In [1]:

```
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.colors as mcolors
import pandas as pd
import random
import math
import time
```

In [2]:

```
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import RandomizedSearchCV, train_test_split
from sklearn.preprocessing import PolynomialFeatures
from sklearn.svm import SVR
from sklearn.metrics import mean_squared_error, mean_absolute_error,r2_score
import datetime
plt.style.use('ggplot')
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
```

In [3]:

```
import os
os.chdir('C:\\Users\\Dell\\Documents\\Assignments_Placement\\1.Vumonic\\datasets')
```

Loading all datasets

In [4]:

```
recoveries_df = pd.read_csv("1.time_series_covid19_recovered_global.csv")
deaths_df = pd.read_csv("2.time_series_covid19_deaths_global.csv")
confirmed_df = pd.read_csv("3.time_series_covid19_confirmed_global.csv")
latest_data = pd.read_csv("5.csse_daily.csv")
medical_data =pd.read_csv("07-12-2020.csv")
```

In [5]:

latest_data.head()

Out[5]:

	FIPS	Admin2	Province_State	Country_Region	Last_Update	Lat	Long_	Cı
_	0 45001.0	Abbeville	South Carolina	US	2020-07-13 04:43:04	34.223334	-82.461707	
,	1 22001.0	Acadia	Louisiana	US	2020-07-13 04:43:04	30.295065	-92.414197	
:	2 51001.0	Accomack	Virginia	US	2020-07-13 04:43:04	37.767072	-75.632346	
;	3 16001.0	Ada	Idaho	US	2020-07-13 04:43:04	43.452658	-116.241552	
	4 19001.0	Adair	lowa	US	2020-07-13 04:43:04	41.330756	-94.471059	

→

In [6]:

confirmed_df.head()

Out[6]:

	Province/State	Country/Region	Lat	Long	1/22/20	1/23/20	1/24/20	1/25/20	1/26/
0	NaN	Afghanistan	33.93911	67.709953	0	0	0	0	
1	NaN	Albania	41.15330	20.168300	0	0	0	0	
2	NaN	Algeria	28.03390	1.659600	0	0	0	0	
3	NaN	Andorra	42.50630	1.521800	0	0	0	0	
4	NaN	Angola	-11.20270	17.873900	0	0	0	0	

5 rows × 178 columns

In [7]:

```
medical_data.head()
```

Out[7]:

	Province_State	Country_Region	Last_Update	Lat	Long_	Confirmed	Deaths	Reco
0	Alabama	US	2020-07-13 04:43:16	32.3182	-86.9023	53587	1121	25
1	Alaska	US	2020-07-13 04:43:16	61.3707	-152.4044	1479	17	
2	American Samoa	US	2020-07-13 04:43:16	-14.2710	-170.1320	0	0	
3	Arizona	US	2020-07-13 04:43:16	33.7298	-111.4312	122467	2237	14
4	Arkansas	US	2020-07-13 04:43:16	34.9697	-92.3731	28367	321	21

```
→
```

In [8]:

```
cols = confirmed_df.keys()
```

Loading only dates in dataframes

In [9]:

```
confirmed = confirmed_df.loc[:, cols[4]:]
deaths = deaths_df.loc[:, cols[4]:]
recoveries = recoveries_df.loc[:, cols[4]:]
```

In [10]:

```
dates = confirmed.keys()
world_cases = []
total_deaths = []
mortality_rate = []
recovery_rate = []
total_recovered = []
total_active = []
```

Get all cases hike from dates for the outbreak

In [11]:

```
for i in dates:
    confirmed_sum = confirmed[i].sum()
    death_sum = deaths[i].sum()
    recovered_sum = recoveries[i].sum()

# confirmed, deaths, recovered, and active
    world_cases.append(confirmed_sum)
    total_deaths.append(death_sum)
    total_recovered.append(recovered_sum)
    total_active.append(confirmed_sum-death_sum-recovered_sum)

# calculate rates
    mortality_rate.append(death_sum/confirmed_sum)
    recovery_rate.append(recovered_sum/confirmed_sum)
```

Daily count of cases

```
In [12]:
```

Moving Average of cases

```
In [13]:
```

```
def moving_average(data, step_size):
    moving_average = []
    for i in range(len(data)):
        if i + step_size < len(data):
            moving_average.append(np.mean(data[i:i+ step_size]))
        else:
            moving_average.append(np.mean(data[i:len(data)]))
    return moving_average</pre>
```

```
In [14]:
```

```
# step size from week will be 7
step = 7

# confirmed cases
world_daily_increase = daily_increase(world_cases)
world_confirmed_avg= moving_average(world_cases, step)
world_daily_increase_avg = moving_average(world_daily_increase, step)
```

```
In [15]:
```

```
# deaths
world_daily_death = daily_increase(total_deaths)
world_death_avg = moving_average(total_deaths, step)
world_daily_death_avg = moving_average(world_daily_death, step)
```

In [16]:

```
# recoveries
world_daily_recovery = daily_increase(total_recovered)
world_recovery_avg = moving_average(total_recovered, step)
world_daily_recovery_avg = moving_average(world_daily_recovery, step)

# active
world_active_avg = moving_average(total_active, step)
```

Reshaping values for ease of graphs

```
In [17]:
```

```
days_since_1_22 = np.array([i for i in range(len(dates))]).reshape(-1, 1)
world_cases = np.array(world_cases).reshape(-1, 1)
total_deaths = np.array(total_deaths).reshape(-1, 1)
total_recovered = np.array(total_recovered).reshape(-1, 1)
```

Creating Forcasting days

```
In [18]:
```

```
days_in_future = 10
future_forcast = np.array([i for i in range(len(dates)+days_in_future)]).reshape(-1, 1)
available_dates = future_forcast[:-10]
```

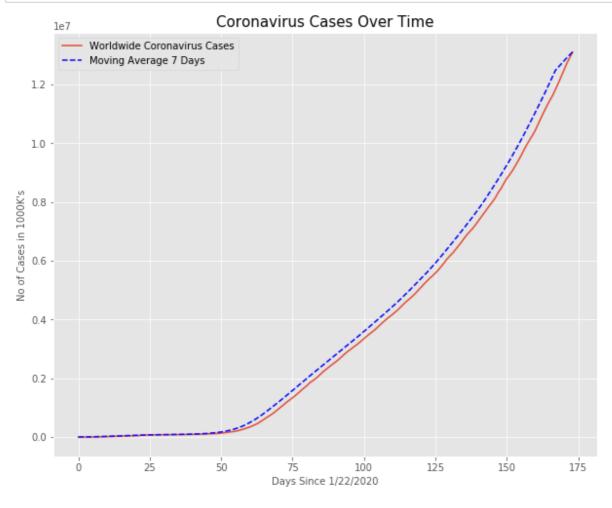
Graphing the number of Weekly Analysis of confirmed cases, active cases, deaths, recoveries

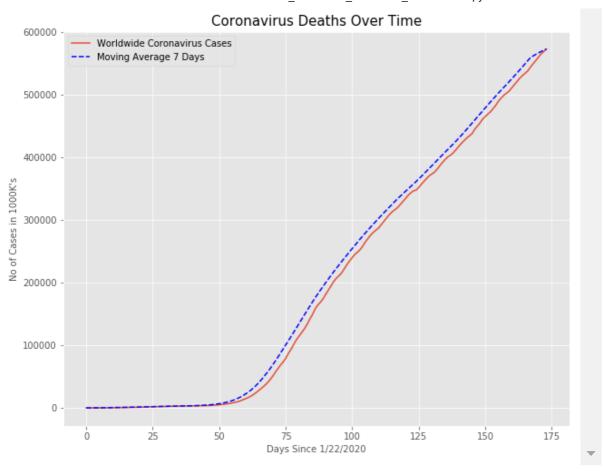
```
In [19]:
```

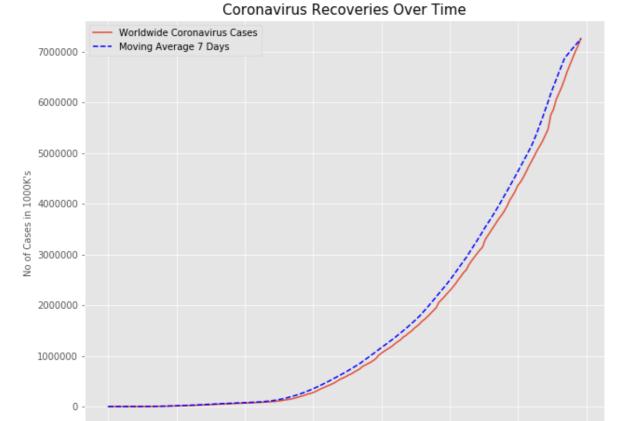
```
# Confirmed/Death/Recovery/Active Cases with moving avergae of it
def case_avg(days,case,avg,string):
    days = days.reshape(1, -1)[0]
    plt.figure(figsize=(10, 8))
    plt.plot(days, case)
    plt.plot(days, avg, '--b')
    plt.title('Coronavirus {} Over Time'.format(string), size=15)
    plt.xlabel('Days Since 1/22/2020', size=10)
    plt.ylabel("No of Cases in 1000K's", size=10)
    plt.legend(['Worldwide Coronavirus Cases', 'Moving Average {} Days'.format(step)], prop
    plt.show()
```

In [20]:

```
case_avg(available_dates, world_cases,world_confirmed_avg,'Cases')
case_avg(available_dates, total_deaths,world_death_avg,'Deaths')
case_avg(available_dates, total_recovered,world_recovery_avg,'Recoveries')
case_avg(available_dates, total_active,world_active_avg,'Active')
```

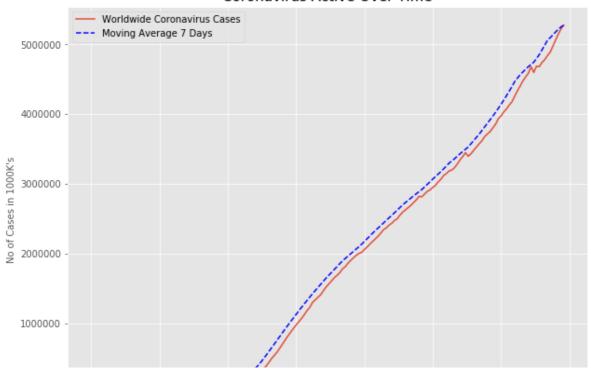






Days Since 1/22/2020

Coronavirus Active Over Time

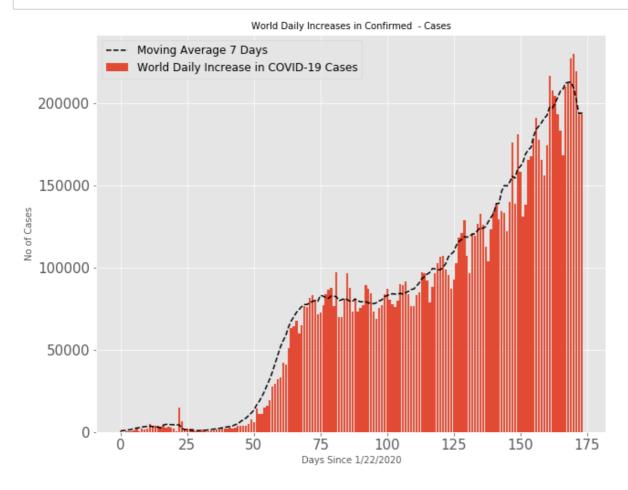


In [21]:

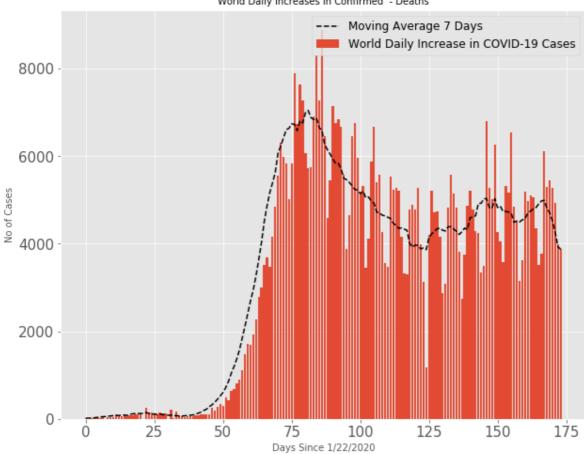
```
# Daily Analysis of Confirmed/Recovery/Death cases with moving average
def daily_graph(days,increase,increase_avg,case,string):
    days = days.reshape(1, -1)[0]
    plt.figure(figsize=(10, 8))
    plt.bar(days, increase)
    plt.plot(days, increase_avg, color='black', linestyle='dashed')
    plt.title('World Daily Increases in {} - {}'.format(case,string), size=10)
    plt.xlabel('Days Since 1/22/2020', size=10)
    plt.ylabel('No of Cases', size=10)
    plt.legend(['Moving Average {} Days'.format(step), 'World Daily Increase in COVID-19 Ca
    plt.xticks(size=15)
    plt.yticks(size=15)
    plt.show()
```

In [22]:

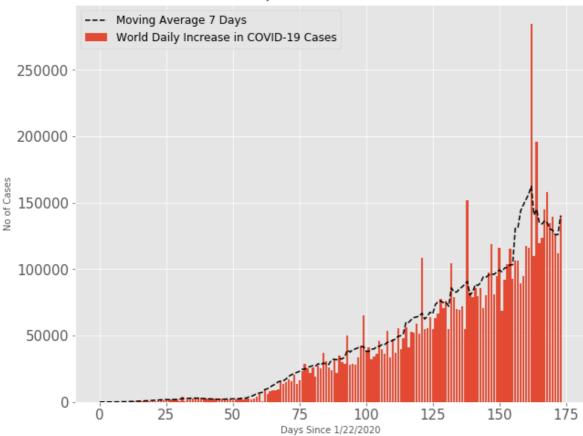
daily_graph(available_dates,world_daily_increase,world_daily_increase_avg,'Confirmed','Case daily_graph(available_dates,world_daily_death,world_daily_death_avg,'Confirmed','Deaths') daily_graph(available_dates,world_daily_recovery,world_daily_recovery_avg,'Confirmed','Reco



World Daily Increases in Confirmed - Deaths







Country wise analysis

In [23]:

```
def country_plot(available_dates, country_cases, country_daily_increase, country_daily_deat
#def country_plot(x, y1, y2, y3, y4, country):
    # step is set as 14 in in the beginning of the notebook
    confirmed_avg = moving_average(country_cases, step)
    confirmed_increase_avg = moving_average(country_daily_increase, step)
    death_increase_avg = moving_average(country_daily_death, step)
    recovery_increase_avg = moving_average(country_daily_recovery, step)
    case_avg(available_dates,country_cases,confirmed_avg,str(country))
    daily_graph(available_dates,country_daily_increase,confirmed_increase_avg,'cases',str(c
    daily_graph(available_dates,country_daily_death,death_increase_avg,'deaths',str(country
    daily_graph(available_dates,country_daily_recovery,recovery_increase_avg,'recovery',str
```

In [24]:

```
# helper function for getting country's cases, deaths, and recoveries

def get_country_info(country_name):
    country_cases = []
    country_deaths = []
    country_recoveries = []

for i in dates:
    country_cases.append(confirmed_df[confirmed_df['Country/Region']==country_name][i].
    country_deaths.append(deaths_df[deaths_df['Country/Region']==country_name][i].sum()
    country_recoveries.append(recoveries_df[recoveries_df['Country/Region']==country_name][i].sum()
    country_recoveries.append(recoveries_df[recoveries_df['Country/Region']==country_name][i].
```

In [25]:

```
def country_visualizations(country_name):
    country_info = get_country_info(country_name)
    country_cases = country_info[0]
    country_deaths = country_info[1]
    country_recoveries = country_info[2]

    country_daily_increase = daily_increase(country_cases)
    country_daily_death = daily_increase(country_deaths)
    country_daily_recovery = daily_increase(country_recoveries)

    country_plot(available_dates, country_cases, country_daily_increase, country_daily_deat
```

In [26]:

```
latest_data.head()
```

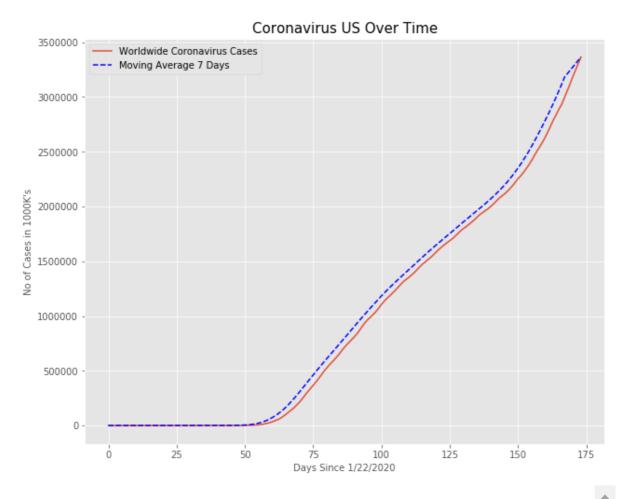
Out[26]:

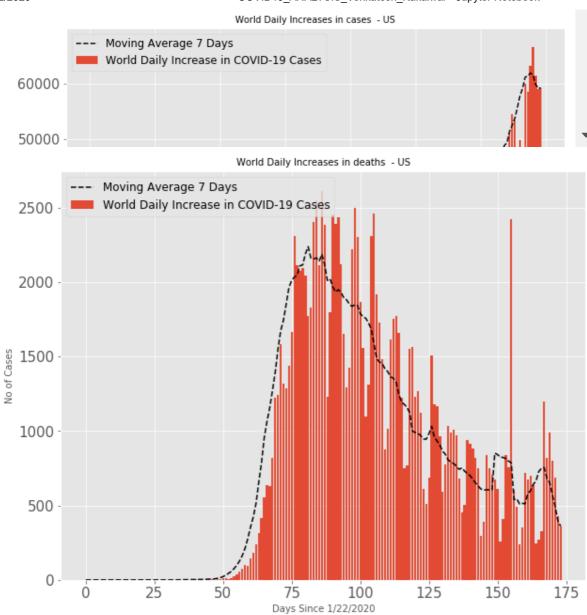
	FIPS	Admin2	Province_State	Country_Region	Last_Update	Lat	Long_	Cı
0	45001.0	Abbeville	South Carolina	US	2020-07-13 04:43:04	34.223334	-82.461707	
1	22001.0	Acadia	Louisiana	US	2020-07-13 04:43:04	30.295065	-92.414197	
2	51001.0	Accomack	Virginia	US	2020-07-13 04:43:04	37.767072	-75.632346	
3	16001.0	Ada	Idaho	US	2020-07-13 04:43:04	43.452658	-116.241552	
4	19001.0	Adair	lowa	US	2020-07-13 04:43:04	41.330756	-94.471059	
4								•

In [27]:

In [28]:

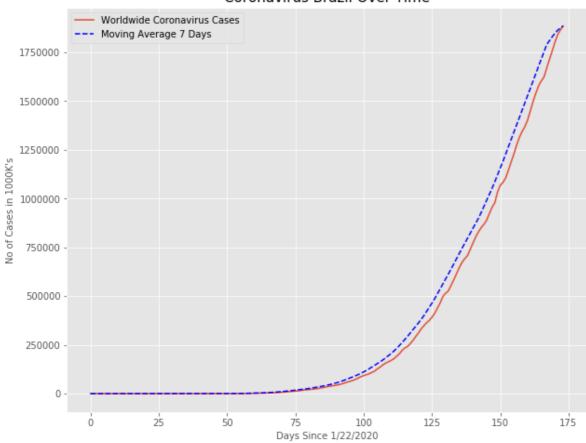
```
countries = latest_data.groupby('Country_Region').sum().sort_values(by = 'Confirmed', ascen
#countries = ['India']
for country in countries:
    print('-*-*'*10,end = '')
    print(country,end = '')
    print('-*-*'*10)
    country_visualizations(country)
```



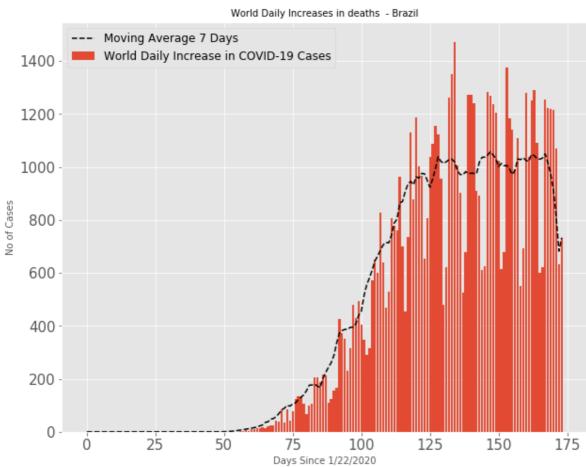


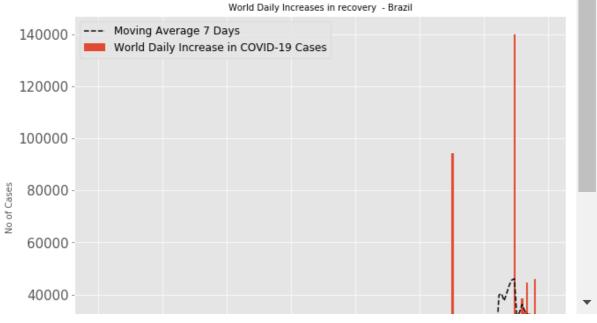


Coronavirus Brazil Over Time

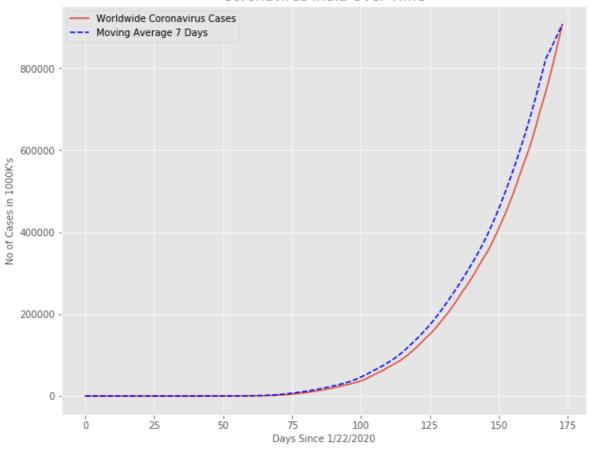


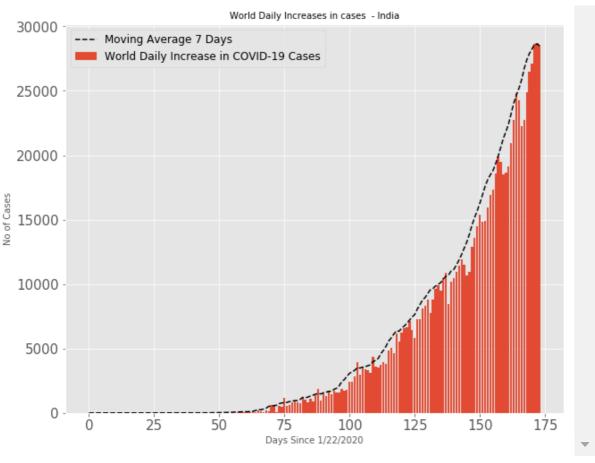


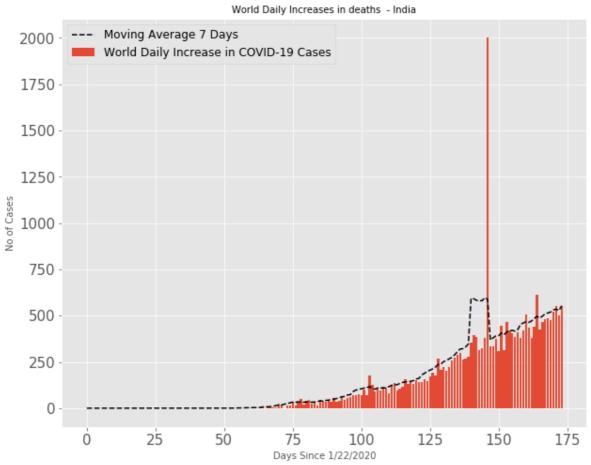


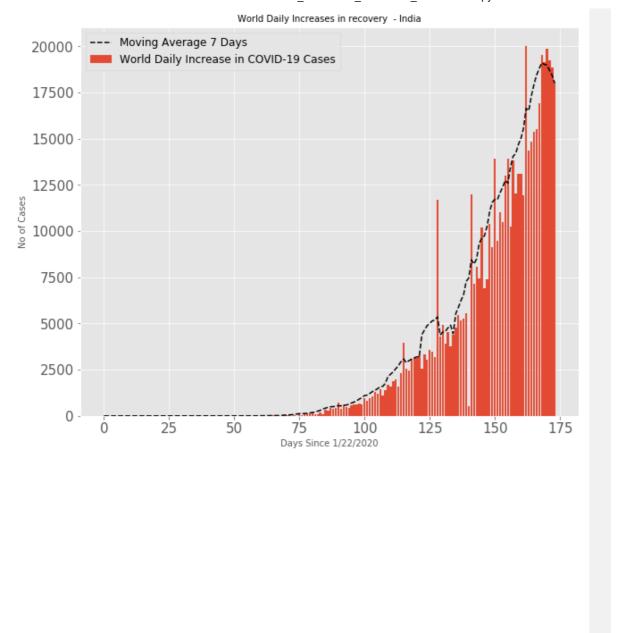


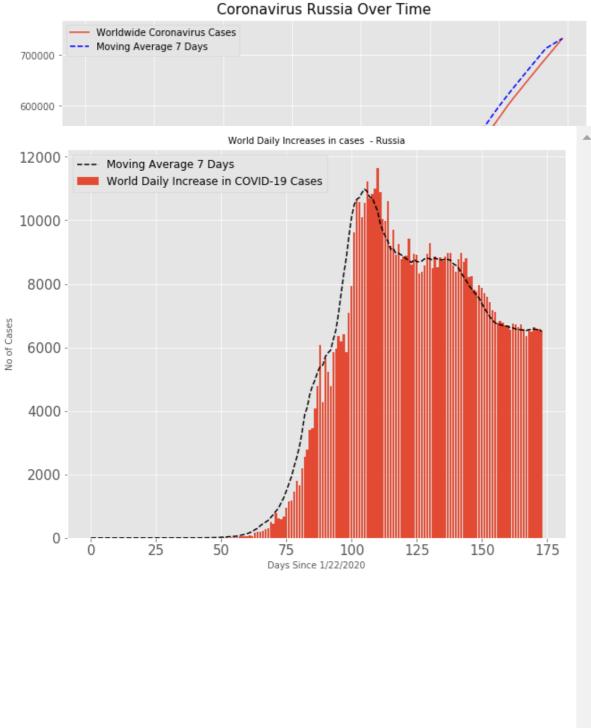
Coronavirus India Over Time



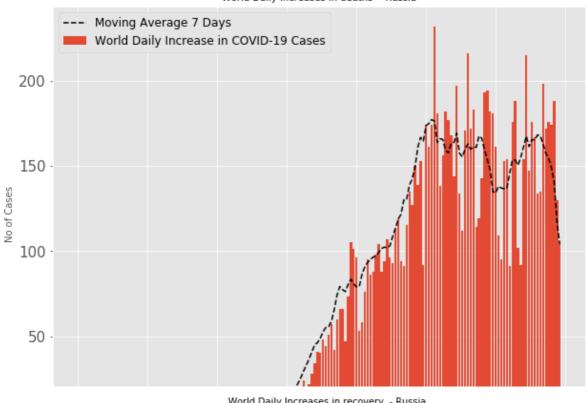




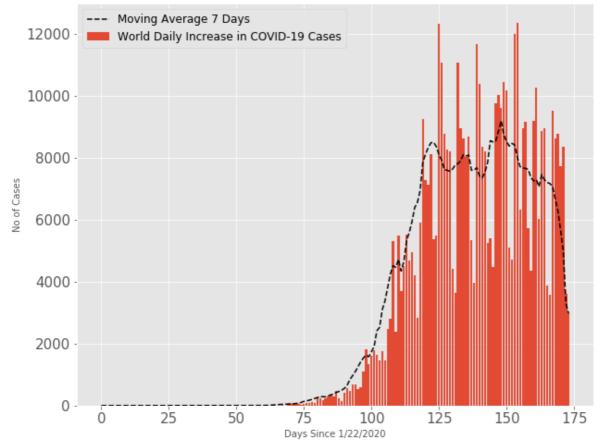


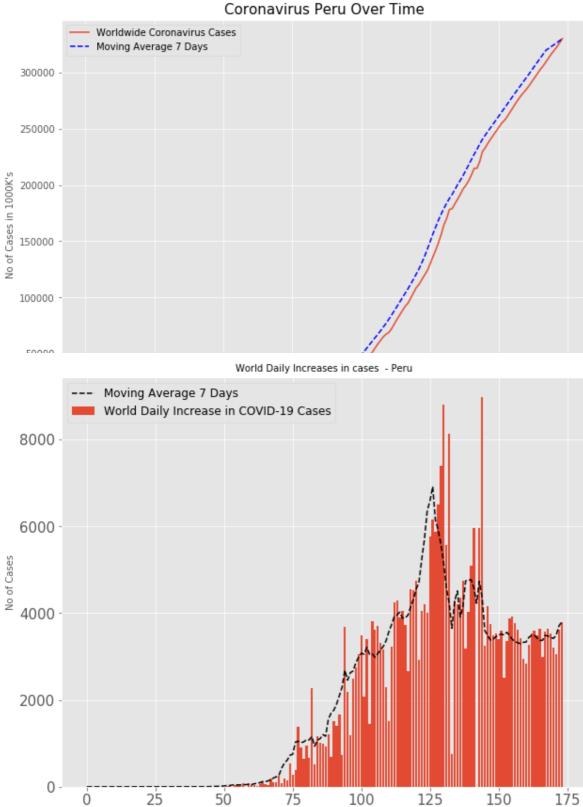


World Daily Increases in deaths - Russia



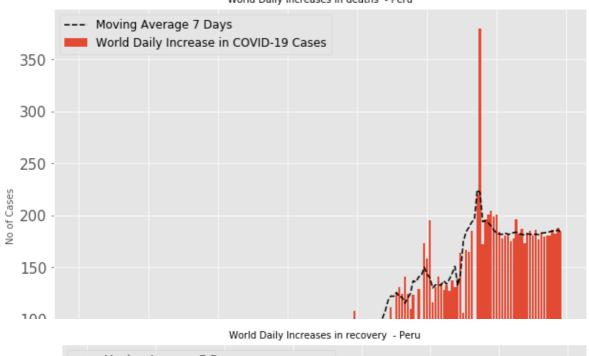


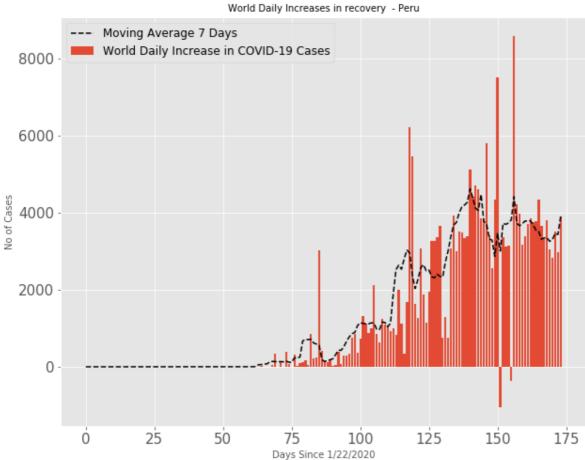


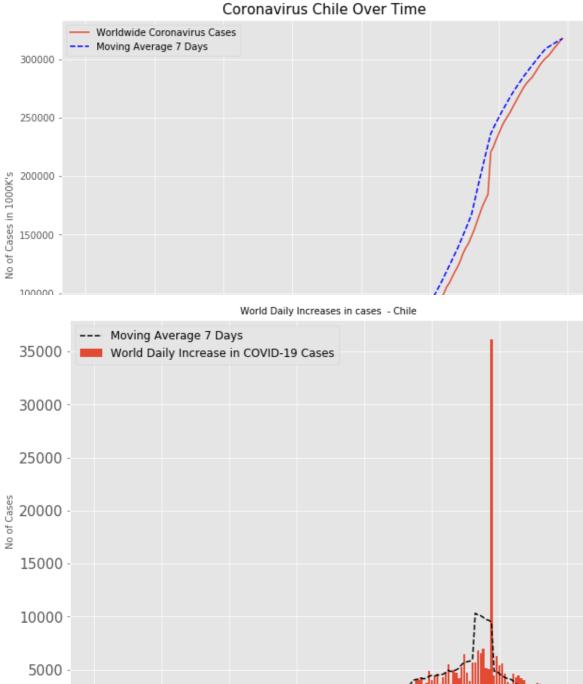


Days Since 1/22/2020

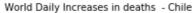
World Daily Increases in deaths - Peru

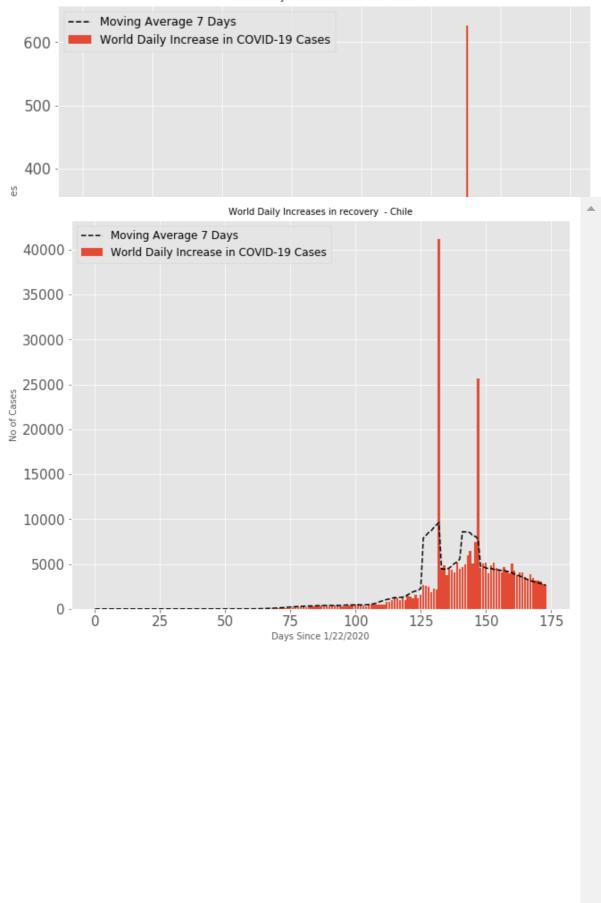






Days Since 1/22/2020

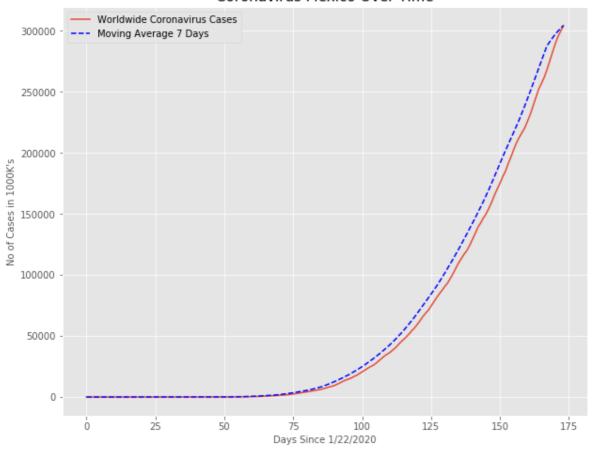


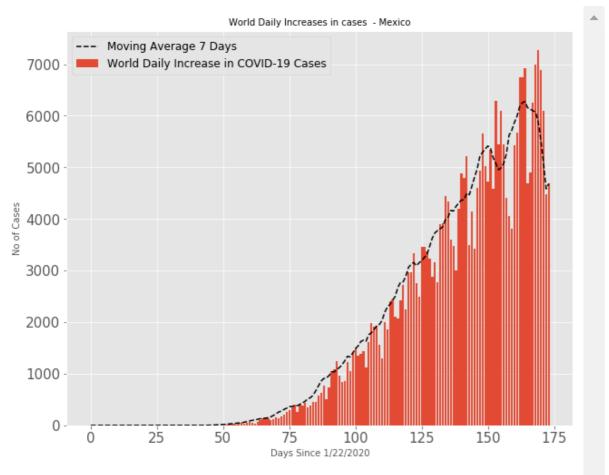


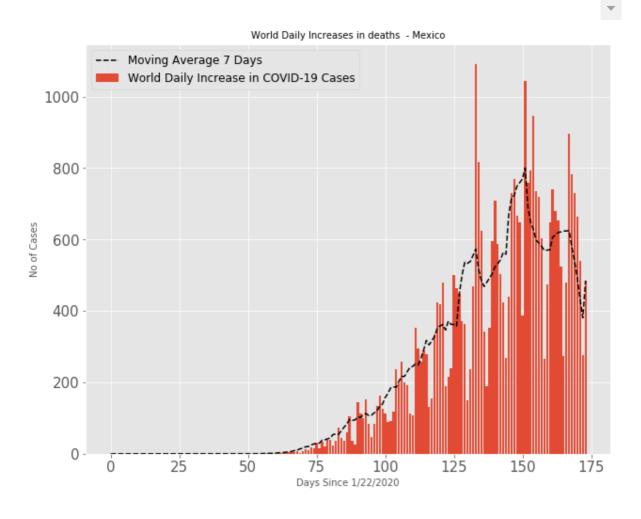
-*-*-*-*-*-*-*-*-*-*-*-*-*-*-*Mexico-*-*-*-*-*-*-*-*-*-*-*-*-*-

**_*_*

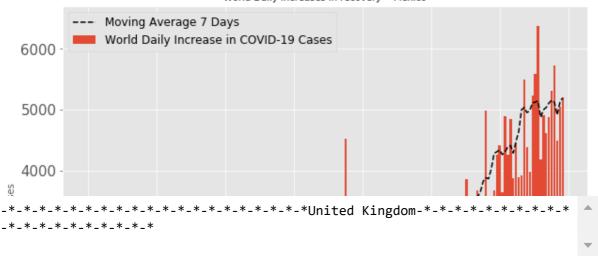
Coronavirus Mexico Over Time



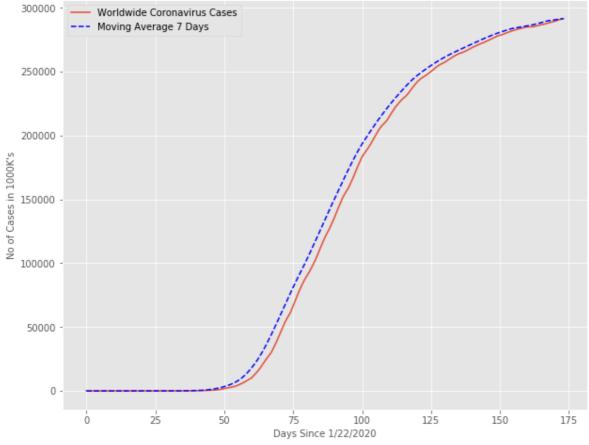


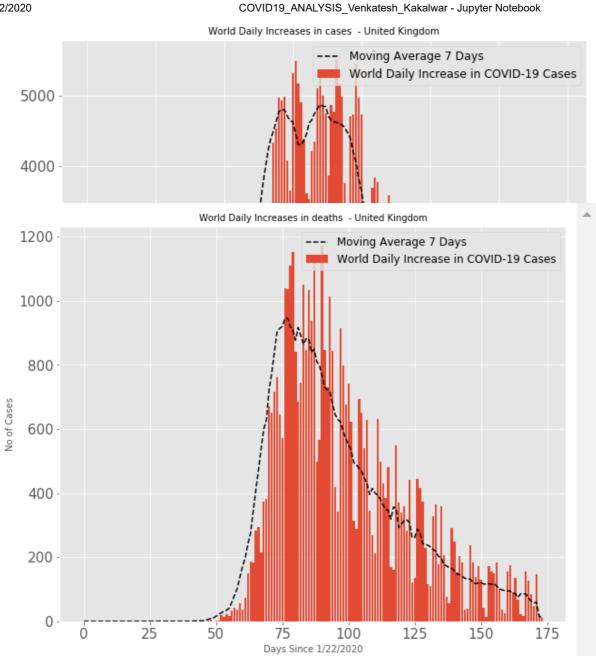


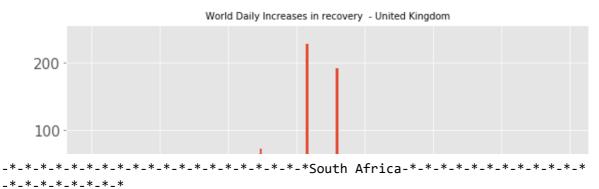
World Daily Increases in recovery - Mexico

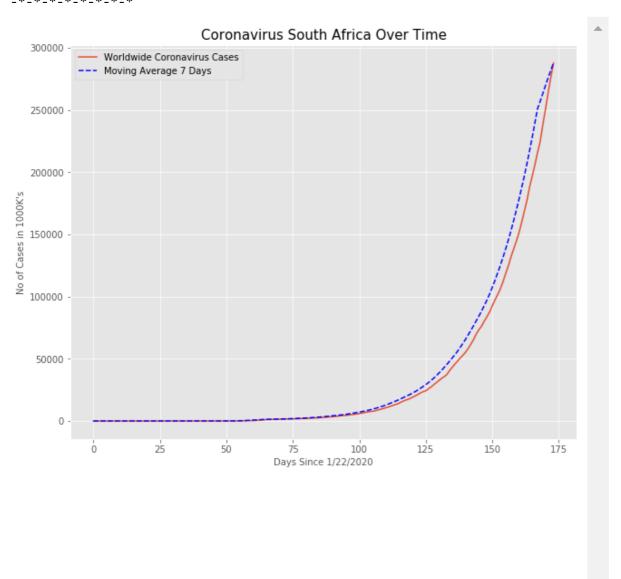


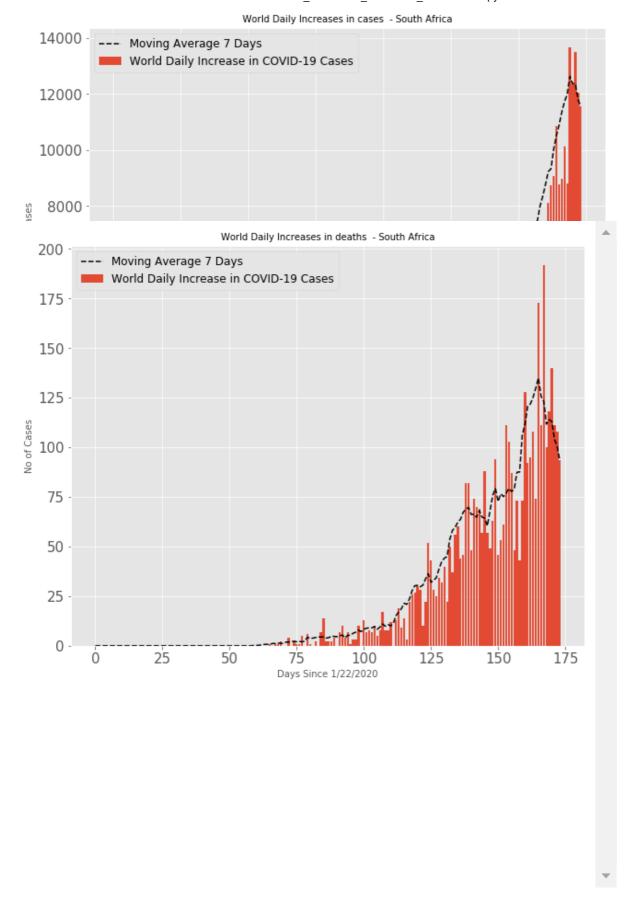




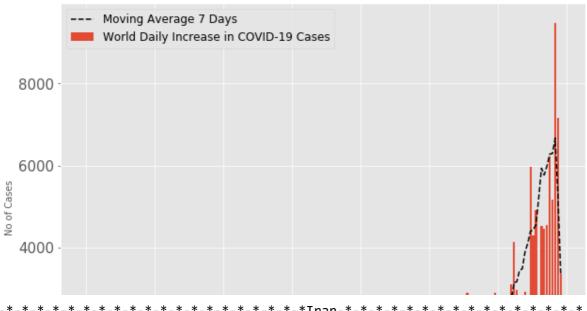




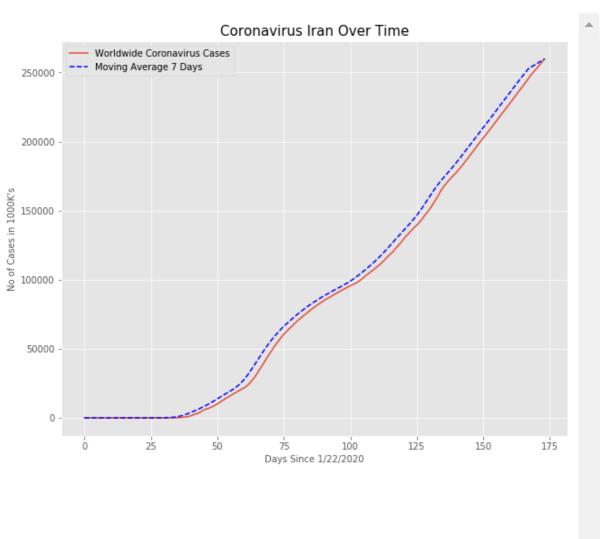




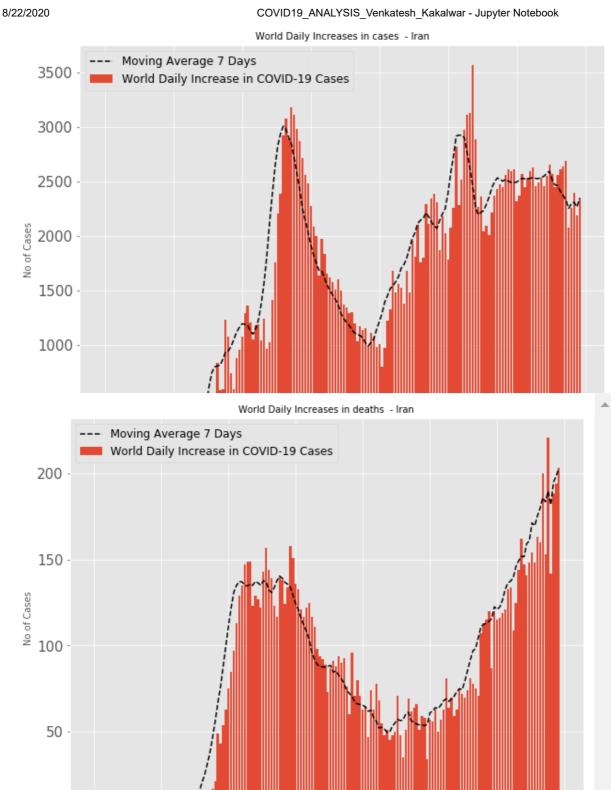
World Daily Increases in recovery - South Africa



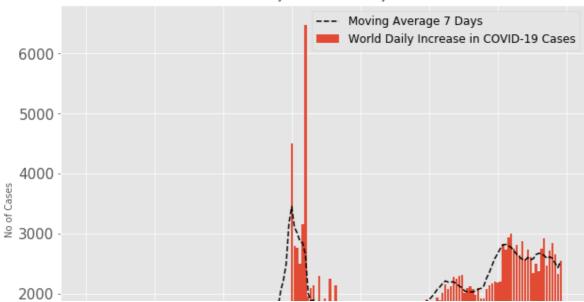




Days Since 1/22/2020



World Daily Increases in recovery - Iran

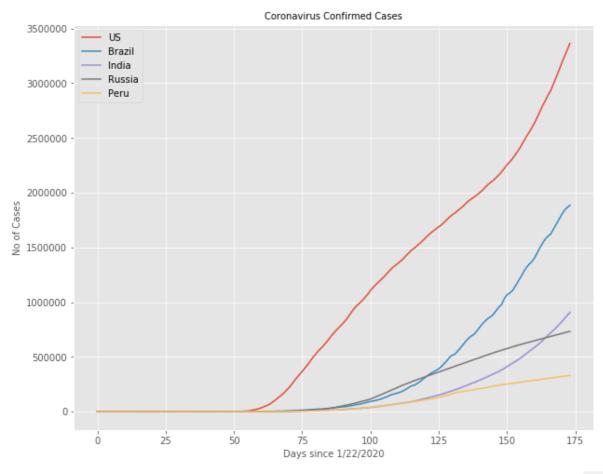


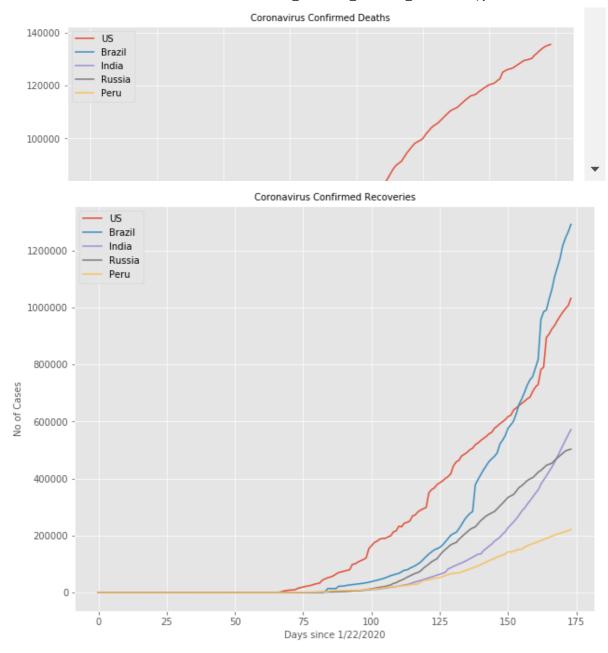
Comparison between top Countries

In [29]:

```
compare_countries = countries[:5]
graph_name = ['Coronavirus Confirmed Cases', 'Coronavirus Confirmed Deaths', 'Coronavirus C

for num in range(3):
    plt.figure(figsize=(10, 8))
    for country in compare_countries:
        plt.plot(get_country_info(country)[num])
    plt.legend(compare_countries, prop={'size': 10})
    plt.xlabel('Days since 1/22/2020', size=10)
    plt.ylabel('No of Cases', size=10)
    plt.title(graph_name[num], size=10)
    plt.xticks(size=10)
    plt.yticks(size=10)
    plt.show()
```





Analysis of latest data

In [30]:

```
latest_data.head()
```

Out[30]:

	FIPS	Admin2	Province_State	Country_Region	Last_Update	Lat	Long_	Cı
0	45001.0	Abbeville	South Carolina	US	2020-07-13 04:43:04	34.223334	-82.461707	
1	22001.0	Acadia	Louisiana	US	2020-07-13 04:43:04	30.295065	-92.414197	
2	51001.0	Accomack	Virginia	US	2020-07-13 04:43:04	37.767072	-75.632346	
3	16001.0	Ada	Idaho	US	2020-07-13 04:43:04	43.452658	-116.241552	
4	19001.0	Adair	lowa	US	2020-07-13 04:43:04	41.330756	-94.471059	
4								

Country data frame sorted by country with increase in cases

In [31]:

```
country_df = latest_data.groupby('Country_Region').sum().sort_values(by = 'Confirmed',ascen
```

In [32]:

```
country_df.reset_index(inplace =True)
```

In [33]:

country_df['Mortality_rate'] = country_df.Deaths/country_df.Confirmed

In [34]:

```
country_df.head(20).style.background_gradient(cmap='PuBu')
```

Out[34]:

	Country_Region	Confirmed	Deaths	Recovered	Active	Mortality_rate
0	US	3304942	135205	1006326	2208772.000000	0.040910
1	Brazil	1864681	72100	1264843	527738.000000	0.038666
2	India	878254	23174	553471	301609.000000	0.026386
3	Russia	726036	11318	500208	214510.000000	0.015589
4	Peru	326326	11870	217111	97345.000000	0.036375
5	Chile	315041	6979	283902	24160.000000	0.022153
6	Mexico	299750	35006	234905	29839.000000	0.116784
7	United Kingdom	291154	44904	1378	244872.000000	0.154228
8	South Africa	276242	4079	134874	137289.000000	0.014766
9	Iran	257303	12829	219993	24481.000000	0.049860
10	Spain	253908	28403	150376	75130.000000	0.111863
11	Pakistan	251625	5266	161917	84442.000000	0.020928
12	Italy	243061	34954	194928	13179.000000	0.143808
13	Saudi Arabia	232259	2223	167138	62898.000000	0.009571
14	Turkey	212993	5363	194515	13115.000000	0.025179
15	France	208015	30007	78513	99495.000000	0.144254
16	Germany	199919	9071	184414	6434.000000	0.045373
17	Bangladesh	183795	2352	93614	87829.000000	0.012797
18	Colombia	145362	5426	61186	78750.000000	0.037327
19	Canada	109348	8829	72954	27566.000000	0.080742

Province_State data frame sorted by Province_State with increase in cases

```
In [35]:
```

```
province_df = latest_data.groupby('Province_State').sum().sort_values(by = 'Confirmed',asce
province_df.reset_index(inplace =True)
```

In [36]:

```
province_df[province_df['Confirmed'] == 0].index
```

Out[36]:

Int64Index([547], dtype='int64')

In [37]:

```
province_df.drop(province_df[province_df['Confirmed'] == 0].index,inplace=True)
```

In [38]:

province_df.tail()

Out[38]:

	Province_State	Confirmed	Deaths	Recovered	Active
542	Northwest Territories	5	0	5	0.0
543	Anguilla	3	0	3	0.0
544	Saint Pierre and Miquelon	2	0	1	1.0
545	Tibet	1	0	1	0.0
546	Vichada	1	0	1	0.0

In [39]:

province_df['Mortality_rate'] = province_df.Deaths/province_df.Confirmed

In [40]:

province_df.head(20).style.background_gradient(cmap='PuBu')

Out[40]:

	Province_State	Confirmed	Deaths	Recovered	Active	Mortality_rate	
0	New York	401706	32350	0	369356.000000	0.080532	
1	Sao Paulo	371997	17848	220974	133175.000000	0.047979	
2	California	324543	7051	0	317492.000000	0.021726	
3	Florida	269811	4242	0	265569.000000	0.015722	
4	Texas	262762	3216	0	259546.000000	0.012239	
5	Maharashtra	254427	10289	140325	103813.000000	0.040440	
6	England	249510	40234	0	209276.000000	0.161252	
7	Metropolitana	238360	5814	219896	12650.000000	0.024392	
8	Moscow	229357	4143	164560	60654.000000	0.018064	
9	Lima	175631	5530	0	170101.000000	0.031486	
10	New Jersey	175298	15525	0	159773.000000	0.088563	
11	Illinois	155048	7388	0	147660.000000	0.047650	
12	Tamil Nadu	138470	1966	89532	46972.000000	0.014198	
13	Ceara	136785	6868	110224	19693.000000	0.050210	
14	Rio de Janeiro	129684	11415	110061	8208.000000	0.088022	
15	Para	125714	5289	112237	8188.000000	0.042072	
16	Arizona	122467	2237	0	120231.000000	0.018266	
17	Georgia	116935	3003	0	113932.000000	0.025681	
18	Delhi	112494	3371	89968	19155.000000	0.029966	
19	Massachusetts	111597	8325	0	103272.000000	0.074599	

COUNTRY wise Bar Chart Visualizations for COVID-19

In [41]:

```
# As already sorted with increasing cases we can directly take the cases
country_confirmed_cases = list(country_df['Confirmed'])
unique_countries = list(country_df['Country_Region'].unique())
```

In [42]:

```
def plot_bar_graphs(x, y, title):
    plt.figure(figsize=(15, 8))
    plt.barh(x, y, color='blue')
    plt.title(title, size=10)
    plt.xticks(size=10)
    plt.yticks(size=10)
    plt.show()
```

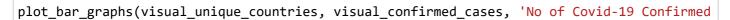
In [43]:

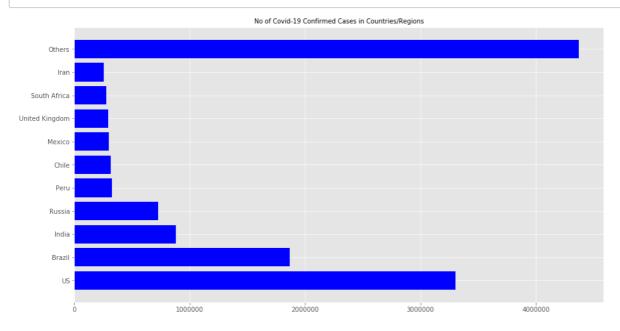
```
# Only show 15 countries with the most confirmed cases, the rest are grouped into the other
visual_unique_countries = []
visual_confirmed_cases = []
others = np.sum(country_confirmed_cases[10:])

for i in range(len(country_confirmed_cases[:10])):
    visual_unique_countries.append(unique_countries[i])
    visual_confirmed_cases.append(country_confirmed_cases[i])

visual_unique_countries.append('Others')
visual_confirmed_cases.append(others)
```

In [44]:





PROVINCE wise Bar Chart Visualizations for COVID-19

In [45]:

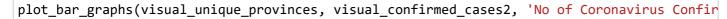
```
# As already sorted with increasing cases we can directly take the cases
province_confirmed_cases=list(province_df['Confirmed'])
unique_provinces = list(province_df.Province_State.unique())
```

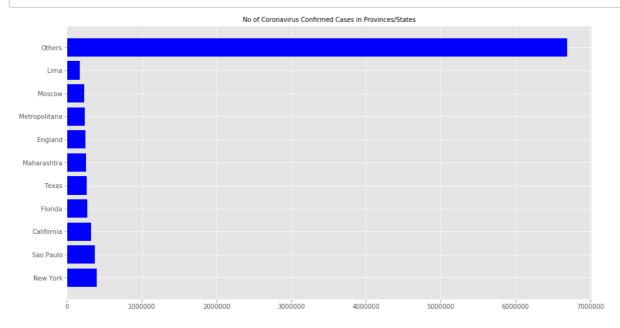
In [46]:

```
visual_unique_provinces = []
visual_confirmed_cases2 = []
others = np.sum(province_confirmed_cases[10:])
for i in range(len(province_confirmed_cases[:10])):
    visual_unique_provinces.append(unique_provinces[i])
    visual_confirmed_cases2.append(province_confirmed_cases[i])

visual_unique_provinces.append('Others')
visual_confirmed_cases2.append(others)
```

In [47]:





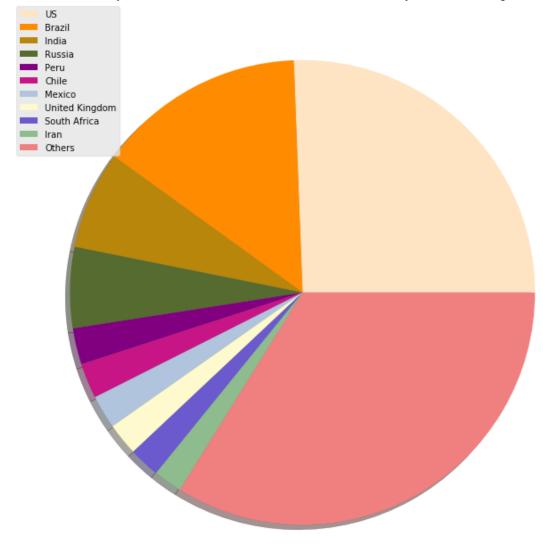
Pie Chart Visualizations for COVID-19

In [48]:

In [49]:

plot_pie_charts(visual_unique_countries, visual_confirmed_cases, 'Top 10 Covid-19 Confirmed

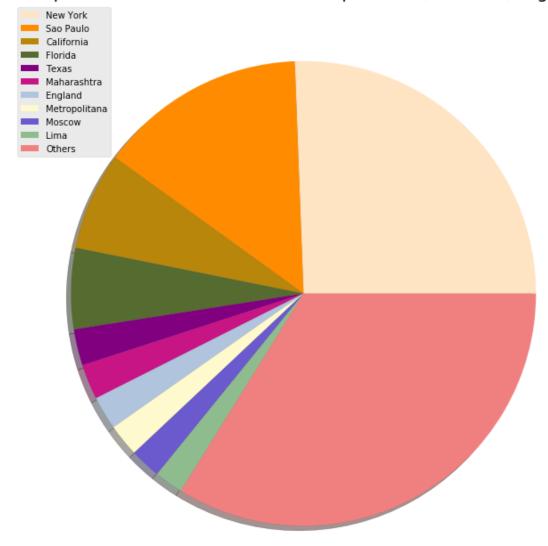




In [50]:

plot_pie_charts(visual_unique_provinces, visual_confirmed_cases, 'Top 10 Covid-19 Confirmed

Top 10 Covid-19 Confirmed Cases per State/Province/Region



In [51]:

```
# Plotting countries with regional data using a pie chart
def get_provinces_pie_chart(country_name):
   temp_df = latest_data[latest_data['Country_Region'] == country_name].groupby('Province_
   temp_df = temp_df.sort_values('Confirmed',ascending= False).reset_index()
   if temp_df.shape[0] > 5:
        province_name = list(temp_df.iloc[:5,0])
        province_count = list(temp_df.iloc[:5,4])
        province_name.append('others')
        province count.append(sum(list(temp df.iloc[5:,4])))
        title='Covid-19 Confirmed Cases in {}'.format(str(country_name))
        plot_pie_charts(province_name,province_count, title)
   else:
        province_name = list(temp_df.iloc[:,0])
        province_count = list(temp_df.iloc[:,4])
        title='Covid-19 Confirmed Cases in {}'.format(str(country_name))
        plot_pie_charts(province_name,province_count, title)
```

In [52]:

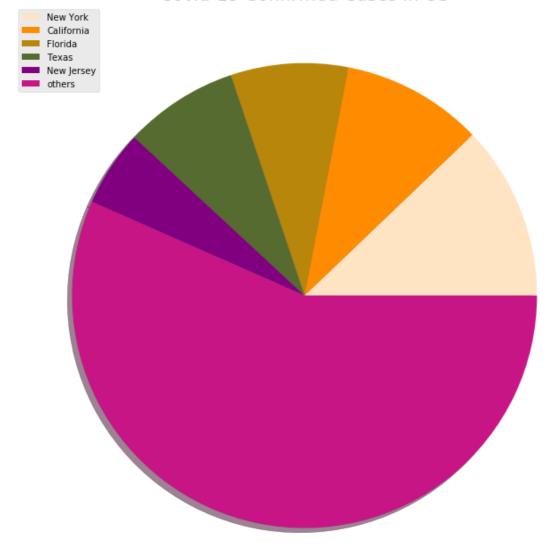
countries

Out[52]:

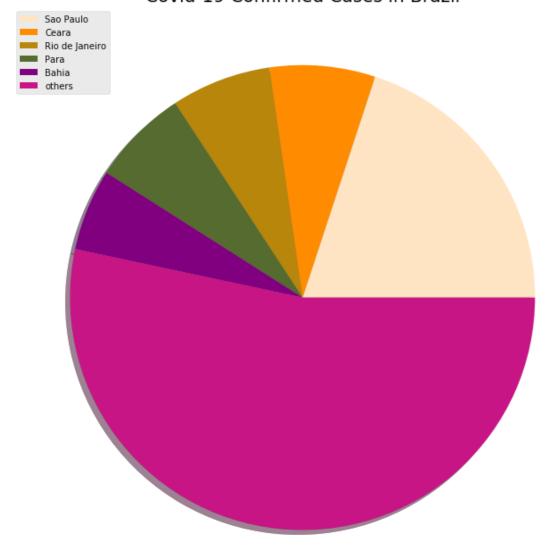
```
['US',
'Brazil',
'India',
'Russia',
'Peru',
'Chile',
'Mexico',
'United Kingdom',
'South Africa',
```

In [53]:

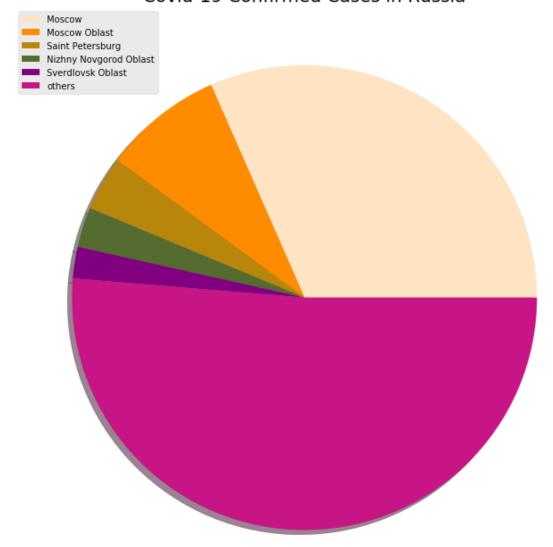
Covid-19 Confirmed Cases in US



Covid-19 Confirmed Cases in Brazil

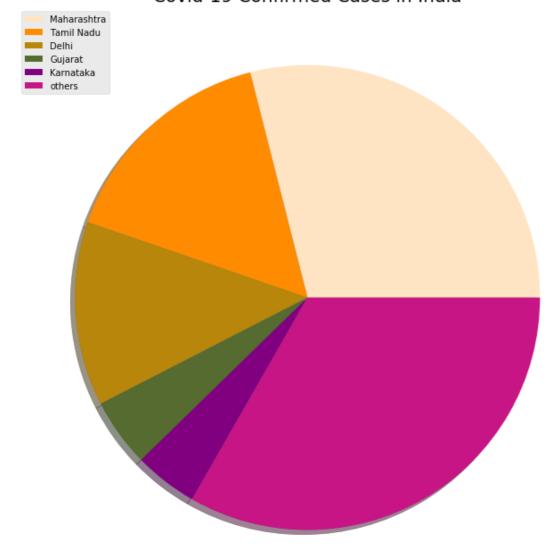


Covid-19 Confirmed Cases in Russia



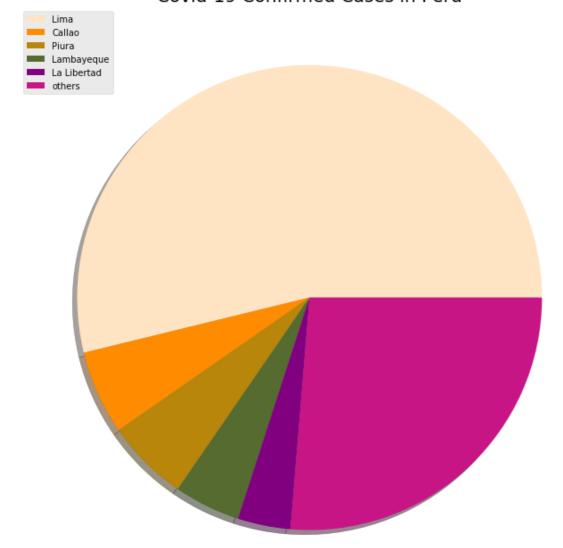
**_*_*_*_*_*_*_*_*_*_*_*_*_*_*_*India_*_*_*_*_*_*_*_*_*_*_*_*

Covid-19 Confirmed Cases in India

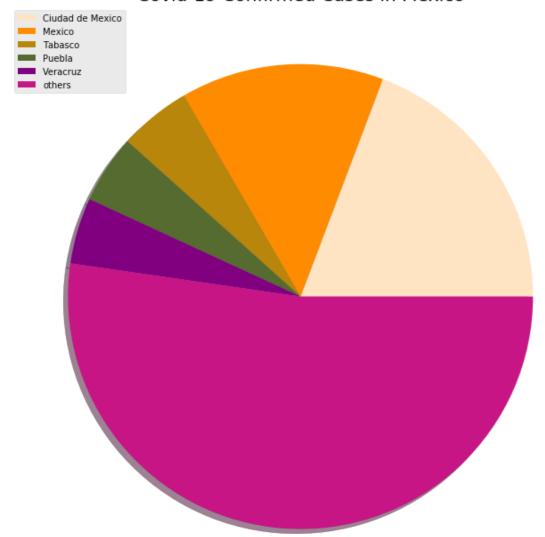


**_*_*_*_*_*_*_*_*_*_*_*_*_*_*_*Peru-*_*_*_*_*_*_*_*_*

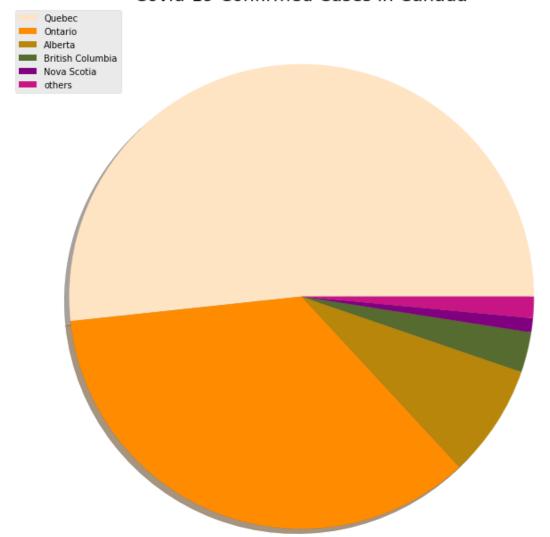
Covid-19 Confirmed Cases in Peru



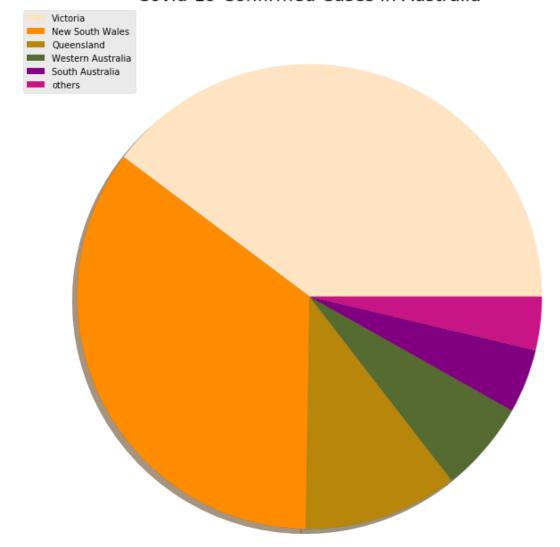






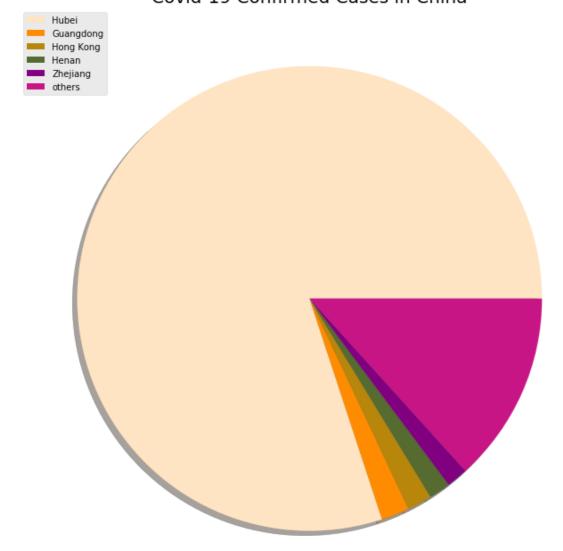


Covid-19 Confirmed Cases in Australia

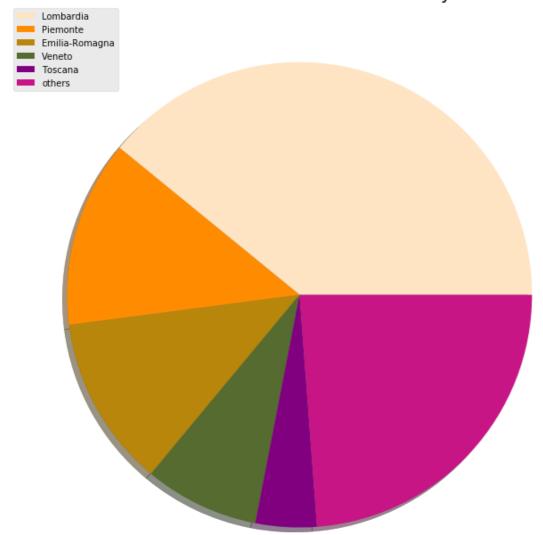


-*-*-*-*-*-*-*-*-*-*-*-*-*-*-*China-*-*-*-*-*-*-*-*-*-*-*-*-*-

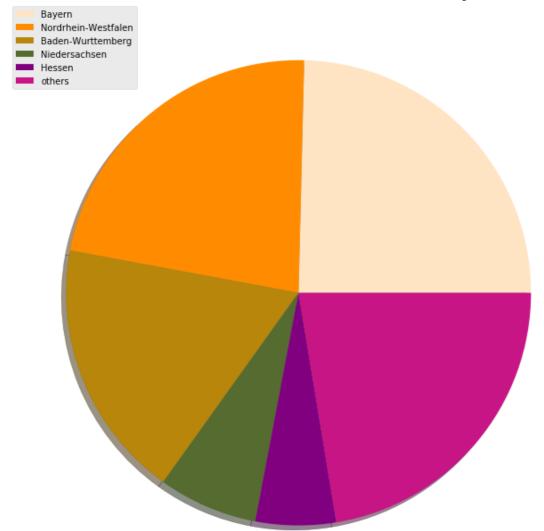




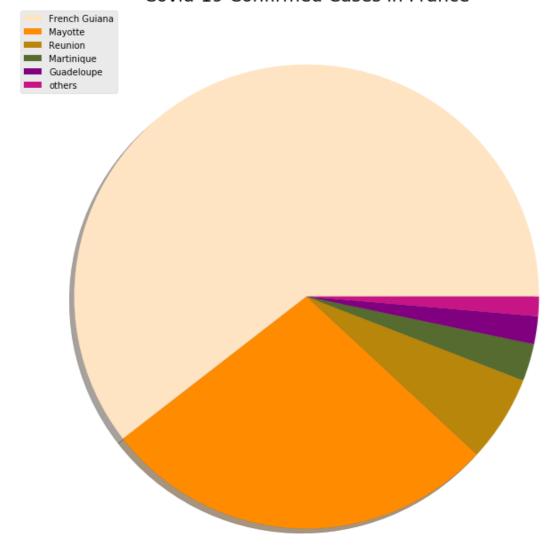




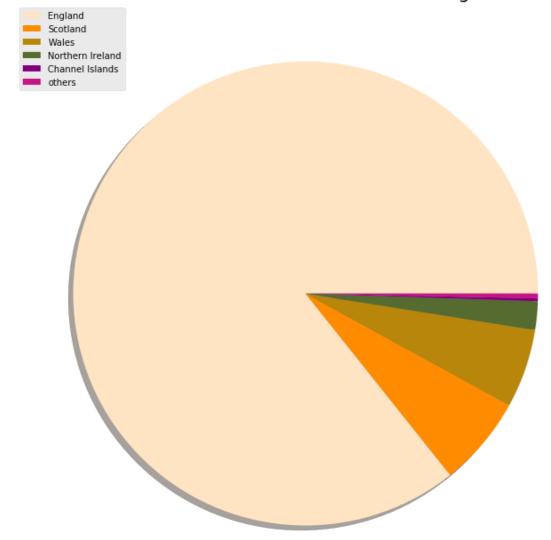
Covid-19 Confirmed Cases in Germany







Covid-19 Confirmed Cases in United Kingdom



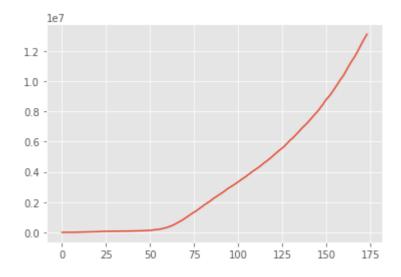
Prediction for confirmed cases over worldwide

In [54]:

plt.plot(days_since_1_22,world_cases)

Out[54]:

[<matplotlib.lines.Line2D at 0x16b3c136648>]



In [55]:

X_train_confirmed, X_test_confirmed, y_train_confirmed, y_test_confirmed = train_test_split

Finding best parameters using Randomized Search as it is faster then the Grid Search

In [56]:

```
# use this to find the optimal parameters for SVR
#c = [0.01, 0.1, 1]
#gamma = [0.01, 0.1, 1]
#epsilon = [0.01, 0.1, 1]
#shrinking = [True, False]
#degree = [3, 4, 5]

#svm_grid = {'C': c, 'gamma' : gamma, 'epsilon': epsilon, 'shrinking' : shrinking, 'degree'

#svm = SVR(kernel='poly')
#svm_search = RandomizedSearchCV(svm, svm_grid, scoring='neg_mean_squared_error', cv=3, ret
#svm_search.fit(X_train_confirmed, y_train_confirmed)'''
```

In [57]:

```
#svm_search.best_params_
```

Building SVM model

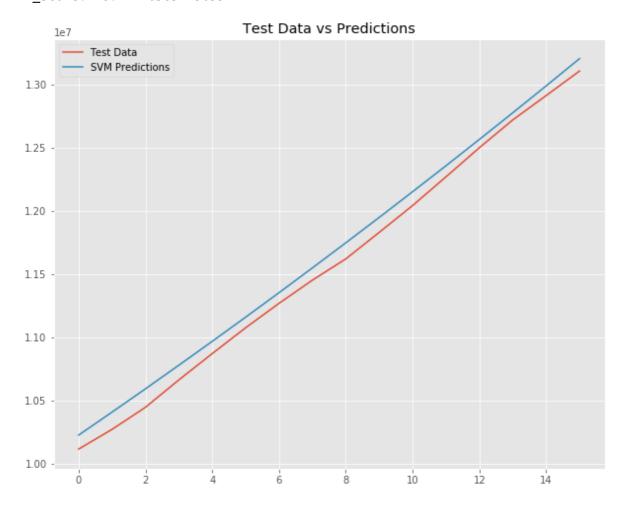
In [58]:

```
svm_confirmed = SVR(shrinking=True, kernel='poly',gamma=0.01, epsilon=1,degree=3, C=0.1)
svm_confirmed.fit(X_train_confirmed, y_train_confirmed)
svm_pred = svm_confirmed.predict(future_forcast)
```

In [59]:

```
svm_test_pred = svm_confirmed.predict(X_test_confirmed)
plt.figure(figsize=(10,8))
plt.title('Test Data vs Predictions')
plt.plot(y_test_confirmed)
plt.plot(svm_test_pred)
plt.legend(['Test Data', 'SVM Predictions'])
print('MAE:', mean_absolute_error(svm_test_pred, y_test_confirmed))
print('MSE:',mean_squared_error(svm_test_pred, y_test_confirmed)/1000000000)
print('r2_score:',r2_score(svm_test_pred, y_test_confirmed)*100)
```

MAE: 100525.51082913461 MSE: 10.685429589228875 r2_score: 98.72116585113653

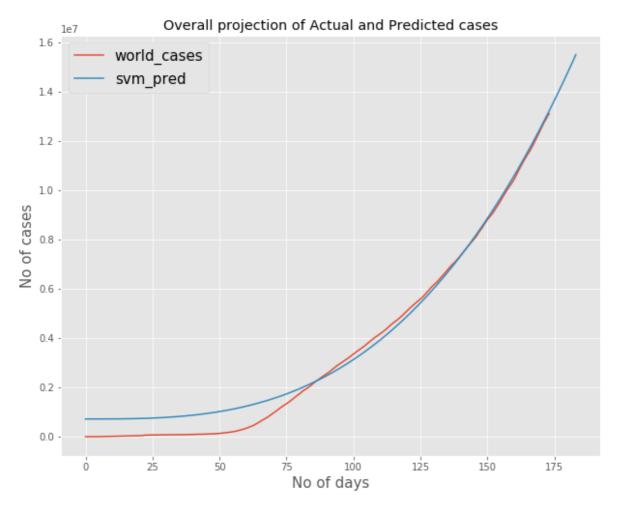


In [60]:

```
plt.figure(figsize=(10,8))
plt.plot(days_since_1_22,world_cases)
plt.plot(future_forcast,svm_pred)
plt.title('Overall projection of Actual and Predicted cases')
plt.legend(['world_cases','svm_pred'],fontsize = 15)
plt.xlabel('No of days',size = 15)
plt.ylabel('No of cases',size=15)
```

Out[60]:

Text(0, 0.5, 'No of cases')



In [61]:

```
# transform our data for polynomial regression
poly = PolynomialFeatures(degree=5)
poly_X_train_confirmed = poly.fit_transform(X_train_confirmed)
poly_X_test_confirmed = poly.fit_transform(X_test_confirmed)
poly_future_forcast = poly.fit_transform(future_forcast)
```

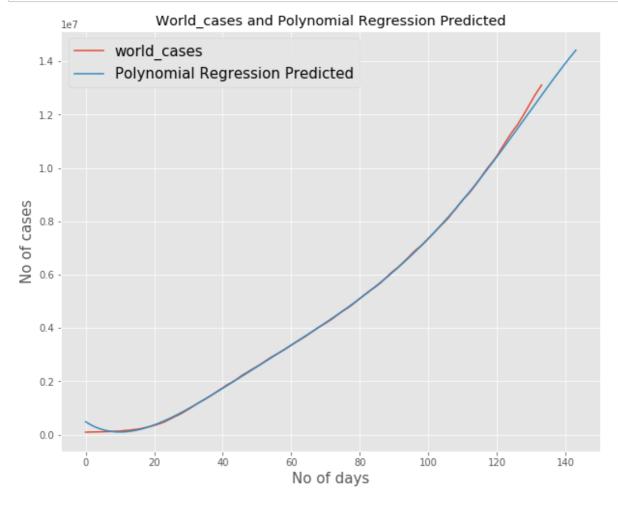
In [62]:

```
# polynomial regression
linear_model = LinearRegression(normalize=True, fit_intercept=False)
linear_model.fit(poly_X_train_confirmed, y_train_confirmed)
test_linear_pred = linear_model.predict(poly_X_test_confirmed)
linear_pred = linear_model.predict(poly_future_forcast)
print('MAE:', mean_absolute_error(test_linear_pred, y_test_confirmed))
print('MSE:',mean_squared_error(test_linear_pred, y_test_confirmed)/1000000000)
print('R2_SCORE:',r2_score(test_linear_pred, y_test_confirmed)*100)
```

MAE: 178039.9687996395 MSE: 45.46661105469835 R2 SCORE: 93.14031942255382

In [63]:

```
plt.figure(figsize=(10,8))
plt.title('World_cases and Polynomial Regression Predicted')
plt.plot(world_cases[40:])
plt.plot(linear_pred[40:])
plt.legend(['world_cases','Polynomial Regression Predicted'],fontsize = 15)
plt.xlabel('No of days',size = 15)
plt.ylabel('No of cases',size=15)
plt.show()
```



Conclusion :- SVM out performs in case of Confirmed Cases in every aspect (MSE,r2_score)

```
In [64]:
```

```
world_cases[-1][0] # current last case count
```

Out[64]:

13104391

In [65]:

```
#predicted confirmed case counts
future_case_count= [int(val) for val in svm_pred[-10:]]
future_case_count
```

Out[65]:

[13419976, 13640254, 13863064, 14088420, 14316337, 14546830, 14779912, 15015599, 15253904, 15494843]

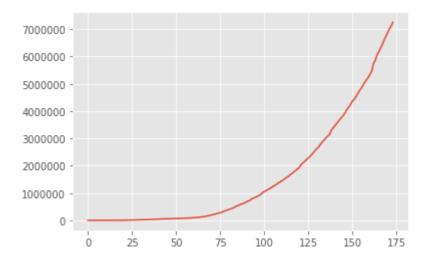
Prediction for Recovered cases over worldwide

In [66]:

```
plt.plot(days_since_1_22,total_recovered)
```

Out[66]:

[<matplotlib.lines.Line2D at 0x16b3c1e99c8>]



In [67]:

```
X_train_recovered, X_test_recovered, y_train_recovered, y_test_recovered = train_test_split
```

In [68]:

```
# use this to find the optimal parameters for SVR
#c = [0.01, 0.1, 1]
#gamma = [0.01, 0.1, 1]
#epsilon = [0.01, 0.1, 1]
#degree = [3, 4, 5]

#svm_grid = {'C': c, 'gamma' : gamma, 'epsilon': epsilon, 'shrinking' : shrinking, 'degree'

#svm = SVR(kernel='poly')
#svm_search = RandomizedSearchCV(svm, svm_grid, cv=3, n_jobs=-1, n_iter=10, verbose=1)
#svm_search.fit(X_train_confirmed, y_train_confirmed)'''
```

In [69]:

```
#svm_search.best_params_
```

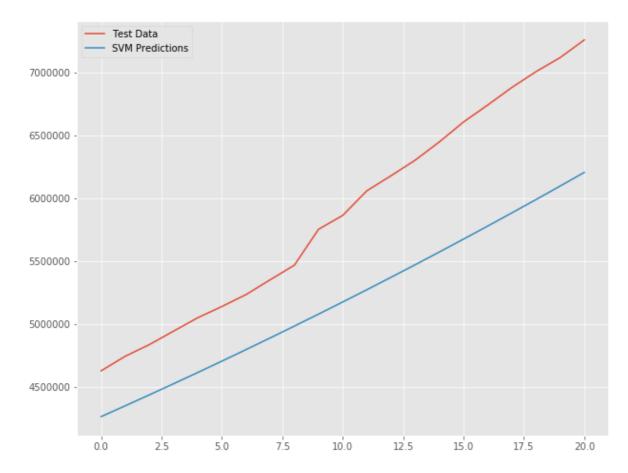
In [70]:

```
svm_recovered = SVR(shrinking=True, kernel='poly',gamma=0.01, epsilon=1,degree=3, C=0.1)
svm_recovered.fit(X_train_recovered, y_train_recovered)
svm_pred = svm_recovered.predict(future_forcast)
```

In [71]:

```
svm_test_pred = svm_recovered.predict(X_test_recovered)
plt.figure(figsize=(10,8))
plt.plot(y_test_recovered)
plt.plot(svm_test_pred)
plt.legend(['Test Data', 'SVM Predictions'])
print('MAE:', mean_absolute_error(svm_test_pred, y_test_recovered))
print('MSE:',mean_squared_error(svm_test_pred, y_test_recovered)/1000000000)
```

MAE: 688314.9931560386 MSE: 535.1556065192208



In [72]:

```
# transform our data for polynomial regression
poly = PolynomialFeatures(degree=6)
poly_X_train_recovered = poly.fit_transform(X_train_recovered)
poly_X_test_recovered = poly.fit_transform(X_test_recovered)
poly_future_forcast = poly.fit_transform(future_forcast)
```

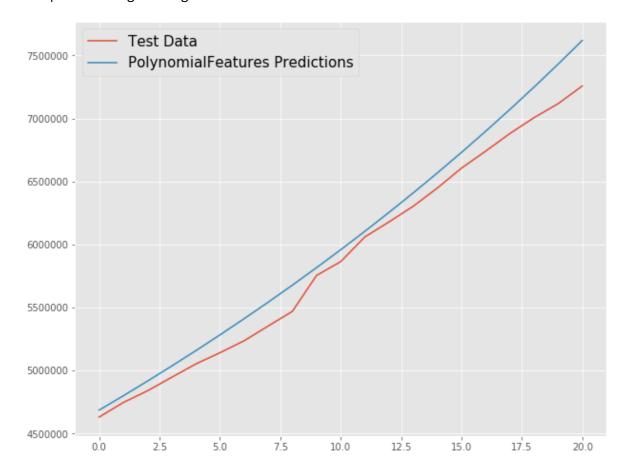
In [73]:

```
# polynomial regression
linear_model = LinearRegression(normalize=True, fit_intercept=False)
linear_model.fit(poly_X_train_recovered, y_train_recovered)
test_linear_pred = linear_model.predict(poly_X_test_recovered)
linear_pred = linear_model.predict(poly_future_forcast)
print('MAE:', mean_absolute_error(test_linear_pred, y_test_recovered))
print('MSE:', mean_squared_error(test_linear_pred, y_test_recovered)/1000000000)
print('R2_SCORE:',r2_score(test_linear_pred, y_test_recovered)*100)
plt.figure(figsize=(10,8))
plt.plot(y_test_recovered)
plt.plot(test_linear_pred)
plt.legend(['Test_Data', 'PolynomialFeatures_Predictions'],fontsize = 15)
```

MAE: 142199.317376865 MSE: 27.189698333070922 R2_SCORE: 96.52960865966634

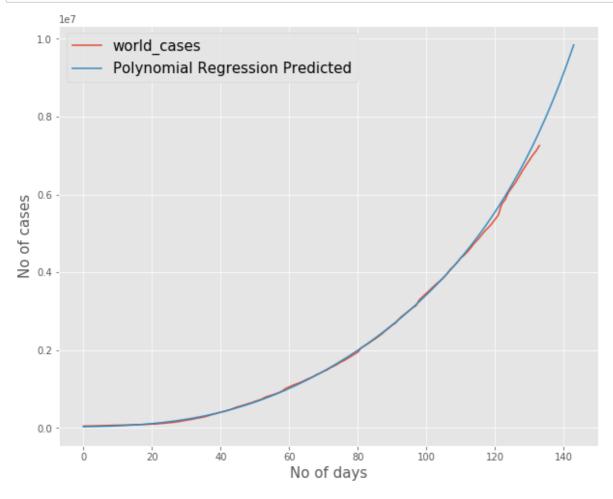
Out[73]:

<matplotlib.legend.Legend at 0x16b3c8aae88>



In [74]:

```
plt.figure(figsize=(10,8))
plt.plot(total_recovered[40:])
plt.plot(linear_pred[40:])
plt.legend(['world_cases','Polynomial Regression Predicted'],fontsize = 15)
plt.xlabel('No of days',size = 15)
plt.ylabel('No of cases',size=15)
plt.show()
```

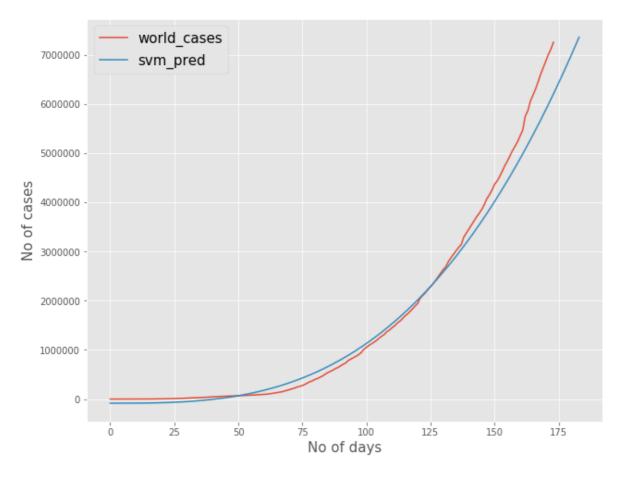


In [75]:

```
plt.figure(figsize=(10,8))
plt.plot(days_since_1_22,total_recovered)
plt.plot(future_forcast,svm_pred)
plt.legend(['world_cases','svm_pred'],fontsize = 15)
plt.xlabel('No of days',size = 15)
plt.ylabel('No of cases',size=15)
```

Out[75]:

Text(0, 0.5, 'No of cases')



Here Polynomial Regression Out Performs SVM as it can be seen from Graph and MSE values

```
In [76]:
total_deaths[-1][0]
Out[76]:
573003
In [77]:
#predicted recovery case counts
future_recover_count = linear_model.predict(poly_future_forcast[-10:])
future_recover_count = [int(val[0]) for val in future_recover_count]
future_recover_count
Out[77]:
[7812342,
 8011859,
 8217415,
 8429239,
 8647565,
 8872636,
 9104702,
9344022,
9590862,
9845498]
```

Prediction for Death cases over worldwide

```
In [78]:
```

```
In [79]:
```

```
nd the optimal parameters for SVR
1]
0.1, 1]

: c, 'gamma' : gamma}

L='poly')
ndomizedSearchCV(svm, svm_grid, scoring='neg_mean_squared_error', cv=3, return_train_score=TX_train_confirmed, y_train_confirmed)'''
```

```
In [80]:
```

```
#svm_search.best_params_
```

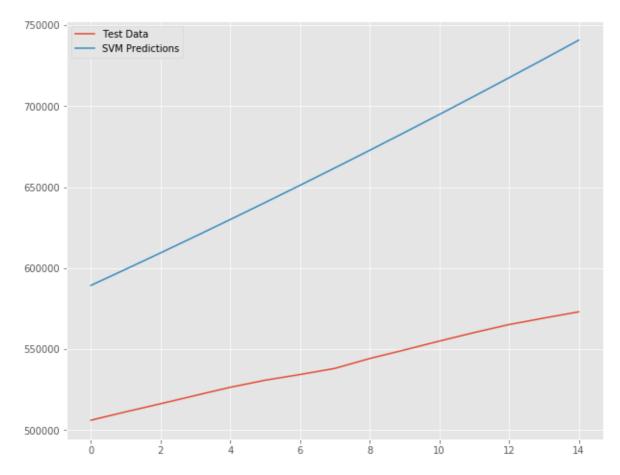
In [81]:

```
svm_death = SVR(shrinking=True, kernel='poly',gamma=0.01, epsilon=1,degree=3, C=0.1)
svm_death.fit(X_train_death, y_train_death)
svm_pred = svm_death.predict(future_forcast)
```

In [82]:

```
svm_test_pred = svm_death.predict(X_test_death)
plt.figure(figsize=(10,8))
plt.plot(y_test_death)
plt.plot(svm_test_pred)
plt.legend(['Test_Data', 'SVM_Predictions'])
print('MAE:', mean_absolute_error(svm_test_pred, y_test_death))
print('MSE:',mean_squared_error(svm_test_pred, y_test_death)/1000000000)
```

MAE: 122993.56283879364 MSE: 15.800004721292286

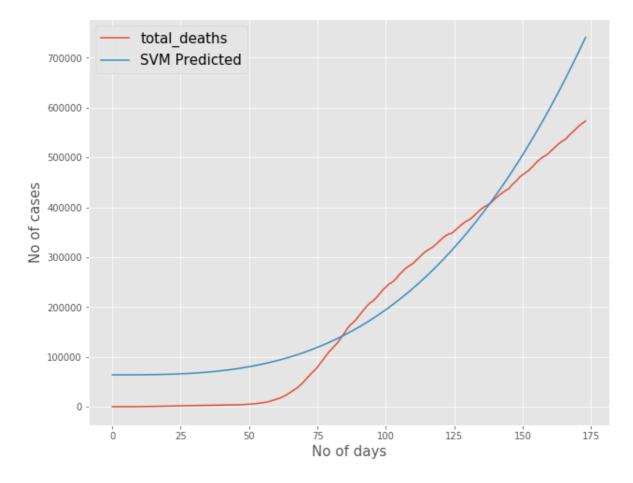


In [83]:

```
svm_pred = svm_death.predict(days_since_1_22)
print('MAE:', mean_absolute_error(svm_pred, total_deaths))
print('MSE:',mean_squared_error(svm_pred, total_deaths)/1000000000)
print('R2_score : ',r2_score(svm_pred, total_deaths)*100)
plt.figure(figsize=(10,8))
plt.plot(total_deaths)
plt.plot(svm_pred)
plt.legend(['total_deaths','SVM Predicted'],fontsize = 15)
plt.xlabel('No of days',size = 15)
plt.ylabel('No of cases',size=15)
plt.show()
```

MAE: 55486.47787262399 MSE: 4.002608608149228

R2_score: 89.30641934846717



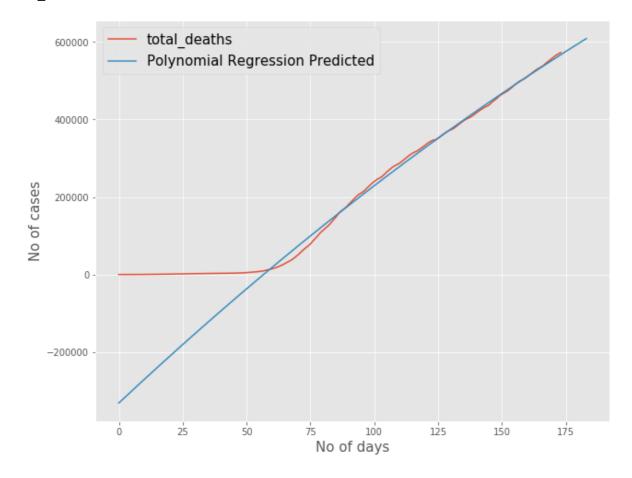
In [84]:

```
# transform our data for polynomial regression
poly = PolynomialFeatures(degree=2)
poly_X_train_death = poly.fit_transform(X_train_death)
poly_X_test_death = poly.fit_transform(X_test_death)
poly_future_forcast = poly.fit_transform(future_forcast)
```

In [85]:

```
# polynomial regression
linear_model = LinearRegression(normalize=True, fit_intercept=False)
linear_model.fit(poly_X_train_death, y_train_death)
test_linear_pred = linear_model.predict(poly_X_test_death)
linear_pred = linear_model.predict(poly_future_forcast)
print('MAE:', mean_absolute_error(test_linear_pred, y_test_death))
print('MSE:',mean_squared_error(test_linear_pred, y_test_death)/100000000)
print('R2_SCORE:',r2_score(test_linear_pred, y_test_death)*100)
plt.figure(figsize=(10,8))
plt.plot(total_deaths)
plt.plot(linear_pred)
plt.legend(['total_deaths','Polynomial Regression Predicted'],fontsize = 15)
plt.xlabel('No of days',size = 15)
plt.ylabel('No of cases',size=15)
plt.show()
```

MAE: 2907.487223150271 MSE: 0.13332335671676102 R2_SCORE: 96.1271025753039



Here Polynomial Regression Out Performs SVM as it can be seen from Graph and MSE values

```
In [86]:
total deaths[-1][0] # recent death case count
Out[86]:
573003
In [87]:
#predicted death case counts
future_death_count = linear_model.predict(poly_future_forcast[-10:])
future_death_count = [int(val[0]) for val in future_death_count]
future_death_count
Out[87]:
[571373,
 575571,
 579758,
 583933,
 588097,
 592250,
 596392,
 600522,
 604641,
 608749]
```

Top countries effected with virus

```
In [88]:
print('Top countries effected with virus are : ',countries)
Top countries effected with virus are : ['US', 'Brazil', 'India', 'Russia',
'Peru', 'Chile', 'Mexico', 'United Kingdom', 'South Africa', 'Iran']
In [89]:
start = '1/22/2020'
start date = datetime.datetime.strptime(start, '%m/%d/%Y')
future_forcast_dates = []
for i in range(len(future forcast)):
    future_forcast_dates.append((start_date + datetime.timedelta(days=i)).strftime('%m/%d/%
future_forcast_dates[-10:]
Out[89]:
['07/14/2020',
 '07/15/2020',
 '07/16/2020'
 '07/17/2020',
 '07/18/2020',
 '07/19/2020',
 '07/20/2020',
 '07/21/2020',
 '07/22/2020',
 '07/23/2020']
```

Prediction of Confirmed ,Recoveries,Deaths Case Counts

In [90]:

Out[90]:

	Cases	Recoveries	Deaths	Moratality_rate	Recovery_rate
Dates					
07/14/2020	13419976	7812342	571373	0.042576	0.582143
07/15/2020	13640254	8011859	575571	0.042197	0.587369
07/16/2020	13863064	8217415	579758	0.041820	0.592756
07/17/2020	14088420	8429239	583933	0.041448	0.598310
07/18/2020	14316337	8647565	588097	0.041079	0.604035
07/19/2020	14546830	8872636	592250	0.040713	0.609936
07/20/2020	14779912	9104702	596392	0.040352	0.616019
07/21/2020	15015599	9344022	600522	0.039993	0.622288
07/22/2020	15253904	9590862	604641	0.039638	0.628748
07/23/2020	15494843	9845498	608749	0.039287	0.635405