

M06 Homework

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```
In [ ]: import pandas as pd
import numpy as np
from glob import glob
import re
import nltk
import plotly_express as px
from lib.textparser import TextParser
from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer, TfidfTransformer

from numpy.linalg import norm
from scipy.spatial.distance import pdist
import scipy.cluster.hierarchy as sch
import matplotlib.pyplot as plt

In [ ]: import configparser
config = configparser.ConfigParser()
config.read("../env.ini")
data_home = config['DEFAULT']['data_home']
output_dir = config['DEFAULT']['output_dir']
data_prefix = 'austen-melville'

In [ ]: OHCO = ['book_id', 'chap_id', 'para_num', 'sent_num', 'token_num']
bags = dict(
    SENTS = OHCO[:4],
    PARAS = OHCO[:3],
    CHAPS = OHCO[:2],
    BOOKS = OHCO[:1]
)

In [ ]: LIB = pd.read_csv(f"{output_dir}/{data_prefix}-LIB.csv").set_index('book_id')
CORPUS = pd.read_csv(f"{output_dir}/{data_prefix}-CORPUS.csv").set_index(OHCO)
VOCAB = pd.read_csv(f"{output_dir}/{data_prefix}-VOCAB.csv").set_index('term_str').dropna()

In [ ]: austen_dates = {
    "EMMA": 1815,
    "LADY SUSAN": 1794,
    "LOVE AND FREINDSHIP SIC": 1790,
    "MANSFIELD PARK": 1814,
    "NORTHANGER ABBEY": 1803,
    "PERSUASION": 1818,
    "PRIDE AND PREJUDICE": 1813,
    "SENSE AND SENSIBILITY": 1811
}

LIB['year'] = LIB['title'].map(austen_dates)
LIB['label'] = LIB['year'].astype(str) + '_' + LIB['title']
LIB = LIB.query("author == 'AUSTEN, JANE'")

LIB

Out [ ]:
source_file_path  author  title  chap_regex  book_len  n_chaps  year  label
book_id
105 /Users/michaelvaden/GithubRepos/DS5001-Workpla... AUSTEN, JANE  PERSUASION  ^Chapter(s+)\d+$ 83624 24 1818.0 1818.0_PERSUASION
121 /Users/michaelvaden/GithubRepos/DS5001-Workpla... AUSTEN, JANE  NORTHANGER ABBEY  ^CHAPTER(s+)\d+$ 77601 31 1803.0 1803.0_NORTHANGER ABBEY
141 /Users/michaelvaden/GithubRepos/DS5001-Workpla... AUSTEN, JANE  MANSFIELD PARK  ^CHAPTER(s+)\d+$ 160378 48 1814.0 1814.0_MANSFIELD PARK
158 /Users/michaelvaden/GithubRepos/DS5001-Workpla... AUSTEN, JANE  EMMA  ^s*CHAPTER(s+)\d+$ 160926 55 1815.0 1815.0_EMMA
161 /Users/michaelvaden/GithubRepos/DS5001-Workpla... AUSTEN, JANE  SENSE AND SENSIBILITY  ^CHAPTER(s+)\d+$ 119873 50 1811.0 1811.0_SENSE AND SENSIBILITY
946 /Users/michaelvaden/GithubRepos/DS5001-Workpla... AUSTEN, JANE  LADY SUSAN  ^s*[IVXLCM]+s*$ 23116 41 1794.0 1794.0_LADY SUSAN
1212 /Users/michaelvaden/GithubRepos/DS5001-Workpla... AUSTEN, JANE  LOVE AND FREINDSHIP SIC  ^s*LETTER .* to .*$ 33265 24 1790.0 1790.0_LOVE AND FREINDSHIP SIC
1342 /Users/michaelvaden/GithubRepos/DS5001-Workpla... AUSTEN, JANE  PRIDE AND PREJUDICE  ^Chapter(s+)\d+$ 122126 61 1813.0 1813.0_PRIDE AND PREJUDICE

In [ ]: def create_bag_of_words(CORPUS, bag):
    BOW = CORPUS.groupby(bag+['term_str']).term_str.count().to_frame('n')
    return BOW

idf_method = 'standard'

def get_TFIDF(BOW, tf_method):
    DTCM = BOW.n.unstack(fill_value=0)

    DF = DTCM.astype('bool').sum()
    N = len(DTCM)

    if tf_method == 'sum':
        TF = DTCM.T / DTCM.T.sum()
```

```

elif tf_method == 'max':
    TF = DTCM.T / DTCM.T.max()

elif tf_method == 'log':
    TF = np.log2(1 + DTCM.T)

elif tf_method == 'raw':
    TF = DTCM.T

elif tf_method == 'double_norm':
    TF = DTCM.T / DTCM.T.max()

elif tf_method == 'binary':
    TF = DTCM.T.astype('bool').astype('int')

TF = TF.T

if idf_method == 'standard':
    IDF = np.log2(N / DF)

elif idf_method == 'max':
    IDF = np.log2(DF.max() / DF)

elif idf_method == 'smooth':
    IDF = np.log2((1 + N) / (1 + DF)) + 1

return TF * IDF, DF * IDF

```

In []: `CORPUS = CORPUS.loc[LIB.index]`

In []: `TFIDF = get_TFIDF(create_bag_of_words(CORPUS, bags['CHAPS']), 'max')[0]`

TFIDF

Out []:

	term_str	0	1	10	10000	10th	11th	12	12th	1399	13th	...	youthful	youths	ysr	z	zeal	zealous	zealously	zephyr	zigzags	!20
book_id	chap_id																					
105	1	0.0	0.119092	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.043417	0.0	0.0	0.067611	0.000000	0.000000	0.0	0.0	0.0	
	2	0.0	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.0	0.0	0.000000	0.000000	0.061689	0.0	0.0	0.0	
	3	0.0	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.0	0.0	0.000000	0.036998	0.000000	0.0	0.0	0.0	
	4	0.0	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.0	0.0	0.000000	0.000000	0.000000	0.0	0.0	0.0	
	5	0.0	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.0	0.0	0.000000	0.000000	0.000000	0.0	0.0	0.0	
...	
1342	57	0.0	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.0	0.0	0.000000	0.000000	0.000000	0.0	0.0	0.0	
	58	0.0	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.0	0.0	0.000000	0.000000	0.000000	0.0	0.0	0.0	
	59	0.0	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.0	0.0	0.000000	0.000000	0.000000	0.0	0.0	0.0	
	60	0.0	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.0	0.0	0.000000	0.000000	0.000000	0.0	0.0	0.0	
	61	0.0	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.0	0.0	0.000000	0.000000	0.000000	0.0	0.0	0.0	

334 rows x 14745 columns

In []: `DFIDF = get_TFIDF(create_bag_of_words(CORPUS, bags['CHAPS']), 'max')[1].to_frame('DFIDF')`

DFIDF

Out []:

	DFIDF
term_str	
0	8.383704
1	14.767409
10	20.396225
10000	8.383704
10th	14.767409
...	...
zealous	34.792451
zealously	14.767409
zephyr	8.383704
zigzags	8.383704
!20000	8.383704

14745 rows x 1 columns

In []: `part_of_speech = ['NN', 'NNS', 'VB', 'VBD', 'VBG', 'VBN', 'VBP', 'VBZ', 'JJ', 'JJR', 'JJS', 'RB', 'RBR', 'RBS']`

`Collapsed_TFIDF = TFIDF.loc[:, VOCAB.query("max_pos in @part_of_speech").join(DFIDF).sort_values('DFIDF', ascending=False)\`
`.head(1000).reset_index()['term_str'].to_list().reset_index().groupby(['book_id']).mean().drop('chap_id', axis=1)`

`Collapsed_TFIDF`

```
Out [ ]: term_str    respect    forward    greatest    stay    thinking    assure    marriage    fortune    believed    new ... impression    unable    deep    sensations    forti
book_id
105    0.006140    0.009262    0.003233    0.010584    0.011340    0.010205    0.011468    0.013589    0.008962    0.010591 ... 0.006734    0.004557    0.004729    0.007360    0.002720
121    0.004111    0.006748    0.007326    0.008810    0.008241    0.008954    0.005660    0.010054    0.009190    0.016896 ... 0.002037    0.001299    0.006226    0.002145    0.010875
141    0.009049    0.009500    0.010356    0.009412    0.010821    0.004308    0.007412    0.006946    0.008376    0.009582 ... 0.006074    0.003468    0.005070    0.007835    0.002605
158    0.011402    0.008882    0.007157    0.011392    0.014596    0.014604    0.009475    0.010334    0.011059    0.009415 ... 0.003549    0.000696    0.006773    0.006615    0.002374
161    0.006497    0.006474    0.009488    0.008128    0.005347    0.009847    0.013288    0.012934    0.008158    0.010615 ... 0.003638    0.011270    0.002284    0.001596    0.007865
946    0.010156    0.010188    0.015570    0.005638    0.001255    0.005513    0.021950    0.005375    0.009788    0.000000 ... 0.004818    0.001120    0.001906    0.005358    0.010265
1212   0.003126    0.001501    0.010051    0.000884    0.003517    0.011138    0.016403    0.011411    0.000000    0.005096 ... 0.010533    0.002869    0.000339    0.002735    0.004974
1342   0.009518    0.006757    0.005553    0.012313    0.007312    0.013715    0.020465    0.014328    0.010656    0.007232 ... 0.003633    0.008668    0.005856    0.001252    0.002505
```

8 rows x 1000 columns

```
In [ ]: L0 = Collapsed_TFIDF.astype('bool').astype('int') # Binary (Pseudo L)
L1 = Collapsed_TFIDF.apply(lambda x: x / x.sum(), 1) # Probabilistic
L2 = Collapsed_TFIDF.apply(lambda x: x / norm(x), 1) # Pythagorean, AKA Euclidean

PAIRS = pd.DataFrame(index=pd.MultiIndex.from_product([LIB.index.tolist(), LIB.index.tolist()]).reset_index())

# Keep only unique pairs of different books
PAIRS = PAIRS[PAIRS.level_0 < PAIRS.level_1].set_index(['level_0', 'level_1'])

# Name index cols
PAIRS.index.names = ['doc_a', 'doc_b']
```

```
In [ ]: PAIRS['cityblock'] = pdist(Collapsed_TFIDF, 'cityblock')
PAIRS['euclidean'] = pdist(Collapsed_TFIDF, 'euclidean')
PAIRS['cosine'] = pdist(Collapsed_TFIDF, 'cosine')
PAIRS['jaccard'] = pdist(L0, 'jaccard')
PAIRS['dice'] = pdist(L0, 'dice')
PAIRS['js'] = pdist(L1, 'jensenshannon')
```

```
In [ ]: def hac(sims, linkage_method='complete', color_thresh=.3, figsize=(10, 10)):

    # Generate the clustering
    tree = sch.linkage(sims, method=linkage_method)

    # Get labels for the leaves
    labels = LIB.label.values

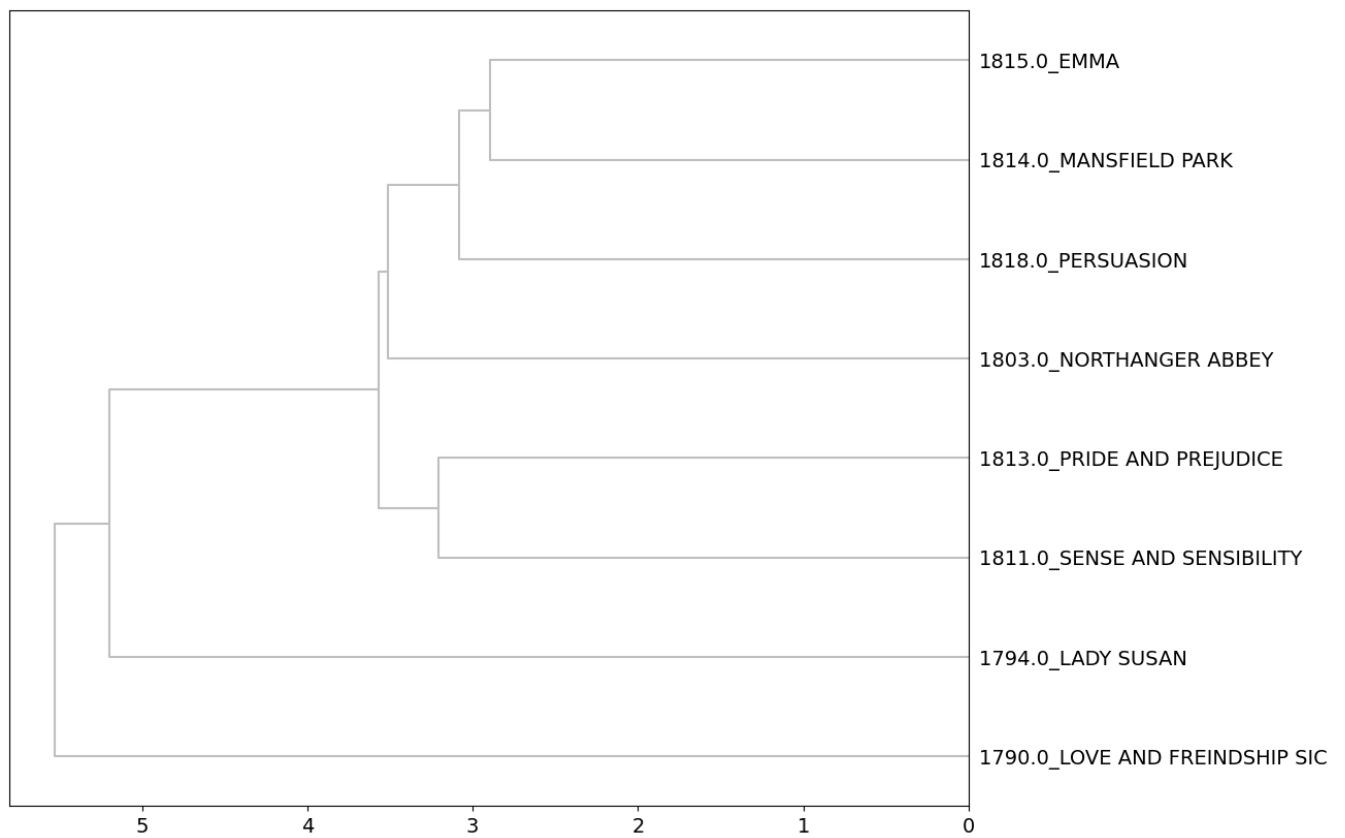
    # Create a figure
    plt.figure()
    fig, axes = plt.subplots(figsize=figsize)

    # Create a dendrogram with the tree
    dendrogram = sch.dendrogram(tree,
                                labels=labels,
                                orientation="left",
                                count_sort=True,
                                distance_sort=True,
                                above_threshold_color='.75',
                                color_threshold=color_thresh
                                )

    # Change the appearance of ticks, tick labels, and gridlines
    plt.tick_params(axis='both', which='major', labelsize=14)
```

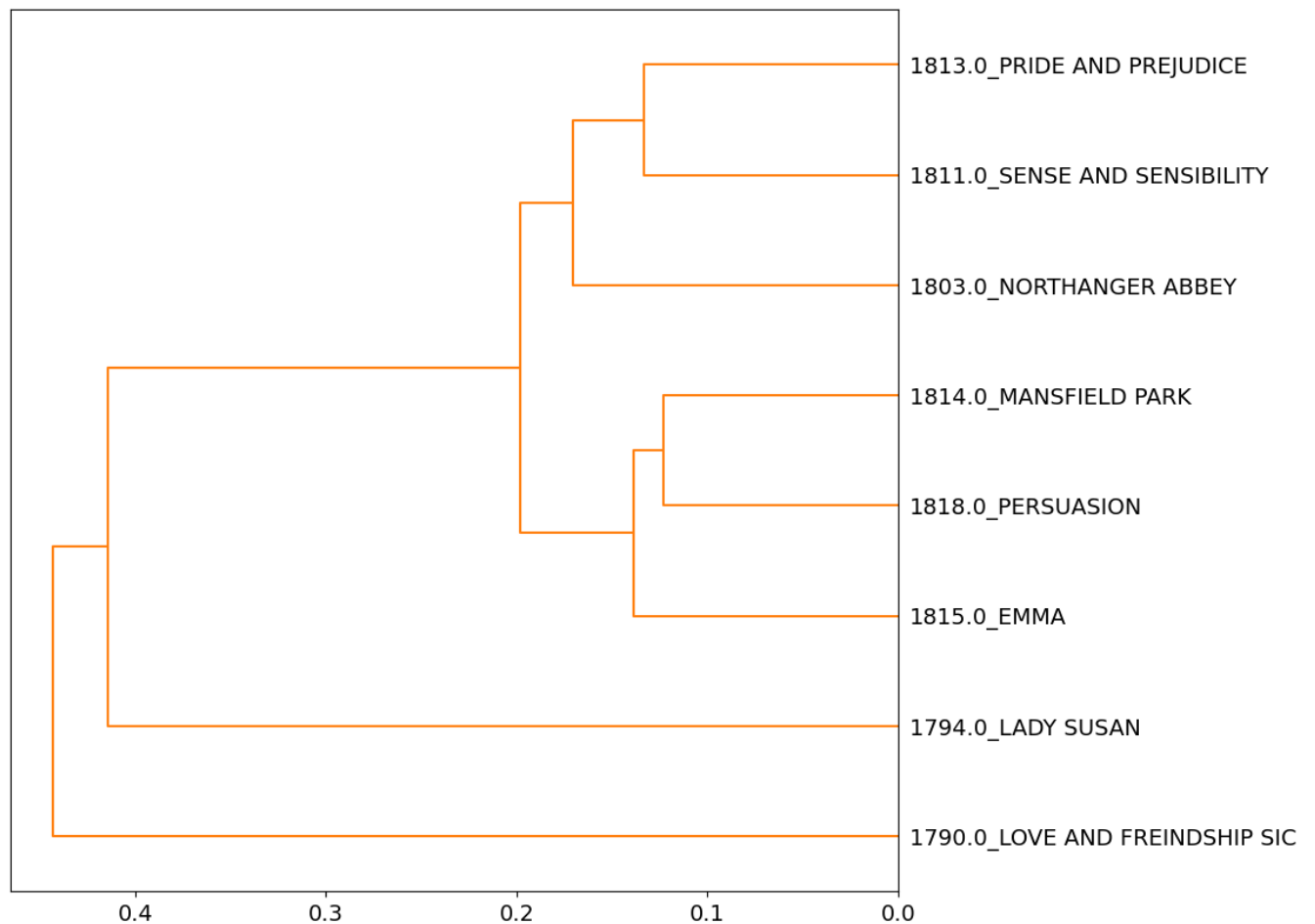
```
In [ ]: hac(PAIRS.cityblock, linkage_method='weighted', color_thresh=1.2)
```

<Figure size 640x480 with 0 Axes>



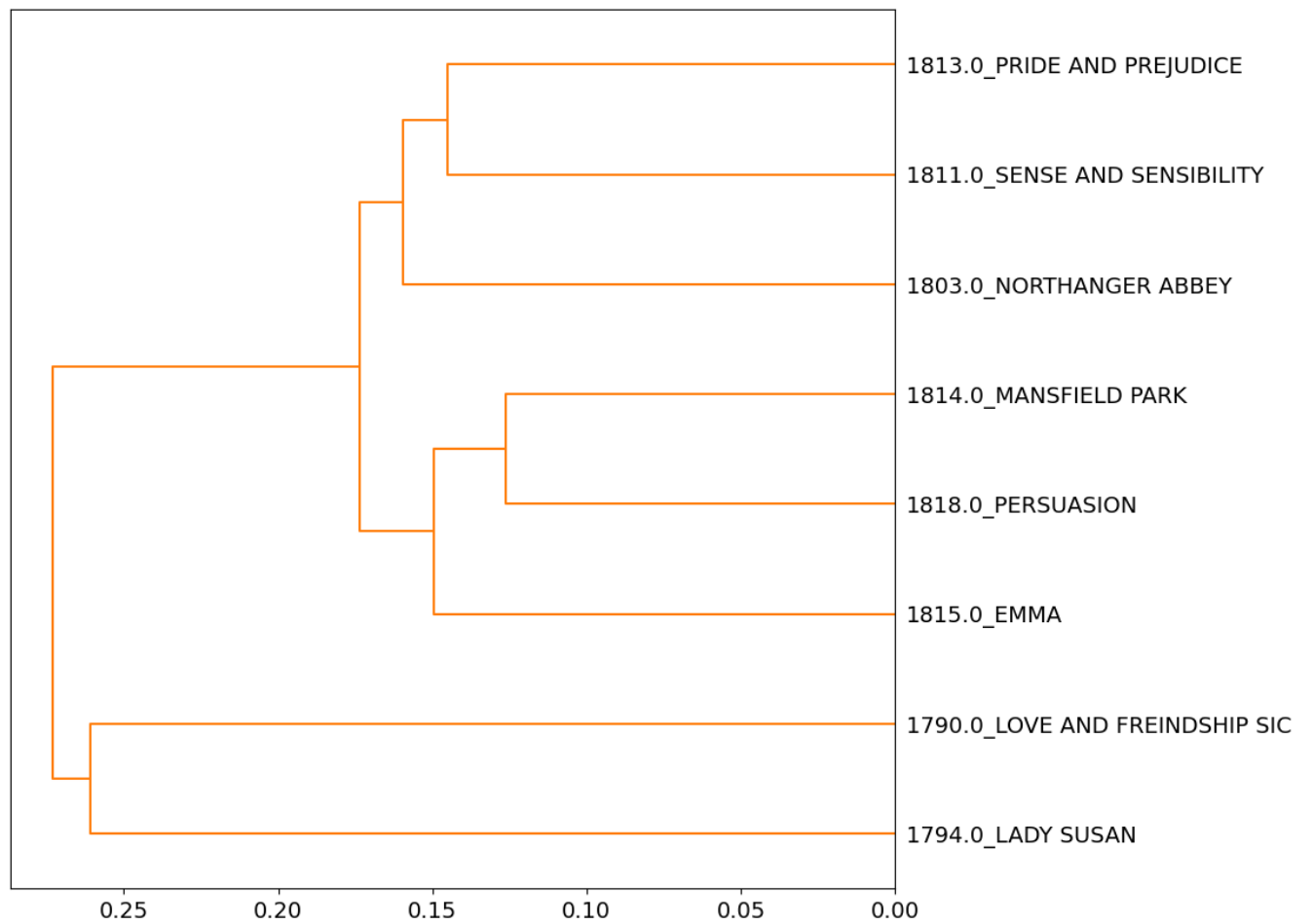
```
In [ ]: hac(PAIRS.cosine, linkage_method='ward', color_thresh=.85);
```

<Figure size 640x480 with 0 Axes>

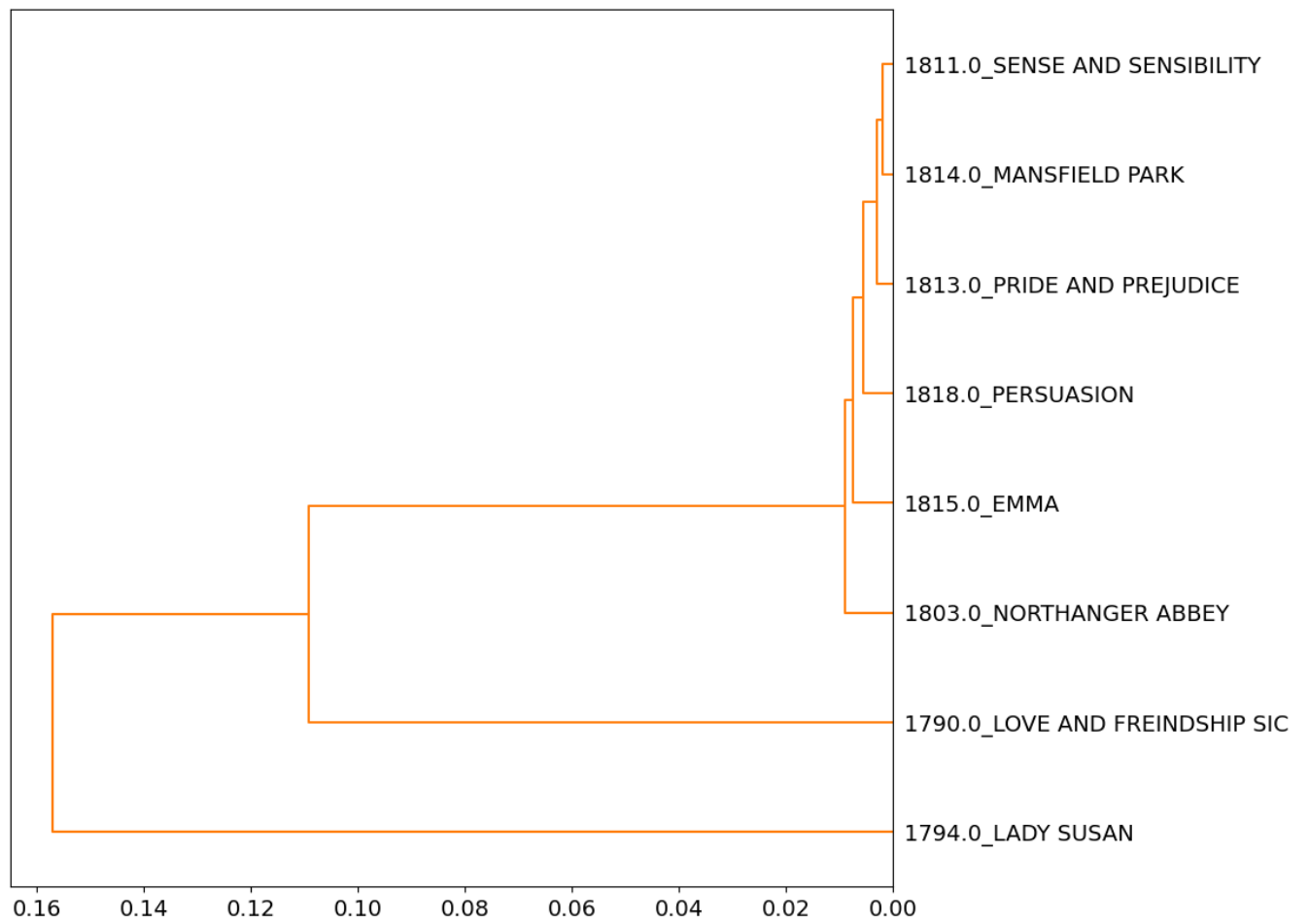


```
In [ ]: hac(PAIRS.euclidean, linkage_method='ward', color_thresh=.85);
```

<Figure size 640x480 with 0 Axes>

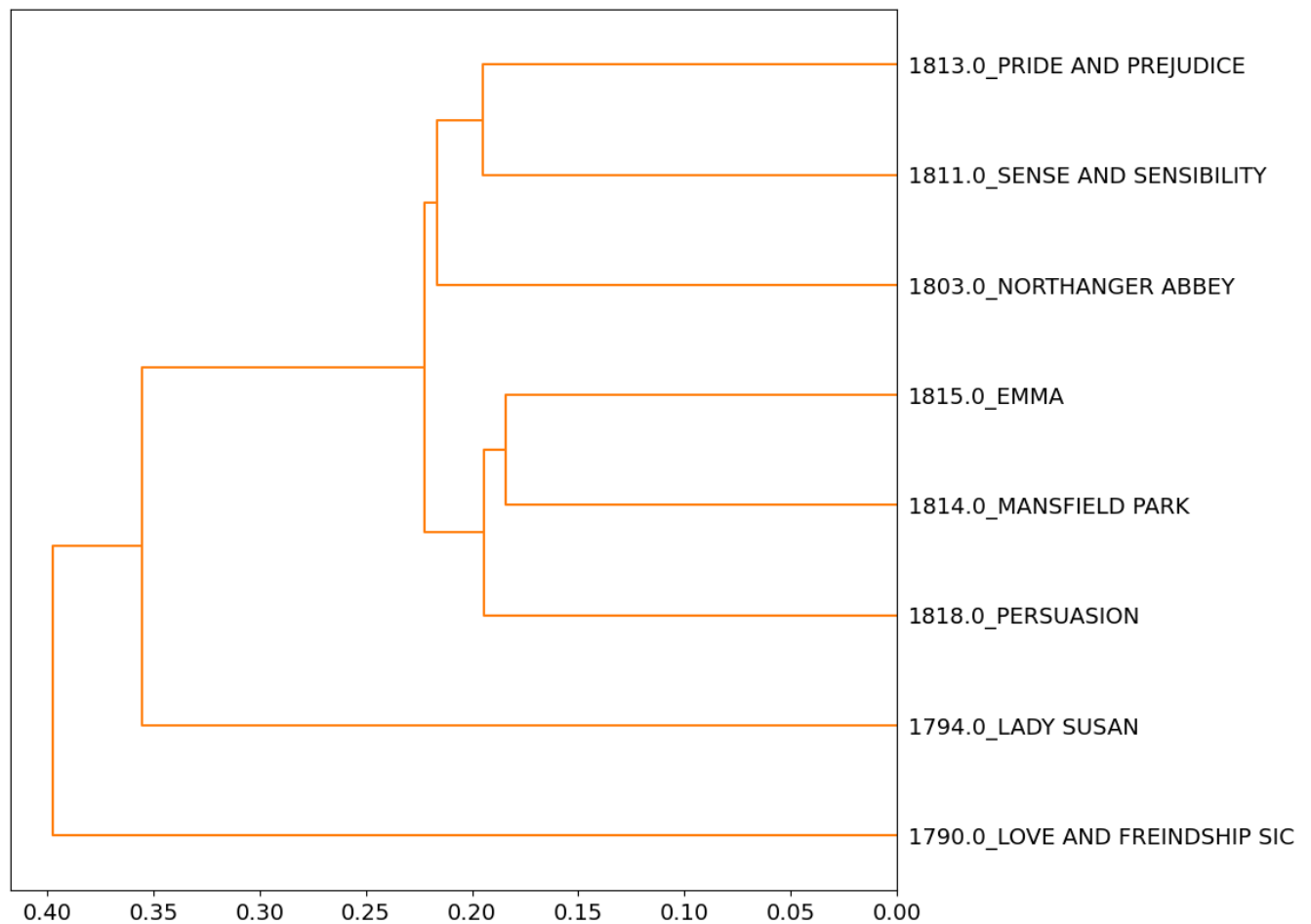


```
In [ ]: hac(PAIRS.jaccard, linkage_method='weighted', color_thresh=.85);  
<Figure size 640x480 with 0 Axes>
```



```
In [ ]: hac(PAIRS.js, linkage_method='weighted', color_thresh=.85);
```

<Figure size 640x480 with 0 Axes>



1. What are the top 10 nouns by DFIDF, sorted in descending order? Include plural nouns, but don't include proper nouns.

```
In [ ]: VOCAB.query("max_pos == 'NN' or max_pos == 'NNS'").join(DFIDF).sort_values('DFIDF', ascending=False).head(10)['DFIDF'].to_frame()
```

```
Out [ ]:
```

term_str	DFIDF
respect	177.266344
marriage	177.261968
fortune	177.261968
question	177.258990
ladies	177.258990
behaviour	177.240001
farther	177.240001
advantage	177.217644
girl	177.209470
voice	177.209470

2. Grouping your TFIDF results by book, and taking the mean TFIDF of all terms per book, what is Austen's most "significant" book? This value is computed from the TFIDF matrix your function returned.

```
In [ ]: TFIDF_means = TFIDF.loc[:, VOCAB.join(DFIDF, how='right').reset_index()['term_str'].to_list()] \
        .reset_index().groupby(['book_id']).mean().drop('chap_id', axis=1)

TFIDF_means.T.mean().to_frame('TFIDF').join(LIB, how='left').sort_values(by='TFIDF', ascending=False).head(1)[['title', 'TFIDF']]
```

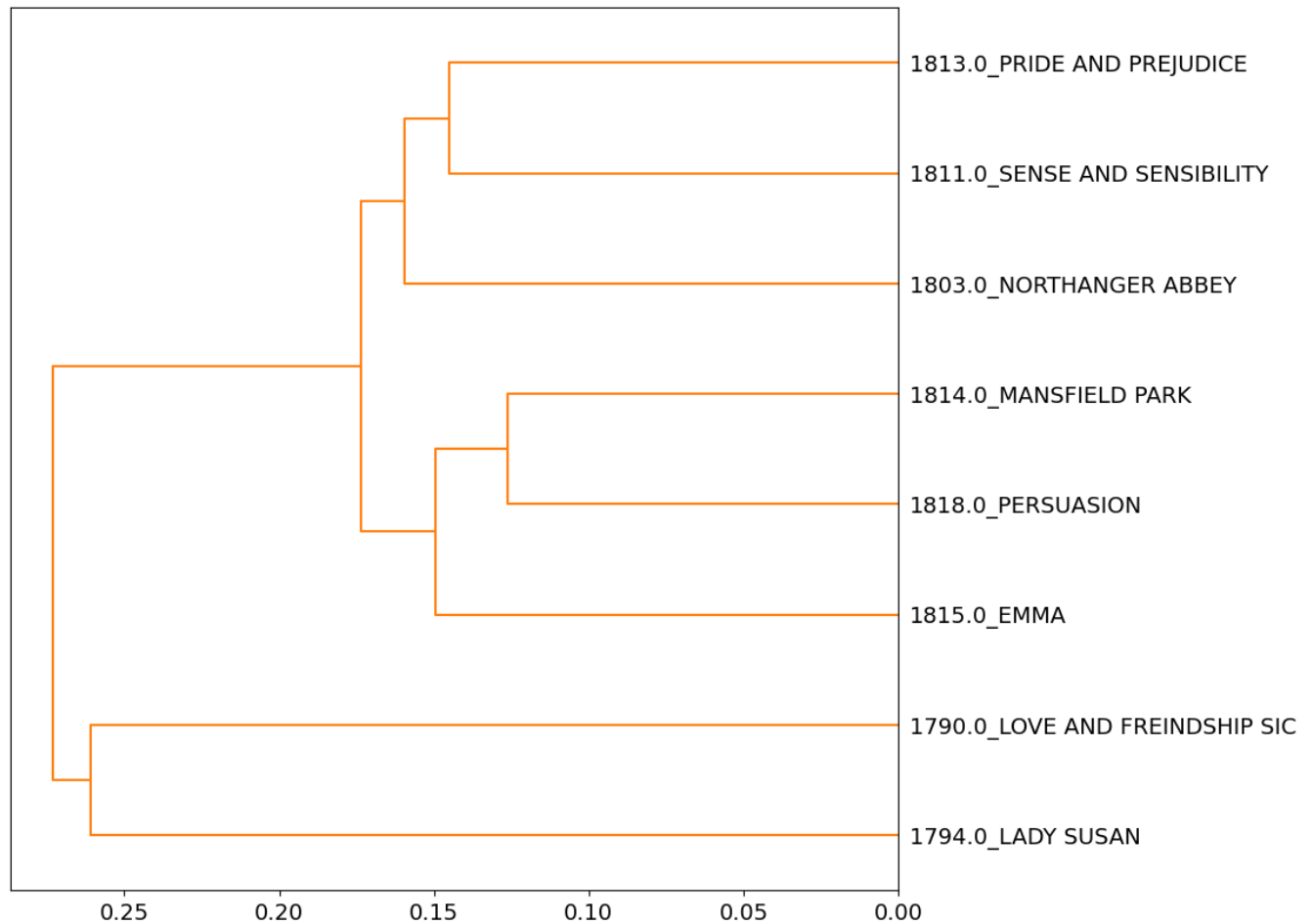
```
Out [ ]:
```

book_id	title	TFIDF
121	NORTHANGER ABBEY	0.001851

3. Using the dendrograms you generated, which distance measure most clearly distinguishes Austen's two youthful works from her later works? That is, which measure show the greatest separation between the first two work and the rest? Note that the two youthful works were published before 1800.

```
In [ ]: hac(PAIRS.euclidean, linkage_method='ward', color_thresh=.85);
```

<Figure size 640x480 with 0 Axes>



As we can see from the dendrograms created, Austen's earliest works Lady Susan (946) and Love and Friendship (1212) are most clearly distinguished from her later works through **Euclidean** distance. Shown above, we can see the first split separates the two earliest works from the rest of Austen's works.

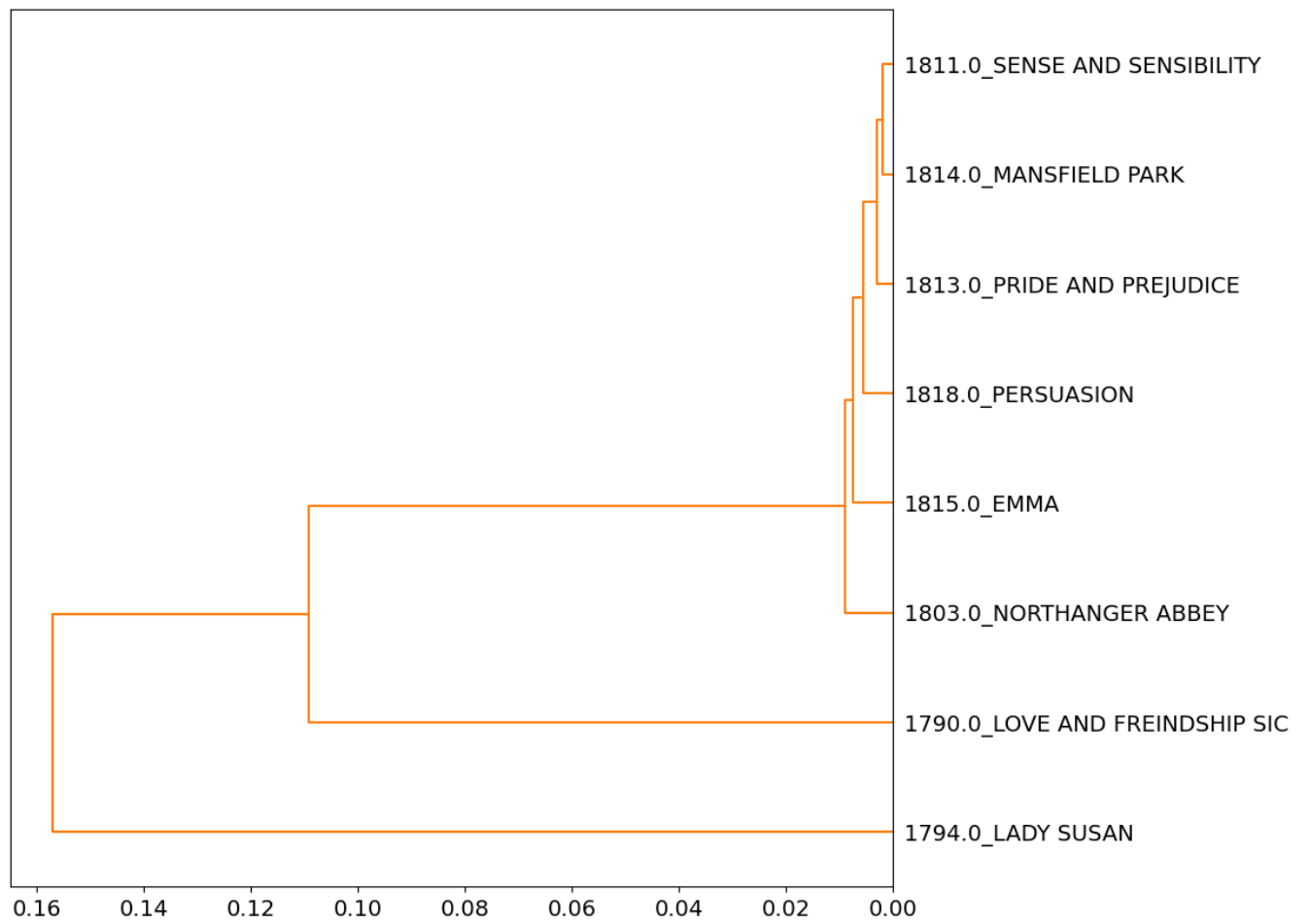
4. Do any of the distance measures produce dendrograms with works sorted in the exact order of their publication years?

No, none of the dendrograms with the given distance and linkage measure combinations produce the works exactly sorted. However, **cityblock** comes the closest to doing so.

5. Some literary critics believe that Northanger Abbey is, among Austen's mature works, the one that most resembles her juvenalia, i.e. her two works written as a young adult. Which distance measure dendrograms appear to corroborate this thesis? In other words, do any of them show that Northanger Abbey is closer to her juvenalia than the her other adult works?

```
In [ ]: hac(PAIRS.jaccard, linkage_method='weighted', color_thresh=.85);
```

<Figure size 640x480 with 0 Axes>



yes, the **jaccard** distance measure dendrogram shows that Northanger Abbey is closest to Austen's two earliest works!