

## Second Dataset

February 21, 2021

```
[12]: import numpy as np
import pandas as pd
import pickle
from statsmodels import tsa
import statsmodels.api as sm
from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
from statsmodels import multivariate
from statsmodels import regression
import scipy.stats as stats
from statsmodels.sandbox.regression import gmm
from statsmodels.sandbox.regression.gmm import GMM
import statsmodels.stats.diagnostic as smd
from statsmodels.tsa.adfvalues import mackinnonp, mackinnoncrit
from statsmodels.tsa.stattools import adfuller
import statsmodels.tsa.api as smt
from statsmodels.tsa.vector_ar.hypothesis_test_results import CausalityTestResults
from statsmodels.tsa.vector_ar.var_model import VAR, VARProcess, VARResults
from statsmodels.tsa.vector_ar.vecm import VECM, coint_johansen, select_order
import statsmodels.tsa.arima_model as am
from statsmodels.regression.rolling import RollingOLS

from tabulate import tabulate

import datetime as dt
from dateutil.relativedelta import relativedelta
from datetime import timedelta

import seaborn as sns
import matplotlib.pyplot as plt
from matplotlib.dates import DateFormatter, MinuteLocator
from matplotlib.ticker import PercentFormatter

import os
import warnings
```

```
from scipy.optimize import minimize, brute
from arch import arch_model
```

```
warnings.filterwarnings("ignore")
```

```
[13]: covid_data = pd.read_csv('/Users/garik/Datathon/owid-covid-data.csv')
eu = covid_data[covid_data.continent == 'Europe']
```

```
[14]: countries = eu.location.unique()
deaths = []
cases = []
dates = []
for c in countries:
    temp_df = eu[eu.location == c]
    dates.append(temp_df.date.values)
    cases.append(temp_df.total_cases_per_million.values)
    deaths.append(temp_df.total_deaths_per_million.values)
```

```
[15]: ans1 = pd.DataFrame([dates[0],cases[0],deaths[0]]).T
ans1.columns = ['Date',countries[0] + '_Cases',countries[0] + '_Deaths']

for i in range(1,len(countries)):
    ans2 = pd.DataFrame([dates[i],cases[i],deaths[i]]).T
    ans2.columns = ['Date',countries[i] + '_Cases',countries[i] + '_Deaths']
    ans2.index = ans2['Date']
    ans1 = ans1.join(ans2,on = 'Date',rsuffix = 'r')

ans1 = ans1.drop(['Dater'],axis = 1)
ans1.iloc[0] = ans1.iloc[0].fillna(0)
ans1 = ans1.ffill()
```

```
[16]: cl = ans1.iloc[:,1::2]
import scipy.cluster.hierarchy as shc

plt.figure(figsize=(18, 7))
plt.title("Cases Dendograms")
dend = shc.dendrogram(shc.linkage(cl.T, method='ward'),truncate_mode='lastp',
                        orientation='top', show_leaf_counts=True,
                        no_labels = False,labels = cl.columns,leaf_rotation =90)

cluster_1 = [1,4,9,24,28,36,45]
cl_1 = list(np.array(countries[cluster_1]) + "_Cases")

cluster_2 = np.arange(0,cl.shape[1])
```

```

cluster_2= cluster_2[list(map(lambda x: x not in cluster_1,cluster_2))]
cl_2 = list(np.array(countries[cluster_2]) + "_Cases")

series_1 = ans1[list(np.array(countries[cluster_1]) + "_Cases")]

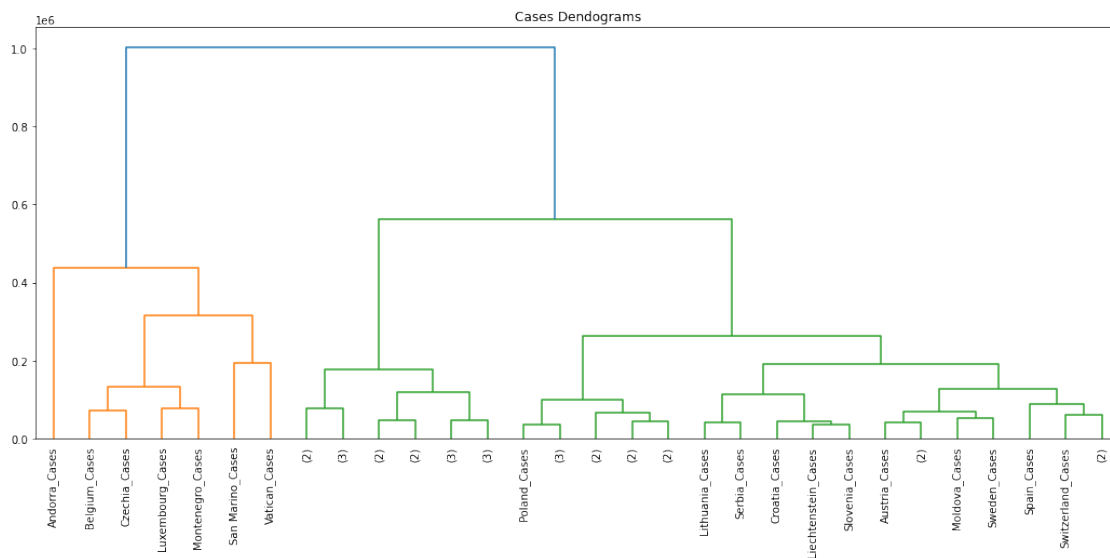
plt.figure(figsize = (15,4))
plt.plot(series_1)
plt.legend(cl_1)
plt.show()

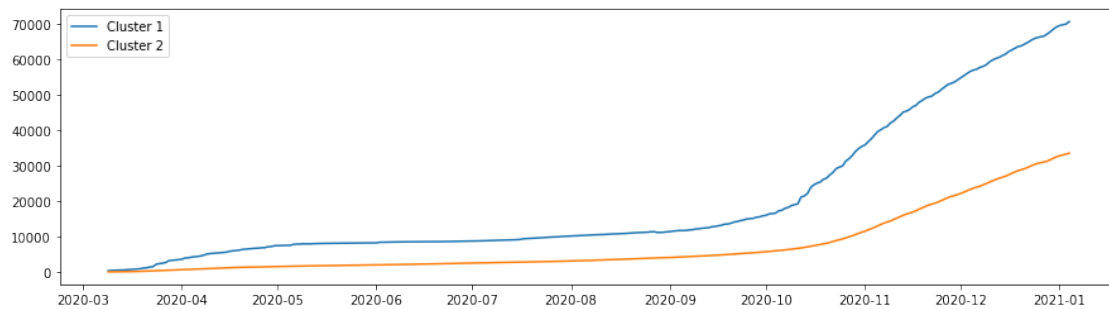
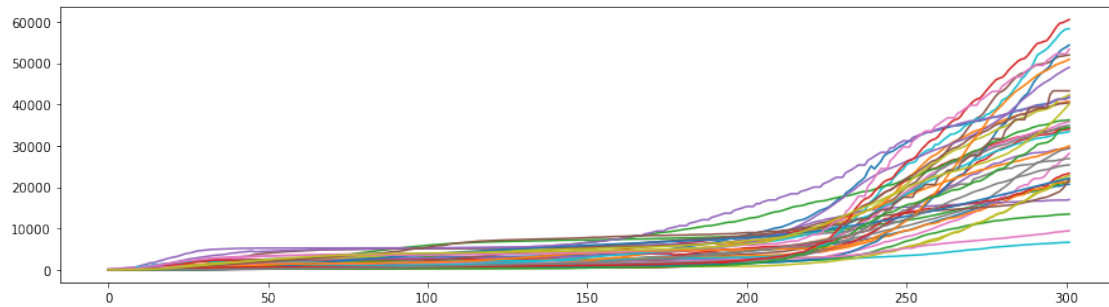
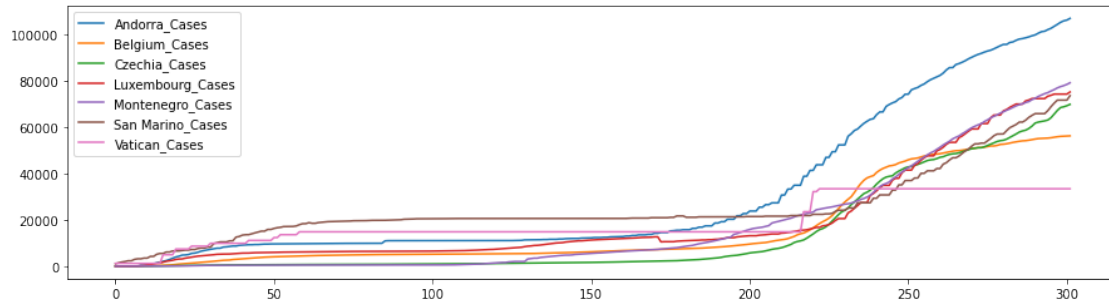
series_2 = ans1[list(np.array(countries[cluster_2]) + "_Cases")]
plt.figure(figsize = (15,4))
plt.plot(series_2)
plt.show()

series_1.index = pd.to_datetime(ans1.Date)
series_2.index = pd.to_datetime(ans1.Date)

plt.figure(figsize = (15,4))
plt.plot(series_1.mean(axis = 1))
plt.plot(series_2.mean(axis = 1))
plt.legend(["Cluster 1","Cluster 2"])
plt.show()

```





[22]: ans1

	Date	Albania_Cases	Albania_Deaths	Andorra_Cases	Andorra_Deaths	\
0	2020-03-09	0.695	0.000	12.942	0.000	
1	2020-03-10	3.475	0.000	12.942	0.000	
2	2020-03-11	4.170	0.347	12.942	0.000	
3	2020-03-12	7.992	0.347	12.942	0.000	
4	2020-03-13	11.467	0.347	12.942	0.000	
..	...	...	...	...	...	
297	2020-12-31	20264.091	410.383	104173.947	1087.168	
298	2021-01-01	20264.091	410.383	105054.035	1087.168	
299	2021-01-02	20498.645	413.510	105688.216	1087.168	

300	2021-01-03	20653.972	414.553	106024.720	1087.168
301	2021-01-04	20718.257	416.638	106762.441	1087.168

	Austria_Cases	Austria_Deaths	Belarus_Cases	Belarus_Deaths	\
0	14.545	0.000	0.635	0.000	
1	20.208	0.000	0.952	0.000	
2	27.314	0.000	0.952	0.000	
3	33.532	0.111	1.270	0.000	
4	55.960	0.111	2.857	0.000	
..	...	...	...	...	
297	40062.067	690.842	20560.631	150.699	
298	40294.790	695.172	20765.831	151.651	
299	40449.236	696.727	20967.115	152.604	
300	40612.009	702.167	21161.520	153.556	
301	40794.324	705.831	21359.312	154.614	

	Belgium_Cases	...	Sweden_Cases	Sweden_Deaths	Switzerland_Cases	\
0	20.622	...	32.280	0.000	43.214	
1	23.038	...	41.983	0.099	56.733	
2	27.093	...	61.391	0.099	75.336	
3	27.093	...	76.342	0.198	75.336	
4	48.233	...	91.393	0.297	131.606	
..	...	...	...	...	...	
297	55782.349	...	43307.982	864.122	52260.654	
298	55937.056	...	43307.982	864.122	52260.654	
299	56012.986	...	43307.982	864.122	52260.654	
300	56085.637	...	43307.982	864.122	52260.654	
301	56161.222	...	43307.982	864.122	53377.399	

	Switzerland_Deaths	Ukraine_Cases	Ukraine_Deaths	United Kingdom_Cases	\
0	0.231	0.023	0.000	9.280	
1	0.347	0.023	0.000	13.095	
2	0.462	0.023	0.000	19.164	
3	0.462	0.023	0.000	26.368	
4	1.271	0.069	0.023	33.438	
..	...	...	...	...	
297	883.343	24854.872	440.872	36770.923	
298	890.161	25080.282	444.439	37558.390	
299	893.049	25203.779	445.834	38410.598	
300	895.129	25315.386	448.852	39223.092	
301	914.887	25418.144	450.727	40091.062	

	United Kingdom_Deaths	Vatican_Cases	Vatican_Deaths
0	0.044	1236.094	0
1	0.103	1236.094	0
2	0.103	1236.094	0
3	0.133	1236.094	0

4	0.147	1236.094	0
..	...	...	...
297	1084.495	33374.536	0
298	1093.554	33374.536	0
299	1100.109	33374.536	0
300	1106.811	33374.536	0
301	1112.851	33374.536	0

[302 rows x 93 columns]

```
[67]: cl = ans1.iloc[:,2::2]
import scipy.cluster.hierarchy as shc
from scipy.cluster.hierarchy import fcluster

plt.figure(figsize=(18, 7))
plt.title("Dendrogram: Deaths per Million")

Z = shc.linkage(cl.T, method='ward')
dend = shc.dendrogram(Z,truncate_mode='lastp',
                      orientation='top', show_leaf_counts=True,
                      no_labels = False,leaf_rotation =90,labels =_
                      ↪list(map(lambda x: x[:-7],cl.columns)))

k = 2
names_cl = fcluster(Z, k, criterion='maxclust')

cluster_1 = cl.columns[names_cl == 1]
cl_1 = list(cluster_1)
cluster_2 = cl.columns[names_cl == 2]
cl_2 = list(cluster_2)

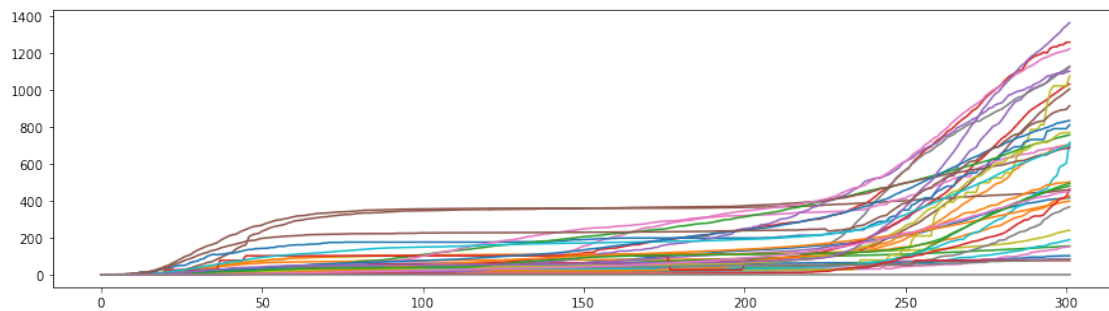
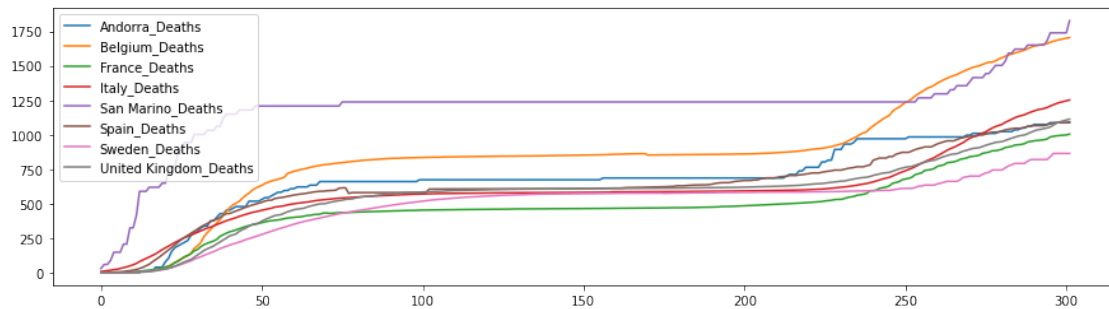
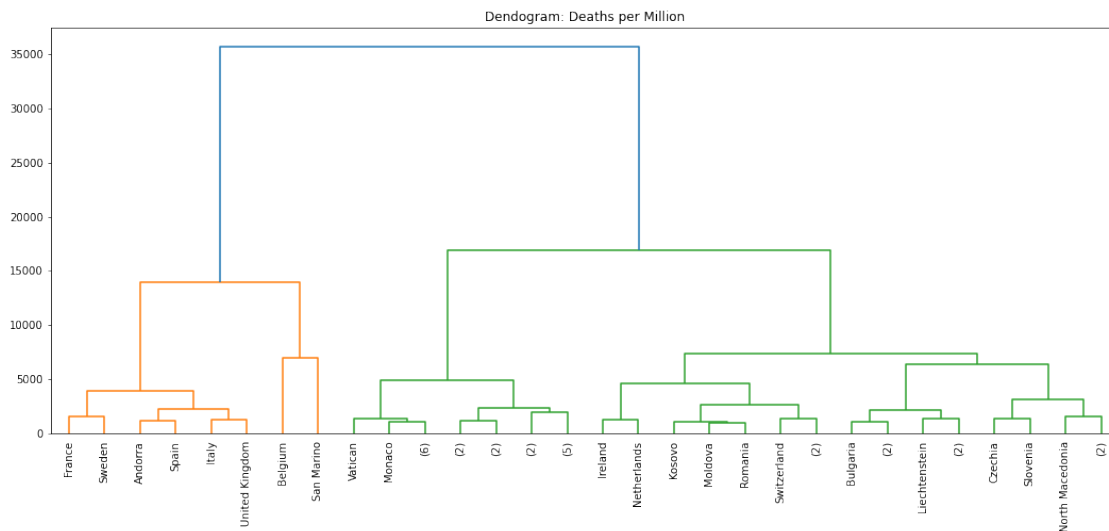
series_1 = ans1[list(cluster_1)]

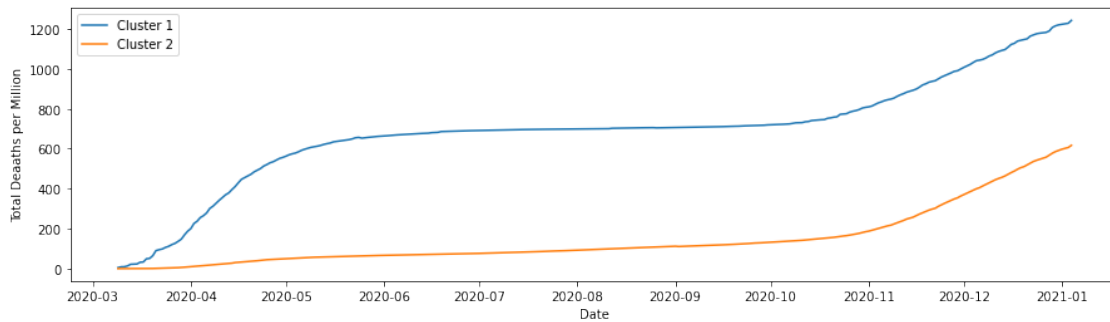
plt.figure(figsize = (15,4))
plt.plot(series_1)
plt.legend(cl_1)
plt.show()

series_2 = ans1[list(cluster_2)]
plt.figure(figsize = (15,4))
plt.plot(series_2)
plt.show()

series_1.index = pd.to_datetime(ans1.Date)
series_2.index = pd.to_datetime(ans1.Date)
```

```
plt.figure(figsize = (15,4))
plt.plot(series_1.mean(axis = 1))
plt.plot(series_2.mean(axis = 1))
plt.legend(["Cluster 1","Cluster 2"])
plt.ylabel("Total Deaaths per Million")
plt.xlabel("Date")
plt.show()
```





```
[63]: data = dict(type = 'choropleth',
                locations = pd.Series(list(map(lambda x: x[:-7],cl.columns))),
                locationmode = 'country names',
                colorscale= 'Portland',
                text= ['IND','NEP','CHI','PAK','BAN','BHU', 'MYN','SLK'],
                z=pd.Series(names_cl),
                colorbar = {'title':'Country Clusters', 'len':200,'lenmode':
                ↪'pixels' })
```

```
[56]: names_cl
```

```
[56]: array([2, 1, 2, 2, 1, 2, 2, 2, 2, 2, 2, 2, 2, 1, 2, 2, 2, 2, 2, 1, 2, 2,
            2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 1, 2, 2, 2, 1, 1, 2, 2,
            1, 2], dtype=int32)
```

```
[60]: import plotly.graph_objs as gobj
import pandas as pd
from plotly.offline import download_plotlyjs,init_notebook_mode,plot,iplot
init_notebook_mode(connected=True)
```

```
[66]: data = dict(type = 'choropleth',
                locations = pd.Series(list(map(lambda x: x[:-7],cl.columns))),
                locationmode = 'country names',
                colorscale= 'Portland',
                text= ['IND','NEP','CHI','PAK','BAN','BHU', 'MYN','SLK'],
                z=pd.Series(names_cl),
                colorbar = {'title':'Country Clusters', 'len':200,'lenmode':
                ↪'pixels' })

layout = dict(geo = {'scope':'europe'})
col_map = gobj.Figure(data = [data],layout = layout)
```



```
ipplot(col_map)
```

### 0.0.1 -30,-29,...29,30 days, graphs

```
[68]: ans1.index = pd.to_datetime(ans1.Date)

cases_data = ans1.iloc[:,1::2]
cols = cases_data.columns
cols_new = list(map(lambda x: x[:-6],cols))
cases_data.columns = cols_new

death_data = ans1.iloc[:,2::2]
cols = death_data.columns
cols_new = list(map(lambda x: x[:-7],cols))
death_data.columns = cols_new

response = pd.read_csv('country_response_measures.csv',parse_dates=True)
response.name = 'response'

response.date_start = pd.to_datetime(response.date_start)
response.date_end = pd.to_datetime(response.date_end)

response = response[response.date_start > min(cases_data.index)]

def compare_measures(response,by_data = death_data, m = 'MassGatherAll',du = 1,
    ↪30,dd = 30):

    mass = response[response.Response_measure == m]
    mass_countries = mass.Country.unique()

    policy_matrix = pd.DataFrame(columns = mass_countries, index= pd.
    ↪to_datetime(by_data.index)).fillna(0)

    for c in mass_countries:
        df_temp = mass[mass.Country == c]
        try:
            policy_matrix.loc[df_temp.date_start.iloc[0],c] = 1
        except:
            pass
        try:
            policy_matrix.loc[df_temp.date_end.iloc[0],c] = -1
        except:
            pass
```

```

policy_matrix = policy_matrix.dropna()
changes_before_after = pd.DataFrame(index = np.arange(-dd,du+1),columns =
↳policy_matrix.columns)
    for c in policy_matrix.columns:
        row = policy_matrix[c]
        pdate = row[row == 1].index[0]
        temp_change = by_data.loc[pdate - relativedelta(days = du):
                                pdate + relativedelta(days = dd),c].
↳pct_change()*100

        if(temp_change.shape[0] <= du+dd):
            changes_before_after.loc[0 - temp_change[:pdate].shape[0]:-1,c] =
↳temp_change[:pdate].values

            changes_before_after.loc[0: 0 + temp_change[pdate:].shape[0],c] =
↳temp_change[pdate:].values
        else:

            changes_before_after.loc[0 - temp_change[:pdate].shape[0]:
                                -1,c] = temp_change[:
↳pdate-relativedelta(days = 1)].values

            changes_before_after.loc[0: 0 + temp_change[pdate:].shape[0],c] =
↳temp_change[pdate:].values

    return changes_before_after.replace([np.inf, -np.inf], np.nan)

```

```

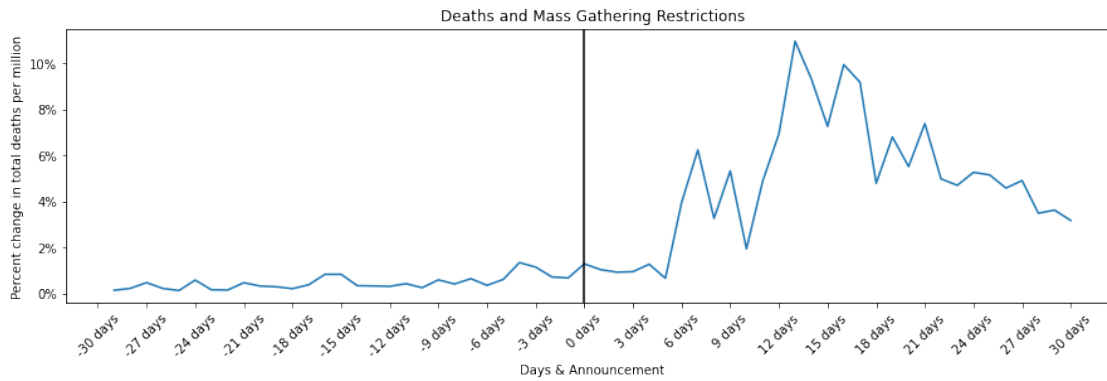
[101]: ev_name = 'MassGatherAll'
title = 'Mass Gathering Restrictions'
dd = 30
du = 30

plt.figure(figsize = (15,4))
md = compare_measures(response,by_data = death_data, m = ev_name)
y = md.ffill().quantile(0.5,axis = 1)
y.plot()

plt.axvline(x = 0 , c = 'black')
plt.title("Deaths and "+ title)
plt.ylabel('Percent change in total deaths per million')
plt.xlabel("Days & Announcement")
plt.xticks(ticks = np.arange(-dd,du+3,3),
           labels = list(map(lambda x: str(x)+ ' days',np.arange(-dd,du+3,3))),
           rotation=45)

```

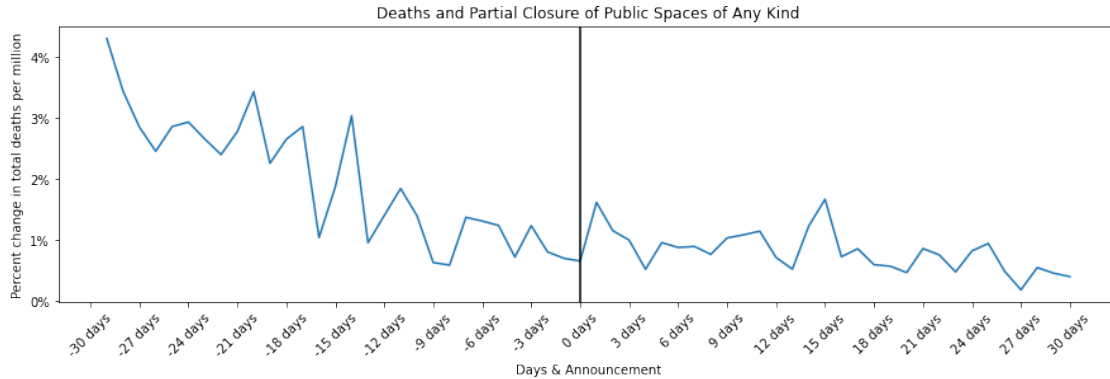
```
plt.gca().set_yticklabels(['{:0f}%'.format(x) for x in plt.gca().get_yticks()])
#plt.legend(['10% quantile',
            #'50% quantile',
            #'90% quantile'])
plt.show()
```



```
[103]: ev_name = 'ClosPubAnyPartial'
title = 'Partial Closure of Public Spaces of Any Kind'
dd = 30
du = 30

plt.figure(figsize = (15,4))
md = compare_measures(response,by_data = death_data, m = ev_name)
y = md.ffill().quantile(0.5,axis = 1)
y.plot()

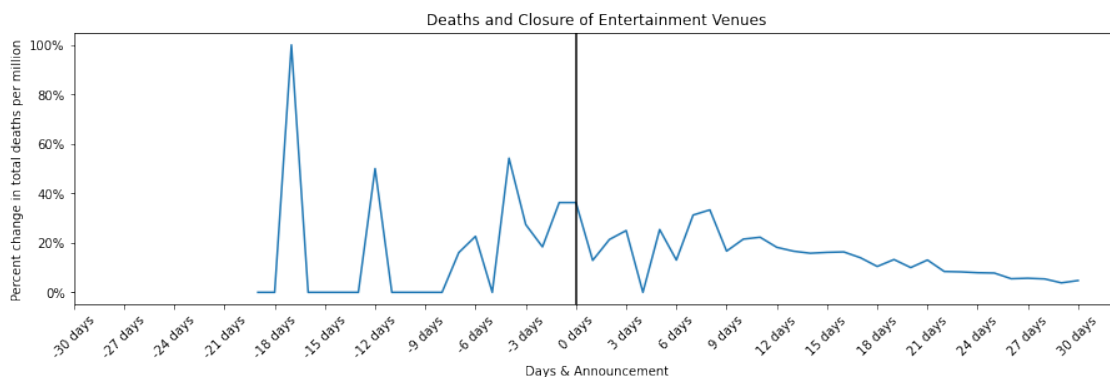
plt.axvline(x = 0 , c = 'black')
plt.title("Deaths and " + title)
plt.ylabel('Percent change in total deaths per million')
plt.xlabel("Days & Announcement")
plt.xticks(ticks = np.arange(-dd,du+3,3),
           labels = list(map(lambda x: str(x)+ ' days',np.arange(-dd,du+3,3))),
           rotation=45)
plt.gca().set_yticklabels(['{:0f}%'.format(x) for x in plt.gca().get_yticks()])
#plt.legend(['10% quantile',
            #'50% quantile',
            #'90% quantile'])
plt.show()
```



```
[104]: ev_name = 'EntertainmentVenues'
title = 'Closure of Entertainment Venues'

plt.figure(figsize = (15,4))
md = compare_measures(response,by_data = death_data, m = ev_name)
y = md.ffill().quantile(0.5,axis = 1)
y.plot()

plt.axvline(x = 0 , c = 'black')
plt.title("Deaths and "+ title)
plt.ylabel('Percent change in total deaths per million')
plt.xlabel("Days & Announcement")
plt.xticks(ticks = np.arange(-dd,du+3,3),
           labels = list(map(lambda x: str(x)+ ' days',np.arange(-dd,du+3,3))),
           rotation=45)
plt.gca().set_yticklabels(['{:0.0f}%'.format(x) for x in plt.gca().get_yticks()])
#plt.legend(['10% quantile',
#           #'50% quantile',
#           #'90% quantile'])
plt.show()
```

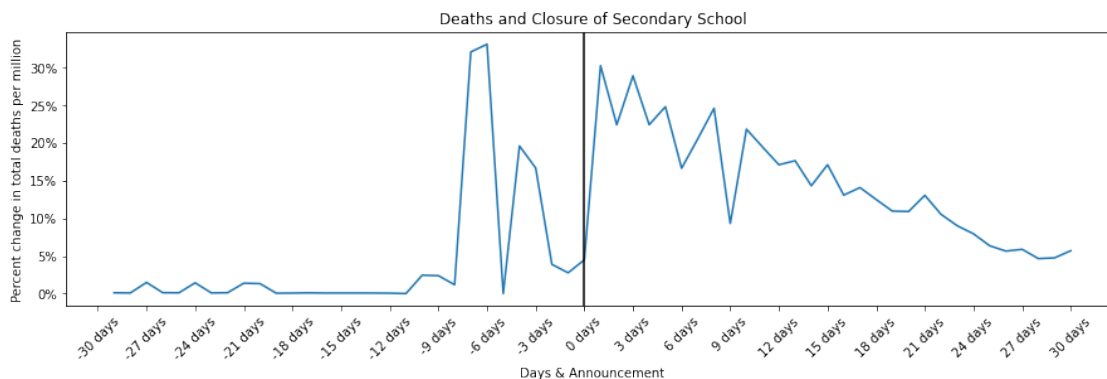


```
[105]: ev_name = 'ClosSec'
title = 'Closure of Secondary School'

plt.figure(figsize = (15,4))
md = compare_measures(response,by_data = death_data, m = ev_name)
y = md.ffill().quantile(0.5,axis = 1)
y.plot()

plt.axvline(x = 0 , c = 'black')
plt.title("Deaths and "+ title)
plt.ylabel('Percent change in total deaths per million')
plt.xlabel("Days & Announcement")
plt.xticks(ticks = np.arange(-dd,du+3,3),
           labels = list(map(lambda x: str(x)+ ' days',np.arange(-dd,du+3,3))),
           rotation=45)
plt.gca().set_yticklabels(['{: .0f}%'.format(x) for x in plt.gca().get_yticks()])
#plt.legend(['10% quantile',
#           #'50% quantile',
#           #'90% quantile'])

plt.show()
```



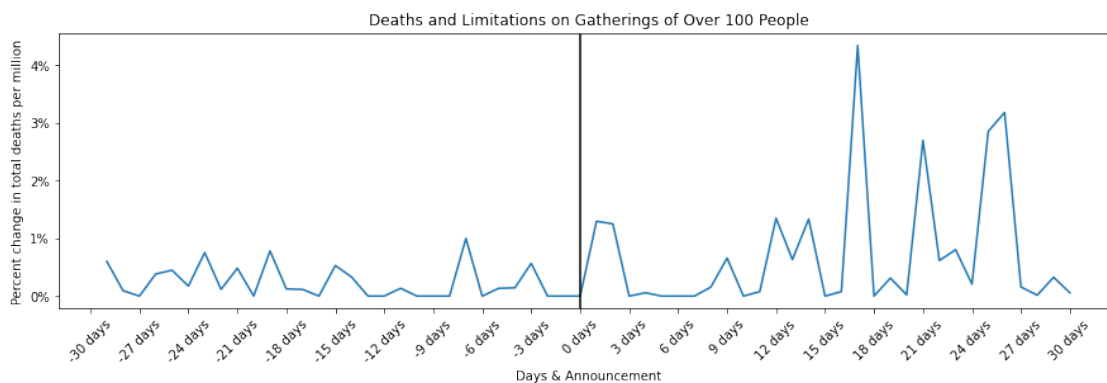
```
[114]: ev_name = 'IndoorOver100'
title = 'Limitations on Gatherings of Over 100 People'

plt.figure(figsize = (15,4))
md = compare_measures(response,by_data = death_data, m = ev_name)
y = md.ffill().quantile(0.5,axis = 1)
y.plot()
```

```

plt.axvline(x = 0 , c = 'black')
plt.title("Deaths and "+ title)
plt.ylabel('Percent change in total deaths per million')
plt.xlabel("Days & Announcement")
plt.xticks(ticks = np.arange(-dd,du+3,3),
           labels = list(map(lambda x: str(x)+ ' days',np.arange(-dd,du+3,3))),
           rotation=45)
plt.gca().set_yticklabels(['{:0.0f}%'.format(x) for x in plt.gca().get_yticks()])
#plt.legend(['10% quantile',
#           #'50% quantile',
#           #'90% quantile'])
plt.show()

```



```

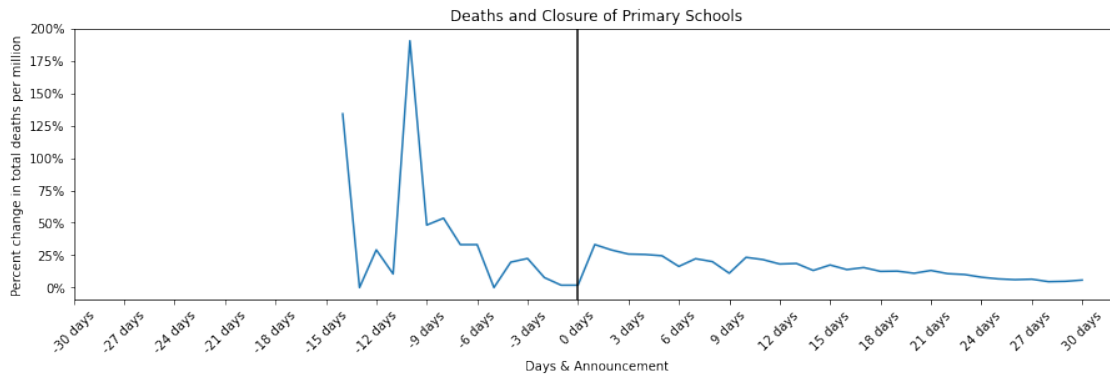
[115]: ev_name = 'ClosPrim'
title = 'Closure of Primary Schools'

plt.figure(figsize = (15,4))
md = compare_measures(response,by_data = death_data, m = ev_name)
y = md.ffill().quantile(0.5,axis = 1)
y.plot()

plt.axvline(x = 0 , c = 'black')
plt.title("Deaths and "+ title)
plt.ylabel('Percent change in total deaths per million')
plt.xlabel("Days & Announcement")
plt.xticks(ticks = np.arange(-dd,du+3,3),
           labels = list(map(lambda x: str(x)+ ' days',np.arange(-dd,du+3,3))),
           rotation=45)
plt.gca().set_yticklabels(['{:0.0f}%'.format(x) for x in plt.gca().get_yticks()])
#plt.legend(['10% quantile',
#           #'50% quantile',
#           #'90% quantile'])

```

```
plt.show()
```

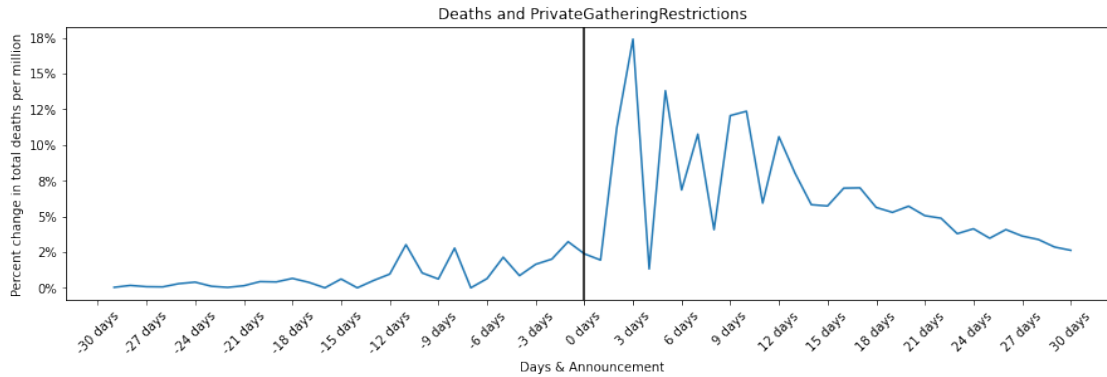


```
[117]: ev_name = 'PrivateGatheringRestrictions'
title = 'PrivateGatheringRestrictions'

plt.figure(figsize = (15,4))
md = compare_measures(response,by_data = death_data, m = ev_name)
y = md.ffill().quantile(0.5,axis = 1)
y.plot()

plt.axvline(x = 0 , c = 'black')
plt.title("Deaths and "+ title)
plt.ylabel('Percent change in total deaths per million')
plt.xlabel("Days & Announcement")
plt.xticks(ticks = np.arange(-dd,du+3,3),
           labels = list(map(lambda x: str(x)+ ' days',np.arange(-dd,du+3,3))),
           rotation=45)
plt.gca().set_yticklabels(['{:0f}%'.format(x) for x in plt.gca().get_yticks()])
#plt.legend(['10% quantile',
#           #'50% quantile',
#           #'90% quantile'])

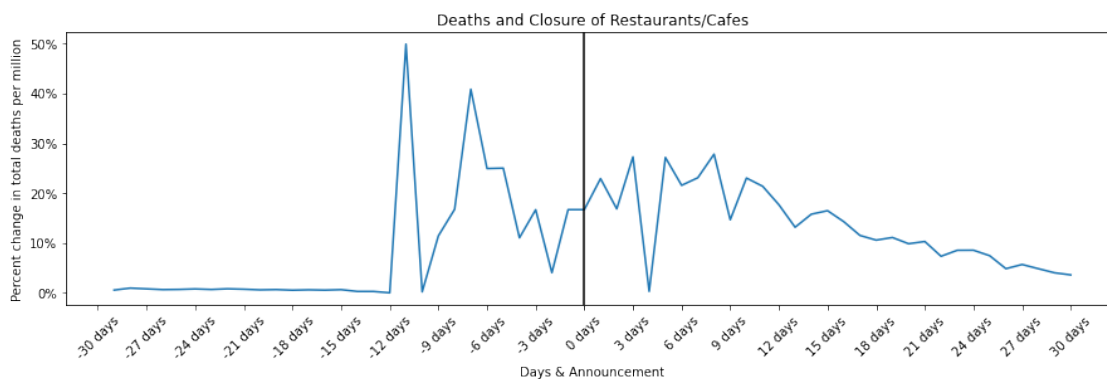
plt.show()
```



```
[119]: ev_name = 'RestaurantsCafes'
title = 'Closure of Restaurants/Cafes'

plt.figure(figsize = (15,4))
md = compare_measures(response,by_data = death_data, m = ev_name)
y = md.ffill().quantile(0.5,axis = 1)
y.plot()

plt.axvline(x = 0 , c = 'black')
plt.title("Deaths and "+ title)
plt.ylabel('Percent change in total deaths per million')
plt.xlabel("Days & Announcement")
plt.xticks(ticks = np.arange(-dd,du+3,3),
           labels = list(map(lambda x: str(x)+ ' days',np.arange(-dd,du+3,3))),
           rotation=45)
plt.gca().set_yticklabels(['{:0.0f}%'.format(x) for x in plt.gca().get_yticks()])
#plt.legend(['10% quantile',
#           #'50% quantile',
#           #'90% quantile'])
plt.show()
```





## 0.1 Clusters & Measures

```
[106]: response['days_between'] = list(map(lambda x: x.
      ↪days, (response['date_end']-response['date_start'])))

[107]: cluster_1_response = response[response['Country'].isin(c1)]
      cluster_2_response = response[response['Country'].isin(c2)]

[108]: analysis = (pd.DataFrame(cluster_1_response.Response_measure.value_counts()/
      ↪len(c1)).
      join(cluster_2_response.Response_measure.value_counts()/len(c2), rsuffix = '2'))
      analysis.columns = ['Cluster 1 av measure', 'Cluster 2 av measure']

      analysis = analysis.join(cluster_1_response.groupby('Response_measure').
      ↪days_between.mean())
      analysis = analysis.join(cluster_2_response.groupby('Response_measure').
      ↪days_between.mean(), rsuffix = '2')
      analysis.columns = ['Cluster 1 av measure', 'Cluster 2 av measure', 'Cluster 1 av_
      ↪days', 'Cluster 2 av days']

[109]: analysis = analysis.sort_values(by = 'Cluster 2 av measure', ascending = False)
      analysis['diff_measure'] = analysis['Cluster 2 av measure'] - analysis['Cluster_
      ↪1 av measure']
      analysis['diff_days'] = analysis['Cluster 2 av days'] - analysis['Cluster 1 av_
      ↪days']

[110]: display(analysis.sort_values(by = 'diff_measure', ascending = False)[:10])
      display(analysis.sort_values(by = 'diff_measure', ascending = False)[-10:])
```

	Cluster 1 av measure	Cluster 2 av measure \
IndoorOver100	0.125	0.578947
OutdoorOver500	0.125	0.552632
IndoorOver50	0.250	0.526316
ClosPrim	0.625	0.894737
MassGatherAllPartial	0.375	0.578947
ClosHigh	0.750	0.921053
StayHomeGen	0.250	0.421053
MassGather50Partial	0.250	0.368421
BanOnAllEventsPartial	0.125	0.236842
PrivateGatheringRestrictions	0.875	0.973684

	Cluster 1 av days	Cluster 2 av days \
IndoorOver100	19.000000	46.100000

OutdoorOver500	123.000000	44.578947
IndoorOver50	74.000000	53.933333
ClosPrim	98.600000	82.615385
MassGatherAllPartial	66.333333	81.312500
ClosHigh	116.333333	89.000000
StayHomeGen	8.000000	40.785714
MassGather50Partial	26.500000	57.400000
BanOnAllEventsPartial	23.000000	55.833333
PrivateGatheringRestrictions	98.333333	56.550000

	diff_measure	diff_days
IndoorOver100	0.453947	27.100000
OutdoorOver500	0.427632	-78.421053
IndoorOver50	0.276316	-20.066667
ClosPrim	0.269737	-15.984615
MassGatherAllPartial	0.203947	14.979167
ClosHigh	0.171053	-27.333333
StayHomeGen	0.171053	32.785714
MassGather50Partial	0.118421	30.900000
BanOnAllEventsPartial	0.111842	32.833333
PrivateGatheringRestrictions	0.098684	-41.783333

	Cluster 1 av measure	Cluster 2 av measure \
MasksMandatoryClosedSpaces	0.625	0.421053
OutdoorOver1000	0.500	0.263158
MasksVoluntaryClosedSpaces	0.375	0.131579
BanOnAllEvents	0.875	0.552632
NonEssentialShopsPartial	0.875	0.526316
SocialCircle	0.500	0.078947
StayHomeOrder	0.875	0.394737
Teleworking	1.000	0.447368
StayHomeOrderPartial	0.875	0.289474
RegionalStayHomeOrderPartial	0.250	NaN

	Cluster 1 av days	Cluster 2 av days \
MasksMandatoryClosedSpaces	24.000000	119.333333
OutdoorOver1000	95.000000	73.000000
MasksVoluntaryClosedSpaces	34.666667	81.500000
BanOnAllEvents	78.500000	78.818182
NonEssentialShopsPartial	26.200000	51.625000
SocialCircle	NaN	NaN
StayHomeOrder	47.428571	42.250000
Teleworking	70.250000	72.875000
StayHomeOrderPartial	27.666667	16.000000
RegionalStayHomeOrderPartial	26.500000	NaN

	diff_measure	diff_days
MasksMandatoryClosedSpaces	-0.203947	95.333333

OutdoorOver1000	-0.236842	-22.000000
MasksVoluntaryClosedSpaces	-0.243421	46.833333
BanOnAllEvents	-0.322368	0.318182
NonEssentialShopsPartial	-0.348684	25.425000
SocialCircle	-0.421053	NaN
StayHomeOrder	-0.480263	-5.178571
Teleworking	-0.552632	2.625000
StayHomeOrderPartial	-0.585526	-11.666667
RegionalStayHomeOrderPartial	NaN	NaN

```
[111]: display(analysis.sort_values(by = 'diff_days',ascending = False)[:10])
display(analysis.sort_values(by = 'diff_days',ascending = False)[-10:])
```

	Cluster 1 av measure	Cluster 2 av measure \
MasksMandatoryClosedSpaces	0.625	0.421053
IndoorOver1000	0.250	0.263158
ClosureOfPublicTransportPartial	0.125	0.157895
MasksVoluntaryClosedSpaces	0.375	0.131579
BanOnAllEventsPartial	0.125	0.236842
StayHomeGen	0.250	0.421053
MassGather50Partial	0.250	0.368421
RegionalStayHomeOrder	0.125	0.052632
IndoorOver100	0.125	0.578947
MassGatherAll	1.000	0.947368

	Cluster 1 av days	Cluster 2 av days \
MasksMandatoryClosedSpaces	24.000000	119.333333
IndoorOver1000	7.000000	65.800000
ClosureOfPublicTransportPartial	28.000000	78.000000
MasksVoluntaryClosedSpaces	34.666667	81.500000
BanOnAllEventsPartial	23.000000	55.833333
StayHomeGen	8.000000	40.785714
MassGather50Partial	26.500000	57.400000
RegionalStayHomeOrder	16.000000	46.000000
IndoorOver100	19.000000	46.100000
MassGatherAll	53.000000	79.350000

	diff_measure	diff_days
MasksMandatoryClosedSpaces	-0.203947	95.333333
IndoorOver1000	0.013158	58.800000
ClosureOfPublicTransportPartial	0.032895	50.000000
MasksVoluntaryClosedSpaces	-0.243421	46.833333
BanOnAllEventsPartial	0.111842	32.833333
StayHomeGen	0.171053	32.785714
MassGather50Partial	0.118421	30.900000
RegionalStayHomeOrder	-0.072368	30.000000
IndoorOver100	0.453947	27.100000
MassGatherAll	-0.052632	26.350000

	Cluster 1 av measure	Cluster 2 av measure \
ClosPubAnyPartial	1.000	1.078947
PlaceOfWorshipPartial	0.500	0.394737
OutdoorOver500	0.125	0.552632
StayHomeGenPartial	0.125	0.131579
TeleworkingPartial	0.375	0.263158
AdaptationOfWorkplacePartial	0.250	0.105263
MasksVoluntaryAllSpacesPartial	0.125	0.105263
SocialCircle	0.500	0.078947
StayHomeRiskGPartial	0.125	0.026316
RegionalStayHomeOrderPartial	0.250	NaN

	Cluster 1 av days	Cluster 2 av days \
ClosPubAnyPartial	141.0	71.875000
PlaceOfWorshipPartial	119.0	46.750000
OutdoorOver500	123.0	44.578947
StayHomeGenPartial	112.0	28.750000
TeleworkingPartial	151.0	16.000000
AdaptationOfWorkplacePartial	115.0	NaN
MasksVoluntaryAllSpacesPartial	NaN	70.666667
SocialCircle	NaN	NaN
StayHomeRiskGPartial	26.0	NaN
RegionalStayHomeOrderPartial	26.5	NaN

	diff_measure	diff_days
ClosPubAnyPartial	0.078947	-69.125000
PlaceOfWorshipPartial	-0.105263	-72.250000
OutdoorOver500	0.427632	-78.421053
StayHomeGenPartial	0.006579	-83.250000
TeleworkingPartial	-0.111842	-135.000000
AdaptationOfWorkplacePartial	-0.144737	NaN
MasksVoluntaryAllSpacesPartial	-0.019737	NaN
SocialCircle	-0.421053	NaN
StayHomeRiskGPartial	-0.098684	NaN
RegionalStayHomeOrderPartial	NaN	NaN

```
[120]: ev_name = 'MasksMandatoryClosedSpaces'
title = 'Mandatory Masks'
dd = 30
du = 30

plt.figure(figsize = (15,4))
md = compare_measures(response,by_data = death_data, m = ev_name)
y = md.ffill().quantile(0.5,axis = 1)
y.plot()
```

```

plt.axvline(x = 0 , c = 'black')
plt.title("Deaths and "+ title)
plt.ylabel('Percent change in total deaths per million')
plt.xlabel("Days & Announcement")
plt.xticks(ticks = np.arange(-dd,du+3,3),
           labels = list(map(lambda x: str(x)+ ' days',np.arange(-dd,du+3,3))),
           rotation=45)
plt.gca().set_yticklabels(['{:0.0f}%'.format(x) for x in plt.gca().get_yticks()])
#plt.legend(['10% quantile',
#           #'50% quantile',
#           #'90% quantile'])
plt.show()

```

