## KED2022 – Miniprojekt

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#### Einleitung

#### **Thema**

- Geschlechterunterschiede bei 1. August-Reden:
  - Anzahl Ansprachen pro Jahr nach Geschlecht
  - Vokabular: welche Worte werden von welchem Geschlecht wie oft verwendet?
  - Wortgebrauch zwischen den Geschlechtern: Gemeinsamkeiten und Unterschiede

#### **Motivation**

- Aktualität
- Historische Entwicklung
- Rhetorik

# Anzahl Ansprachen pro Jahr nach Geschlecht

#### Code am Beispiel weiblicher Sprecherinnen

```
#make Python understand gender
m = "male"
f = "female"
Geschlecht = [m,f]
Python
```

```
#function to filter by metadata gender (female)
def filter_func_female(doc):
    return doc._.meta.get("Geschlecht") == 'f'
def filter_func_X1(doc):
    return doc._.meta.get("Jahr") > 2000

# create new corpus after applying filter function
subcor_female = textacy.corpus.Corpus(de, data=corpus_speeches_test.get(filter_func_female))

Python
```

```
#Export-corpus-as-csv-dataset

#in-this-case-we-used-the-female-dataset

#merge-metadata-and-actual-content-for-each-document-in-the-corpus

#-ugly, verbose-syntax-to-merge-two-dictionaries

data = [{**doc._.meta, ***{ 'text': doc.text} } - for-doc-in-subcor_female]

#-export-corpus-as-csv

f_csv-= '../KED2022/materials/data/dataset_speeches_f.csv'

textacy.io.csv.write_csv(data, -f_csv, -fieldnames=data[0].keys())

#-csv-format-is-the-best-to-load-in-scattertext

data[0]
```

```
#read dataset (female) from csv file
f_csv = '../KED2022/materials/data/dataset_speeches_f.csv'
df = pd.read_csv(f_csv)

# filter out non-german texts or very short texts
df_sub = df[(df['Sprache'] === 'de') & (df['text'].str.len() > 10)]

# make new column containing all relevant metadata (showing in plot later on)
df_sub['descripton'] = df_sub[['Redner', 'Partei', 'Jahr']].astype(str).agg(', '.join, axis=1)

# sneak peek of dataset
df_sub.head()
```

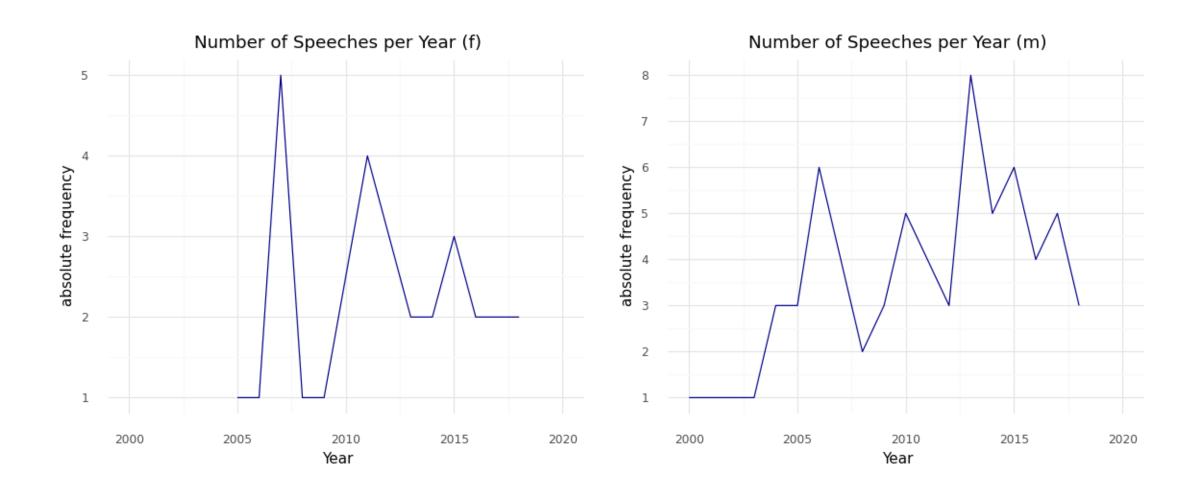
```
#create corpus for female dataset
def get_texts_from_csv(f_csv, text_column):
    Read dataset from a csv file and sequentially stream the rows,
    including metadata.
    -#-read-dataframe
    df = pd.read_csv(f_csv)
    # keep only documents that have text
    filtered_df = df[df[text_column].notnull()]
    # iterate over rows in dataframe
    for idx, row in filtered_df.iterrows():
        text = row[text_column].replace('\n', '')
        metadata = row.to_dict()
        del metadata[text_column]
        yield (text, metadata)
f_csv = '../KED2022/materials/data/dataset_speeches_f.csv'
texts = get_texts_from_csv(f_csv, text_column='text')
corpus_speeches_f = textacy.Corpus(de, data=texts)
```

```
# define what groups are formed and what terms should be included
# here, groups by year and words are lowercased (incl. stop words)
tokenized_docs, groups = textacy.io.unzip(
       (textacy.extract.utils.terms_to_strings(textacy.extract.words(doc, filter_stops=False), by="lower"),
  doc._.meta["Jahr"])
  for doc in corpus_speeches_f)
# here relative term frequency
vectorizer = textacy.representations.vectorizers.GroupVectorizer(
       tf_type='linear', # absolute term frequency
  -----dl_type="linear", # normalized by document length
  ••••••vocabulary_grps=range(1950, 2019)) # limit to years from 1950 to 2019
# create group-term-matrix with with frequency counts
grp_term_matrix = vectorizer.fit_transform(tokenized_docs, groups)
# create dataframe from matrix
df terms = pd.DataFrame.sparse.from spmatrix(grp term matrix, index=vectorizer.grps list, columns=vectorizer.terms list)
df_terms['year'] = df_terms.index
# change shape of dataframe
df tidy = df terms.melt(id vars='year', var name="term", value name="frequency")
df_tidy
```

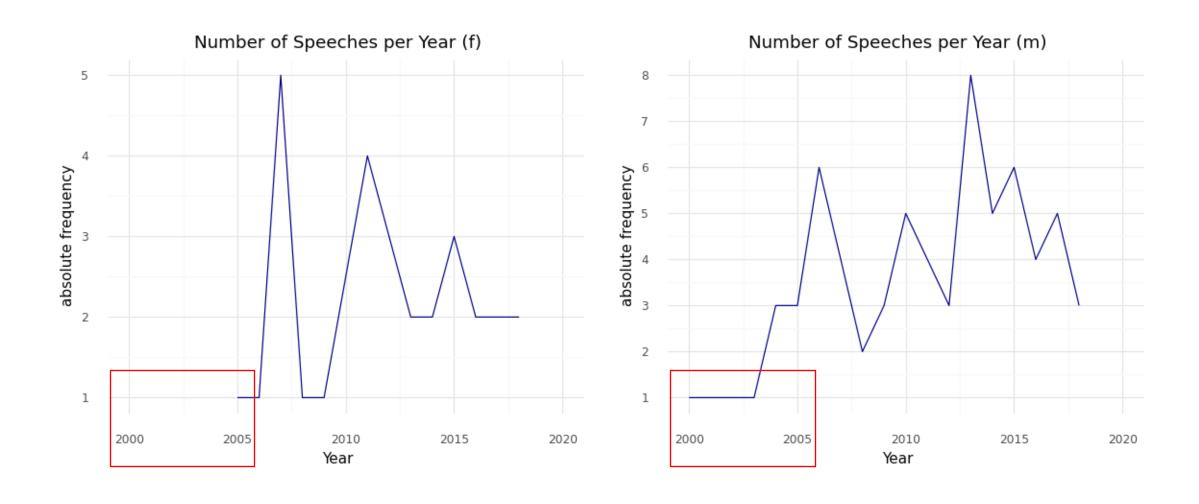
```
docs_per_year = df_sub.groupby('Jahr').agg({'text': "count" }).reset_index().rename(columns={'text':'count'})

(ggplot(docs_per_year, aes(x='Jahr', y='count'))
+ geom_line(color='darkblue')
+ labs(title = "Number of Speeches per Year (f)", x = "Year", y = "absolute frequency")
+ xlim(2000,2020)
+ scale_y_continuous(breaks=range(0, 11))
+ theme_minimal())
```

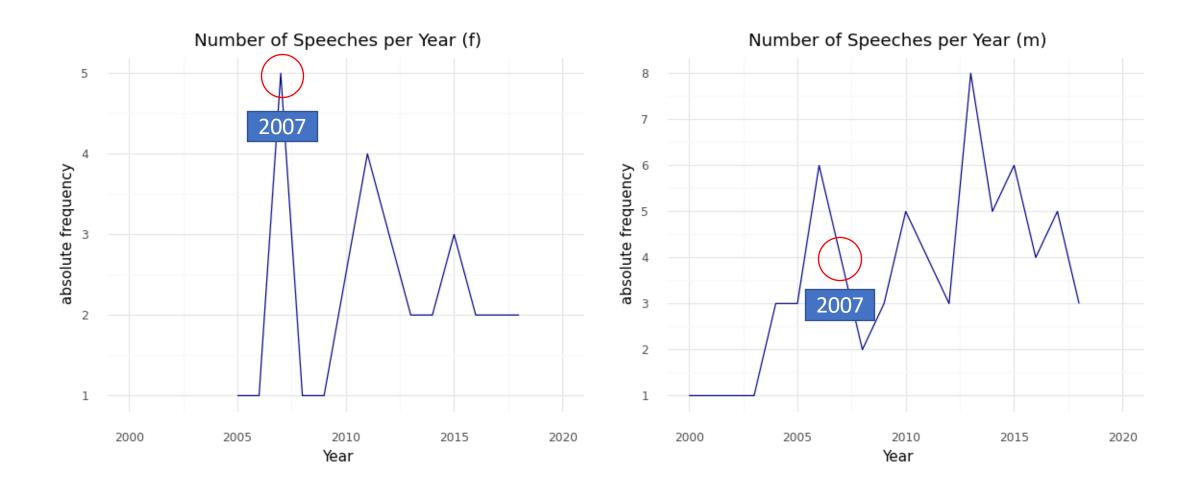
### Resultate: Anzahl Ansprachen pro Geschlecht



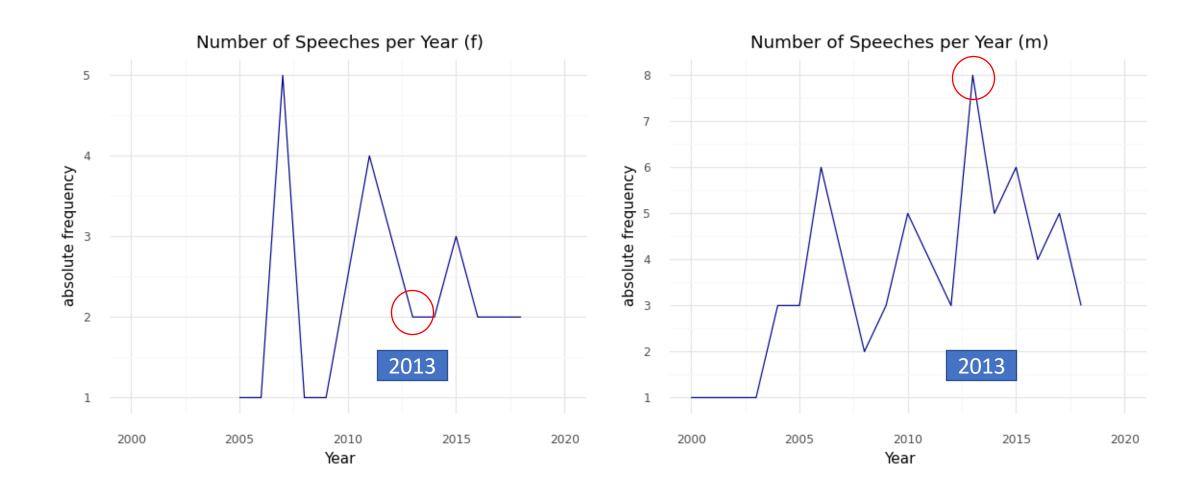
### Resultate: Anzahl Ansprachen pro Geschlecht



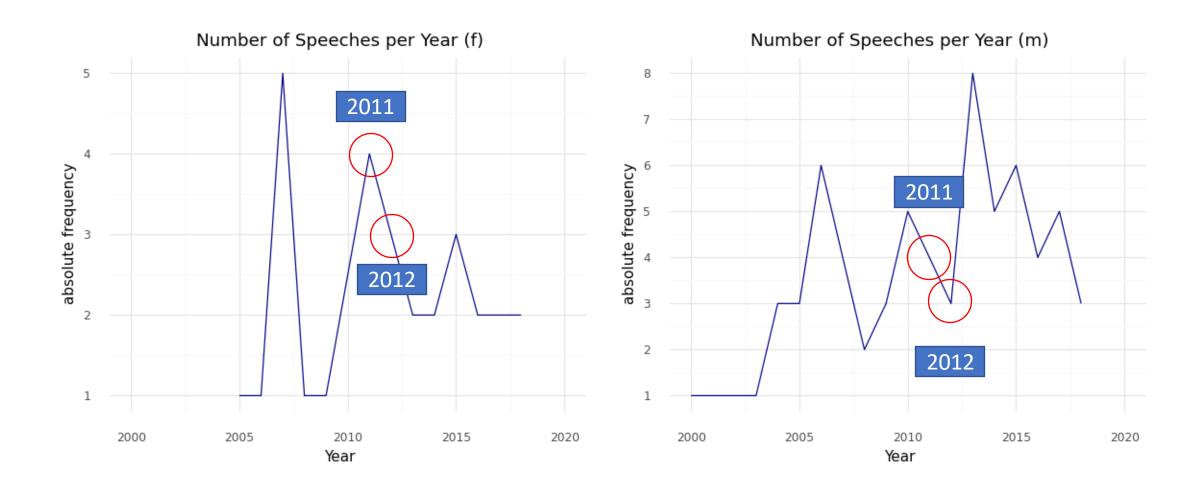
#### Resultate: Peak Frauen



#### Resultate: Peak Männer



#### Resultate: Gleichverteilung

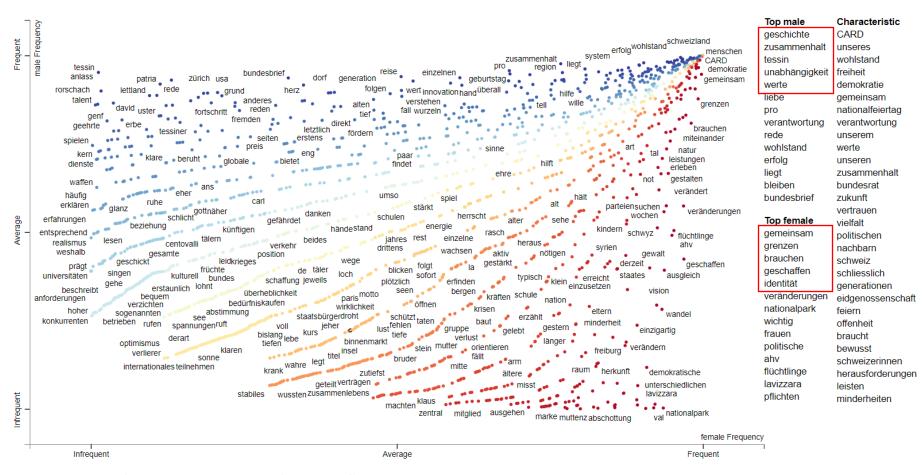


### Wortgebrauch nach Geschlecht

#### Code: Wortgebrauch nach Geschlecht

```
censor tags = set(['CARD']) # tags to ignore in corpus, e.g. numbers
de stopwords = spacy.lang.de.stop words.STOP WORDS # default stop words
custom stopwords = set(['[', ']', '%', '*', '•', '2.', '19.', '21.', '9.'])
de stopwords = de stopwords.union(custom stopwords) # extend with custom stop words
# create corpus from dataframe
corpus speeches = st.CorpusFromPandas(df sub, # dataset
                             category col='Geschlecht', # index differences by ...
                             text col='text',
                             nlp=de, # German model
                             feats from_spacy_doc=st.FeatsFromSpacyDoc(tag_types_to_censor=censor_tags, use_lemmas=False),
                            ).build().get stoplisted unigram corpus(de stopwords)
html = st.produce scattertext explorer(corpus speeches,
            category='m', # set attribute to divide corpus into two parts
            category name='male',
            not category name='female',
            metadata=df sub['descripton'],
            width in pixels=1000,
            minimum term frequency=5, # drop terms occurring less than 5 times
            save svg button=True,
# write visualization to html file
fname = '/home/valmey00/documents/KED2022/materials/data/gender differences2.html'
open(fname, 'wb').write(html.encode('utf-8'))
```

#### Resultat: Wortgebrauch nach Geschlecht



# Wortgebrauch zwischen den Geschlechtern

#### Code: Männer

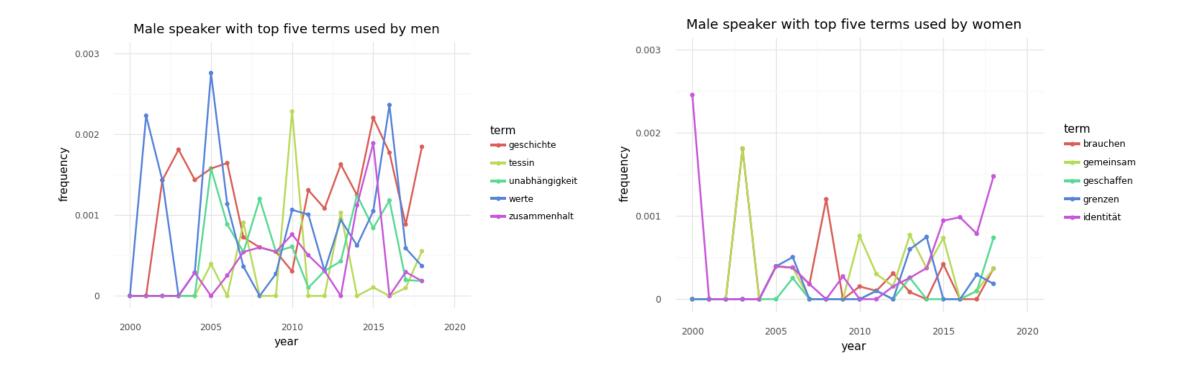
```
# filter the dataset for the five most used terms in speeches read by a female speaker
terms = ["geschichte", "zusammenhalt", "tessin", "unabhängigkeit", "werte"]

df_terms = df_tidy[df_tidy['term'].isin(terms)]

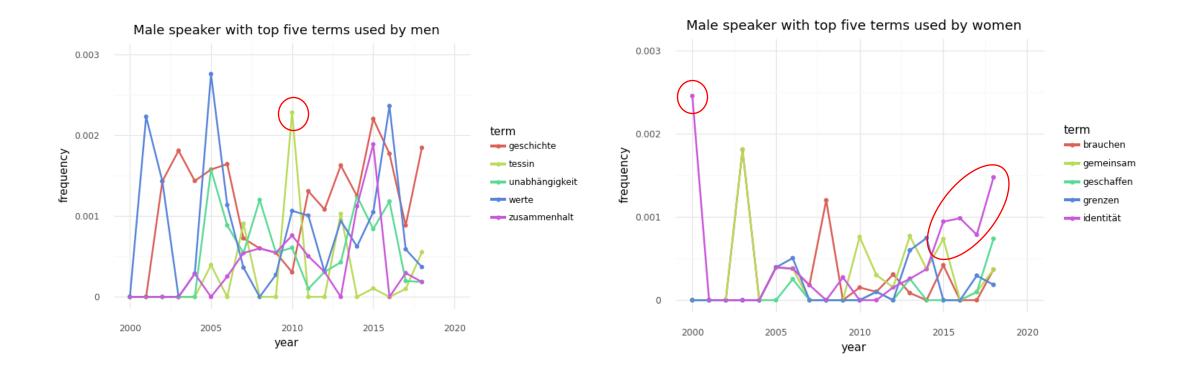
# plot the relative frequency for the terms above
(ggplot(df_terms, aes(x='year', y='frequency', color='term'))

+ ggtitle('Male speaker with top five terms used by men') # give plot a name to differentiate from male plot
+ geom_point() # show individual points
+ stat_smooth(method='lowess', span=0.15, se=False) # overlay points with a smoothed line
+ ylim(0,0.003)
+ xlim(2000,2020) # change x-axis numbers to match the data
+ theme_minimal()) # make the plot look nicer
```

#### Resultate: Männer



#### Resultate: Männer



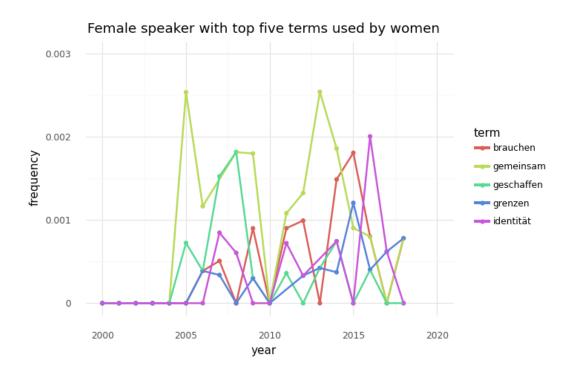
#### Code: Frauen

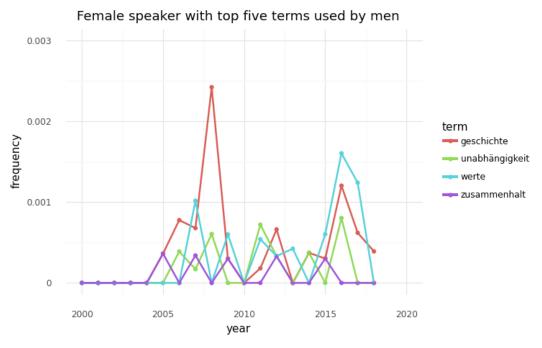
```
# filter the dataset for the five most used terms in speeches read by a female speaker
terms = ["gemeinsam", "grenzen", "brauchen", "geschaffen", "identität"]

df_terms = df_tidy[df_tidy['term'].isin(terms)]

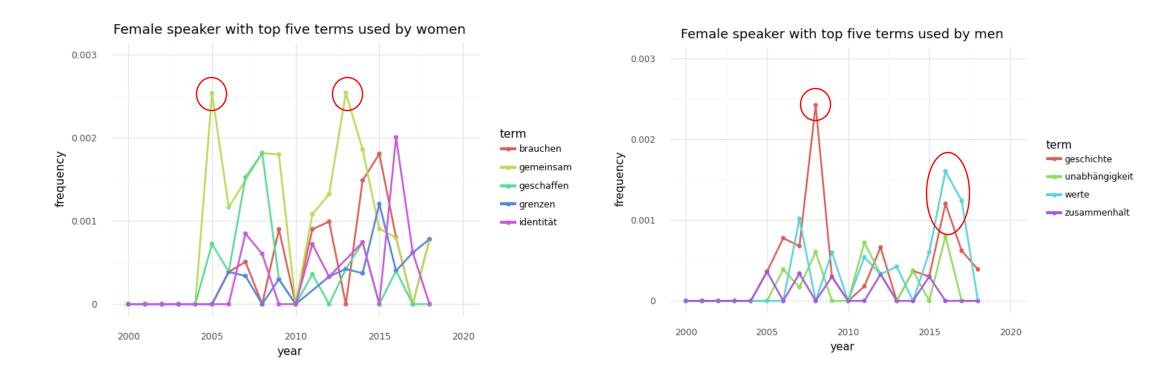
# plot the relative frequency for the terms above
(ggplot(df_terms, aes(x='year', y='frequency', color='term'))
+ ggtitle('Female speaker with top five terms used by women') # give plot a name to differentiate from male plot
+ geom_point() # show individual points
+ stat_smooth(method='lowess', span=0.15, se=False) # overlay points with a smoothed line
+ ylim(0,0.003)
+ xlim(2000,2020) # change x-axis numbers to match the data
+ theme_minimal()) # make the plot look nicer
```

#### Resultate: Frauen





#### Resultate: Frauen



## Fragen?

Danke für eure Aufmerksamkeit!