

INTRODUCTION

Philosophy has seen many eccentric and world-changing personalities. From the time humans have started recording history, we have written accounts of their writings, sayings and written thoughts. The school of philosophy has changed with time. We have different schools of thoughts of ancient times, popularized by the philosophers like Plato and Aristotle. However, as times grew and the focus of humans shifted from monarchies to democracies, new schools of thoughts like capitalism and communism grew. In this project my focus is to analyze different schools of thoughts, identify the underlying 'topics' in different schools of thoughts and predict some articles as to which school of thought they align with.

Downloading Packages

```
In [1]: import matplotlib.pyplot as plt
import numpy as np
import matplotlib.dates as mdates
from datetime import datetime
import pandas as pd
from nltk.corpus import stopwords
from nltk.tokenize import word_tokenize
from nltk.stem import PorterStemmer
from collections import Counter
from nltk.probability import FreqDist
from gensim.models import LdaModel
import nltk
import os
import string
import numpy as np
import pandas as pd
import pickle
import re
import math
import nltk.time
import gzip
import pickle as cpickle
from gensim import models
from gensim.models import Word2Vec
import logging
import nltk
from gensim import corpora, models, similarities
stopwords_en=stopwords.words("english")
```

Exploring data

```
In [2]: philo_data_csv=pd.read_csv('C://Users/Devika/Documents/Columbia/Term 2/Applied Data Science/philoso
phy_data.csv')
```

```
In [3]: philo_data_csv.head()
```

Out[3]:

	title	author	school	sentence_spacy	sentence_str	original_publication_date	corpus_edition_date	sentence_length	sentence_loww
0	Plato - Complete Works	Plato	plato	What's new, Socrates, to make you leave your ...	What's new, Socrates, to make you leave your ...	-350	1997	125	what's r socrates, to m you leave you
1	Plato - Complete Works	Plato	plato	Surely you are not prosecuting anyone before t...	Surely you are not prosecuting anyone before t...	-350	1997	69	surely you are prosecuting any before
2	Plato - Complete Works	Plato	plato	The Athenians do not call this a prosecution b...	The Athenians do not call this a prosecution b...	-350	1997	74	the athenian not call th prosecution
3	Plato - Complete Works	Plato	plato	What is this you say?	What is this you say?	-350	1997	21	what is this s
4	Plato - Complete Works	Plato	plato	Someone must have indicted you, for you are no...	Someone must have indicted you, for you are no...	-350	1997	101	someone m have indicted for you are i

Code Information

This data consists of many pieces of information about the authors, their school of their and their thoughts and sayings. It also contains the era of the philosophers and the latest edition of the corpus with their thoughts. The sentences are also lemmatized for ease of use and the sentence length is also given. To begin, I will create a simple timeline of the period in which each philosopher existed. First, I will duplicate the dataframe as it will come in handy.

```
In [4]: philo_data=philo_data_csv.copy()
```

```
In [5]: temp_df=philo_data_csv.sort_values(by=['original_publication_date'],ascending=True)
temp_df=pd.DataFrame(temp_df.loc[:, ['author','original_publication_date']].drop_duplicates().values)
dates = temp_df.iloc[:,1]
names = temp_df.iloc[:,0]
#Removing negative value as python does not accept negative years as dates
names=names[4:]
dates=dates[4:].astype(str)
dates = [datetime.strptime(d, "%Y") for d in dates]
```

```
In [6]: # Choose some nice levels
levels = np.tile([-5, 5, -3, 3, -1, 1],
               int(np.ceil(len(dates)/6)))[1:len(dates)]

# Create figure and plot a stem plot with the date
fig, ax = plt.subplots(figsize=(20, 8), constrained_layout=True)
ax.set(title="Philosophers in different eras")

markerline, stemline, baseline = ax.stem(dates, levels,
                                         linefmt="C3-", basefmt="k-",
                                         use_line_collection=True)

plt.setp(markerline, mec="k", mfc="w", zorder=3)

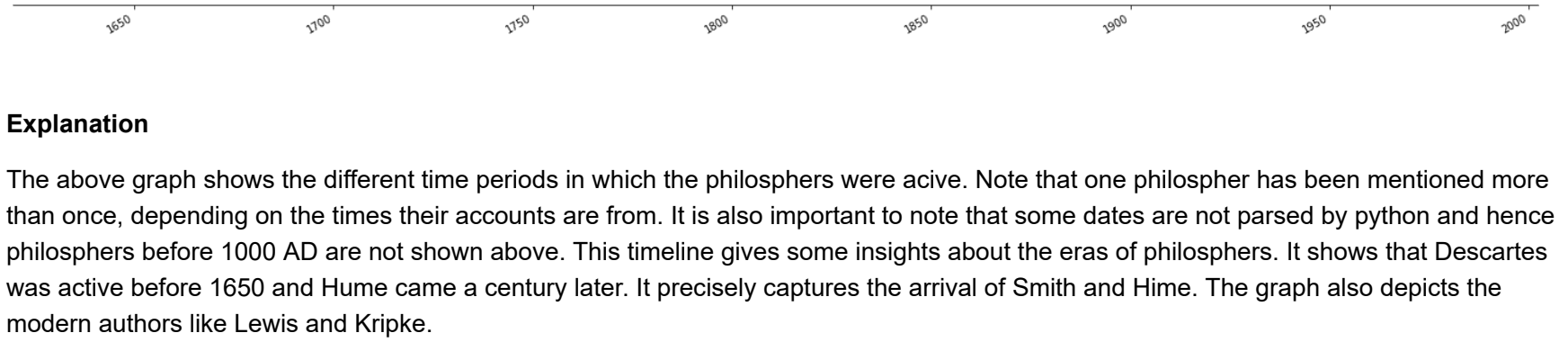
# Shift the markers to the baseline by replacing the y-data by zeros.
markerline.set_ydata(np.zeros(len(dates)))

# annotate lines
vert = np.array(['top', 'bottom'])[:(levels > 0).astype(int)]
for d, l, r, va in zip(dates, levels, names, vert):
    ax.annotate(r, xy=(d, l), xytext=(-3, np.sign(l)*3),
               textcoords="offset points", va=va, ha="right")

# format x axis with 100 month intervals
ax.get_xaxis().set_major_locator(mdates.YearLocator(50))
ax.get_xaxis().set_major_formatter(mdates.DateFormatter("%Y"))
plt.setp(ax.get_xticklabels(), rotation=30, ha="right")

# remove y axis and spines
ax.get_yaxis().set_visible(False)
for spine in ["left", "top", "right"]:
    ax.spines[spine].set_visible(False)

ax.margins(y=0.1)
plt.show()
```



Explanation

The above graph shows the different time periods in which the philosophers were active. Note that one philosopher has been mentioned more than once, depending on the times their accounts are from. It is also important to note that some dates are not parsed by python and hence philosophers before 1000 AD are not shown above. This timeline gives some insights about the eras of philosophers. It shows that Descartes was active before 1650 and Hume came a century later. It precisely captures the arrival of Smith and Hume. The graph also depicts the modern authors like Lewis and Kripke.

```
In [7]: temp_df=philo_data_csv.sort_values(by=['original_publication_date'],ascending=True)
temp_df=pd.DataFrame(temp_df.loc[:, ['author','original_publication_date']].drop_duplicates().values)
dates = temp_df.iloc[:,1]
names = temp_df.iloc[:,0]
#Removing negative value as python does not accept negative years as dates
names=names[4:]
dates=dates[4:].astype(str)
dates = [datetime.strptime(d, "%Y") for d in dates]
```

```
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                                         linefmt="C3-", basefmt="k-",
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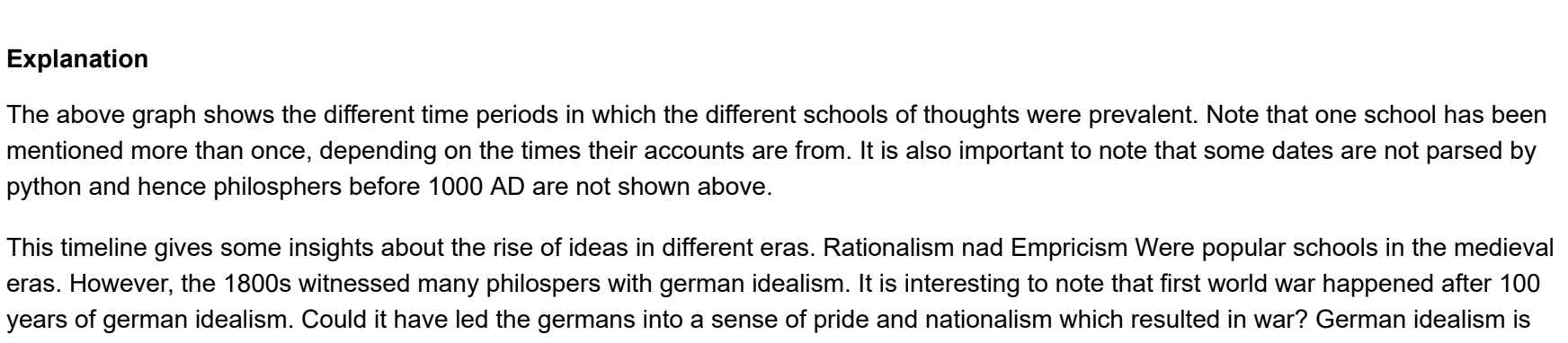
# Shift the markers to the baseline by replacing the y-data by zeros.
markerline.set_ydata(np.zeros(len(dates)))

# annotate lines
vert = np.array(['top', 'bottom'])[:(levels > 0).astype(int)]
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# remove y axis and spines
ax.get_yaxis().set_visible(False)
for spine in ["left", "top", "right"]:
    ax.spines[spine].set_visible(False)

ax.margins(y=0.1)
plt.show()
```



Explanation

The above graph shows the different time periods in which the different schools of thoughts were prevalent. Note that one school has been mentioned more than once, depending on the times their accounts are from. It is also important to note that some dates are not parsed by python and hence philosophers before 1000 AD are not shown above. This timeline gives some insights about the eras of philosophers. It shows that Descartes was active before 1650 and Hume came a century later. It precisely captures the arrival of Smith and Hume. The graph also depicts the modern authors like Lewis and Kripke.

Data Analysis

Code Information

Now I will analyse the different schools of thoughts. I will roll up the data to make the data unique at author-sentence level. I will use this data to first analyze the different topics selected school and then use LSI algorithm for predicting the most related school for a selected sentence or paragraph.

Topic Selection

```
In [8]: import ipywidgets as widgets
```

```
In [9]: w=widgets.Dropdown(
    options=['plato','aristotle','empiricism','rationalism','analytic','continental','phenomenology','german_idealism','communism','capitalism','stoicism','nietzsche','feminism'],
    value='plato',
    description="Select a school of thought to find the topics in it:",
)
```

```
In [10]: def on_change(change):
    if change['type'] == 'change' and change['name'] == 'value':
        print("changed to %s" % change['new'])
        return change['new']
```

Code Information

User can select any school of thought and the corresponding topics would be shown to the user.

```
In [11]: w.observe(on_change)
#observe(selected_topics)
display(w)
```

```
In [12]: from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.decomposition import TruncatedSVD
rsl_df = philo_data_csv[philo_data_csv['school']=='school'].value
vectorizer = TfidfVectorizer(stop_words='english',
                             max_features= 1000, # keep top 1000 terms
                             max_df = 0.5,
                             smooth_idf=True)

X = vectorizer.fit_transform(rsl_df['sentence_lowered'])

#X.SVD # check shape of the document-term matrix

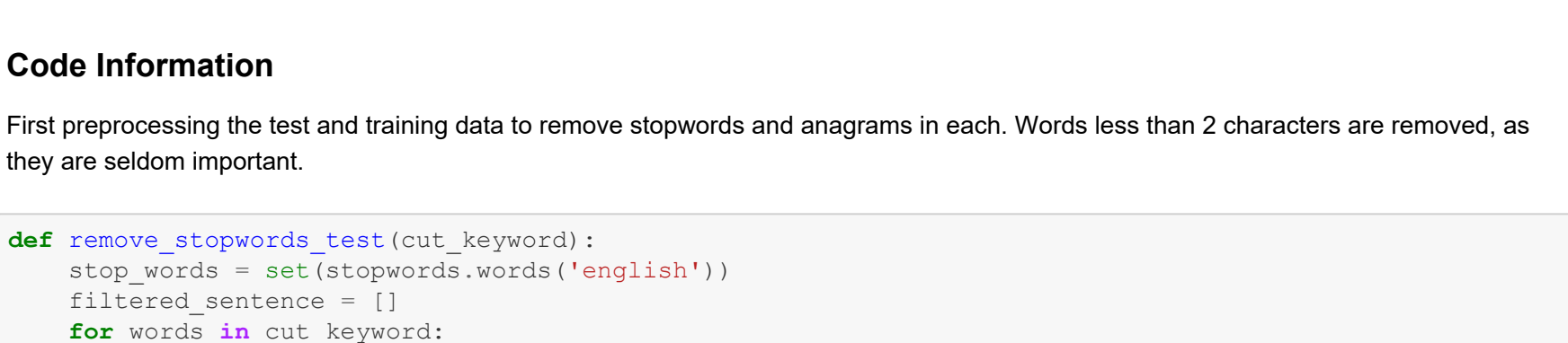
svd_model = TruncatedSVD(n_components=10, algorithm='randomized', n_iter=100, random_state=122)

svd_model.fit(X)

#len(svd_model.components_)
terms = vectorizer.get_feature_names()
wordcloud_string=""
for i, comp in enumerate(svd_model.components_):
    terms, comp = zip(terms, comp)
    sorted_terms = sorted(terms, key=lambda x:x[1], reverse=True)[:7]
    for t in sorted_terms:
        wordcloud_string+=t[0]+' '
```

```
In [13]: import matplotlib.pyplot as plt
from wordcloud import WordCloud
# Create and generate a word cloud image:
wordcloud = WordCloud(max_font_size=50, max_words=1000, background_color="white").generate(wordcloud_string)

# Display the generated image:
plt.imshow(wordcloud, interpolation='bilinear')
plt.axis("off")
plt.show()
```



Explanation

The above figure shows the words in the top topics of the selected school by the user. Analysing different schools gives an insight into the basic philosophy of each school. Plato and Aristotle's writings are more about men and understanding. Rationalists' writing are more about good and evil and love, whereas feminists writing are about women and workers etc.

Predicting school of new sentences

Code Information

First preprocessing the test and training data to remove stopwords and anagrams in each. Words less than 2 characters are removed, as they are seldom important.

```
In [14]: def remove_stopwords_test(cut_keyword):
    stop_words = set(stopwords.words('english'))
    filtered_sentence = []
    for words in cut_keyword:
        if words not in stop_words and len(words)>=3:
            filtered_sentence.append(words)
    return filtered_sentence
```

```
In [15]: def remove_stopwords_train(cut_keyword):
    stop_words = set(stopwords.words('english'))
    filtered_sentence = []
    for l in cut_keyword:
        temp=[]
        for w in l:
            if w not in stop_words and len(w)>=3:
                temp.append(w)
        filtered_sentence.append(temp)
    return filtered_sentence
```

Code Information

Since the combined dataset of all the authors is huge, training will take a lot of time. Therefore only a select few sentences of each school is used for training the data.

```
In [16]: philo_data['sentence_lowered']=philo_data['sentence_lowered'].apply(str)
philo_data['sentence_lowered']=philo_data['sentence_lowered'].apply(lambda x: x.lower()).reset_index()
philo_data['paragraph']=philo_data['sentence_lowered'].str[:100000]
```

Code Information

The data below is test data. It can be changed and schools of new sentences can be predicted. Just for fun, snippets of Trump's tweet is used in test data for finding the school associated with it. It can be changed to find school of newer sentences.

```
In [25]: test_data="The concept of global warming was created by and for the Chinese in order to make US manufac
turing non-competitive. Sorry losers and haters, but my I.Q. is one of the highest - and you all know i
t! Please don't feel so stupid or insecure, it's not your fault. 26,000 unreported sexual assaults in t
he military-only 238 convictions."
```

```
In [26]: training_data=philo_data['paragraph'].iloc[0:]
philo_data_school=philo_data['school'].iloc[0:]

philo_data_sim=pd.DataFrame()
philo_data_sim['school']=philo_data['school'].iloc[0:]
philo_data_sim['paragraph']=philo_data['paragraph'].iloc[0:]

#keyword=remove_stopwords([text for text in test_data_lower.split()])
tokenized = nltk.word_tokenize(test_data)
keywords = [word for word,pos in nltk_pos_tag(tokenized) if 'NN' in pos or 'JJ' in pos ]
keyword = list(dict.fromkeys(keyword))
texts = [[word for word,pos in nltk_pos_tag(nltk.word_tokenize(doc)) if 'NN' in pos or 'JJ' in pos] for
doc in training_data]
#texts = [[text.lower(i) for text in doc.split(i)] for doc in training_data]
#texts = [jtokenize(str(text)) for text in training_data]
texts=remove_stopwords_train(texts)
#texts=remove_anagram(texts)

#texts=remove_stopwords_lol(texts)
dictionary = corpora.Dictionary(texts)
feature_cnt = len(dictionary.token2id)
corpus = [dictionary.doc2bow(text) for text in texts]
tfidf = models.TfidfModel(corpus)
corpus_tfidf = tfidf[corpus]
lsi = models.LsiModel(corpus_tfidf, id2word=dictionary) # initialize an LSI transformation
kw_vector = dictionary.doc2bow(keyword)
vec_lsi = lsi[kw_vector] # convert the query to LSI space

index = similarities.MatrixSimilarity(lsi[corpus]) # transform corpus to LSI space and index it

sim = index[vec_lsi]
philo_data_sim['sim']=sim
philo_data_sim=philo_data_sim.sort_values(by=['sim'],ascending=False)
if(philo_data_sim['sim'].iloc[0]==0):
    print("This text is not similar to any philosophical text this system has encountered till now.")
else:
    print("The text is similar to",philo_data_sim['school'].iloc[0],"school of thought with a simil
arity of",philo_data_sim['sim'].iloc[0])
#for i in range(len(sim)):
    #print("keyword is similar to text -> %d: %2f" % (philo_data_sim.iloc[i+1], sim[i]))
```

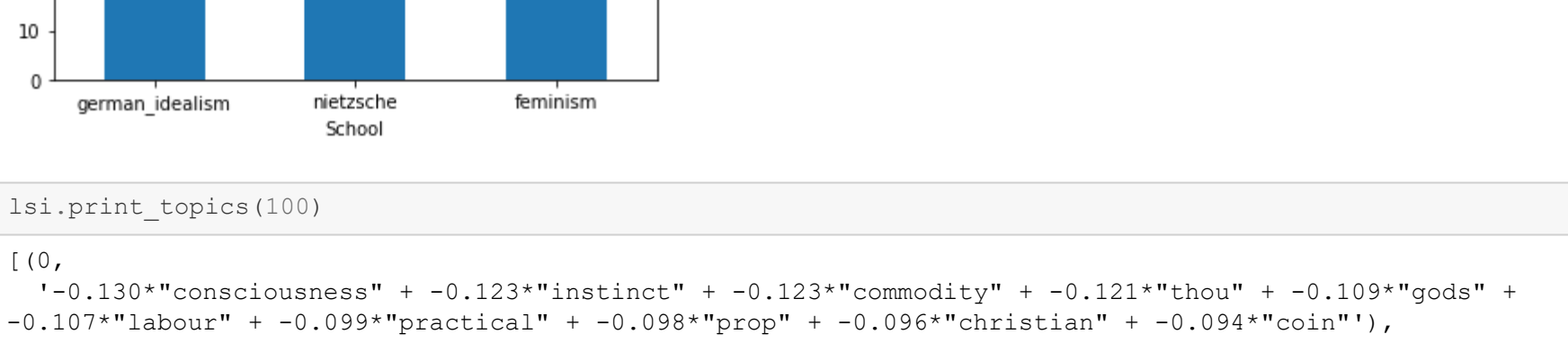
The text is similar to german_idealism school of thought with a similarity of 0.7626275

Code Information

The top 3 school of thoughts are shown below as per the words and topics identified by LSI in the training data and test data.

```
In [32]: df = pd.DataFrame({'School':philo_data_sim['school'][:3], 'Similarity':philo_data_sim['sim'][:3]
})
ax = df.plot.bar(x='School', y='Similarity', rot=0)
ax.set(title='Top 3 similar schools of thoughts to your words')
```

Out[32]: (Text(0.5, 'Top 3 similar schools of thoughts to your words'))



```
In [21]: lsi.print_topics(100)
```

Out[21]:

```
[0,
'-0.130*''consciousness'' + -0.123*''instinct'' + -0.123*''commodity'' + -0.121*''thou'' + -0.109*''gods'' +
-0.107*''labour'' + -0.099*''practical'' + -0.098*''prop'' + -0.096*''christian'' + -0.094*''coin'''),
(1,
'-0.362*''commodity'' + -0.280*''coin'' + -0.273*''labour'' + -0.228*''commodities'' + -0.215*''silver'' + -
0.215*''linen'' + -0.175*''coat'' + -0.173*''gold'' + -0.125*''quantity'' + 0.120*''consciousness'''),
(2,
'-0.204*''thou'' + 0.204*''gods'' + -0.187*''consciousness'' + 0.156*''thy'' + -0.148*''commodity'' + 0.142*''m
eletus'' + 0.132*''pious'' + 0.129*''thee'' + 0.117*''socrates'' + 0.115*''unto'''),
(3,
'-0.306*''prop'' + -0.263*''affirmation'' + -0.231*''contraries'' + -0.195*''white'' + -0.169*''negation'' +
-0.144*''qualification'' + -0.128*''gods'' + -0.127*''socrates'' + -0.120*''meletus'' + -0.117*''relatives'''),
(4,
'-0.493*''prop'' + -0.247*''disease'' + 0.183*''attributes'' + 0.170*''god'' + -0.155*''medicine'' + -0.149*''m
edical'' + 0.147*''practical'' + -0.136*''gods'' + 0.127*''finite'' + 0.122*''attribute'''),
(5,
'-0.412*''disease'' + -0.262*''medicine'' + -0.252*''medical'' + -0.155*''prop'' + -0.150*''hospital'' + 0.14
1*''gods'' + -0.136*''doctor'' + -0.126*''disease'' + 0.124*''instinct'' + -0.103*''viz'''),
(6,
'-0.393*''practical'' + -0.327*''critique'' + -0.187*''ak.'' + -0.175*''maxim'' + -0.156*''speculative'' + -
0.148*''e.'' + 0.142*''prop'' + -0.126*''disease'' + 0.124*''instinct'' + -0.103*''viz'''),
(7,
'-0.318*''prop'' + -0.244*''affirmation'' + -0.223*''contraries'' + 0.205*''disease'' + -0.184*''white'' + -0.
154*''negation'' + -0.139*''qualification'' + 0.139*''gods'' + 0.128*''medicine'' + 0.123*''medical'''),
(8,
'-0.236*''christian'' + 0.232*''prop'' + -0.215*''gospels'' + -0.205*''christianity'' + 0.149*''thou'' + -0.1
22*''instinct'' + 0.121*''thy'' + -0.116*''jewish'' + -0.108*''instincts'' + -0.107*''saviour'''),
(9,
'-0.323*''thou'' + -0.240*''thy'' + -0.215*''thee'' + 0.215*''meletus'' + 0.198*''pious'' + -0.182*''unto'' +
0.153*''euthyphro'' + 0.142*''athens'' + 0.138*''gentlemen'' + 0.121*''gods'''),
(10,
'-0.214*''consent'' + -0.181*''women'' + -0.147*''parents'' + -0.137*''father'' + 0.129
*''sex'' + -0.125*''hath'' + 0.117*''rousseau'' + 0.110*''thou'' + -0.103*''land'''),
(11,
'-0.200*''movements'' + -0.179*''refection'' + -0.154*''husserl'' + 0.137*''behaviour'' + 0.135*''instinct'' +
0.131*''unconscious'' + -0.126*''perception'' + -0.114*''phenomenology'' + -0.107*''red'' + -0.106*''stimul
i'''),
(12,
'-0.385*''coin'' + 0.307*''commodity'' + -0.269*''silver'' + 0.265*''linen'' + 0.230*''coat'' + 0.190*''commod
ities'' + -0.175*''gold'' + 0.156*''yards'' + -0.141*''bullion'' + -0.130*''shillings''')]
```

Explanation

The model suggests Trump's tweets are similar to German Idealism philosophy. This is pretty interesting. On analysing the words in German Idealism, one can find words like unity, duty, judgement, and freedom etc. This might be the reason behind Trump's school prediction as German Idealism.