## Group 4: Kiet Ly, Mary Monroe, and Shaswati Mukherjee

### In [191]:

```
combine_df = []
for file_ in ['/Users/smukherjee5/cs109_final_project/cs109a/data/tweets_nlp_1_2
.csv','/Users/smukherjee5/cs109_final_project/cs109a/data/tweets_nlp_2_2.csv','/
Users/smukherjee5/cs109_final_project/cs109a/data/tweets_nlp_3_2.csv']:
    df = pd.read_csv(file_,index_col=None, header=0,keep_default_na=False)
    combine_df.append(df)
all_tweets = pd.concat(combine_df, axis = 0, ignore_index = True)
display(train_base_tweets_df.head(2))
print('train_tweets_shape:',train_base_tweets_df.shape)
print('test_tweets_shape:',test_base_tweets_df.shape)
```

	retweet_count	favorite_count	num_hashtags	num_urls	num_mentions	user_type	sen
54933	0	0	0	1	0	0	
80644	0	0	0	1	0	0	

train tweets shape: (80574, 22) test tweets shape: (39686, 22)

Ref: <a href="https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0197775">https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0197775</a>

(https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0197775)

https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3813439/

(https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3813439/)

https://rdrr.io/cran/guanteda/man/textstat lexdiv.html (https://rdrr.io/cran/guanteda/man/textstat lexdiv.html)

# **Lexical Diversity Features**

Listed below are some sophisticated techniques that have been developed to measure LD.

ttr- The Type-Token Ratio represents the most used and intuitive way to measure lexical diversity on the basis of word repetition patterns. TTR consists in expressing the number of different words "as a proportion of the total number of words." The higher the probability that a new word token is also a new word type, the closer the TTR is to 1, and the greater the lexical diversity of that text. In the case of lexical diversity measurement, a common strategy used to cope with sample size dependency consists in finding an adequate mathematical expression of the type count slowdown in order to counterbalance its effect on the TTR.

In the case of lexical diversity measurement, a common strategy used to cope with sample size dependency consists in finding an adequate mathematical expression of the type count slowdown in order to counterbalance its effect on the TTR.

root\_ttr - Various attempts were made in this regard: some studies assumed that the ratio fall is proportional to the square root of the token count

log\_ttr - Attempt to 'linearize' the same ratio fall through various logarithmic transformations

mtld - Strategy that has so far successfully dealt with the sample size dependency of the TTR or any TTR-based measure consists in controlling for sample upsizing through fixed size sampling procedures.

HD-D - index derived directly from the hypergeometric distribution

Yule-k - . The measure of lexical repetition constitutes one of the variables used to determine the lexical diversity of texts. Although most of the constants for lexical richness actually depend on text length, Yule's characteristic is considered to be highly reliable for being text length independent

```
In [199]:
```

```
from collections import Counter, defaultdict
from math import sqrt, log
from nltk import word tokenize
import numpy as np
import pandas as pd
import os
def ttr(text):
    tok_text = word_tokenize(text)
    return len(set(tok text)) / len(tok text)
def root ttr(text):
    return sqrt(ttr(text))
def corrected ttr(text):
    tok text = word tokenize(text)
    return sqrt(len(set(tok text)) / (2 * len(tok text)))
def log ttr(text):
    tok text = word_tokenize(text)
    if log(len(tok text),2) == 0:
        print(text)
        return 0
    return log(len(set(tok text)),2) / log(len(tok text),2)
def uber index(text):
    tok text = word tokenize(text)
    if log(len(tok_text),2) != 0 and log(len(set(tok_text)),2) != 0:
        return (log(len(tok_text),2) ** 2) / (log(len(set(tok_text)),2) / log(le
n(tok_text),2))
    else:
        return 0
def yule s k(text):
    tok text = word tokenize(text)
    token counter = Counter(tok.upper() for tok in tok text)
    m1 = sum(token_counter.values())
    m2 = sum([freq ** 2 for freq in token counter.values()])
    if m2-m1 != 0:
        i = (m1*m1) / (m2-m1)
        k = 10000 / i
        return k
```

### In [200]:

```
#Copyright 2017 John Frens
#
#Permission is hereby granted, free of charge, to any person obtaining a copy of
this software and associated documentation files (the "Software"), to deal in th
e Software without restriction, including without limitation the rights to use.
```

```
copy, modify, merge, publish, distribute, sublicense, and/or sell copies of the
Software, and to permit persons to whom the Software is furnished to do so, subj
ect to the following conditions:
#The above copyright notice and this permission notice shall be included in all
copies or substantial portions of the Software.
#THE SOFTWARE IS PROVIDED "AS IS", WITHOUT WARRANTY OF ANY KIND, EXPRESS OR IMPL
IED, INCLUDING BUT NOT LIMITED TO THE WARRANTIES OF MERCHANTABILITY, FITNESS FOR
A PARTICULAR PURPOSE AND NONINFRINGEMENT. IN NO EVENT SHALL THE AUTHORS OR COPYR
IGHT HOLDERS BE LIABLE FOR ANY CLAIM, DAMAGES OR OTHER LIABILITY, WHETHER IN AN
ACTION OF CONTRACT, TORT OR OTHERWISE, ARISING FROM, OUT OF OR IN CONNECTION WIT
H THE SOFTWARE OR THE USE OR OTHER DEALINGS IN THE SOFTWARE.
import string
# Global trandform for removing punctuation from words
remove punctuation = str.maketrans('', '', string.punctuation)
# MTLD internal implementation
def mtld_calc(word_array, ttr_threshold):
    current ttr = 1.0
    token count = 0
    type_count = 0
    types = set()
    factors = 0.0
    for token in word array:
        token = token.translate(remove punctuation).lower() # trim punctuation,
make lowercase
        token count += 1
        if token not in types:
            type_count +=1
            types.add(token)
        current ttr = type count / token count
        if current ttr <= ttr threshold:</pre>
            factors += 1
            token count = 0
            type count = 0
            types = set()
            current_ttr = 1.0
    excess = 1.0 - current ttr
    excess val = 1.0 - ttr threshold
    factors += excess / excess val
    if factors != 0:
        return len(word array) / factors
    return -1
# MTLD implementation
def mtld(word_array, ttr_threshold=0.72):
    word_array = word_tokenize(word_array)
```

```
return (mtld_calc(word_array, ttr_threshold) + mtld_calc(word_array[::-1], t
tr_threshold)) / 2
# HD-D internals
\# x! = x(x-1)(x-2)...(1)
def factorial(x):
    if x <= 1:
        return 1
    else:
        return x * factorial(x - 1)
# n choose r = n(n-1)(n-2)...(n-r+1)/(r!)
def combination(n, r):
   r fact = factorial(r)
   numerator = 1.0
   num = n-r+1.0
   while num < n+1.0:
        numerator *= num
        num += 1.0
    return numerator / r fact
# hypergeometric probability: the probability that an n-trial hypergeometric exp
eriment results
\# in exactly x successes, when the population consists of N items, k of which a
re classified as successes.
# (here, population = N, population successes = k, sample = n, sample successes
= X
\# h(x; N, n, k) = [kCx] * [N-kCn-x] / [NCn]
def hypergeometric(population, population successes, sample, sample successes):
    return (combination(population successes, sample successes) *\
            combination(population - population successes, sample - sample succe
sses)) /\
            combination(population, sample)
# HD-D implementation
def hdd(word array, sample size=42.0):
   word array = word tokenize(word array)
    # Create a dictionary of counts for each type
    type_counts = {}
    for token in word array:
        token = token.translate(remove punctuation).lower() # trim punctuation,
make lowercase
        if token in type counts:
            type counts[token] += 1.0
        else:
            type_counts[token] = 1.0
    # Sum the contribution of each token - "If the sample size is 42, the mean c
ontribution of any given
    # type is 1/42 multiplied by the percentage of combinations in which the ty
pe would be found." (McCarthy & Jarvis 2010)
```

```
for token_type in type_counts.keys():
        contribution = (1.0 - hypergeometric(len(word array), sample size, type
counts[token_type], 0.0)) / sample_size
        hdd value += contribution
    return hdd value
In [201]:
LD_feature_functions = [ttr, root_ttr, corrected_ttr, log_ttr, uber_index, yule_
s k, mtld, hdd]
In [202]:
def save df(path,df):
    df.to csv(path,index=False)
for input file in ['/Users/smukherjee5/cs109 final project/cs109a/data/tweets nl
p 1 2.csv','/Users/smukherjee5/cs109 final project/cs109a/data/tweets nlp 2 2.cs
v','/Users/smukherjee5/cs109 final project/cs109a/data/tweets nlp 3 2.csv']:
    print(input file)
    df = pd.read_csv(input_file,index_col=None, header=0,keep_default_na=False)
    for function in LD_feature_functions:
        df["LD-"+function.__name__] = [function(text) for text in df["text"]]
    filename = input file.replace('.csv','')+' ld.csv'
    save df(filename, df)
/Users/smukherjee5/cs109 final project/cs109a/data/tweets nlp 1 2.cs
Zzzzzzzzzzzzzz
88.8
Goodnight
Beach
Earthquake
كىرك
divergent
3 - 1
Gm
Нарру
Hahahahaha
Gm
Surprise
ا لـحمـدالله
GN
GN
Bored
mbc3♥
我非常真很累死了得不得了
```

hdd\_value = 0.0

你好,台南!

Hobbaloopa testing test no Nothing wiggy drake georgina lucy chloe billi billie sam lydia gemma popey diags ronnie korn haha motherboy MKiss - -Afraaaaah Hungry Starbucks later GoodMorning Done Ηi Traffic \*\*\*\* Barney 12/13/14 Tomorrow Hahahahahah Fireball Ηi Magnum Ηi Bored Hello Lucy Ηi GoodMorning Hahahahahaha GoodNight• •School -\hat{Q}-\hat{ boring Earthquake Okay Gabbers

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Childish
Ugly
Hungry
And
Bored
Pizzzzzaaaaaaaaaaaaa
Yogurt
Wtf..
Bittersweet
Awesome
IDGAF
College
EARTHQUAKKKKE
EEE-
fdhf
QUICKLY
Idk
SEDAL
MOM
ASDFGHJKL
mngbfvdcsa
kk
Milestones
Jantelagen
すかれた。。
Requiem
Exhausted
Ngawur
Froh
Insomnia
Chromaggia
hi
confirmed
memes
DOOOOOOOOOWEEEEEEEOOOOOOO
Why
BOTH
WOW
Alice
Omggggg
Ηi
'Merica
God
GooDAy
Christmas
HEYYEAYAYAYA
Ηi
Pineapples
```

Ηi Fuck.. DENIS tired Integrity Redbullbeatbattle AMANDA JCCO Jwu Busuga Night Morning Summer Earthquake Hala Night 1D Night Ayala Bae Traffic Soon Night Kimmy Sleepy Sick jgh Out deactivate jwu jwu Earthquake Earthquake 台北再見 Franalations Okaayyyyyyy ^\_^ Hypocrite DLC Jerk KTT \*buy GUYS Gn Defense Permitted 2+2=5 Wahooooooo 不改變就是等著被改變!

```
ight
Ignorance
vegefail
vegemite
getting
noooooooo
finally
November
anmalecrossing
/Users/smukherjee5/cs109_final_project/cs109a/data/tweets_nlp_2_2.cs
V
Babe
Earthquake
Cooking
D.O.S
Ć
Night
7-5
Huge
Bored
Fresh
Chill
Huf
Ass
Winning
Earthquake
Skins
Hah
Morning
R.a.f
Jackass
Inbetweeners
Kid
Gym
Lovely
Slaggggg
Roast
Lovely
Cams
Faggot
Christmasssss
Friendsssss
Night
Malletsheisidieksiaka
Rough
Twice
Ready
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Beeg
Taming
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ill
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Haircut
Refreshed
Keen
Swear
Jackson
Test
Boo
Test
ImL
Test
Test
Tweet
Umm
Apple
Slayyyy
Divergent
Bankroll
GOLAZO
CHICHADIOS
LMFAOOOO
2-0
Lmfao
3-0
Lmfaooo
CHICHARO
G000000000L
Lmfao
Duke
EARTHQUAKE
G000000L
LMFAO
G000000000LAZO
Lmfao
GOOOL
Fuck
NAWFSIDE
Minimalism
hi
hi
hi
www.lifeofearth.blogspot.com
www.helpmybusiness.com
www.PissedOffNetworkMarketer.com
www.magic-keywords.co.uk
23:23
```

```
mionzinho*
quem*
T+
seque*
fussanu..xD
CS
www.freeme2day.com
SAINDO
sono.aiaiai
Offf
Off-line
Saindo^^\xb2
80tweets..hahahaha
Nhainnnn
eutosemfazernadaaki.com
off-LINES
Off=lines
Off-lines
Offs
Off-lines
justin
/Users/smukherjee5/cs109 final project/cs109a/data/tweets nlp 3 2.cs
trend
www.speedbit.com.br/s1.htm
hi
\u3068\u3053\u308d\u304c\u3069\u3063\u3053\u3044
\uff65\uff6a\uff65\uff09
hey
\u753a\u7530
essa*
\u30a2\u30af\u30a2\u30d5\u30ec\u30c3\u30b7\u30e5\u306f\u30c1\u30e5\u
30fc\u30d6\u306e\u4e2d\u306b\u3069\u3093\u306a\u98a8\u306b\u8a70\u30
7e\u3063\u3066\u3044\u30
\uff65\uff6a\uff65\uff09
\u30de\u30b8\u30e7\u30de\u30b8\u30e7\u306e\u30df\u30f3\u30c8\u8272\u
306e\u30cd\u30a4\u30eb\u3001\u610f\u5916\u306b\u86cd\u5149\u307d\u30
44\u8272\u3060\u3063\u30
hahahahaha
\u4f1a\u3048\u306a\u3044\u3072\u3068\u306b\u4f1a\u3044\u305f\u3044
\xac\xac
Sushiiii
working
\u308f\u305f\u3057\u3082\u30bb\u30fc\u30eb\u884c\u304d\u305f\u3044\u
306a\u3042
\u5148\u65e5\u30e4\u30d5\u30aa\u30af\u3067\u8ca0\u3051\u3066\u843d\u
3068\u305b\u306a\u304b\u3063\u305f\u3082\u306e\u3068\u540c\u3058\u30
82\u306e\u304c\u672c\u65
\u5915\u65b9\u3068\u3093\u3073\u304c\u98db\u3093\u3067\u305f\u306a\u
3042\u3002\u6771\u4eac\u3067\u898b\u308b\u306e\u306f\u521d\u3081\u30
66\u306e\u3088\u3046\u30
BED
```

```
\u592a\u3089\u306a\u304f\u3066\u590f\u30d0\u30c6\u306b\u52b9\u304f\u
98df\u3079\u7269\u63a2\u3057\u3066\u305f\u3089\u3001\u4eca\u65e5\u30
d5\u30a3\u30d5\u30a3\u30
bored..myspacin..meeboin..BLAH
|\u5f61\uff7b\uff6f|
\u753a\u7530
\u685c\u6728\u753a\u306b\u51fa\u304b\u3051\u305f\u30892\u5e74\u9593\
u9650\u5b9a\u30aa\u30fc\u30d7\u30f3\u3068\u3044\u3046\u30d0\u30d3\u3
0fc\u30ba\u6a2a\u6d5c\u3
internet
\u3072\u3068\u591a\u3059\u304e
\u592a\u3089\u306a\u304f\u3066\u590f\u30d0\u30c6\u306b\u52b9\u304f\u
98df\u3079\u7269\u63a2\u3057\u3066\u305f\u3089\u3001\u4eca\u65e5\u30
d5\u30a3\u30d5\u30a3\u30
Sushiiii
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G0000000000000000000000000L
-q
Tchau
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Т
Andr
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3,000
\u662f \u662f \u662f \u9019\u4e0d \u771f \u6b63\u4eba \u751f \u5b83\u800c \u800c \u8
u662f\u85dd\u8853\u4f5c\u70ba\u771f\u6b63\u7684\u751f\u6d3b\uff0c\u6
211\u8a8d\u70ba\u4ecd\u7
flungo
holaa
hellooooo
Parksonnnnnnn
www.learntocut.com
www.Learntocut.com
www.Learntocut.com
\u0e2a\u0e16\u0e32\u0e19\u0e17\u0e35\u0e48\u0e17\u0e48\u0e2d\u0e07\u
0e40\u0e17\u0e35\u0e48\u0e22\u0e27\u0e0a\u0e37\u0e48\u0e2d\u0e19\u0e
48\u0e32\u0e2b\u0e21\u0e
\u0e28\u0e32\u0e25\u0e17\u0e49\u0e32\u0e27\u0e21\u0e2b\u0e32\u0e1e\u
0e23\u0e2b\u0e21\u0e42\u0e23\u0e07\u0e41\u0e23\u0e21\u0e40\u0e2d\u0e
```

23\u0e32\u0e27\u0e31\u0e \u0e2a\u0e16\u0e32\u0e19\u0e17\u0e35\u0e48\u0e17\u0e48\u0e2d\u0e07\u 0e40\u0e17\u0e35\u0e48\u0e22\u0e27\u0e02\u0e2d\u0e07\u0e27\u0e31\u0e 22\u0e23\u0e38\u0e48\u0e \u0e2a\u0e16\u0e32\u0e19\u0e17\u0e35\u0e48\u0e17\u0e48\u0e2d\u0e07\u 0e40\u0e17\u0e35\u0e48\u0e22\u0e27\u0e17\u0e35\u0e48\u0e2a\u0e33\u0e 04\u0e31\u0e0d\u0e02\u0e \u0e27\u0e31\u0e14\u0e42\u0e1e\u0e18\u0e34\u0e4c\u0e2a\u0e16\u0e32\u 0e19\u0e17\u0e35\u0e48\u0e17\u0e48\u0e2d\u0e07\u0e40\u0e17\u0e35\u0e 48\u0e22\u0e27\u0e17\u0e \u0e1e\u0e32\u0e41\u0e21\u0e48\u0e40\u0e17\u0e35\u0e48\u0e22\u0e27\u 0e2a\u0e16\u0e32\u0e19\u0e17\u0e35\u0e48\u0e17\u0e48\u0e2d\u0e07\u0e 40\u0e17\u0e35\u0e48\u0e \u0e2a\u0e16\u0e32\u0e19\u0e17\u0e35\u0e48\u0e17\u0e49\u0e2d\u0e07\u 0e40\u0e17\u0e35\u0e48\u0e22\u0e27\u0e17\u0e35\u0e48\u0e07\u0e14\u0e 07\u0e32\u0e21\u0e22\u0e \u0e39\u0e39^^ Ηi Ηi oi pov ulla Oie tediooooo lol 0 0 Lol Α lol lol 0 NeverSayNever Shaheen Over 36 1 Lol MemelilerBAYDODOyuTakibEdiyor Music AuS 118/1 LOL EnGland chori Fuck.. Whatifindattractive \*leaves\* OPEN20:00 IGotSooooul Shaheen

Sure

```
0
Lol
yes
LOL
-Wolf
lol
lol
Lol
*music*
3rdSingle
Strong
Err:509
best
*IN
*shouts*
In [203]:
combine_df = []
```

```
combine_df = []
for file_ in ['/Users/smukherjee5/cs109_final_project/cs109a/data/tweets_nlp_1_2
    _ld.csv','/Users/smukherjee5/cs109_final_project/cs109a/data/tweets_nlp_2_2_ld.c
sv','/Users/smukherjee5/cs109_final_project/cs109a/data/tweets_nlp_3_2_ld.csv']:
    df = pd.read_csv(file_,index_col=None, header=0,keep_default_na=False)
    combine_df.append(df)
all_tweets = pd.concat(combine_df, axis = 0, ignore_index = True)
```

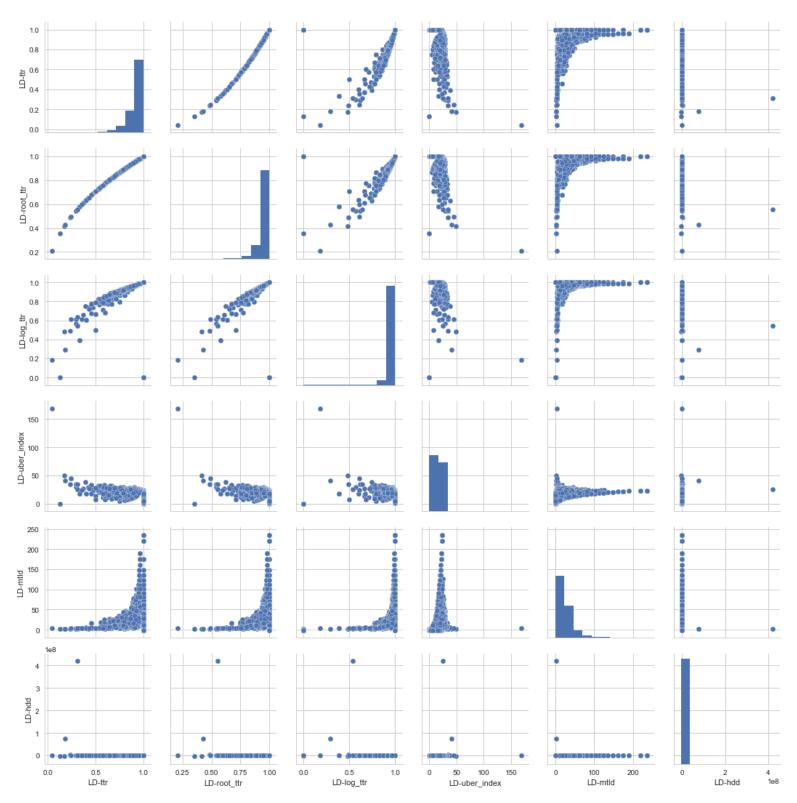
### In [221]:

?

### In [223]:

```
import seaborn as sns;
tweets_ld = all_tweets[['LD-ttr', 'LD-root_ttr', 'LD-log_ttr','LD-uber_index','L
D-yule_s_k','LD-mtld','LD-hdd']]
tweets_ld_sample = resample(tweets_ld, n_samples=5000, replace=False)
g = sns.pairplot(tweets_ld_sample)
print('Scatter Matrix for Lexical Diversity')
g.savefig('twitter-ld')
```

# Scatter Matrix for Lexical Diversity



```
In [226]:
```

```
from sklearn.feature_selection import VarianceThreshold
def variance_threshold_selector(data, threshold=0.5):
    #https://stackoverflow.com/questions/39812885/retain-feature-names-after-sci
kit-feature-selection
    selector = VarianceThreshold(threshold)
    selector.fit(data)
    return data[data.columns[selector.get_support(indices=True)]]
```

### In [273]:

```
tweets_all_var = all_tweets_df[['retweet_count', 'favorite_count', 'num_hashtags
', 'num_urls', 'num_mentions',\
                                  'user type', 'sentiment negative', 'sentiment n
eutral', 'sentiment positive',\
                                 'ratio pos', 'ratio neg', 'ratio neu', 'token c
ount','url_token_ratio', \
                                  'ant', 'disgust', 'fear', 'joy', 'sadness', 'sur
prise', 'trust', 'jaccard', 'LD-ttr', 'LD-root ttr', 'LD-corrected ttr', 'LD-log ttr
','LD-uber index','LD-yule s k','LD-mtld','LD-hdd']]
tweets all var[['LD-yule s k']] = tweets all var[['LD-yule s k']].fillna(0)
def convert float(val):
    try:
        return float(val)
    except ValueError:
        return 0
tweets all var['LD-yule s k']=tweets all var['LD-yule s k'].apply(lambda x: conv
ert float(x))
tweets all var
features = variance threshold selector(tweets all var, threshold=(.95*.1)).colum
ns
```

### In [274]:

features

#### Out[274]:

```
Index(['retweet_count', 'favorite_count', 'num_hashtags', 'num_urls'
, 'num_mentions', 'user_type', 'sentiment_negative', 'sentiment_neut
ral', 'sentiment_positive', 'token_count', 'ant', 'fear', 'joy', 'tr
ust', 'LD-uber_index', 'LD-yule_s_k', 'LD-mtld', 'LD-hdd'], dtype='o
bject')
```

### In [ ]: