# **Machine learning assignment**

Covid-19 Outbreak Prediction

Dataset: <a href="https://www.kaggle.com/sudalairajkumar/novel-corona-virus-2019-dataset">https://www.kaggle.com/sudalairajkumar/novel-corona-virus-2019-dataset</a> (<a href="https://www.kaggle.com/sudalairajkumar/novel-corona-virus-2019-dataset">https://www.kaggle.com/sudalairajkumar/novel-corona-virus-2019-dataset</a>)

Objective: -to predict the spread of corona virus across the region for China, US and Malaysia -to analyses the growth rates and types of mitigation applied based on China, US and Malaysia

### In [2]:

```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import matplotlib.pyplot as plt
import seaborn as sns
import statsmodels as sm
from sklearn.impute import SimpleImputer
sns.set()
%matplotlib inline
pd.options.plotting.backend
pd.plotting.register_matplotlib_converters()
```

# **Data cleaning**

In this part of notebook, I take csv files covid 19 data

Clean data from covid\_19\_data.csv file and only select region for China, US and Malaysia

```
In [3]:
```

```
covid = pd.read_csv('covid_19_data.csv')
```

# In [4]:

```
#see data
covid.head()
```

# Out[4]:

|   | SNo | ObservationDate | Province/State | Country/Region | Last<br>Update     | Confirmed | Deaths | Rec |
|---|-----|-----------------|----------------|----------------|--------------------|-----------|--------|-----|
| 0 | 1   | 01/22/2020      | Anhui          | Mainland China | 1/22/2020<br>17:00 | 1.0       | 0.0    |     |
| 1 | 2   | 01/22/2020      | Beijing        | Mainland China | 1/22/2020<br>17:00 | 14.0      | 0.0    |     |
| 2 | 3   | 01/22/2020      | Chongqing      | Mainland China | 1/22/2020<br>17:00 | 6.0       | 0.0    |     |
| 3 | 4   | 01/22/2020      | Fujian         | Mainland China | 1/22/2020<br>17:00 | 1.0       | 0.0    |     |
| 4 | 5   | 01/22/2020      | Gansu          | Mainland China | 1/22/2020<br>17:00 | 0.0       | 0.0    |     |
| 4 |     |                 |                |                |                    |           |        | •   |

# In [5]:

```
# data information
covid.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 172480 entries, 0 to 172479
Data columns (total 8 columns):
```

Column Non-Null Count Dtype ---------0 SNo 172480 non-null int64 1 ObservationDate 172480 non-null object 2 Province/State 124597 non-null object 3 Country/Region 172480 non-null object 4 Last Update 172480 non-null object 5 Confirmed 172480 non-null float64 Deaths 172480 non-null 6 float64

172480 non-null

dtypes: float64(3), int64(1), object(4)

memory usage: 10.5+ MB

Recovered

# In [6]:

```
# check if there exist a missing value
mis = covid.isnull().sum()
mis[mis>0]
```

float64

# Out[6]:

Province/State 47883

dtype: int64

### In [8]:

#Only Province/State have a missing value. I can impute it because this variable is nec
essary for visualizing a data.

imputer = SimpleImputer(strategy='constant')#here I use constant because I cannot put a
nother Province/State
#that we do not know or that does not correspond to his country/region
impute\_covid = pd.DataFrame(imputer.fit\_transform(covid), columns=covid.columns)
impute\_covid.head()

### Out[8]:

|   | SNo | ObservationDate | Province/State | Country/Region | Last<br>Update     | Confirmed | Deaths | Rec |
|---|-----|-----------------|----------------|----------------|--------------------|-----------|--------|-----|
| 0 | 1   | 01/22/2020      | Anhui          | Mainland China | 1/22/2020<br>17:00 | 1         | 0      |     |
| 1 | 2   | 01/22/2020      | Beijing        | Mainland China | 1/22/2020<br>17:00 | 14        | 0      |     |
| 2 | 3   | 01/22/2020      | Chongqing      | Mainland China | 1/22/2020<br>17:00 | 6         | 0      |     |
| 3 | 4   | 01/22/2020      | Fujian         | Mainland China | 1/22/2020<br>17:00 | 1         | 0      |     |
| 4 | 5   | 01/22/2020      | Gansu          | Mainland China | 1/22/2020<br>17:00 | 0         | 0      |     |
| 4 |     |                 |                |                |                    |           |        | •   |

# In [9]:

```
#convert ObservationDate and Last Update object to datetime
#convert confirmed, recovered, death to numeric
impute_covid['ObservationDate'] = pd.to_datetime(impute_covid['ObservationDate'])
impute_covid['Last Update'] = pd.to_datetime(impute_covid['Last Update'])
impute_covid['Confirmed'] = pd.to_numeric(impute_covid['Confirmed'], errors='coerce')
impute_covid['Recovered'] = pd.to_numeric(impute_covid['Recovered'], errors='coerce')
impute_covid['Deaths'] = pd.to_numeric(impute_covid['Deaths'], errors='coerce')
```

#### In [10]:

```
#check
#impute_covid.to_csv('covid_19_data_clean.csv', index=False)
impute_covid.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 172480 entries, 0 to 172479
Data columns (total & columns):
```

Data columns (total 8 columns):

| #     | Column             | Non-Null Count    | Dtype                     |
|-------|--------------------|-------------------|---------------------------|
|       |                    |                   |                           |
| 0     | SNo                | 172480 non-null   | object                    |
| 1     | ObservationDate    | 172480 non-null   | <pre>datetime64[ns]</pre> |
| 2     | Province/State     | 172480 non-null   | object                    |
| 3     | Country/Region     | 172480 non-null   | object                    |
| 4     | Last Update        | 172480 non-null   | <pre>datetime64[ns]</pre> |
| 5     | Confirmed          | 172480 non-null   | float64                   |
| 6     | Deaths             | 172480 non-null   | float64                   |
| 7     | Recovered          | 172480 non-null   | float64                   |
| dtyp  | es: datetime64[ns] | ](2), float64(3), | object(3)                 |
| memoi | ry usage: 10.5+ M  | 3                 |                           |

we are finishing data cleaning

Feature Statistics and Visualization impute\_covid: we are going to visualize this data and make some statistics to find the relevant information.

# In [11]:

```
# see again data table
impute_covid.head(3)
```

#### Out[11]:

|   | SNo | ObservationDate | Province/State | Country/Region | Last<br>Update             | Confirmed | Deaths | Reco |
|---|-----|-----------------|----------------|----------------|----------------------------|-----------|--------|------|
| 0 | 1   | 2020-01-22      | Anhui          | Mainland China | 2020-<br>01-22<br>17:00:00 | 1.0       | 0.0    |      |
| 1 | 2   | 2020-01-22      | Beijing        | Mainland China | 2020-<br>01-22<br>17:00:00 | 14.0      | 0.0    |      |
| 2 | 3   | 2020-01-22      | Chongqing      | Mainland China | 2020-<br>01-22<br>17:00:00 | 6.0       | 0.0    |      |
| 4 |     |                 |                |                |                            |           |        | •    |

# In [12]:

```
# we compute the active_confirmed
impute_covid['active_confirmed'] = impute_covid['Confirmed'].values - \
(impute_covid['Deaths'].values+impute_covid['Recovered'].values)
```

```
In [13]:
```

```
#check if all is ok
impute_covid.isnull().sum()[impute_covid.isnull().sum()>0]

Out[13]:
Series([], dtype: int64)

In [14]:
#ok we have no problem see table data
impute_covid.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 172480 entries, 0 to 172479
Data columns (total 9 columns):

# Column Non-Null Count Dtype
--- 0 SNo 172480 non-null object

1 ObservationDate 172480 non-null datetime64[ns]
2 Province/State 172480 non-null object

2 Province/State 1/2480 non-null object 3 Country/Region 172480 non-null object

4 Last Update 172480 non-null datetime64[ns]

5 Confirmed 172480 non-null float64
6 Deaths 172480 non-null float64
7 Recovered 172480 non-null float64
8 active\_confirmed 172480 non-null float64

dtypes: datetime64[ns](2), float64(4), object(3)

memory usage: 11.8+ MB

# China

#### In [17]:

```
china = impute_covid[impute_covid['Country/Region'] == 'Mainland China']
chstar_date = china.ObservationDate.min()
chend_date = china.ObservationDate.max()
print('Novel covid-19 China:\n start date = {}\n end date = {}'.format(chstar_date, che
nd_date))
```

```
Novel covid-19 China:

start date = 2020-01-22 00:00:00

end date = 2020-12-06 00:00:00
```

### In [18]:

```
lastChina = china[china['ObservationDate'] == chend_date]
lastChina.head()
```

# Out[18]:

|        | SNo    | ObservationDate | Province/State | Country/Region | Last<br>Update             | Confirmed | Death |
|--------|--------|-----------------|----------------|----------------|----------------------------|-----------|-------|
| 171912 | 171913 | 2020-12-06      | Anhui          | Mainland China | 2020-<br>12-07<br>05:26:14 | 992.0     | 6     |
| 171946 | 171947 | 2020-12-06      | Beijing        | Mainland China | 2020-<br>12-07<br>05:26:14 | 952.0     | 9     |
| 171996 | 171997 | 2020-12-06      | Chongqing      | Mainland China | 2020-<br>12-07<br>05:26:14 | 590.0     | 6     |
| 172038 | 172039 | 2020-12-06      | Fujian         | Mainland China | 2020-<br>12-07<br>05:26:14 | 500.0     | 1     |
| 172043 | 172044 | 2020-12-06      | Gansu          | Mainland China | 2020-<br>12-07<br>05:26:14 | 182.0     | 2     |

**→** 

# In [19]:

Tota confirmed: 86634.0 Total Deaths: 4634.0 Total Recovered: 81718.0 Total active confirmed: 282.0

# In [20]:

lastChina[['Province/State', 'Confirmed', 'Deaths', 'Recovered', 'active\_confirmed']].s
tyle.\
background\_gradient(cmap='viridis')

# Out[20]:

|        | Province/State | Confirmed    | Deaths      | Recovered    | active_confirmed |
|--------|----------------|--------------|-------------|--------------|------------------|
| 171912 | Anhui          | 992.000000   | 6.000000    | 986.000000   | 0.000000         |
| 171946 | Beijing        | 952.000000   | 9.000000    | 939.000000   | 4.000000         |
| 171996 | Chongqing      | 590.000000   | 6.000000    | 583.000000   | 1.000000         |
| 172038 | Fujian         | 500.000000   | 1.000000    | 453.000000   | 46.000000        |
| 172043 | Gansu          | 182.000000   | 2.000000    | 180.000000   | 0.000000         |
| 172061 | Guangdong      | 2004.000000  | 8.000000    | 1960.000000  | 36.000000        |
| 172062 | Guangxi        | 263.000000   | 2.000000    | 260.000000   | 1.000000         |
| 172065 | Guizhou        | 147.000000   | 2.000000    | 145.000000   | 0.000000         |
| 172068 | Hainan         | 171.000000   | 6.000000    | 165.000000   | 0.000000         |
| 172074 | Hebei          | 373.000000   | 6.000000    | 367.000000   | 0.000000         |
| 172075 | Heilongjiang   | 949.000000   | 13.000000   | 936.000000   | 0.000000         |
| 172076 | Henan          | 1295.000000  | 22.000000   | 1266.000000  | 7.000000         |
| 172085 | Hubei          | 68149.000000 | 4512.000000 | 63633.000000 | 4.000000         |
| 172087 | Hunan          | 1020.000000  | 4.000000    | 1016.000000  | 0.000000         |
| 172095 | Inner Mongolia | 336.000000   | 1.000000    | 308.000000   | 27.000000        |
| 172109 | Jiangsu        | 680.000000   | 0.000000    | 674.000000   | 6.000000         |
| 172110 | Jiangxi        | 935.000000   | 1.000000    | 934.000000   | 0.000000         |
| 172111 | Jilin          | 157.000000   | 2.000000    | 155.000000   | 0.000000         |
| 172159 | Liaoning       | 289.000000   | 2.000000    | 287.000000   | 0.000000         |
| 172243 | Ningxia        | 75.000000    | 0.000000    | 75.000000    | 0.000000         |
| 172308 | Qinghai        | 18.000000    | 0.000000    | 18.000000    | 0.000000         |
| 172353 | Shaanxi        | 502.000000   | 3.000000    | 478.000000   | 21.000000        |
| 172354 | Shandong       | 857.000000   | 7.000000    | 841.000000   | 9.000000         |
| 172355 | Shanghai       | 1366.000000  | 7.000000    | 1294.000000  | 65.000000        |
| 172356 | Shanxi         | 222.000000   | 0.000000    | 218.000000   | 4.000000         |
| 172360 | Sichuan        | 812.000000   | 3.000000    | 779.000000   | 30.000000        |
| 172392 | Tianjin        | 301.000000   | 3.000000    | 292.000000   | 6.000000         |
| 172393 | Tibet          | 1.000000     | 0.000000    | 1.000000     | 0.000000         |
| 172463 | Xinjiang       | 980.000000   | 3.000000    | 977.000000   | 0.000000         |
| 172471 | Yunnan         | 221.000000   | 2.000000    | 210.000000   | 9.000000         |
| 172477 | Zhejiang       | 1295.000000  | 1.000000    | 1288.000000  | 6.000000         |

# **United State**

# In [21]:

```
us = impute_covid[impute_covid['Country/Region'] == 'US']
usstar_date = us.ObservationDate.min()
usend_date = us.ObservationDate.max()
print('Novel covid-19 US:\n start date = {}\n end date = {}'.format(usstar_date, usend_date))
```

```
Novel covid-19 US:

start date = 2020-01-22 00:00:00

end date = 2020-12-06 00:00:00
```

# In [23]:

```
lastUs = us[us['ObservationDate'] == usend_date]
lastUs.head()
```

# Out[23]:

|        | SNo    | ObservationDate | Province/State | Country/Region | Last<br>Update             | Confirmed | Deat  |
|--------|--------|-----------------|----------------|----------------|----------------------------|-----------|-------|
| 171896 | 171897 | 2020-12-06      | Alabama        | US             | 2020-<br>12-07<br>05:26:14 | 269877.0  | 3889  |
| 171898 | 171899 | 2020-12-06      | Alaska         | US             | 2020-<br>12-07<br>05:26:14 | 37036.0   | 143   |
| 171923 | 171924 | 2020-12-06      | Arizona        | US             | 2020-<br>12-07<br>05:26:14 | 364276.0  | 6950  |
| 171924 | 171925 | 2020-12-06      | Arkansas       | US             | 2020-<br>12-07<br>05:26:14 | 170924.0  | 2660  |
| 171967 | 171968 | 2020-12-06      | California     | US             | 2020-<br>12-07<br>05:26:14 | 1366673.0 | 19928 |
| 4      |        |                 |                |                |                            |           | •     |

```
In [24]:
```

Total confirmed: 14757000.0 Total Deaths: 282299.0 Total Recovered: 5624444.0

Total active confirmed: 8850257.0

\_\_\_\_\_\_

```
In [25]:
```

```
lastUs[['Province/State', 'Confirmed', 'Deaths', 'Recovered', 'active_confirmed']].styl
e.\
background_gradient(cmap='viridis')
```

# Out[25]:

|        | Province/State               | Confirmed      | Deaths       | Recovered | active_confirmed |
|--------|------------------------------|----------------|--------------|-----------|------------------|
| 171896 | Alabama                      | 269877.000000  | 3889.000000  | 0.000000  | 265988.000000    |
| 171898 | Alaska                       | 37036.000000   | 143.000000   | 0.000000  | 36893.000000     |
| 171923 | Arizona                      | 364276.000000  | 6950.000000  | 0.000000  | 357326.000000    |
| 171924 | Arkansas                     | 170924.000000  | 2660.000000  | 0.000000  | 168264.000000    |
| 171967 | California                   | 1366673.000000 | 19928.000000 | 0.000000  | 1346745.000000   |
| 172002 | Colorado                     | 260581.000000  | 3356.000000  | 0.000000  | 257225.000000    |
| 172003 | Connecticut                  | 127715.000000  | 5146.000000  | 0.000000  | 122569.000000    |
| 172013 | Delaware                     | 39912.000000   | 793.000000   | 0.000000  | 39119.000000     |
| 172016 | Diamond Princess cruise ship | 49.000000      | 0.000000     | 0.000000  | 49.000000        |
| 172017 | District of Columbia         | 23136.000000   | 697.000000   | 0.000000  | 22439.000000     |
| 172033 | Florida                      | 1058074.000000 | 19177.000000 | 0.000000  | 1038897.000000   |
| 172046 | Georgia                      | 501405.000000  | 9806.000000  | 0.000000  | 491599.000000    |
| 172054 | Grand Princess               | 103.000000     | 3.000000     | 0.000000  | 100.000000       |
| 172059 | Guam                         | 6959.000000    | 113.000000   | 0.000000  | 6846.000000      |
| 172073 | Hawaii                       | 18842.000000   | 262.000000   | 0.000000  | 18580.000000     |
| 172091 | Idaho                        | 110510.000000  | 1035.000000  | 0.000000  | 109475.000000    |
| 172092 | Illinois                     | 787573.000000  | 14116.000000 | 0.000000  | 773457.000000    |
| 172093 | Indiana                      | 381617.000000  | 6242.000000  | 0.000000  | 375375.000000    |
| 172096 | Iowa                         | 244691.000000  | 2717.000000  | 0.000000  | 241974.000000    |
| 172123 | Kansas                       | 171364.000000  | 1786.000000  | 0.000000  | 169578.000000    |
| 172128 | Kentucky                     | 200631.000000  | 2072.000000  | 0.000000  | 198559.000000    |
| 172170 | Louisiana                    | 251123.000000  | 6584.000000  | 0.000000  | 244539.000000    |
| 172182 | Maine                        | 13348.000000   | 227.000000   | 0.000000  | 13121.000000     |
| 172189 | Maryland                     | 215027.000000  | 4846.000000  | 0.000000  | 210181.000000    |
| 172190 | Massachusetts                | 256844.000000  | 11004.000000 | 0.000000  | 245840.000000    |
| 172201 | Michigan                     | 426576.000000  | 10321.000000 | 0.000000  | 416255.000000    |
| 172205 | Minnesota                    | 350862.000000  | 4043.000000  | 0.000000  | 346819.000000    |
| 172206 | Mississippi                  | 164931.000000  | 3961.000000  | 0.000000  | 160970.000000    |
| 172207 | Missouri                     | 329420.000000  | 4284.000000  | 0.000000  | 325136.000000    |
| 172212 | Montana                      | 67875.000000   | 736.000000   | 0.000000  | 67139.000000     |
| 172230 | Nebraska                     | 139834.000000  | 1205.000000  | 0.000000  | 138629.000000    |
| 172232 | Nevada                       | 168140.000000  | 2315.000000  | 0.000000  | 165825.000000    |
| 172235 | New Hampshire                | 24888.000000   | 564.000000   | 0.000000  | 24324.000000     |
| 172236 | New Jersey                   | 368016.000000  | 17321.000000 | 0.000000  | 350695.000000    |
| 172237 | New Mexico                   | 108088.000000  | 1749.000000  | 0.000000  | 106339.000000    |
| 172239 | New York                     | 705827.000000  | 34958.000000 | 0.000000  | 670869.000000    |

|        | Province/State              | Confirmed      | Deaths       | Recovered      | active_confirmed |
|--------|-----------------------------|----------------|--------------|----------------|------------------|
| 172250 | North Carolina              | 394990.000000  | 5543.000000  | 0.000000       | 389447.000000    |
| 172251 | North Dakota                | 82981.000000   | 1019.000000  | 0.000000       | 81962.000000     |
| 172254 | Northern Mariana<br>Islands | 109.000000     | 2.000000     | 0.000000       | 107.000000       |
| 172267 | Ohio                        | 475024.000000  | 6959.000000  | 0.000000       | 468065.000000    |
| 172271 | Oklahoma                    | 216486.000000  | 1896.000000  | 0.000000       | 214590.000000    |
| 172275 | Oregon                      | 84496.000000   | 1033.000000  | 0.000000       | 83463.000000     |
| 172288 | Pennsylvania                | 423100.000000  | 11255.000000 | 0.000000       | 411845.000000    |
| 172302 | Puerto Rico                 | 56671.000000   | 1192.000000  | 0.000000       | 55479.000000     |
| 172315 | Recovered                   | 0.000000       | 0.000000     | 5624444.000000 | -5624444.000000  |
| 172319 | Rhode Island                | 62137.000000   | 1413.000000  | 0.000000       | 60724.000000     |
| 172371 | South Carolina              | 232099.000000  | 4566.000000  | 0.000000       | 227533.000000    |
| 172372 | South Dakota                | 85991.000000   | 1110.000000  | 0.000000       | 84881.000000     |
| 172388 | Tennessee                   | 400594.000000  | 4943.000000  | 0.000000       | 395651.000000    |
| 172390 | Texas                       | 1322738.000000 | 23137.000000 | 0.000000       | 1299601.000000   |
| 172427 | Utah                        | 215407.000000  | 939.000000   | 0.000000       | 214468.000000    |
| 172442 | Vermont                     | 5015.000000    | 79.000000    | 0.000000       | 4936.000000      |
| 172446 | Virgin Islands              | 1633.000000    | 23.000000    | 0.000000       | 1610.000000      |
| 172447 | Virginia                    | 255053.000000  | 4200.000000  | 0.000000       | 250853.000000    |
| 172456 | Washington                  | 177447.000000  | 2925.000000  | 0.000000       | 174522.000000    |
| 172459 | West Virginia               | 54997.000000   | 838.000000   | 0.000000       | 54159.000000     |
| 172461 | Wisconsin                   | 441067.000000  | 3952.000000  | 0.000000       | 437115.000000    |
| 172462 | Wyoming                     | 36218.000000   | 266.000000   | 0.000000       | 35952.000000     |

# Malaysia

# In [26]:

```
my = impute_covid[impute_covid['Country/Region'] == 'Malaysia']
mystar_date = my.ObservationDate.min()
myend_date = my.ObservationDate.max()
print('Novel covid-19 Malaysia:\n start date = {}\n end date = {}'.format(mystar_date, myend_date))
Novel covid-19 Malaysia:
```

start date = 2020-01-23 00:00:00 end date = 2020-12-06 00:00:00

```
In [28]:
```

```
lastMy = my[my['ObservationDate'] == myend_date]
lastMy.head()
```

## Out[28]:

|        | SNo    | ObservationDate | Province/State | Country/Region | Last<br>Update             | Confirmed | Death    |
|--------|--------|-----------------|----------------|----------------|----------------------------|-----------|----------|
| 171815 | 171816 | 2020-12-06      | missing_value  | Malaysia       | 2020-<br>12-07<br>05:26:14 | 72694.0   | 382      |
| 4      |        |                 |                |                |                            |           | <b>•</b> |

# In [29]:

Total confirmed: 72694.0 Total Deaths: 382.0 Total Recovered: 61273.0

Total active confirmed: 11039.0

\_\_\_\_\_\_

#### In [30]:

```
lastMy[['Province/State', 'Confirmed', 'Deaths', 'Recovered', 'active_confirmed']].styl
e.\
background_gradient(cmap='viridis')
```

# Out[30]:

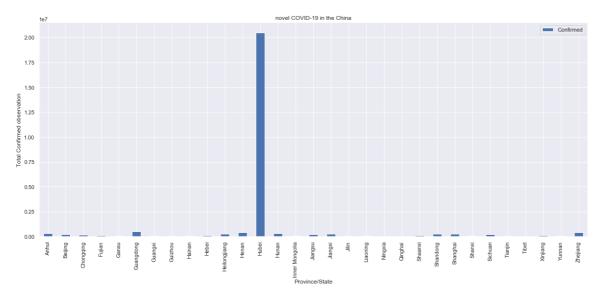
|        | Province/State | Confirmed    | Deaths     | Recovered    | active_confirmed |
|--------|----------------|--------------|------------|--------------|------------------|
| 171815 | missing_value  | 72694.000000 | 382.000000 | 61273.000000 | 11039.000000     |

# In [89]:

```
time_obs = china.groupby('Province/State')['Confirmed'].aggregate([np.sum])
time_obs.columns = ['Confirmed']
time_obs.plot(figsize=(20,8), title='novel COVID-19 in the China', kind='bar')
plt.ylabel('Total Confirmed observation')
```

# Out[89]:

Text(0, 0.5, 'Total Confirmed observation')



1) Predict the spread of corona virus across the region for China, US and Malaysia using Linear Regression

# In [95]:

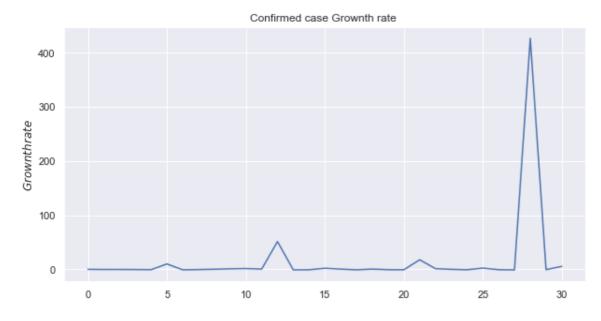
```
x = []
x.append(0)
for i in range(time_obs.shape[0]-1):
    a = time_obs.iloc[i+1,0]-time_obs.iloc[i,0]
    x.append(a/time_obs.iloc[i,0])

grown_rate = time_obs.reset_index()
grown_rate['grownRate'] = x
grown_rate.head()

grown_rate.grownRate.plot(figsize=(10,5))
plt.title('Confirmed case Grownth rate ')
plt.ylabel('$Grownth rate$')
```

# Out[95]:

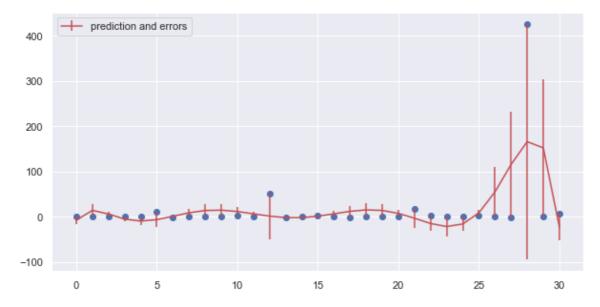
Text(0, 0.5, '\$Grownth rate\$')



# **Linear Regression**

### In [93]:

```
from sklearn.pipeline import make pipeline
from sklearn.preprocessing import PolynomialFeatures
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean absolute error
from sklearn.model selection import train test split
trend_model = make_pipeline(PolynomialFeatures(8), LinearRegression(normalize=True, fit
_intercept=True))
trend_model.fit(np.array(grown_rate.index).reshape((-1,1)), grown_rate['grownRate'])
dt =np.array(grown_rate.index).reshape((-1,1))
fit grown = trend model.predict(dt)
errors = grown_rate['grownRate'] - fit_grown
upperlimits = [True, False]
lowerlimits = [False, True]
plt.figure(figsize=(10,5))
plt.scatter(dt, grown_rate['grownRate'])
plt.errorbar(dt, fit_grown,yerr = errors, color='r', label='prediction and errors')
plt.legend(loc='best')
plt.show()
```



#### In [ ]: