

In [1]:

```
1 import pandas as pd
2 import matplotlib.pyplot as plt
3 import numpy as np
4 import seaborn as sns
5 import datetime
6 import re
7 from collections import Counter
```

D:\Working\Anaconda\lib\site-packages\scipy__init__.py:138: UserWarning: A NumPy version $\geq 1.16.5$ and $< 1.23.0$ is required for this version of SciPy (detected version 1.24.3)

```
warnings.warn(f"A NumPy version  $\geq \{np\_minversion\}$  and  $< \{np\_maxversion\}$  is required for this version of "
```

In [2]:

```
1 data = pd.read_csv("C:/Users/86166/Desktop/v2tone/full.csv")
2 data['DATE'] = pd.to_datetime(data['DATE'])
```

In [3]:

1	data
---	------

Out[3]:

	DATE	DocumentIdentifier	V2Tone	
0	2020-01-03	https://www.independent.co.uk/life-style/gadge...	-0.437318	SCIENCE;MANM
1	2020-01-06	https://cleantechnica.com/2020/01/05/energy-st...	-0.150038	WB_135_TRANSPO
2	2020-01-07	https://calgaryherald.com/business/local-busin...	0.846561	UNGP_FORESTS_I
3	2020-01-07	https://calgarysun.com/business/local-business...	0.847458	UNGP_FORESTS_I
4	2020-01-07	https://nationalpost.com/business/local-busine...	0.842993	UNGP_FORESTS_I
...	
7280	2016-12-23	http://www.einnews.com/pr_news/359300579/avalo...	0.942655	WB_855_LABOR_M
7281	2016-12-25	http://www.dealstreetasia.com/stories/india-li...	0.619638	WB_135_TRANSI
7282	2016-12-27	https://knowridge.com/2016/12/lost-lithium-des...	2.404526	DRUG_TRADE;WB
7283	2016-12-28	http://www.marketwatch.com/story/lithium-explo...	-0.975081	TAX_DISEASE;'
7284	2016-12-28	http://www.prnewswire.com/news-releases/lithiu...	-1.025057	TAX_DISEASE;'

7285 rows × 6 columns



In [4]:

1	Y=pd.read_csv("C:/Users/86166/Desktop/v2tone/lithium_v2tone_2016-2023_rate.csv")
---	--

In [5]:

1	Y = Y[['DATE2', 'Li2CO3 99%']]
2	Y = Y.rename(columns={'DATE2': 'DATE'})
3	Y['DATE'] = pd.to_datetime(Y['DATE'])

In [6]:

```
1 mergedt = pd.merge(data, Y, on='DATE')
```

In [7]:

```
1 mergedt['Y'] = abs(mergedt['Li2CO3 99%'])
```

In [8]:

```
1 sort = mergedt.sort_values('Y', ascending=False)
```

In [9]:

```
1 top = sort[sort['Y']!=1][['DATE', 'Y', 'Li2CO3 99%']].drop_duplicates().nlargest(12, keep='al
```

In [10]:

```
1 top = top.sort_values('DATE')
```

In [11]:

```
1 top
```

Out[11]:

	DATE	Y	Li2CO3 99%
5384	2016-01-12	0.085837	0.085837
5387	2016-01-19	0.041152	0.041152
5469	2016-03-02	0.068259	0.068259
2966	2023-03-09	0.043290	-0.043290
3116	2023-03-31	0.050420	-0.050420
3132	2023-04-03	0.055432	-0.055432
3407	2023-05-08	0.050000	0.050000
3455	2023-05-10	0.062350	0.062350
3488	2023-05-11	0.073333	0.073333
3497	2023-05-12	0.070248	0.070248
3509	2023-05-15	0.062016	0.062016
3533	2023-05-18	0.055172	0.055172

In [12]:

```
1 neg = top[top['Li2CO3 99%']<0]
2 neg
```

Out[12]:

	DATE	Y	Li2CO3 99%
2966	2023-03-09	0.043290	-0.043290
3116	2023-03-31	0.050420	-0.050420
3132	2023-04-03	0.055432	-0.055432

In [13]:

```
1 pos = top.sort_values('Y').tail(4)
2 pos
```

Out[13]:

	DATE	Y	Li2CO3 99%
5469	2016-03-02	0.068259	0.068259
3497	2023-05-12	0.070248	0.070248
3488	2023-05-11	0.073333	0.073333
5384	2016-01-12	0.085837	0.085837

Negative

In [14]:

```
1 neg
```

Out[14]:

	DATE	Y	Li2CO3 99%
2966	2023-03-09	0.043290	-0.043290
3116	2023-03-31	0.050420	-0.050420
3132	2023-04-03	0.055432	-0.055432

In [15]:

```
1  ### t7
2  t7 = set()
3  for i in range(neg.shape[0]):
4      dt = neg.iloc[i]['DATE']
5      j=7
6      t7.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t7)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t7)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t7 = res.most_common(15)
17 res_t7
```

```
{Timestamp('2023-03-24 00:00:00'), Timestamp('2023-03-02 00:00:00'), Timestamp('2023-03-27 00:00:00')}
```

```
[V2Tone--std]: 1.7486565762185693
```

```
[Y--std]: 0.0022471920728855932
```

Out[15]:

```
[('TAX_ETHNICITY', 13),
 ('ENV_MINING', 10),
 ('ECON_STOCKMARKET', 7),
 ('LEGISLATION', 5),
 ('MANMADE_DISASTER_IMPLIED', 4),
 ('GENERAL_GOVERNMENT', 4),
 ('ENV_SOLAR', 4),
 ('TAX_DISEASE', 4),
 ('EDUCATION', 3),
 ('ENV_GREEN', 3),
 ('NATURAL_DISASTER', 3),
 ('SLFID_MINERAL_RESOURCES', 2),
 ('KILL', 2),
 ('DELAY', 2),
 ('GENERAL_HEALTH', 2)]
```

In [16]:

```
1 top3 = pd.DataFrame(dict(res_t7), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic,index=range(4)).head(1),'\n')
```

A

TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	1.187074	1.750585	-0.017365	0.002115

B

ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	1.18267	1.484589	-0.017019	0.002321

C

ECON_STOCKMARKET

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.899151	1.306862	-0.016538	0.002653

In [17]:

```
1  ### t6
2  t6 = set()
3  for i in range(neg.shape[0]):
4      dt = neg.iloc[i]['DATE']
5      j=6
6      t6.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t6)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t6)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t6 = res.most_common(15)
17 res_t6
```

```
{Timestamp('2023-03-25 00:00:00'), Timestamp('2023-03-28 00:00:00'), Timestamp('2023-03-03 00:00:00')}
```

```
[V2Tone--std]: 1.2787033209061482
```

```
[Y--std]: 0.005754809183030011
```

Out[17]:

```
[('ENV_MINING', 20),
 ('TAX_ETHNICITY', 19),
 ('ECON_STOCKMARKET', 13),
 ('MANMADE_DISASTER_IMPLIED', 12),
 ('LEGISLATION', 12),
 ('GENERAL_GOVERNMENT', 12),
 ('KILL', 8),
 ('ENV_GREEN', 8),
 ('GENERAL_HEALTH', 7),
 ('SLFID_MINERAL_RESOURCES', 6),
 ('NATURAL_DISASTER', 5),
 ('ECON_CURRENCY_EXCHANGE_RATE', 4),
 ('TRIAL', 4),
 ('TAX_DISEASE', 4),
 ('ECON_TAXATION', 3)]
```

In [18]:

```
1 top3 = pd.DataFrame(dict(res_t6), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'% (i), '\n', pd.DataFrame(dic, index=range(4)).head(1), '\n')
```

C ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.609352	1.285618	-0.023455	0.006486

B TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.641985	0.884184	-0.02412	0.005921

A ECON_STOCKMARKET

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.713154	1.35368	-0.024184	0.005935

In [19]:

```
1  ### t5
2  t5 = set()
3  for i in range(neg.shape[0]):
4      dt = neg.iloc[i]['DATE']
5      j=5
6      t5.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t5)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t5)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t5 = res.most_common(15)
17 res_t5
```

```
{Timestamp('2023-03-29 00:00:00'), Timestamp('2023-03-26 00:00:00'), Timestamp('2023-03-04 00:00:00')}
```

```
[V2Tone--std]: 1.6336928848964023
```

```
[Y--std]: 7.152448122690996e-18
```

Out[19]:

```
[('TAX_ETHNICITY', 13),
 ('ECON_STOCKMARKET', 9),
 ('LEGISLATION', 6),
 ('ENV_MINING', 6),
 ('NATURAL_DISASTER', 5),
 ('SLFID_MINERAL_RESOURCES', 4),
 ('MANMADE_DISASTER_IMPLIED', 4),
 ('GENERAL_GOVERNMENT', 4),
 ('URBAN', 4),
 ('ELECTION', 4),
 ('RATIFY', 4),
 ('TAX_DISEASE', 3),
 ('ECON_ENTREPRENEURSHIP', 2),
 ('ENV_GREEN', 2),
 ('KILL', 2)]
```

In [20]:

```
1 top3 = pd.DataFrame(dict(res_t5), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic,index=range(4)).head(1),'\n')
```

A TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	1.031674	1.313241	-0.023346	7.222230e-18

A ECON_STOCKMARKET

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	1.225434	1.526008	-0.023346	0.0

A LEGISLATION

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.459598	1.427765	-0.023346	0.0

In [21]:

```
1  ### t4
2  t4 = set()
3  for i in range(neg.shape[0]):
4      dt = neg.iloc[i]['DATE']
5      j=4
6      t4.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t4)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t4)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t4 = res.most_common(15)
17 res_t4
```

```
{Timestamp('2023-03-30 00:00:00'), Timestamp('2023-03-05 00:00:00'), Timestamp('2023-03-27 00:00:00')}
```

```
[V2Tone--std]: 1.579902713696243
```

```
[Y--std]: 0.004644958091178718
```

Out[21]:

```
[('TAX_ETHNICITY', 21),
 ('ENV_MINING', 18),
 ('ECON_STOCKMARKET', 13),
 ('GENERAL_GOVERNMENT', 12),
 ('LEGISLATION', 11),
 ('MANMADE_DISASTER_IMPLIED', 9),
 ('TAX_DISEASE', 8),
 ('KILL', 6),
 ('ENV_GREEN', 6),
 ('GENERAL_HEALTH', 4),
 ('DELAY', 4),
 ('PROTEST', 4),
 ('SLFID_MINERAL_RESOURCES', 4),
 ('NEGOTIATIONS', 4),
 ('NATURAL_DISASTER', 3)]
```

In [22]:

```
1 top3 = pd.DataFrame(dict(res_t4), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic,index=range(4)).head(1),'\n')
```

C TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.809604	1.699252	-0.024388	0.004718

B ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.685218	1.656523	-0.025366	0.00437

A ECON_STOCKMARKET

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.863653	1.088693	-0.025812	0.004158

In [23]:

```
1  ### t3
2  t3 = set()
3  for i in range(neg.shape[0]):
4      dt = neg.iloc[i]['DATE']
5      j=3
6      t3.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t3)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t3)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t3 = res.most_common(15)
17 res_t3
```

```
{Timestamp('2023-03-31 00:00:00'), Timestamp('2023-03-28 00:00:00'), Timestamp('2023-03-06 00:00:00')}
```

```
[V2Tone--std]: 1.8608267110233963
```

```
[Y--std]: 0.015233517257276398
```

Out[23]:

```
[('TAX_ETHNICITY', 33),
 ('ENV_MINING', 31),
 ('GENERAL_GOVERNMENT', 21),
 ('ECON_STOCKMARKET', 21),
 ('MANMADE_DISASTER_IMPLIED', 18),
 ('LEGISLATION', 12),
 ('SLFID_MINERAL_RESOURCES', 11),
 ('TAX_DISEASE', 10),
 ('GENERAL_HEALTH', 10),
 ('NATURAL_DISASTER', 8),
 ('KILL', 8),
 ('ENV_GREEN', 7),
 ('ECON_TAXATION', 7),
 ('ENV_OIL', 5),
 ('ECON_ENTREPRENEURSHIP', 5)]
```

In [24]:

```
1 top3 = pd.DataFrame(dict(res_t3), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic,index=range(4)).head(1),'\n')
```

A TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.29501	1.383319	-0.033592	0.015136

C ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.617851	1.290584	-0.032506	0.014975

B GENERAL_GOVERNMENT

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.461631	1.416169	-0.032854	0.015642

In [25]:

```
1  ### t2
2  t2 = set()
3  for i in range(neg.shape[0]):
4      dt = neg.iloc[i]['DATE']
5      j=2
6      t2.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t2)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t2)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t2 = res.most_common(15)
17 res_t2
```

```
{Timestamp('2023-03-29 00:00:00'), Timestamp('2023-04-01 00:00:00'), Timestamp('2023-03-07 00:00:00')}
```

```
[V2Tone--std]: 1.914955841621569
```

```
[Y--std]: 0.009724598284415749
```

Out[25]:

```
[('TAX_ETHNICITY', 31),
 ('ENV_MINING', 21),
 ('MANMADE_DISASTER_IMPLIED', 13),
 ('ECON_STOCKMARKET', 10),
 ('GENERAL_GOVERNMENT', 9),
 ('LEGISLATION', 7),
 ('NATURAL_DISASTER', 6),
 ('EDUCATION', 5),
 ('TAX_DISEASE', 5),
 ('GENERAL_HEALTH', 4),
 ('MEDICAL', 4),
 ('SLFID_MINERAL_RESOURCES', 4),
 ('URBAN', 4),
 ('ELECTION', 4),
 ('RATIFY', 4)]
```

In [26]:

```
1 top3 = pd.DataFrame(dict(res_t2), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic,index=range(4)).head(1),'\n')
```

A TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.131633	1.468065	-0.01217	0.009655

C ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-0.140201	1.324484	-0.009598	0.00891

B MANMADE_DISASTER_IMPLIED

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-0.908658	2.0961	-0.010021	0.009246

In [27]:

```
1 ### t1
2 t1 = set()
3 for i in range(neg.shape[0]):
4     dt = neg.iloc[i]['DATE']
5     j=1
6     t1.add(dt+datetime.timedelta(days=-j))
7     #day.add(dt+datetime.timedelta(days=1))
8 print(t1)
9 d1 = mergedt.loc[mergedt['DATE'].isin(t1)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t1 = res.most_common(15)
17 res_t1
```

{Timestamp('2023-04-02 00:00:00'), Timestamp('2023-03-08 00:00:00'), Timestamp('2023-03-30 00:00:00')}

[V2Tone--std]: 1.4626233761691043

[Y--std]: 0.007089198913374481

Out[27]:

```
[('ENV_MINING', 16),
 ('TAX_ETHNICITY', 14),
 ('ECON_STOCKMARKET', 14),
 ('GENERAL_GOVERNMENT', 10),
 ('LEGISLATION', 8),
 ('ENV_GREEN', 7),
 ('MANMADE_DISASTER_IMPLIED', 6),
 ('SLFID_MINERAL_RESOURCES', 5),
 ('TAX_DISEASE', 5),
 ('KILL', 5),
 ('DELAY', 4),
 ('NEGOTIATIONS', 4),
 ('CORRUPTION', 3),
 ('ARMEDCONFLICT', 3),
 ('GENERAL_HEALTH', 3)]
```

In [28]:

```
1 top3 = pd.DataFrame(dict(res_t1), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'% (i), '\n', pd.DataFrame(dic, index=range(4)).head(1), '\n')
```

B

ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.307117	1.712552	-0.025084	0.006269

A

TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.300646	1.653841	-0.026889	0.004156

C

ECON_STOCKMARKET

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.7984	0.799391	-0.023557	0.007291

In [29]:

```
1  ### t0
2  t0 = set()
3  for i in range(neg.shape[0]):
4      dt = neg.iloc[i]['DATE']
5      j=0
6      t0.add(dt+datetime.timedelta(days=j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t0)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t0)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t0 = res.most_common(15)
17 res_t0
```

```
{Timestamp('2023-03-09 00:00:00'), Timestamp('2023-03-31 00:00:00'), Timestamp('2023-04-03 00:00:00')}
```

```
[V2Tone--std]: 2.325190058649111
```

```
[Y--std]: 0.004568839967703527
```

Out[29]:

```
[('TAX_ETHNICITY', 29),
 ('ENV_MINING', 20),
 ('TAX_DISEASE', 17),
 ('MANMADE_DISASTER_IMPLIED', 16),
 ('GENERAL_HEALTH', 15),
 ('ECON_STOCKMARKET', 15),
 ('KILL', 12),
 ('GENERAL_GOVERNMENT', 12),
 ('SCIENCE', 11),
 ('EDUCATION', 11),
 ('HEALTH_PANDEMIC', 10),
 ('BAN', 10),
 ('SLFID_MINERAL_RESOURCES', 10),
 ('LEGISLATION', 9),
 ('ARMEDCONFLICT', 8)]
```

In [30]:

```
1 top3 = pd.DataFrame(dict(res_t0), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic,index=range(4)).head(1),'\n')
```

B

TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-0.83113	1.853501	-0.049417	0.004645

A

ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.488366	1.666214	-0.050247	0.004157

C

TAX_DISEASE

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-1.661974	2.104821	-0.048664	0.005024

In [31]:

```
1  ### tpl
2  tpl = set()
3  for i in range(neg.shape[0]):
4      dt = neg.iloc[i]['DATE']
5      j=1
6      tpl.add(dt+datetime.timedelta(days=j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(tpl)
9  d1=mergedt.loc[mergedt['DATE'].isin(tpl)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_tpl = res.most_common(15)
17 res_tpl
```

```
{Timestamp('2023-03-10 00:00:00'), Timestamp('2023-04-01 00:00:00'), Timestamp('2023-04-04 00:00:00')}
```

```
[V2Tone--std]: 1.997702467160171
```

```
[Y--std]: 0.007543756077995352
```

Out[31]:

```
[('TAX_ETHNICITY', 26),
 ('ENV_MINING', 14),
 ('MANMADE_DISASTER_IMPLIED', 13),
 ('GENERAL_GOVERNMENT', 13),
 ('ECON_STOCKMARKET', 11),
 ('LEGISLATION', 11),
 ('GENERAL_HEALTH', 9),
 ('SCIENCE', 9),
 ('TAX_DISEASE', 8),
 ('ENV_OIL', 7),
 ('WOUND', 6),
 ('KILL', 5),
 ('WATER_SECURITY', 5),
 ('ECON_ENTREPRENEURSHIP', 5),
 ('MEDICAL', 5)]
```

In [32]:

```
1 top3 = pd.DataFrame(dict(res_tp1), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'% (i), '\n', pd.DataFrame(dic, index=range(4)).head(1), '\n')
```

A

TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-0.002185	1.863549	-0.021582	0.007646

B

ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-0.239197	1.861236	-0.021087	0.007723

A

MANMADE_DISASTER_IMPLIED

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-1.00774	1.891283	-0.021582	0.007803

In [33]:

```
1 ### sum(t-7 --- t-1)
2 all7 = set()
3 for i in range(neg.shape[0]):
4     dt = neg.iloc[i]['DATE']
5     for j in range(1,8):
6         all7.add(dt+datetime.timedelta(days=-j))
7         #day.add(dt+datetime.timedelta(days=1))
8 print(all7)
9 d1 = mergedt.loc[mergedt['DATE'].isin(all7)].reset_index(drop=True)
10 print('\n[V2Tone--std]:',d1['V2Tone'].std(),'\n[Y--std]:',d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_all7 = res.most_common(15)
17 res_all7
```

```
{Timestamp('2023-03-08 00:00:00'), Timestamp('2023-04-02 00:00:00'), Timestamp('2023-03-07 00:00:00'), Timestamp('2023-03-29 00:00:00'), Timestamp('2023-03-27 00:00:00'), Timestamp('2023-03-26 00:00:00'), Timestamp('2023-04-01 00:00:00'), Timestamp('2023-03-06 00:00:00'), Timestamp('2023-03-30 00:00:00'), Timestamp('2023-03-31 00:00:00'), Timestamp('2023-03-03 00:00:00'), Timestamp('2023-03-04 00:00:00'), Timestamp('2023-03-02 00:00:00'), Timestamp('2023-03-24 00:00:00'), Timestamp('2023-03-28 00:00:00'), Timestamp('2023-03-25 00:00:00'), Timestamp('2023-03-05 00:00:00')}
```

[V2Tone--std]: 1.7833960452128648

[Y--std]: 0.014303938443581764

Out[33]:

```
[('TAX_ETHNICITY', 94),
 ('ENV_MINING', 82),
 ('ECON_STOCKMARKET', 54),
 ('GENERAL_GOVERNMENT', 46),
 ('MANMADE_DISASTER_IMPLIED', 44),
 ('LEGISLATION', 35),
 ('SLFID_MINERAL_RESOURCES', 25),
 ('TAX_DISEASE', 25),
 ('ENV_GREEN', 22),
 ('KILL', 20),
 ('GENERAL_HEALTH', 19),
 ('NATURAL_DISASTER', 18),
 ('DELAY', 13),
 ('EDUCATION', 12),
 ('ECON_TAXATION', 12)]
```

In [34]:

```
1 top5 = pd.DataFrame(dict(res_all17), index=['cnt']).columns[0:5]
2 for i in top5:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic, index=range(4)).head(1),'\n')
```

TAX_ETHNICITY				
	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.355204	1.500698	-0.022558	0.014216
	L	H	H	M
ENV_MINING				
	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.356517	1.45844	-0.022244	0.014273
	M	M	L	HH
ECON_STOCKMARKET				
	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.6603	1.333849	-0.02657	0.012416
	HH	LL	HH	L
GENERAL_GOVERNMENT				
	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.065559	1.420073	-0.025491	0.014266
	H	L	M	H
MANMADE_DISASTER_IMPLIED				
	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-0.383188	1.998733	-0.019974	0.013587
	LL	HH	LL	LL

Positive

In [35]:

1	pos
---	-----

Out[35]:

	DATE	Y	Li2CO3 99%
5469	2016-03-02	0.068259	0.068259
3497	2023-05-12	0.070248	0.070248
3488	2023-05-11	0.073333	0.073333
5384	2016-01-12	0.085837	0.085837

In [36]:

```
1  ### t7
2  t7 = set()
3  for i in range(pos.shape[0]):
4      dt = pos.iloc[i]['DATE']
5      j=7
6      t7.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t7)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t7)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t7 = res.most_common(15)
17 res_t7
```

```
{Timestamp('2016-02-24 00:00:00'), Timestamp('2023-05-04 00:00:00'), Timestamp('2023-05-05 00:00:00'), Timestamp('2016-01-05 00:00:00')}
```

```
[V2Tone--std]: 2.132678325877105
```

```
[Y--std]: 0.001881917041893799
```

Out[36]:

```
[('ENV_MINING', 17),
 ('TAX_ETHNICITY', 14),
 ('GENERAL_GOVERNMENT', 8),
 ('SLFID_MINERAL_RESOURCES', 8),
 ('ECON_STOCKMARKET', 6),
 ('LEGISLATION', 6),
 ('BAN', 5),
 ('MANMADE_DISASTER_IMPLIED', 5),
 ('WATER_SECURITY', 4),
 ('ENV_OIL', 2),
 ('SLFID_NATURAL_RESOURCES', 2),
 ('PUBLIC_TRANSPORT', 2),
 ('ECON_NATIONALIZE', 2),
 ('SHORTAGE', 2),
 ('GENERAL_HEALTH', 2)]
```

In [37]:

```
1 top3 = pd.DataFrame(dict(res_t7), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'% (i), '\n', pd.DataFrame(dic, index=range(4)).head(1), '\n')
```

ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.809867	1.34515	0.003596	0.001294

TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.348388	1.969314	0.003374	0.0016

GENERAL_GOVERNMENT

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.328402	1.364235	0.003474	0.001275

In [38]:

```
1  ### t6
2  t6 = set()
3  for i in range(pos.shape[0]):
4      dt = pos.iloc[i]['DATE']
5      j=6
6      t6.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t6)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t6)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t6 = res.most_common(15)
17 res_t6
```

```
{Timestamp('2016-02-25 00:00:00'), Timestamp('2023-05-05 00:00:00'), Timestamp('2023-05-06 00:00:00'), Timestamp('2016-01-06 00:00:00')}
```

```
[V2Tone--std]: 1.9215851241337818
```

```
[Y--std]: 0.0028451666039426573
```

Out[38]:

```
[('ENV_MINING', 10),
 ('TAX_ETHNICITY', 9),
 ('MANMADE_DISASTER_IMPLIED', 8),
 ('SLFID_MINERAL_RESOURCES', 6),
 ('GENERAL_GOVERNMENT', 6),
 ('LEGISLATION', 5),
 ('DELAY', 4),
 ('ECON_STOCKMARKET', 3),
 ('SHORTAGE', 2),
 ('ENV_OIL', 2),
 ('ECON_NATIONALIZE', 1),
 ('GENERAL_HEALTH', 1),
 ('MEDICAL', 1),
 ('EDUCATION', 1),
 ('DRUG_TRADE', 1)]
```

In [39]:

```
1 top3 = pd.DataFrame(dict(res_t6), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'% (i), '\n', pd.DataFrame(dic, index=range(4)).head(1), '\n')
```

A ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.613783	0.861678	0.00277	0.00292

B TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.520421	1.317213	0.002462	0.00292

C MANMADE_DISASTER_IMPLIED

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-0.456028	2.268122	0.001385	0.002565

In [40]:

```
1  ### t5
2  t5 = set()
3  for i in range(pos.shape[0]):
4      dt = pos.iloc[i]['DATE']
5      j=5
6      t5.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t5)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t5)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t5 = res.most_common(15)
17 res_t5
```

```
{Timestamp('2016-01-07 00:00:00'), Timestamp('2016-02-26 00:00:00'), Timestamp('2023-05-07 00:00:00'), Timestamp('2023-05-06 00:00:00')}
```

```
[V2Tone--std]: 1.725107557477848
```

```
[Y--std]: 0.0
```

Out[40]:

```
[('ENV_MINING', 3),
 ('TAX_ETHNICITY', 3),
 ('ECON_STOCKMARKET', 2),
 ('MANMADE_DISASTER_IMPLIED', 2),
 ('GENERAL_GOVERNMENT', 2),
 ('ENV_OIL', 1),
 ('NEGOTIATIONS', 1),
 ('ECON_TAXATION', 1),
 ('EDUCATION', 1),
 ('RETIREMENTS', 1),
 ('RETIREMENT', 1),
 ('ENV_COAL', 1),
 ('BAN', 1),
 ('MARITIME_INCIDENT', 1),
 ('MARITIME', 1)]
```

In [41]:

```
1 top3 = pd.DataFrame(dict(res_t5), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'% (i), '\n', pd.DataFrame(dic, index=range(4)).head(1), '\n')
```

ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.407442	0.369458	0.0	0.0

TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-1.090314	2.078333	0.0	0.0

ECON_STOCKMARKET

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-1.743462	2.465628	0.0	0.0

In [42]:

```
1  ### t4
2  t4 = set()
3  for i in range(pos.shape[0]):
4      dt = pos.iloc[i]['DATE']
5      j=4
6      t4.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t4)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t4)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t4 = res.most_common(15)
17 res_t4
```

```
{Timestamp('2016-01-08 00:00:00'), Timestamp('2016-02-27 00:00:00'), Timestamp('2023-05-07 00:00:00'), Timestamp('2023-05-08 00:00:00')}
```

```
[V2Tone--std]: 0.9210569881555004
```

```
[Y--std]: 7.152448122690996e-18
```

Out[42]:

```
[('TAX_ETHNICITY', 13),
 ('ECON_STOCKMARKET', 10),
 ('ENV_MINING', 6),
 ('GENERAL_GOVERNMENT', 5),
 ('MANMADE_DISASTER_IMPLIED', 5),
 ('TAX_DISEASE', 4),
 ('ECON_SUBSIDIES', 3),
 ('DELAY', 3),
 ('SLFID_MINERAL_RESOURCES', 3),
 ('GENERAL_HEALTH', 3),
 ('LEGISLATION', 2),
 ('HEALTH_PANDEMIC', 2),
 ('MEDICAL', 2),
 ('STRIKE', 2),
 ('ECON_CURRENCY_EXCHANGE_RATE', 2)]
```

In [43]:

```
1 top3 = pd.DataFrame(dict(res_t4), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic, index=range(4)).head(1),'\n')
```

TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.308304	0.95407	0.05	0.0

ECON_STOCKMARKET

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.237432	0.928135	0.05	7.314236e-18

ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.152858	0.975116	0.05	7.601177e-18

In [44]:

```
1  ### t3
2  t3 = set()
3  for i in range(pos.shape[0]):
4      dt = pos.iloc[i]['DATE']
5      j=3
6      t3.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t3)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t3)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t3 = res.most_common(15)
17 res_t3
```

```
{Timestamp('2023-05-09 00:00:00'), Timestamp('2023-05-08 00:00:00'), Timestamp('2016-01-09 00:00:00'), Timestamp('2016-02-28 00:00:00')}
```

```
[V2Tone--std]: 1.1772566393466533
```

```
[Y--std]: 0.011071471653217443
```

Out[44]:

```
[('TAX_ETHNICITY', 29),
 ('MANMADE_DISASTER_IMPLIED', 16),
 ('ECON_STOCKMARKET', 13),
 ('TAX_DISEASE', 13),
 ('ENV_MINING', 12),
 ('GENERAL_GOVERNMENT', 11),
 ('ECON_ENTREPRENEURSHIP', 9),
 ('GENERAL_HEALTH', 8),
 ('SLFID_MINERAL_RESOURCES', 6),
 ('LEGISLATION', 6),
 ('DELAY', 5),
 ('ENV_GREEN', 4),
 ('ECON_SUBSIDIES', 3),
 ('HEALTH_PANDEMIC', 3),
 ('ECON_CURRENCY_EXCHANGE_RATE', 3)]
```

In [45]:

```
1 top3 = pd.DataFrame(dict(res_t3), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic,index=range(4)).head(1),'\n')
```

B

TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.200528	1.256901	0.037935	0.011067

C

MANMADE_DISASTER_IMPLIED

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-0.170097	1.282165	0.034966	0.010468

A

ECON_STOCKMARKET

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.317429	1.608971	0.044954	0.009589

In [46]:

```
1  ### t2
2  t2 = set()
3  for i in range(pos.shape[0]):
4      dt = pos.iloc[i]['DATE']
5      j=2
6      t2.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t2)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t2)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t2 = res.most_common(15)
17 res_t2
```

```
{Timestamp('2023-05-09 00:00:00'), Timestamp('2016-02-29 00:00:00'), Timestamp('2023-05-10 00:00:00'), Timestamp('2016-01-10 00:00:00')}
```

```
[V2Tone--std]: 1.8939420204016237
```

```
[Y--std]: 0.017768196128823528
```

Out[46]:

```
[('TAX_ETHNICITY', 36),
 ('MANMADE_DISASTER_IMPLIED', 21),
 ('ECON_STOCKMARKET', 19),
 ('ENV_MINING', 17),
 ('TAX_DISEASE', 15),
 ('ECON_ENTREPRENEURSHIP', 11),
 ('GENERAL_GOVERNMENT', 8),
 ('ENV_GREEN', 8),
 ('GENERAL_HEALTH', 7),
 ('LEGISLATION', 6),
 ('AGRICULTURE', 5),
 ('SLFID_MINERAL_RESOURCES', 4),
 ('ECON_TAXATION', 4),
 ('NEW_CONSTRUCTION', 3),
 ('DELAY', 3)]
```

In [47]:

```
1 top3 = pd.DataFrame(dict(res_t2), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'% (i), '\n', pd.DataFrame(dic, index=range(4)).head(1), '\n')
```

TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.393412	1.284264	0.047143	0.017244

MANMADE_DISASTER_IMPLIED

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	-0.357598	1.641804	0.044427	0.017511

ECON_STOCKMARKET

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.893999	1.492231	0.056947	0.012819

In [48]:

```
1  ### t1
2  t1 = set()
3  for i in range(pos.shape[0]):
4      dt = pos.iloc[i]['DATE']
5      j=1
6      t1.add(dt+datetime.timedelta(days=-j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t1)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t1)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t1 = res.most_common(15)
17 res_t1
```

```
{Timestamp('2023-05-11 00:00:00'), Timestamp('2016-03-01 00:00:00'), Timestamp('2023-05-10 00:00:00'), Timestamp('2016-01-11 00:00:00')}
```

```
[V2Tone--std]: 1.7939166523874135
```

```
[Y--std]: 0.01934557761944701
```

Out[48]:

```
[('TAX_ETHNICITY', 31),
 ('ECON_STOCKMARKET', 29),
 ('ENV_MINING', 28),
 ('MANMADE_DISASTER_IMPLIED', 20),
 ('GENERAL_GOVERNMENT', 11),
 ('TAX_DISEASE', 8),
 ('ENV_GREEN', 8),
 ('NEGOTIATIONS', 7),
 ('ECON_TAXATION', 6),
 ('WATER_SECURITY', 6),
 ('ECON_ENTREPRENEURSHIP', 5),
 ('NATURAL_DISASTER', 5),
 ('ETH_INDIGINOUS', 5),
 ('GENERAL_HEALTH', 4),
 ('LEGISLATION', 4)]
```

In [49]:

```
1 top3 = pd.DataFrame(dict(res_t1), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic,index=range(4)).head(1),'\n')
```

B

TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.68426	1.006618	0.063882	0.012937

C

ECON_STOCKMARKET

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	1.276498	1.357922	0.059687	0.021266

A

ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	0.830999	1.450749	0.066399	0.014088

In [50]:

```
1  ### t0 .
2  t0 = set()
3  for i in range(pos.shape[0]):
4      dt = pos.iloc[i]['DATE']
5      j=0
6      t0.add(dt+datetime.timedelta(days=j))
7      #day.add(dt+datetime.timedelta(days=1))
8  print(t0)
9  d1 = mergedt.loc[mergedt['DATE'].isin(t0)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_t0 = res.most_common(15)
17 res_t0
```

```
{Timestamp('2023-05-11 00:00:00'), Timestamp('2023-05-12 00:00:00'), Timestamp('2016-01-12 00:00:00'), Timestamp('2016-03-02 00:00:00')}
```

```
[V2Tone--std]: 1.1899149841724204
```

```
[Y--std]: 0.002926315864684628
```

Out[50]:

```
[('ENV_MINING', 36),
 ('TAX_ETHNICITY', 28),
 ('ECON_STOCKMARKET', 22),
 ('GENERAL_GOVERNMENT', 20),
 ('MANMADE_DISASTER_IMPLIED', 17),
 ('LEGISLATION', 14),
 ('SLFID_MINERAL_RESOURCES', 14),
 ('SCIENCE', 7),
 ('TAX_DISEASE', 7),
 ('WATER_SECURITY', 7),
 ('PROTEST', 6),
 ('ETH_INDIGINOUS', 6),
 ('ECON_ENTREPRENEURSHIP', 5),
 ('KILL', 5),
 ('NATURAL_DISASTER', 5)]
```

In [51]:

```
1 top3 = pd.DataFrame(dict(res_t0), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic,index=range(4)).head(1),'\n')
```

B ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	1.223447	1.154811	0.071343	0.001919

C TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	1.118152	1.052492	0.070995	0.001918

A ECON_STOCKMARKET

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	1.373175	1.246953	0.071907	0.003763

In [52]:

```
1 ### tpl
2 tpl = set()
3 for i in range(pos.shape[0]):
4     dt = pos.iloc[i]['DATE']
5     j=1
6     tpl.add(dt+datetime.timedelta(days=j))
7     #day.add(dt+datetime.timedelta(days=1))
8 print(tpl)
9 d1=mergedt.loc[mergedt['DATE'].isin(tpl)].reset_index(drop=True)
10 print('\n[V2Tone--std]:', d1['V2Tone'].std(), '\n[Y--std]:', d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_tpl = res.most_common(15)
17 res_tpl
```

{Timestamp('2023-05-13 00:00:00'), Timestamp('2016-03-03 00:00:00'), Timestamp('2023-05-12 00:00:00'), Timestamp('2016-01-13 00:00:00')}

[V2Tone--std]: 1.2125766368956326

[Y--std]: 0.035724420187622584

Out[52]:

```
[('ENV_MINING', 25),
 ('TAX_ETHNICITY', 24),
 ('SLFID_MINERAL_RESOURCES', 20),
 ('LEGISLATION', 18),
 ('ECON_STOCKMARKET', 18),
 ('SLFID_NATURAL_RESOURCES', 11),
 ('ELECTION', 10),
 ('SCIENCE', 7),
 ('GENERAL_GOVERNMENT', 7),
 ('TAX_DISEASE', 5),
 ('MANMADE_DISASTER_IMPLIED', 5),
 ('EDUCATION', 5),
 ('PROTEST', 4),
 ('ECON_ENTREPRENEURSHIP', 4),
 ('ECON_TAXATION', 4)]
```



In [53]:

```
1 top3 = pd.DataFrame(dict(res_tp1), index=['cnt']).columns[0:3]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2CO3 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic,index=range(4)).head(1),'\n')
```

A

ENV_MINING

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	1.41703	0.98098	0.042149	0.035124

B

TAX_ETHNICITY

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	1.190814	1.201358	0.038051	0.035755

C

SLFID_MINERAL_RESOURCES

	mean_V2Tone	std_V2Tone	mean_Y	std_Y
0	1.438538	0.666518	0.028099	0.035308

In [54]:

```
1 ### sum(t-7 --- t-1)
2 all7 = set()
3 for i in range(pos.shape[0]):
4     dt = pos.iloc[i]['DATE']
5     for j in range(1,8):
6         all7.add(dt+datetime.timedelta(days=-j))
7         #day.add(dt+datetime.timedelta(days=1))
8 print(all7)
9 d1 = mergedt.loc[mergedt['DATE'].isin(all7)].reset_index(drop=True)
10 print('\n[V2Tone--std]:',d1['V2Tone'].std(),'\n[Y--std]:',d1['Li2CO3 99%'].std())
11 themes=[]
12 for i in (d1['FinalThemes'].values):
13     for j in i.split(','):
14         themes.append(j)
15     res = Counter(themes)
16 res_all7 = res.most_common(15)
17 res_all7
```

```
{Timestamp('2016-02-29 00:00:00'), Timestamp('2023-05-04 00:00:00'), Timestamp('2016-01-06 00:00:00'), Timestamp('2016-02-28 00:00:00'), Timestamp('2023-05-05 00:00:00'), Timestamp('2016-01-11 00:00:00'), Timestamp('2016-01-09 00:00:00'), Timestamp('2023-05-10 00:00:00'), Timestamp('2016-02-26 00:00:00'), Timestamp('2016-02-24 00:00:00'), Timestamp('2016-01-05 00:00:00'), Timestamp('2016-01-10 00:00:00'), Timestamp('2016-03-01 00:00:00'), Timestamp('2016-01-08 00:00:00'), Timestamp('2016-01-07 00:00:00'), Timestamp('2023-05-06 00:00:00'), Timestamp('2023-05-09 00:00:00'), Timestamp('2023-05-11 00:00:00'), Timestamp('2016-02-25 00:00:00'), Timestamp('2016-02-27 00:00:00'), Timestamp('2023-05-07 00:00:00'), Timestamp('2023-05-08 00:00:00')}
```

[V2Tone--std]: 1.8036836053390493

[Y--std]: 0.028589012627860105

Out[54]:

```
[('TAX_ETHNICITY', 82),
 ('ENV_MINING', 65),
 ('ECON_STOCKMARKET', 50),
 ('MANMADE_DISASTER_IMPLIED', 49),
 ('GENERAL_GOVERNMENT', 36),
 ('TAX_DISEASE', 22),
 ('SLFID_MINERAL_RESOURCES', 21),
 ('LEGISLATION', 21),
 ('GENERAL_HEALTH', 16),
 ('ECON_ENTREPRENEURSHIP', 15),
 ('ENV_GREEN', 13),
 ('DELAY', 11),
 ('WATER_SECURITY', 10),
 ('NEGOTIATIONS', 9),
 ('ECON_TAXATION', 9)]
```

In [56]:

```
1 top3 = pd.DataFrame(dict(res_all7), index=['cnt']).columns[0:5]
2 for i in top3:
3     mean_V2Tone = (d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].mean()
4     std_V2Tone=(d1.loc[d1['FinalThemes'].str.contains(i)]['V2Tone'].std()
5     mean_Y=(d1.loc[d1['FinalThemes'].str.contains(i)]['Li2C03 99%'].mean()
6     std_Y = (d1.loc[d1['FinalThemes'].str.contains(i)]['Li2C03 99%'].std()
7     dic = {'mean_V2Tone':mean_V2Tone,'std_V2Tone':std_V2Tone,'mean_Y':mean_Y,'std_Y':std_Y}
8     print('%s'%(i),'\n',pd.DataFrame(dic, index=range(4)).head(1),'\n')
```

TAX_ETHNICITY

mean_V2Tone std_V2Tone mean_Y std_Y
0 0.373404 1.384379 0.038143 0.026663

ENV_MINING

mean_V2Tone std_V2Tone mean_Y std_Y
0 0.743897 1.257733 0.036756 0.03087

ECON_STOCKMARKET

mean_V2Tone std_V2Tone mean_Y std_Y
0 0.900576 1.516554 0.046695 0.026509

MANMADE_DISASTER_IMPLIED

mean_V2Tone std_V2Tone mean_Y std_Y
0 -0.378359 1.792738 0.037951 0.02813

GENERAL_GOVERNMENT

mean_V2Tone std_V2Tone mean_Y std_Y
0 0.26007 1.378118 0.034203 0.029961

$\Sigma (1 - \dots - 1)$

	Sentiment Avg	Stmnt Volatility	Y Avg	Y Volatility
mining	H	L	M	H
SM	H+I	M	HH	LL
MD	LL	HH	L	M
GG	L	L	LL	M
IE	M	M	M	L