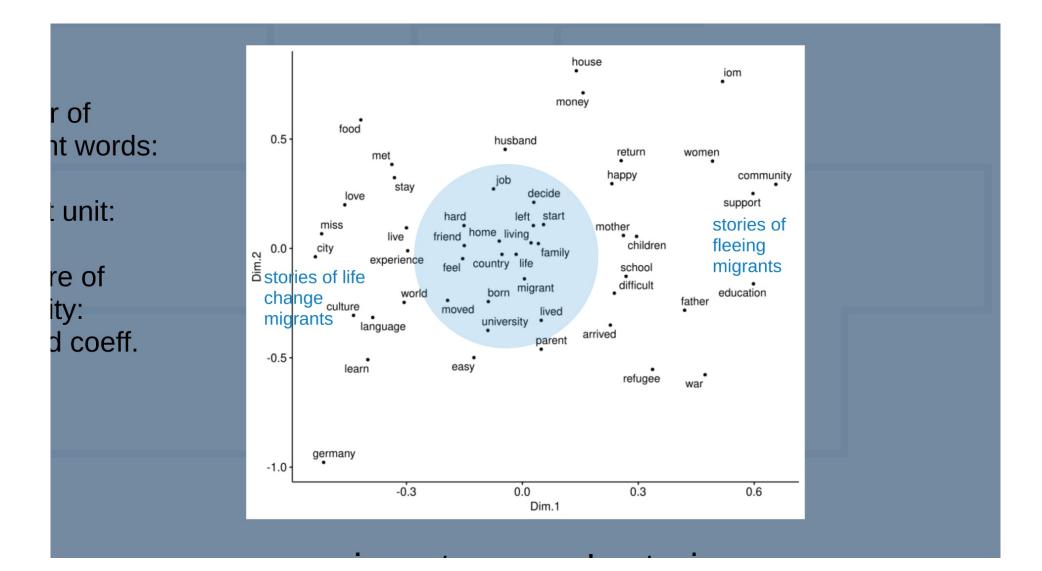
migrant men's stories

migrant women's stories



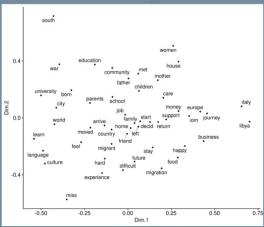






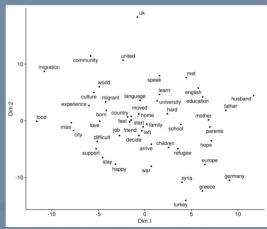
Results II - near and far migrants

Is there a difference between stories of intracontinental and intercontinental migrants?

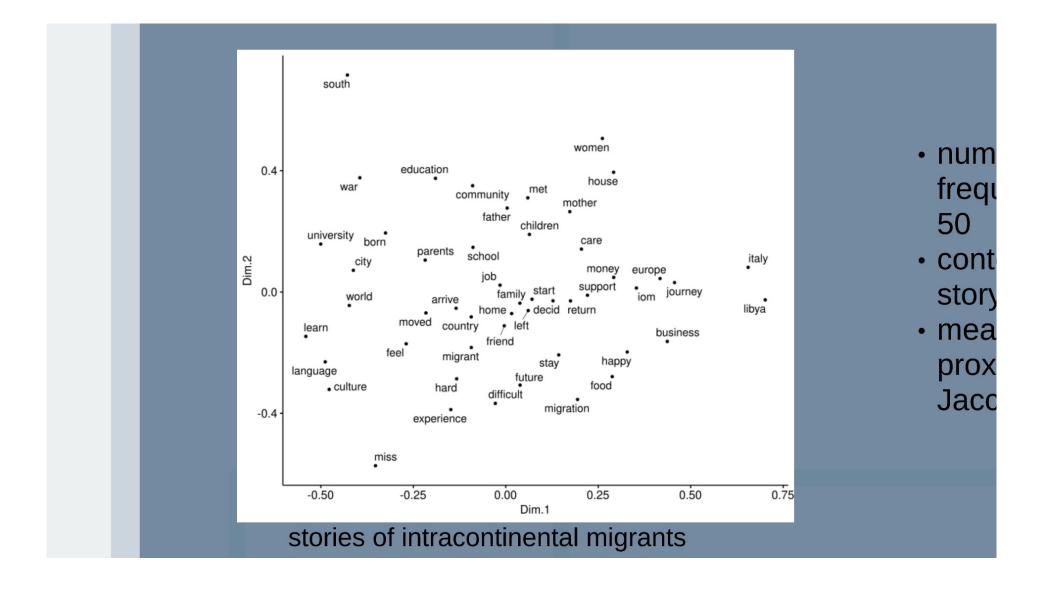


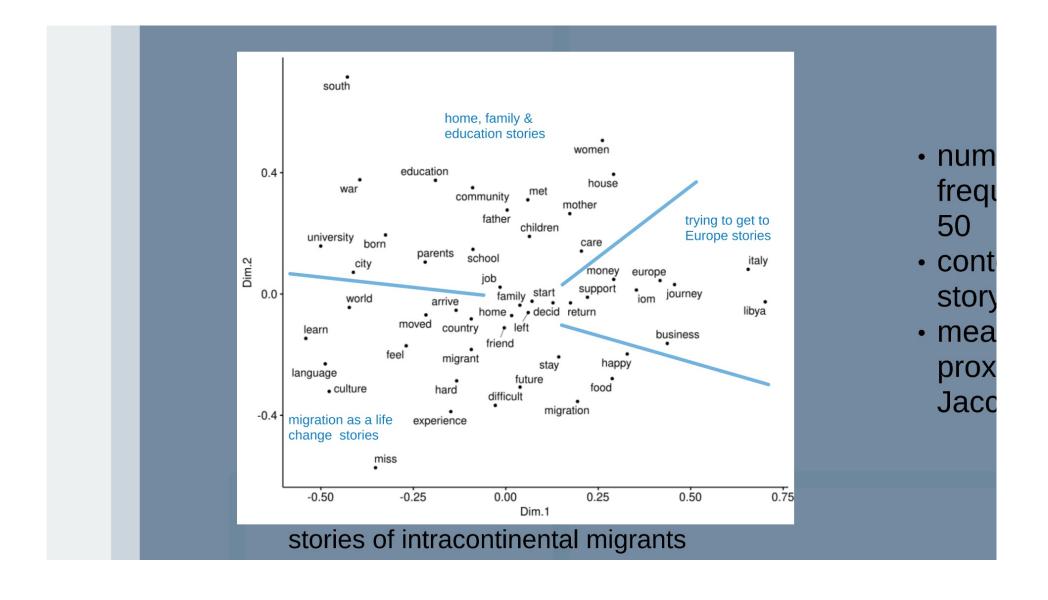
stories of intracontinental migrants

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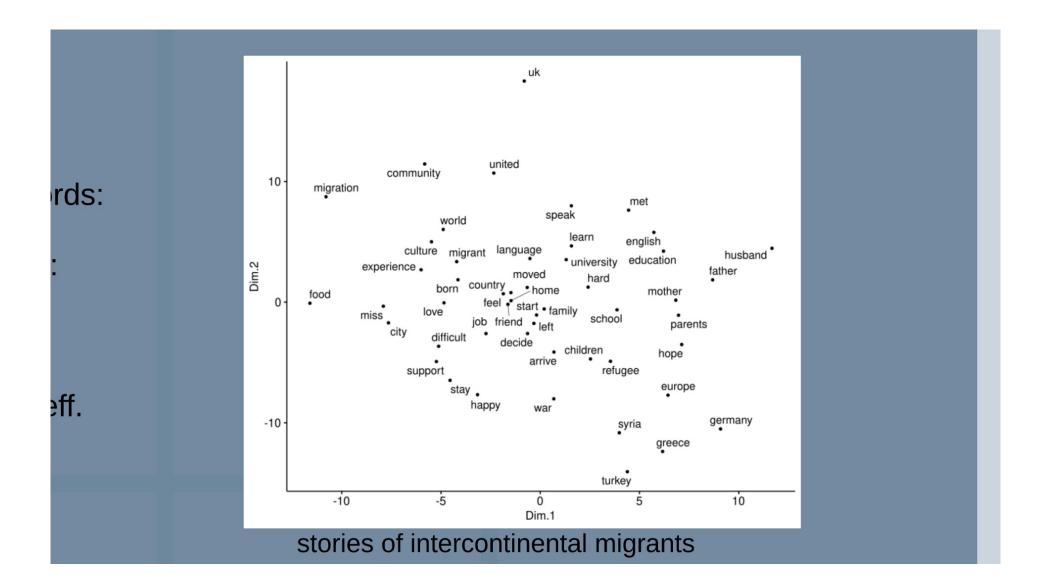


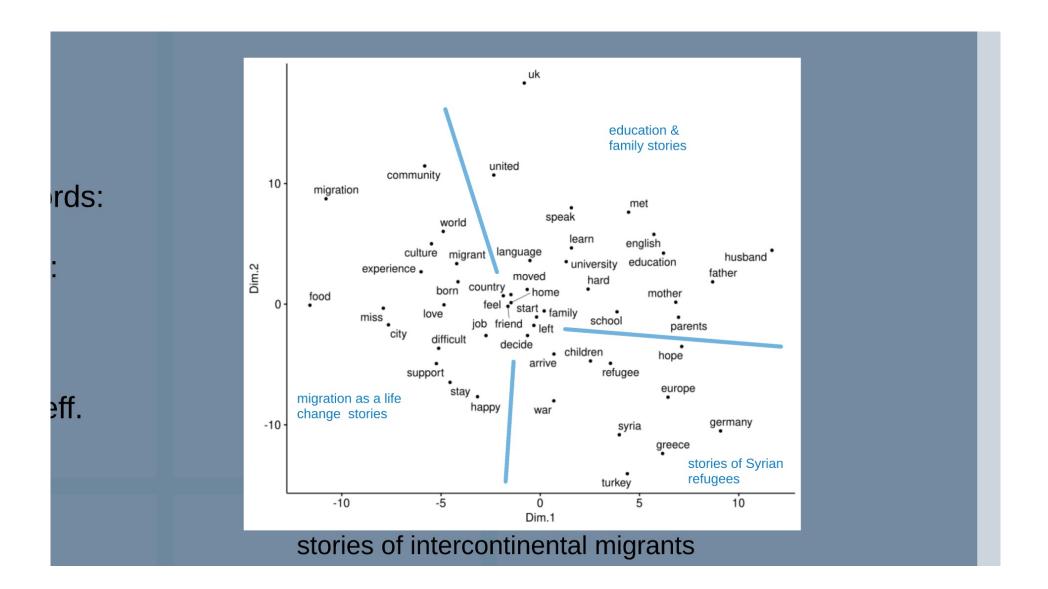
stories of intercontinental migrants





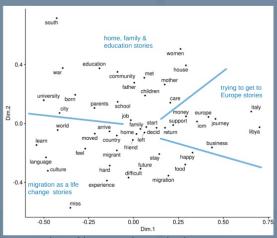
25.





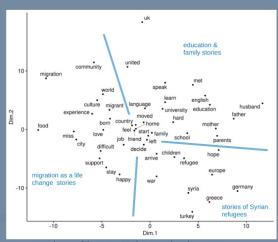
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Is there a difference between stories of intracontinental and intercontinental migrants?



stories of intracontinental migrants

- number of frequent words:
- context unit: story
- measure of proximity: Jaccard coeff.



stories of intercontinental migrants

Interpretation: both categories of migrants share "family & education" and "life change" migrant stories. In this sense, distance does not matter in narrating migration. On the other hand, each category contains a specific group of migrants (Syrian war refugees and African migrants via Libya) who narrated different stories.

Analysis of activist media communication

Data: internet communication of 12 Czech activist organizations and interest groups representing four domains of justice claim-making: trade unionism, feminism, human rights, environmentalism

Method: Counting and displaying co-occurrences of 33 most frequent lexical (semantic) words in texts in each domain







Results: internal split of discursive space in two areas

- the conditions area as the discursive space for verbalizing the existential conditions of subjects of contention,
- **the contention area** referring to the opposition strategies and negotiation with movement opponents.

Conclusion

Non-coding quantitative text analysis in sociology can be used to explore various social phenomena.

A good strategy is to combine textual and nontextual (behavioral, institutional, socio-demographic) data.

The quality of data is more important than the sophistication of analytical techniques.

Interpretation is often not obvious and has a form of hypothesis proposal.

Thank you for your attention!