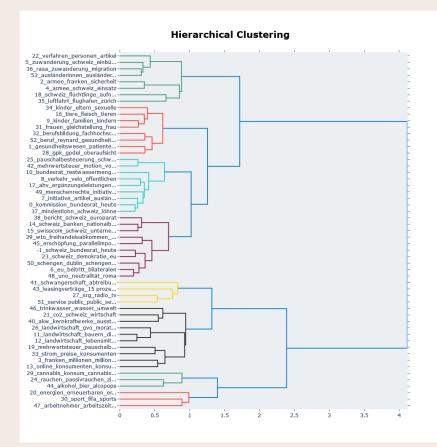
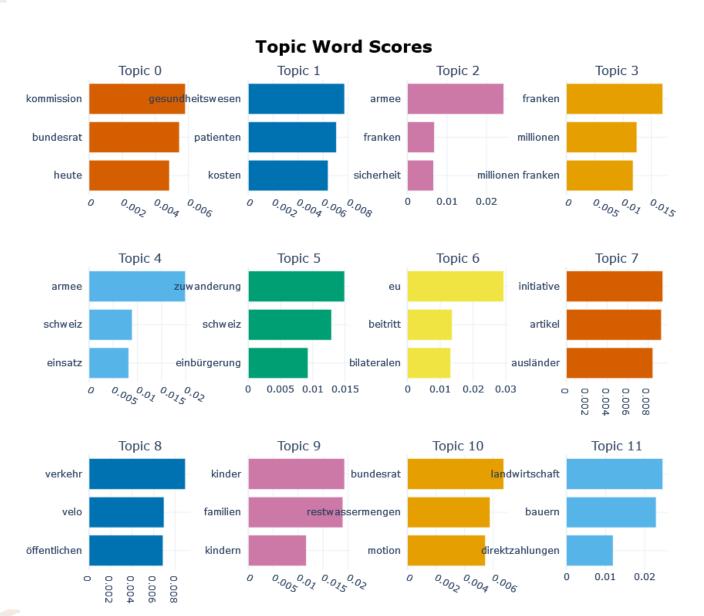


## Report – Key Findings

TEXT MINING EXERCISE 3

## Full Data Analysis on the speeches





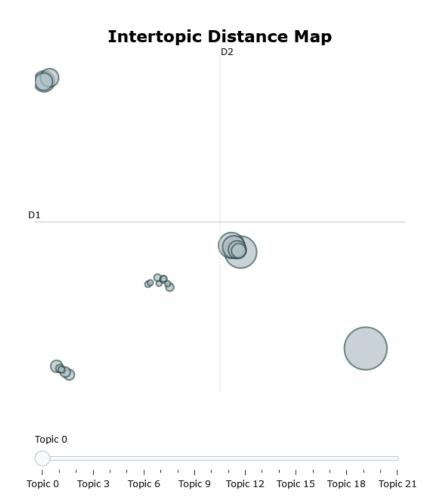
## Looking for similar topics for 'co2' SVP SP

```
[187] topics,similarity = topic_model.find_topics("co2", top_n=5)
topics,similarity = topic_model.find_topics(("co2", top_n=5|)
                                                                                       print(topics)
print(topics)
for top in topic:
                                                                                       for top in topics:
 to = topic model.get topic(top) # lsva: (Leistungsabhängige Schwerverkehrsabgabe)
                                                                                         to = topic model.get topic(top)
 print(to[0])
                                                                                         print(to[0])
                                                                                       [10, 26, 29, 16, 33]
[16, 12, 9, 19, 21]
                                                                                        ('energien', 0.020609364708972882)
('co2', 0.0614439804064923)
                                                                                        'erschöpfung', 0.04168836101149215)
('lsva', 0.01295256603120509)
('energien', 0.017631765368218717)
                                                                                        'akw', 0.02703706663042125)
('luftfahrt', 0.028443119183008677)
                                                                                        ('gpk', 0.019364436399263007)
('forschung', 0.018187095702797373)
                                                                                        ('natur', 0.01427017005899465)
```

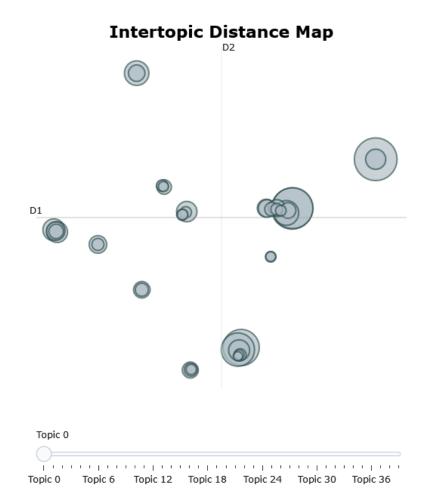
Interestingly, when we search for similar topics in the topic\_model based on SVP-speeches vs. SP-speeches we clearly see a difference. It seems as if SVP 'Co2' related topics focus more on Isva(Scherverkehrsabgabe = shear traffic tax), luftfahrt (aviation) and forschung(research), which are topics more economic related. Whereas, in the SP-speeches similar topics to 'CO2' are very different and 'Erschöpfung'= exhaustion in closely related to Co2.

# Visualize topics, their sizes, and their corresponding words





### SP

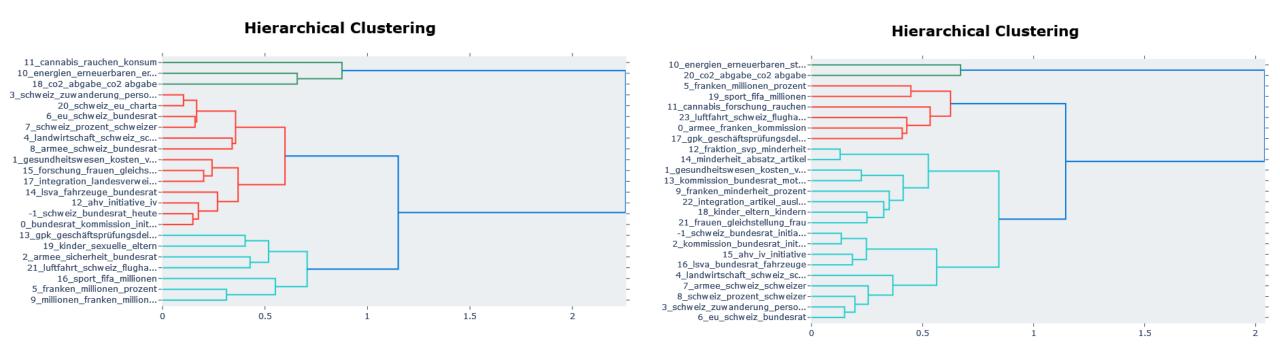


# Visualize topics, their sizes, and their corresponding words

We saw in the previous slide that the topics and their corresponding words are more diverse and spread out, in the SP-speeches.

Whereas the SVP-speeches topics are more less spreaded.

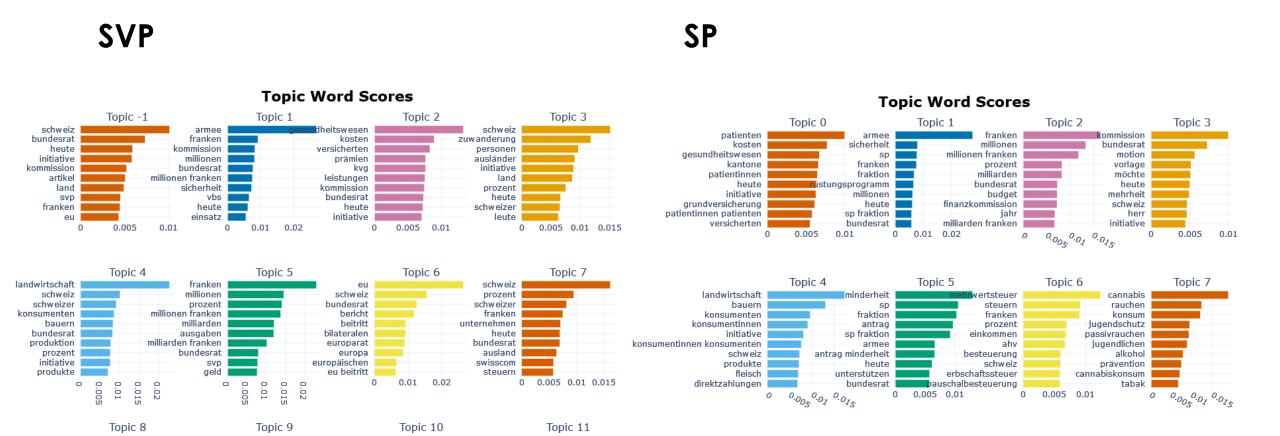
# Hierarchical Clustering: Visualize Topic Hierarchy SP SVP



# Hierarchical Clustering: Visualize Topic Hierarchy

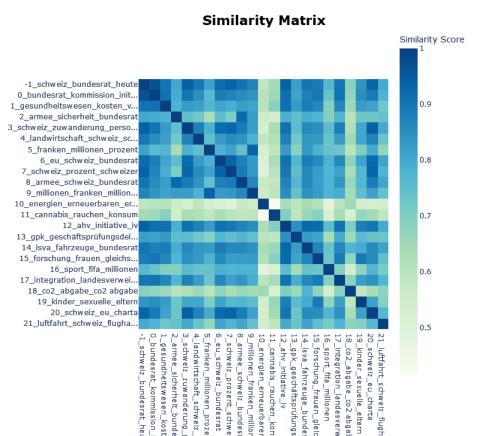
- We see a clear difference in the clustering of the topics.
- But it is clear, depending on the cut of the clustering tree, we can see a grouping of 3 colors (red, green, light blue).
- The green group seems to be about energy/co2 in the SP-speeches but according to the analysis on SVP-speeches also the cannabis consumption topic is a part of that group. This makes the interpretation more difficult.
- Also the interpretation of the red grouping is difficult, as it seems to be about moneyrelated topics in SP-speeches it is very different in the SVP-speeches, maybe about initiatives (Gesundheitswesen/Gleichstellung/AHV/cannabis).
- Then the light blue topic group in the SVP-speeches it seems to be more about money-related topics and in the SP-speeches more about initiatives.
- If we look into the groups and pick for example, the topic 5 it falls in both cases SVP/SP-speeches under a group related to money.
- According to the title and the group if falls into, topic 11 in SVP-speech in my opinion is quite a misfit.

### Visualize a barchart of selected topics

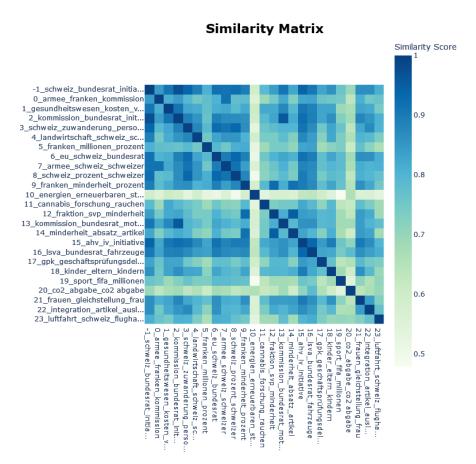


## Visualize Topic Similarity

SVP



#### SP



How well performs the BERTopic?
What keywords do you think have been used to filter the speeches? How long did we have?

- The keywords used in the German data are mainly nouns
- BERTopic performed pretty well but it does not take away the stopwords so one has to manually remove them or use NLTK.
- It took about a day.

### Problems and difficulties

Somewhat troubles interpreting the results. More than
plotting and seeing the differences in the parties is not
really possible. Maybe it would have been also more
interesting to have the dates of the speeches in order to
get the topics over time. This would make the data
analysis somewhat more interesting and meaningful.