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, BERTopic and NMF

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

1. Meta Information Features

```
In [2]: import cwordtm
        from cwordtm import *
        [nltk_data] Downloading package stopwords to
        [nltk_data]
                        C:\Users\User\AppData\Roaming\nltk_data...
        [nltk_data]
                     Package stopwords is already up-to-date!
        [nltk_data] Downloading package wordnet to
        [nltk_data] C:\Users\User\AppData\Roaming\nltk_data...
        [nltk_data] Package wordnet is already up-to-date!
        [nltk_data] Downloading package punkt to
        [nltk_data]
                      C:\Users\User\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\User\AppData\Roaming\nltk_data...
        [nltk_data] Package averaged_perceptron_tagger is already up-to-
        [nltk_data]
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 (wrapped, assigned=('__module__', '__name__', '__qualname__', '__doc__', '__annotations_
             '), updated=('__dict__',))
            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, num topics, chi=False, embed=True)
               LDA (textfile, num_topics, chi=False)
               NMF (textfile, num_topics, chi=False)
               btm_process (doc_file, num_topics=10, source=0, text_col='text', cat=0, chi=False, group=True,
            eval=False, *, timing=False, code=0)
               lda_process (doc_file, num_topics=10, 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, num_topics=10, 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=Tru
            e, underscore numbers=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:
Loading [MathJax]/extensions/Safe.js d (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']
Loading [MathJax]/extensions/Safe.js
                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)
```

```
In [6]: # Show execution time
    df = util.load_text("BBC/BBC News Train.csv", timing=True)
    Finished 'load_text' in 0.0360 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
```

```
# 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]
            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 [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

Loading [MathJax]/extensions/Safe.js

Load BBC News

```
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
                               object
    Text
              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)
```

C:\Dev\Anaconda3\envs\aiml\lib\site-packages\wordcloud\wordcloud.py:106: MatplotlibDeprecationWar
ning: The get_cmap function was deprecated in Matplotlib 3.7 and will be removed two minor releas
es later. Use ``matplotlib.colormaps[name]`` or ``matplotlib.colormaps.get_cmap(obj)`` instead.
 self.colormap = plt.cm.get_cmap(colormap)



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

C:\Dev\Anaconda3\envs\aiml\lib\site-packages\wordcloud\wordcloud.py:106: MatplotlibDeprecationWar
ning: The get_cmap function was deprecated in Matplotlib 3.7 and will be removed two minor releas
es 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 t hat external auditors arthur andersen had approved worldcom s accounting in early 2001 and 2002. she said andersen had given a green light 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 fo llowing the discovery of an \$11bn (£5.7bn) accounting fraud.',

'prosecution lawyers have argued that mr ebbers orchestrated a series of accounting tricks at wo rldcom ordering employees to hide expenses and inflate revenues to meet wall street earnings est imates.',

'the university of california said the trial in the case is scheduled to begin in october 2006. it joined the lawsuit in december 2001alleging massive insider trading and fraud claiming it h ad 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 ar e worth less than they used to be.']

5. Topic Modeling

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

LDA Modeling

```
In [16]: doc_file = "BBC/BBC News Train.csv"
         lda = tm.lda process(doc file, source=1, text col='Text', eval=True, timing=True)
         Corpus 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:
            '0.006*"said" + 0.006*"wa" + 0.004*"ha" + 0.003*"film" + 0.003*"best" + '
            '0.002*"year" + 0.002*"award" + 0.002*"sale" + 0.002*"mr" + 0.002*"new"'),
            '0.006*"wa" + 0.004*"said" + 0.004*"ha" + 0.003*"mr" + 0.002*"vear" + '
           '0.002*"new" + 0.002*"people" + 0.001*"say" + 0.001*"world" + 0.001*"game"').
           '0.005*"said" + 0.003*"ha" + 0.003*"wa" + 0.002*"mr" + 0.002*"new" + '
           '0.002*"year" + 0.001*"people" + 0.001*"world" + 0.001*"government" + '
           '0.001*"blair"'),
          (3,
            '0.008*"said" + 0.005*"wa" + 0.005*"mr" + 0.004*"ha" + 0.002*"people" + '
           '0.002*"labour" + 0.002*"election" + 0.002*"year" + 0.002*"new" +
           '0.002*"brown"'),
          (4,
            '0.006*"wa" + 0.005*"said" + 0.004*"ha" + 0.003*"year" + 0.002*"game" + '
           '0.002*"film" + 0.001*"new" + 0.001*"win" + 0.001*"time" + 0.001*"england"'),
            '0.005*"said" + 0.004*"wa" + 0.004*"ha" + 0.003*"year" + 0.002*"mr" + '
            '0.002*"bn" + 0.002*"government" + 0.001*"new" + 0.001*"world" + '
            '0.001*"number"'),
            '0.004*"ha" + 0.004*"said" + 0.004*"wa" + 0.002*"year" + 0.002*"mr" + '
           '0.001*"dollar" + 0.001*"game" + 0.001*"film" + 0.001*"time" + 0.001*"say"'),
           '0.007*"said" + 0.006*"wa" + 0.004*"ha" + 0.003*"mr" + 0.002*"year" + '
           '0.002*"people" + 0.002*"new" + 0.001*"world" + 0.001*"game" + 0.001*"time"'),
            '0.006*"ha" + 0.006*"said" + 0.004*"wa" + 0.003*"year" + 0.002*"mr" + '
           '0.002*"people" + 0.002*"new" + 0.002*"time" + 0.002*"game" + '
            '0.001*"company"'),
          (9,
            '0.007*"said" + 0.005*"wa" + 0.004*"ha" + 0.002*"year" + 0.002*"people" + '
           '0.002*"new" + 0.001*"mobile" + 0.001*"uk" + 0.001*"game" + 0.001*"mr"')]
         Model Evaluation Scores:
           Coherence: 0.6684181983634158
           Perplexity: -11.226267428136477
           Topic diversity: 0.0007368078040197184
           Topic size distribution: 0.0017825311942959
         Finished 'lda_process' in 63.5579 secs
In [17]: # LDA Model Visualization
         import pyLDAvis
         pyLDAvis.display(lda.vis_data)
Out[17]:
```

BERTopic Modeling

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

```
Corpus loaded!
Text preprocessed!
Text trained!
```

```
Topics from BERTopic Model:
Topic 0: said | mr | wa | ha | year | people | government | new | election | say
Topic 1: wa | game | year | ha | england | said | win | time | half | player
Topic 2: music | band | album | wa | chart | song | year | single | said | singer
Topic 3: film | best | award | oscar | actor | wa | star | director | actress | aviator
Topic 4: mail | virus | spam | anti | security | site | said | user | spyware | attack
```

Topic 5: yukos | russian | russia | tax | gazprom | oil | company | ha | bn | court Topic 6: doping | test | kenteris | iaaf | conte | greek | drug | thanou | sprinter | athens

Topic 7: tv | bbc | series | channel | audience | rating | television | drama | said | wa

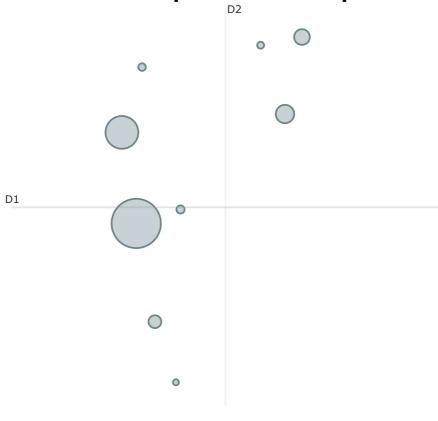
Topic 8: file | peer | sharing | pp | to | network | said | firm | apple | piracy

Model Evaluation Scores:

Coherence: 0.6400456129165745

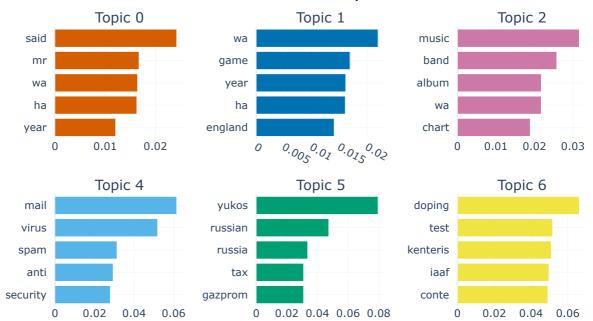
BERTopic Model Visualization:

Intertopic Distance Map

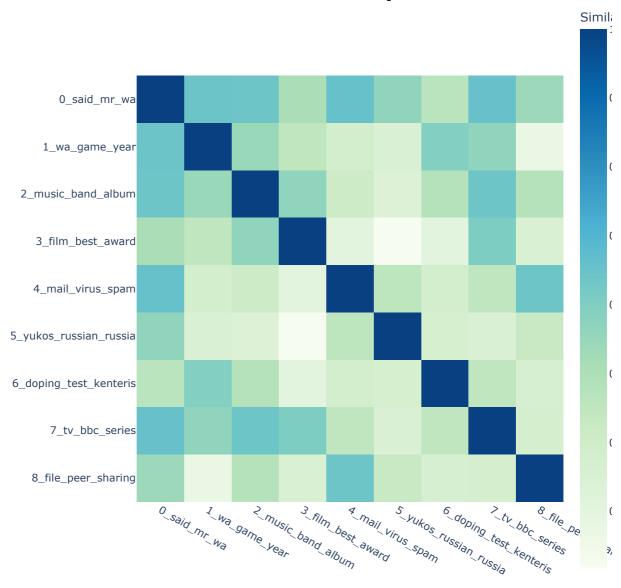




Topic Word Scores



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 169.5788 secs

NMF Modeling

```
In [19]: nmf = tm.nmf_process(doc_file, num_topics=8, source=1, text_col='Text', eval=True, timing=True, colored
```

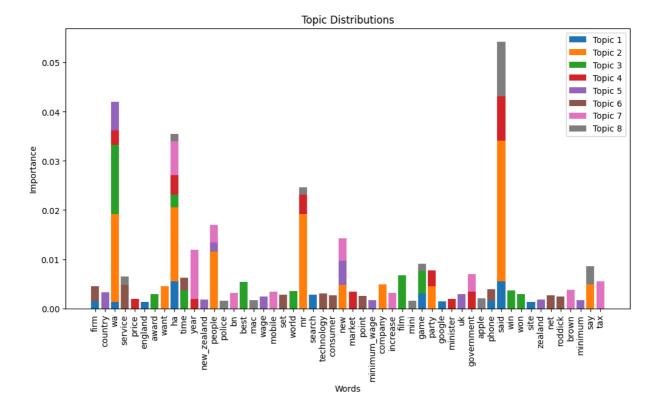
```
Corpus loaded!
            Text preprocessed!
            Text trained!
            Topics-Words from NMF Model:
            Topic 1:
            ha (0.005502)
            said (0.005488)
            game (0.003028)
            search (0.002788)
            phone (0.001550)
            firm (0.001429)
            google (0.001412)
            wa (0.001345)
            england (0.001339)
            site (0.001289)
            Topic 2:
            said (0.028586)
            mr (0.019235)
            wa (0.017877)
            ha (0.015009)
            people (0.011592)
            say (0.004923)
            company (0.004858)
            new (0.004716)
            want (0.004489)
            party (0.004476)
            Topic 3:
            wa (0.014089)
            film (0.006790)
            best (0.005350)
            game (0.004527)
            time (0.003683)
            win (0.003658)
            world (0.003584)
            award (0.002960)
            won (0.002953)
            ha (0.002610)
            Topic 4:
            said (0.009010)
            ha (0.003972)
            mr (0.003871)
            government (0.003408)
            market (0.003403)
            party (0.003221)
            wa (0.002955)
            minister (0.001978)
            price (0.001934)
            year (0.001903)
            Topic 5:
            wa (0.005697)
            new (0.004982)
            country (0.003291)
            uk (0.002917)
            wage (0.002376)
            zealand (0.001834)
            new_zealand (0.001834)
            people (0.001753)
            minimum (0.001704)
            minimum_wage (0.001660)
            Topic 6:
            service (0.004724)
            technology (0.003066)
            firm (0.003040)
            set (0.002814)
            net (0.002721)
            consumer (0.002667)
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time (0.002501)

roddick (0.002433) phone (0.002372)

Topic 7:
year (0.010072)
ha (0.006910)
tax (0.005475)
new (0.004573)
brown (0.003770)
people (0.003628)
government (0.003540)
mobile (0.003380)
increase (0.003180)
bn (0.003099)

Topic 8: said (0.011184) say (0.003642) apple (0.002075) service (0.001739) mac (0.001658) mini (0.001602) game (0.001500) police (0.001495) mr (0.001495) ha (0.001485)



```
Model Evaluation Scores:
              Coherence: 0.5600512699799133
              Topic diversity: 0.0007071142751343485
              Topic size distribution: 0.0010509721492380452
            Finished 'nmf_process' in 36.3509 secs
            def nmf process(doc file, num topics=10, 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 num topics: The number of topics to be modeled, default to 10
                :type num_topics: int, optional
                :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, num topics, 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!")
                nmf.fit()
                print("Text trained!")
                nmf.show_topics_words()
                nmf.viz()
                if eval:
                    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 modeled, default to 10
                :vartype num_topics: int
                :ivar textfile: The filename of the text file to be processed
Loading [MathJax]/extensions/Safe.js |xtfile: 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
                    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, num_topics, chi=False):
    """Constructor method.
                    self.textfile = textfile
                    self.num_topics = num_topics
                    self.chi = chi
                    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]) \
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                                                     for doc in docs]
```

```
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.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[int(word_id)]
                            print("%s (%.6f)" %(word, prob))
                        print()
                def viz(self):
                    """Plot the topic distributions as a stacked bar chart for the built NMF model.
                    # Build a list of word ids from the built topics
                    word ids = []
                    for topic id in range(self.num topics):
                        topic_words = self.model.show_topic(topic_id, topn=10)
                        for word_id, _ in topic_words:
                            word ids.append(int(word id))
                    word ids = list(set(word ids))
                    # Create a topic distribution table
                    topic_dist = np.zeros((self.num_topics, len(word_ids)))
                    for topic id in range(self.num topics):
                        topic_words = self.model.show_topic(topic_id, topn=10)
                        for word id, prob in topic words:
                            topic_dist[topic_id, word_ids.index(int(word_id))] = prob
                    # Build a list of distinct words from the word id list
                    word_list = []
                    for i in range(len(word_ids)):
                        word_list.append(self.dictionary[word_ids[i]])
                    # Plot the topic distributions
                    plt.figure(figsize=(12, 6))
                    bottom = np.zeros(len(word_list))
                    for i, topic in enumerate(topic dist):
                        plt.bar(word_list, topic, width=0.8, bottom=bottom, label=f"Topic {i+1}")
                        bottom += topic
                    plt.xticks(range(len(word_list)), word_list, rotation=90)
                    plt.title("Topic Distributions")
                    plt.xlabel("Words")
                    plt.ylabel("Importance")
                    plt.legend(loc="upper right")
                    plt.show()
                def evaluate(self):
Loading [MathJax]/extensions/Safe.js utes 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_list = ['tor', 'oth', 'ket', 'map', 'mip',\
                             'gos', 'nth', 'pau', 'epi', 'apo']
                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
                    df = df.groupby(['book_no', 'chapter'])\
                                     .agg({'text': lambda x: ' '.join(x)})\
                             .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').
                :param textfile: The prescribed text file from which the text is loaded,
                    default to None
                :tvpe textfile: str
Loading [MathJax]/extensions/Safe.js col: The name of the text column to be extracted, default to 'text'
                :type text_col: str, optional
```

```
:return: The list of documents loaded
:rtype: list
"""

# docs = pd.read_csv(textfile)
docs = util.load_text(textfile)
return list(docs[text_col])
```