CWordTM Toolkit Usage on BBC News

This Jupyter notebook demonstrates how to use the package "CWordTM" on the BBC News:

- 1. Meta Information Features
- 2. Utility Features
- 3. Text Visualization Word Cloud
- 4. Text Summarization
- 5. Topic Modeling LDA and BERTopic

```
In [1]: import warnings
warnings.filterwarnings('ignore')
```

1. Meta Information Features

```
In [2]: import cwordtm
         from cwordtm import *
        [n]{tk\_data}] \ \ Downloading \ package \ stopwords \ to
         [nltk_data]
                         C:\Users\johnnyc\AppData\Roaming\nltk_data...
        [nltk_data]
                       Package stopwords is already up-to-date!
        [{\tt nltk\_data}] \ {\tt Downloading} \ {\tt package} \ {\tt wordnet} \ {\tt to}
         [nltk_data]
                         C:\Users\johnnyc\AppData\Roaming\nltk_data...
         [nltk_data]
                       Package wordnet is already up-to-date!
         [nltk_data] Downloading package punkt to
        [nltk_data]
                        C:\Users\johnnyc\AppData\Roaming\nltk_data...
        [nltk_data] Package punkt is already up-to-date!
         [nltk_data] Downloading package averaged_perceptron_tagger to
        [nltk_data]
                         C:\Users\johnnyc\AppData\Roaming\nltk_data...
         [nltk_data]
                       Package averaged_perceptron_tagger is already up-to-
        [nltk_data]
                           date!
In [3]: cwordtm.__version__
         '0.6.3'
Out[3]:
In [4]: # Show brief module information
        print(meta.get_module_info())
```

```
The member information of the module 'cwordtm'
1. Submodule meta:
    addin (func)
    addin_all (modname='cwordtm')
    addin_all_functions (submod)
    get_function (mod_name, submodules, func_name)
    get_module_info (detailed=False)
    get_submodule_info (submodname, detailed=False)
    import_module (name, package=None)
    wraps \ \overline{(wrapped, assigned=('\_module\_', '\_name\_', '\_qualname\_', '\_doc\_', '\_annotations\_')}, \ updated=('\_diame\_', '\_doc\_', '\_annotations\_'), \ updated=('\_diame\_', '\_annotations\_')
ct__',))
2. Submodule pivot:
    stat (df, chi=False, *, timing=False, code=0)
3. Submodule quot:
    extract_quotation (text, quot_marks, *, timing=False, code=0)
    match_text (target, sent_tokens, lang, threshold, n=5, *, timing=False, code=0)
    match_verse (i, ot_list, otdf, df, book, chap, verse, lang, threshold, *, timing=False, code=0)
    show_quot (target, source='ot', lang='en', threshold=0.5, *, timing=False, code=0)
    tokenize (sentence, *, timing=False, code=0)
4. Submodule ta:
    get_sent_scores (sentences, diction, sent_len, *, timing=False, code=0) -> dict
    get_sentences (docs, lang='en', *, timing=False, code=0)
    get_summary (sentences, sent_weight, threshold, sent_len, *, timing=False, code=0)
    pos_tag (tokens, tagset=None, lang='eng', *, timing=False, code=0)
    preprocess_sent (text, *, timing=False, code=0)
    sent_tokenize (text, language='english', *, timing=False, code=0)
    summary_chi (docs, weight=1.5, sent_len=8, *, timing=False, code=0)
    summary_en (docs, sent_len=8, *, timing=False, code=0)
    word_tokenize (text, language='english', preserve_line=False, *, timing=False, code=0)
5. Submodule tm:
    BTM (textfile, chi=False, num_topics=15, embed=True)
    LDA (textfile, chi=False, num_topics=15)
    NMF (textfile, chi=False, num_topics=15)
    btm_process (doc_file, source=0, text_col='text', cat=0, chi=False, group=True, eval=False, *, timing=False, code
=0)
    lda_process (doc_file, source=0, text_col='text', cat=0, chi=False, group=True, eval=False, *, timing=False, code
=0)
    load_bible (textfile, cat=0, group=True, *, timing=False, code=0)
    load_text (textfile, text_col='text', *, timing=False, code=0)
    ngrams (sequence, n, *, timing=False, code=0, **kwargs)
    nmf_process (doc_file, source=0, text_col='text', cat=0, chi=False, group=True, eval=False, *, timing=False, code
=0)
    pprint (object, stream=None, indent=1, width=80, depth=None, *, compact=False, sort_dicts=True, underscore_number
s=False, timing=False, code=0)
    process_text (doc, *, timing=False, code=0)
6. Submodule util:
    add_chi_vocab (*, timing=False, code=0)
    bible_cat_info (lang='en', *, timing=False, code=0)
chi_sent_terms (text, *, timing=False, code=0)
    chi_stops (*, timing=False, code=0)
    clean_sentences (sentences, *, timing=False, code=0)
clean_text (df, text_col='text', *, timing=False, code=0)
    extract (df, testament=-1, category='', book=0, chapter=0, verse=0, *, timing=False, code=0)
    extract2 (df, filter='', *, timing=False, code=0)
    get diction (docs, *, timing=False, code=0)
    get_diction_chi (docs, *, timing=False, code=0)
get_diction_en (docs, *, timing=False, code=0)
    get_list (df, column='book', *, timing=False, code=0)
    get_sent_terms (text, *, timing=False, code=0)
    get_text (df, text_col='text', *, timing=False, code=0)
    get_text_list (df, text_col='text', *, timing=False, code=0)
group_text (df, column='chapter', *, timing=False, code=0)
    is_chi (*, timing=False, code=0)
    load_text (filepath, nr=0, info=False, *, timing=False, code=0)
    load_word (ver='web.csv', nr=0, info=False, *, timing=False, code=0)
    preprocess_text (text, *, timing=False, code=0)
    remove_noise (text, noise_list, *, timing=False, code=0)
    set_lang (lang='en', *, timing=False, code=0)
    word_tokenize (text, language='english', preserve_line=False, *, timing=False, code=0)
7. Submodule version:
8. Submodule viz:
    \label{local_condition} chi\_wordcloud \ (\mbox{docs, figsize=(15, 10), bg='white', image=0, *, timing=False, code=0)}
    plot_cloud (wordcloud, figsize, *, timing=False, code=0)
     show_wordcloud (docs, clean=False, figsize=(12, 8), bg='white', image=0, *, timing=False, code=0)
```

```
In [5]: # Show detailed module information of a submodule
print(meta.get_submodule_info("viz", detailed=True))
```

```
The function(s) of the submodule 'cwordtm.viz':
def chi_wordcloud(docs, figsize=(15, 10), bg='white', image=0):
    """Prepare and show a Chinese wordcloud
   :param docs: The collection of Chinese documents for preparing a wordcloud,
       default to None
    :type docs: pandas.DataFrame
    :param figsize: Size (width, height) of word cloud, default to (15, 10)
    :type figsize: tuple, optional
    :param bg: The background color (name) of the wordcloud, default to 'white'
    :type bg: str, optional
    :param image: The filename of the presribed image as the mask of the wordcloud,
       or 1/2/3/4 for using an internal image (heart / disc / triangle / arrow),
       default to 0 (No image mask)
    :type image: int or str, optional
   util.set_lang('chi')
   diction = util.get_diction(docs)
   masks = ['heart.jpg', 'disc.jpg', 'triangle.jpg', 'arrow.jpg']
   if image == 0:
       mask = None
    elif image in [1, 2, 3, 4]: # Internal image file
       img file = files('cwordtm.images').joinpath(masks[image-1])
       mask = np.array(Image.open(img_file))
   elif isinstance(image, str) and len(image) > 0:
       mask = np.array(Image.open(image))
   else:
       mask = None
   font_file = files('cwordtm.data').joinpath('msyh.ttc')
   wordcloud = WordCloud(background_color=bg, colormap='Set2',
                          mask=mask, font_path=str(font_file)) \
                    .generate_from_frequencies(frequencies=diction)
   plot cloud(wordcloud, figsize=figsize)
def plot_cloud(wordcloud, figsize):
     ""Plot the prepared 'wordcloud'
    :param wordcloud: The WordCloud object for plotting, default to None
    :type wordcloud: WordCloud object
    :param figsize: Size (width, height) of word cloud, default to None
    :type figsize: tuple
   plt.figure(figsize=figsize)
   plt.imshow(wordcloud)
   plt.axis("off");
def show_wordcloud(docs, clean=False, figsize=(12, 8), bg='white', image=0):
    """Prepare and show a wordcloud
    :param docs: The collection of documents for preparing a wordcloud,
       default to None
    :type docs: pandas.DataFrame
    :param clean: The flag whether text preprocessing is needed,
       default to False
    :type clean: bool, optional
    :param figsize: Size (width, height) of word cloud, default to (12, 8)
    :type figsize: tuple, optional
    :param bg: The background color (name) of the wordcloud, default to 'white'
    :type bg: str, optional
    :param image: The filename of the presribed image as the mask of the wordcloud,
       or 1/2/3/4 for using an internal image (heart / disc / triangle / arrow),
       default to 0 (No image mask)
    :type image: int or str, optional
   masks = ['heart.jpg', 'disc.jpg', 'triangle.jpg', 'arrow.jpg']
   if image == 0:
       mask = None
    elif image in [1, 2, 3, 4]: # Internal image file
       img_file = files('cwordtm.images').joinpath(masks[image-1])
        mask = np.array(Image.open(img_file))
   elif isinstance(image, str) and len(image) > 0:
       mask = np.array(Image.open(image))
    else:
       mask = None
   if isinstance(docs, pd.DataFrame):
        docs = ' '.join(list(docs.text.astype(str)))
    elif isinstance(docs, pd.Series):
```

```
docs = ' '.join(list(docs.astype(str)))
            elif isinstance(docs, list) or isinstance(docs, np.ndarray):
                docs = ' '.join(str(doc) for doc in docs)
            if clean:
                docs = util.preprocess_text(docs)
            wordcloud = WordCloud(background_color=bg, colormap='Set2', mask=mask) \
                            .generate(docs)
            plot_cloud(wordcloud, figsize=figsize)
In [6]: # Show execution time
        df = util.load_text("BBC/BBC News Train.csv", timing=True)
        Finished 'load_text' in 0.0281 secs
In [7]: # Execute and show code
        df = util.load_text("BBC/BBC News Train.csv", code=1)
        def load_text(filepath, nr=0, info=False):
            """Loads and returns the text from the prescribed file path ('filepath').
            :param filepath: The prescribed filepath from which the text is loaded,
                default to None
            :type filepath: str
            :param nr: The number of rows of text to be loaded; 0 represents all rows,
                default to 0
            :type nr: int, optional
            :param info: The flag whether the dataset information is shown,
                default to False
            :type info: bool, optional
            :return: The collection of text with the prescribed number of rows loaded
            :rtype: pandas.DataFrame
            # print("Loading file '%s' ..." %filepath)
            if filepath.lower().endswith('csv'):
                nrows = None
                if nr > 0: nrows = nr
                df = pd.read_csv(filepath, nrows=nrows, encoding='utf-8')
            else:
                noise_list = ['\u3000', '- ', '•']
                tf = open(filepath, encoding='utf-8')
                lines = [remove_noise(line, noise_list) for line in tf.readlines()]
                lines = list(filter(None, lines))
                df = pd.DataFrame({'text': lines})
                if nr > 0: df = df.iloc[:nr]
            if info:
                print("\nDataset Information:")
                df.info()
            return df
        >> cwordtm.util.remove_noise
        def remove_noise(text, noise_list):
             """Removes a list of substrings in noise_list from the input text.
            :param text: The input text, default to None
            :type text: str
            :param noise_list: The list of substrings to be removed, default to ""
            :type noise_list: list, optional
            :return: The text with the prescribed substrings removed
            :rtype: str
            text = text.rstrip()
            for noise in noise_list:
                text = text.replace(noise, '')
            return text
In [8]: # Show code without execution
        df = util.load_text("BBC/BBC News Train.csv", code=2)
```

```
def load_text(filepath, nr=0, info=False):
             """Loads and returns the text from the prescribed file path ('filepath').
            :param filepath: The prescribed filepath from which the text is loaded,
                default to None
            :type filepath: str
            :param nr: The number of rows of text to be loaded; 0 represents all rows,
                default to 0
            :type nr: int, optional
            :param info: The flag whether the dataset information is shown,
                default to False
            :type info: bool, optional
            :return: The collection of text with the prescribed number of rows loaded
            :rtype: pandas.DataFrame
            # print("Loading file '%s' ..." %filepath)
            if filepath.lower().endswith('csv'):
                nrows = None
                if nr > 0: nrows = nr
                df = pd.read_csv(filepath, nrows=nrows, encoding='utf-8')
            else:
                noise_list = ['\u3000', '- ', '•']
                tf = open(filepath, encoding='utf-8')
                lines = [remove_noise(line, noise_list) for line in tf.readlines()]
                lines = list(filter(None, lines))
                df = pd.DataFrame({'text': lines})
                if nr > 0: df = df.iloc[:nr]
                print("\nDataset Information:")
                df.info()
            return df
        >> cwordtm.util.remove_noise
        def remove_noise(text, noise_list):
             ""Removes a list of substrings in noise_list from the input text.
            :param text: The input text, default to None
            :type text: str
            :param noise_list: The list of substrings to be removed, default to ""
            :type noise_list: list, optional
            :return: The text with the prescribed substrings removed
            :rtype: str
            text = text.rstrip()
            for noise in noise list:
                text = text.replace(noise, '')
            return text
In [9]: # Add timing and code reveal features to some other function
        from importlib_resources import files
        files = meta.addin(files)
        files(code=2)
        @package_to_anchor
        def files(anchor: Optional[Anchor] = None) -> Traversable:
            Get a Traversable resource for an anchor.
            return from_package(resolve(anchor))
```

2. Utility Features

Load BBC News

```
In [10]: bbc_file = "BBC/BBC News Train.csv"
df = util.load_text(bbc_file, info=True)
```

```
Dataset Information:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1490 entries, 0 to 1489
Data columns (total 3 columns):
               Non-Null Count Dtype
#
    Column
---
0
     ArticleId 1490 non-null
                                int64
                1490 non-null
 1
     Text
                                object
2
               1490 non-null
    Category
                                object
dtypes: int64(1), object(2)
memory usage: 35.0+ KB
```

Preprocessing Text

```
In [11]: text_list = util.get_text_list(df.iloc[:500], text_col='Text')
    text = util.preprocess_text(text_list)
```

3. Text Visualization - Word Cloud

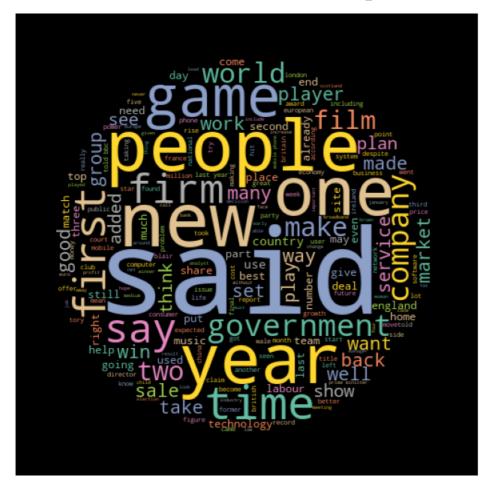
```
In [12]: # White background with no image mask
viz.show_wordcloud(text)
```

D:\Dev\Anaconda3\lib\site-packages\wordcloud\wordcloud.py:106: MatplotlibDeprecationWarning: The get_cmap function w as deprecated in Matplotlib 3.7 and will be removed two minor releases later. Use ``matplotlib.colormaps[name]`` or ``matplotlib.colormaps.get_cmap(obj)`` instead.



In [13]: # Black background with the prescribed image as the mask
 viz.show_wordcloud(text, bg='black', image='images/disc.png')

D:\Dev\Anaconda3\lib\site-packages\wordcloud\wordcloud.py:106: MatplotlibDeprecationWarning: The get_cmap function w
as deprecated in Matplotlib 3.7 and will be removed two minor releases later. Use ``matplotlib.colormaps[name]`` or
``matplotlib.colormaps.get_cmap(obj)`` instead.
self.colormap = plt.cm.get_cmap(colormap)



4. Text Summarization

```
In [14]: news = df.iloc[:5]['Text'] # "df" stores previously loaded text
ta.summary_en(news, sent_len=5)
```

Out[14]: ['but ms cooper who now runs her own consulting business told a jury in new york on wednesday that external audito rs arthur andersen had approved worldcom s accounting in early 2001 and 2002. she said andersen had given a green 1 ight to the procedures and practices used by worldcom.',

'cynthia cooper worldcom s ex-head of internal accounting alerted directors to irregular accounting practices at the us telecoms giant in 2002. her warnings led to the collapse of the firm following the discovery of an \$11bn (£5.7bn) accounting fraud.',

'prosecution lawyers have argued that mr ebbers orchestrated a series of accounting tricks at worldcom ordering em ployees to hide expenses and inflate revenues to meet wall street earnings estimates.',

'the university of california said the trial in the case is scheduled to begin in october 2006. it joined the lawsu it in december 2001alleging massive insider trading and fraud claiming it had lost \$145m on its investments in the company.',

'the bbc s david willey in rome says one reason for that result is the changeover from the lira to the euro in 2001 which is widely viewed as the biggest reason why their wages and salaries are worth less than they used to be.']

5. Topic Modeling

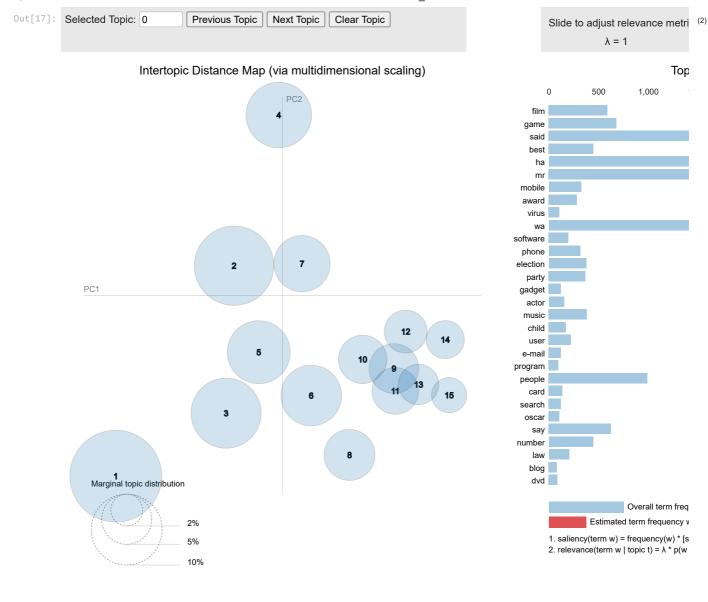
```
In [15]: import warnings
warnings.filterwarnings('ignore')
```

LDA Model

```
In [16]: doc_file = "BBC/BBC News Train.csv"
lda = tm.lda_process(doc_file, source=1, text_col='Text', eval=True, timing=True)
```

```
Cornus loaded!
Text preprocessed!
Text trained!
If no visualization is shown,
  you may execute the following commands to show the visualization:
   > import pyLDAvis
    > pyLDAvis.display(lda.vis_data)
Visualization prepared!
Topics from LDA Model:
[(0,
  .
'0.005*"said" + 0.004*"ha" + 0.003*"wa" + 0.002*"year" + 0.002*"film" + '
  '0.002*"new" + 0.001*"world" + 0.001*"test" + 0.001*"law" + 0.001*"share"'),
 (1,
  '0.007*"said" + 0.006*"wa" + 0.003*"new" + 0.003*"year" + 0.003*"mr" + '
  '0.003*"ha" + 0.002*"sale" + 0.002*"uk" + 0.001*"world" +
  '0.001*"government"'),
  '0.006*"said" + 0.006*"wa" + 0.004*"ha" + 0.004*"film" + 0.004*"best" + '
  '0.003*"year" + 0.002*"mr" + 0.002*"award" + 0.002*"new" + 0.002*"actor"'),
  '0.002*"said" + 0.002*"wa" + 0.002*"mr" + 0.002*"say" + 0.002*"election" +
  '0.001*"party" + 0.001*"year" + 0.001*"home" + 0.001*"ha" + 0.001*"suspect"'),
  '0.005*"said" + 0.003*"ha" + 0.003*"wa" + 0.002*"mr" + 0.001*"game" + '
  '0.001*"id" + 0.001*"card" + 0.001*"id_card" + 0.001*"howard" +
  '0.001*"child"'),
 (5,
  '0.006*"said" + 0.006*"wa" + 0.004*"ha" + 0.003*"mr" + 0.003*"film" + '
  '0.002*"people" + 0.002*"year" + 0.001*"phone" + 0.001*"best" +
  '0.001*"award"'),
  '0.010*"wa" + 0.007*"said" + 0.004*"ha" + 0.002*"year" + 0.002*"game" + '
  '0.002*"mr" + 0.001*"time" + 0.001*"new" + 0.001*"england" + 0.001*"bn"'),
  '0.006*"said" + 0.005*"ha" + 0.005*"wa" + 0.003*"people" + 0.003*"year" + '
  '0.002*"mr" + 0.002*"film" + 0.002*"world" + 0.002*"technology" +
  '0.002*"new"'),
  '0.005*"said" + 0.005*"ha" + 0.003*"wa" + 0.003*"mr" + 0.002*"game" + '
  '0.002*"people" + 0.002*"year" + 0.001*"election" + 0.001*"new" +
  '0.001*"number"'),
  '0.009*"said" + 0.005*"ha" + 0.005*"wa" + 0.005*"mr" + 0.003*"vear" + '
  '0.003*"people" + 0.002*"new" + 0.002*"government" + 0.002*"bn" +
  '0.002*"say"'),
 (10,
  .
0.004*"said" + 0.004*"ha" + 0.003*"wa" + 0.003*"game" + 0.002*"year" + '
'0.002*"mr" + 0.002*"time" + 0.001*"child" + 0.001*"dvd" + 0.001*"world"'),
  '0.006*"said" + 0.004*"wa" + 0.003*"ha" + 0.002*"virus" + 0.002*"mr" + '
  '0.002*"software" + 0.002*"people" + 0.002*"e-mail" + 0.001*"program" + '
  '0.001*"new"'),
 (12,
  '0.004*"said" + 0.003*"wa" + 0.002*"ha" + 0.002*"number" + 0.002*"year" + '
  '0.001*"time" + 0.001*"game" + 0.001*"people" + 0.001*"user" +
  '0.001*"service"'),
 (13,
  '0.005*"said" + 0.004*"ha" + 0.004*"wa" + 0.003*"game" + 0.002*"company" + '
'0.002*"mr" + 0.002*"say" + 0.002*"new" + 0.002*"year" + 0.002*"world"'),
  '0.005*"said" + 0.004*"ha" + 0.003*"wa" + 0.002*"mobile" + 0.002*"year" + '
  '0.002*"people" + 0.002*"phone" + 0.002*"music" + 0.002*"gadget" +
  '0.001*"new"')]
Model Evaluation Scores:
  Coherence: 0.6423303672893046
  Perplexity: -11.29942365009797
  Topic diversity: 0.0007698825601324054
  Topic size distribution: 0.0022087244616234127
Finished 'lda_process' in 72.6000 secs
import pyLDAvis
pyLDAvis.display(lda.vis_data)
```

In [17]: # LDA Model Visualization



BERTopic Model

In [18]: btm = tm.btm_process(doc_file, source=1, text_col='Text', eval=True, timing=True)

Corpus loaded!

Text preprocessed!

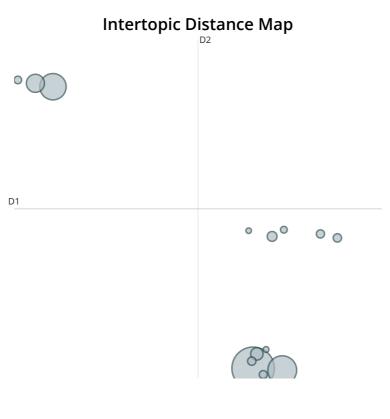
Some weights of the model checkpoint at bert-base-uncased were not used when initializing BertModel: ['cls.predictions.bias', 'cls.predictions.transform.dense.bias', 'cls.predictions.transform.LayerNorm.weight', 'cls.predictions.transform.dense.weight', 'cls.predictions.transform.bias', 'cls.seq_relationship.weight', 'cls.seq_relationship.bias', 'cls.predictions.decoder.weight']

- This IS expected if you are initializing BertModel from the checkpoint of a model trained on another task or with another architecture (e.g. initializing a BertForSequenceClassification model from a BertForPreTraining model).
- This IS NOT expected if you are initializing BertModel from the checkpoint of a model that you expect to be exactly identical (initializing a BertForSequenceClassification model from a BertForSequenceClassification model).

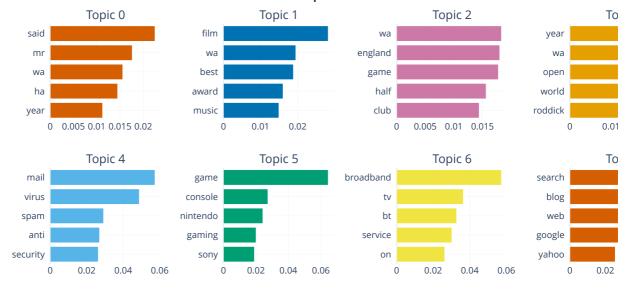
Text trained!

```
Topics from BERTopic Model:
Topic 0: said | mr | wa | ha | year | government | election | labour | bn | party
Topic 1: film | wa | best | award | music | year | said | star | ha | actor
Topic 2: wa | england | game | half | club | ha | player | time | team | said
Topic 3: year | wa | open | world | roddick | champion | old | win | final | ha
Topic 4: mail | virus | spam | anti | security | site | spyware | user | said | attack
Topic 5: game | console | nintendo | gaming | sony | gamers | title | xbox | halo | player
Topic 6: broadband \mid tv \mid bt \mid service \mid on \mid speed \mid net \mid customer \mid people \mid uk
Topic 7: search | blog | web | google | yahoo | people | search_engine | said | user | desktop
Topic 8: phone | mobile | camera | mobile_phone | people | technology | handset | camera_phone | use | said
Topic 9: yukos | russian | russia | gazprom | tax | oil | company | bn | khodorkovsky | ha
Topic 10: doping | test | kenteris | iaaf | conte | greek | drug | thanou | sprinter | athens
Topic 11: technology | digital | gadget | device | electronics | consumer | ce | content | consumer electronics | pe
ople
Topic 12: file | peer | sharing | pp | to | network | said | apple | piracy | firm
Topic 13: mac | mini | mac_mini | pc | computer | commodore | apple | laptop | machine | said
Model Evaluation Scores:
  Coherence: 0.6576968143443093
```

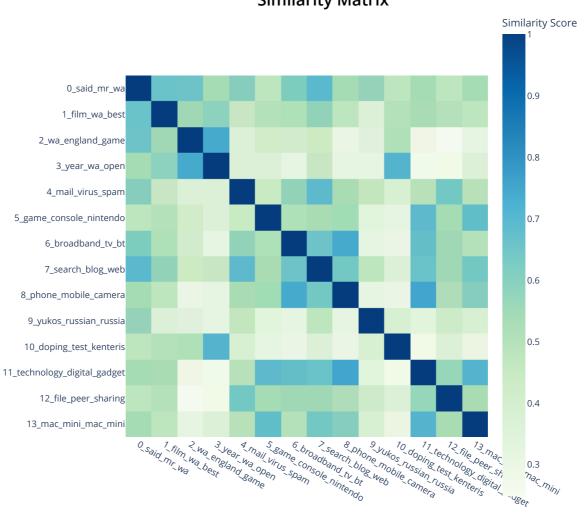
BERTopic Model Visualization:



Topic Word Scores



Similarity Matrix



```
If no visualization is shown,
  you may execute the following commands one-by-one:
    btm.model.visualize_topics()
    btm.model.visualize_barchart()
    btm.model.visualize_heatmap()
Finished 'btm_process' in 125.4787 secs
```

NMF Model (Code Reveal Only)

```
In [19]: nmf = tm.nmf_process(doc_file, source=1, text_col='Text', eval=True, timing=True, code=1)
```

```
Corpus loaded!
Text preprocessed!
Text trained!
Topics-Words from NMF Model:
Topic 1:
best (0.002166)
ireland (0.002122)
year (0.002009)
world (0.001799)
second (0.001739)
sale (0.001706)
england (0.001704)
france (0.001452)
open (0.001429)
net (0.001112)
Topic 2:
new (0.008297)
said (0.004078)
music (0.003227)
radio (0.002460)
home (0.001941)
plan (0.001780)
bbc (0.001735)
digital (0.001379)
wale (0.001273)
right (0.001243)
Topic 3:
people (0.007195)
said (0.004758)
new (0.003756)
work (0.003563)
music (0.003526)
make (0.003161)
wage (0.003143)
message (0.002981)
mobile (0.002900)
just (0.002647)
Topic 4:
mr (0.007335)
election (0.003462)
blair (0.002948)
said (0.002806)
film (0.002547)
government (0.002333)
say (0.001954)
mr_blair (0.001914)
year (0.001768)
leader (0.001592)
Topic 5:
said (0.013868)
mr (0.009683)
wa (0.007273)
year (0.005412)
best (0.005406)
mobile (0.004518)
film (0.003990)
ha (0.003776)
award (0.003349)
phone (0.002821)
Topic 6:
mr (0.013487)
say (0.011946)
wa (0.010159)
new (0.007728)
labour (0.006834)
election (0.006424)
blair (0.005604)
sale (0.004935)
time (0.004765)
public (0.004529)
Topic 7:
ha (0.016264)
wa (0.011105)
world (0.005790)
week (0.003646)
just (0.003440)
year (0.003370)
say (0.003143)
```

increase (0.003122)

```
party (0.002995)
good (0.002885)
Topic 8:
said (0.025548)
ha (0.010240)
wa (0.006689)
player (0.003609)
firm (0.003538)
child (0.002757)
court (0.002600)
case (0.002584)
apple (0.002461)
legal (0.002217)
Topic 9:
ha (0.005434)
bn (0.004653)
said (0.004541)
firm (0.003779)
film (0.003175)
yukos (0.002776)
company (0.002345)
tax (0.002188)
sale (0.002109)
market (0.001831)
Topic 10:
people (0.024035)
number (0.010615)
uk (0.008284)
like (0.007742)
way (0.007092)
work (0.005744)
think (0.005550)
make (0.005467)
year (0.005191)
right (0.005052)
Topic 11:
wa (0.005261)
year (0.002492)
ha (0.002367)
wage (0.002358)
service (0.002216)
increase (0.002037)
time (0.002032)
market (0.001938)
net (0.001856)
roddick (0.001849)
Topic 12:
award (0.001496)
eu (0.001285)
ha (0.001274)
wa (0.001236)
company (0.001179)
member (0.001145)
film (0.001132)
added (0.001117)
cash (0.001055)
party (0.001026)
Topic 13:
wa (0.013839)
mr (0.008215)
said (0.004927)
year (0.003752)
new (0.003511)
tax (0.003459)
brown (0.003376)
bn (0.002963)
ha (0.002892)
game (0.002420)
Topic 14:
wa (0.008144)
game (0.006728)
film (0.004328)
win (0.002837)
actor (0.002401)
director (0.002263)
new_zealand (0.001860)
zealand (0.001860)
actress (0.001860)
```

jamie (0.001800)

```
Topic 15:
service (0.005827)
said (0.005159)
technology (0.004060)
music (0.003307)
digital (0.002579)
home (0.002461)
company (0.002447)
uk (0.002176)
firm (0.002176)
network (0.002022)
Model Evaluation Scores:
  Coherence: 0.5495309484736951
  Topic diversity: 0.000711098456347347
  Topic size distribution: 0.0011337868480725624
Finished 'nmf process' in 42.7166 secs
def nmf_process(doc_file, source=0, text_col='text', cat=0, chi=False, group=True, eval=False):
    """Pipelines the NMF modeling.
    :param doc_file: The filename of the prescribed text file to be loaded,
        default to None
    :type doc file: str
    :param source: The source of the prescribed document file ('doc_file'),
        where 0 refers to internal store of the package and 1 to external file,
        default to 0
    :type source: int, optional
    :param text col: The name of the text column to be extracted, default to 'text'
    :type text_col: str, optional
    :param cat: The category indicating a subset of the Scripture to be loaded, where
        0 stands for the whole Bible, 1 for OT, 2 for NT, or one of the ten categories ['tor', 'oth', 'ket', 'map', 'mip', 'gos', 'nth', 'pau', 'epi', 'apo'] (See the package's internal file 'data/book_cat.csv'), default to 0
    :type cat: int or str, optional
    :param chi: The flag indicating whether the text is processed as Chinese (True)
        or English (False), default to False
    :type chi: bool, optional
    :param group: The flag indicating whether the loaded text is grouped by chapter,
        default to True
    :type group: bool, optional
    :param eval: The flag indicating whether the model evaluation results will be shown,
        default to False
    :type eval: bool, optional
    :return: The pipelined NMF
    :rtype: cwordtm.tm.NMF object
    nmf = NMF(doc_file, chi)
    if source == 0:
        nmf.docs = load_bible(nmf.textfile, cat=cat, group=group)
        nmf.docs = load_text(nmf.textfile, text_col=text_col)
    print("Corpus loaded!")
    if chi:
        nmf.preprocess_chi()
        nmf.preprocess()
    print("Text preprocessed!")
    print("Text trained!")
    nmf.show_topics_words()
        print("\nModel Evaluation Scores:")
        nmf.evaluate()
    return nmf
>> cwordtm.tm.NMF
class NMF:
     ""The NMF object for Non-negative Matrix Factorization (NMF) modeling.
    :cvar num_topics: The number of topics to be built from the modeling,
        default to 10.
    :vartype num_topics: int
    :ivar textfile: The filename of the text file to be processed
    :vartype textfile: str
    :ivar chi: The flag indicating whether the processed text is in Chinese or not,
        True stands for Traditional Chinese or False for English
```

```
:vartype chi: bool
:ivar num_topics: The number of topics set for the topic model
 :vartype num_topics: int
:ivar docs: The collection of the original documents to be processed
 :vartype docs: pandas.DataFrame or list
 :ivar pro_docs: The collection of documents, in form of list of lists of words
          after text preprocessing
:vartype pro docs: list
:ivar dictionary: The dictionary of word ids with their tokenized words % \left( 1\right) =\left( 1\right) \left( 1\right) \left(
          from preprocessed documents ('pro_docs')
 :vartype dictionary: gensim.corpora.Dictionary
 :ivar corpus: The list of documents, where each document is a list of tuples
           (word id, word frequency in the particular document)
 :vartype corpus: list
 :ivar model: The NMF model object
 :vartype model: gensim.models.Nmf
def __init__(self, textfile, chi=False, num_topics=15):
    """Constructor method.
           self.textfile = textfile
           self.chi = chi
           self.num_topics = num_topics
           self.docs = None
           self.pro docs = None
           self.dictionary = None
           self.corpus = None
           self.model = None
def preprocess(self):
                ""Process the original English documents (cwordtm.tm.NMF.docs)
           by invoking cwordtm.tm.process_text, and build a dictionary
           and a corpus from the preprocessed documents for the NMF model.
           self.pro docs = [process text(doc) for doc in self.docs]
           for i, doc in enumerate(self.pro_docs):
                      self.pro_docs[i] += ["_".join(w) for w in ngrams(doc, 2)]
# self.pro_docs[i] += ["_".join(w) for w in ngrams(doc, 3)]
           # Create a dictionary and corpus for the NMF model
           self.dictionary = corpora.Dictionary(self.pro_docs)
           self.corpus = [self.dictionary.doc2bow(doc) for doc in self.pro_docs]
def preprocess_chi(self):
               ""Process the original Chinese documents (cwordtm.tm.NMF.docs)
           by tokenizing text, removing stopwords, and building a dictionary
           and a corpus from the preprocessed documents for the NMF model.
           # Build stop words
           stop_file = files('cwordtm.data').joinpath("tc_stopwords_2.txt")
           stopwords = [k[:-1] for k in open(stop_file, encoding='utf-8')\
                                               .readlines() if k != '']
           # Tokenize"the text using Jieba
           dict_file = files('cwordtm.data').joinpath("user_dict_4.txt")
           jieba.load_userdict(str(dict_file))
           docs = [jieba.cut(doc) for doc in self.docs]
           # Replace special characters
           docs = [[word.replace('\u3000', ' ') for word in doc] \
                                                                                          for doc in docs]
           # Remove stop words
           self.pro_docs = [' '.join([word for word in doc if word not in stopwords]) \
                                                                                                    for doc in docsl
           self.pro_docs = [doc.split() for doc in self.pro_docs]
           # Create a dictionary and corpus
           self.dictionary = corpora.Dictionary(self.pro_docs)
           self.corpus = [self.dictionary.doc2bow(doc) for doc in self.pro_docs]
def fit(self):
               ""Build the NMF model with the created corpus and dictionary.
           self.model = models.Nmf(self.corpus,
                                                                              num_topics=self.num_topics)
```

```
def show topics words(self):
         ""Shows the topics with their keywords from the built NMF model.
        print("\nTopics-Words from NMF Model:")
        for topic_id in range(self.model.num_topics):
            topic_words = self.model.show_topic(topic_id, topn=10)
            print(f"Topic {topic_id+1}:")
            for word_id, prob in topic_words:
                # word = self.dictionary.id2token[int(word_id)]
                word = self.dictionary[int(word_id)]
                print("%s (%.6f)" %(word, prob))
            print()
    def evaluate(self):
         ""Computes and outputs the coherence score, topic diversity,
        and topic size distribution.
        # Compute coherence score
        coherence_model = CoherenceModel(model=self.model,
                                          texts=self.pro_docs,
                                         dictionary=self.dictionary,
                                         coherence='c_v')
        print(f" Coherence: {coherence_model.get_coherence()}")
        # Compute topic diversity
        topic_sizes = [len(self.model[self.corpus[i]]) for i in range(len(self.corpus))]
        total_docs = sum(topic_sizes)
        topic_diversity = sum([(size/total_docs)**2 for size in topic_sizes])
        print(f" Topic diversity: {topic_diversity}")
        # Compute topic size distribution
        # topic_sizes = [len(self.model[self.corpus[i]]) for i in range(len(self.corpus))]
        topic_size_distribution = max(topic_sizes) / sum(topic_sizes)
        print(f" Topic size distribution: {topic_size_distribution}\n")
>> cwordtm.tm.load_bible
def load bible(textfile, cat=0, group=True):
    """Loads and returns the Bible Scripture from the prescribed internal
    file ('textfile').
    :param textfile: The package's internal Bible text from which the text is loaded,
        either World English Bible ('web.csv') or Chinese Union Version (Traditional)
        ('cuv.csv'), default to None
    :type textfile: str
    :param cat: The category indicating a subset of the Scripture to be loaded, where
       0 stands for the whole Bible, 1 for OT, 2 for NT, or one of the ten categories ['tor', 'oth', 'ket', 'map', 'mip', 'gos', 'nth', 'pau', 'epi', 'apo'] (See the package's internal file 'data/book_cat.csv'), default to 0
    :type cat: int or str, optional
    :param group: The flag indicating whether the loaded text is grouped by chapter,
       default to True
    :type group: bool, optional
    :return: The collection of Scripture loaded
    :rtype: pandas.DataFrame
    # textfile = "web.csv"
    scfile = files('cwordtm.data').joinpath(textfile)
    print("Loading Bible '%s' ..." %scfile)
    df = pd.read_csv(scfile)
    cat = str(cat)
    if cat == '1' or cat == 'ot':
       df = util.extract(df, testament=0)
    elif cat == '2' or cat == 'nt':
       df = util.extract(df, testament=1)
    elif cat in cat list:
        df = util.extract(df, category=cat)
    if group:
        # Group verses into chapters
       .reset_index()
    df.text = df.text.str.replace(' ', '')
    return list(df.text)
>> cwordtm.tm.load text
def load_text(textfile, text_col='text'):
    """Loads and returns the list of documents from the prescribed file ('textfile').
```