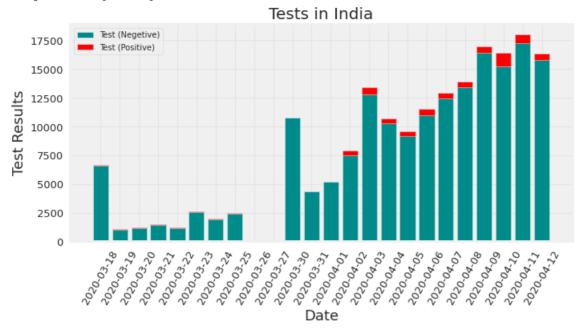
```
import requests
import json
import seaborn as sns
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.colors as mcolors
import pandas as pd
import random
import math
import time
from sklearn.linear_model import LinearRegression, BayesianRidge
from sklearn.model_selection import RandomizedSearchCV, train_test_split
from sklearn.preprocessing import PolynomialFeatures
from sklearn.tree import DecisionTreeRegressor
from sklearn.svm import SVR
from sklearn.metrics import mean_squared_error, mean_absolute_error
import datetime
import operator
from datetime import datetime
from dateutil.parser import parse
plt.style.use('fivethirtyeight')
%matplotlib inline
population = pd.read_csv('population_india_census2011.csv')
import json
from pandas.io.json import json_normalize
```

### General Stats of India:

Гэ

```
df_india_test = pd.io.json.json_normalize(requests.get('https://api.rootnet.in/covid19-in/stats/testing/l
   /usr/local/lib/python3.6/dist-packages/ipykernel launcher.py:1: FutureWarning: pandas.io.json.json n
       """Entry point for launching an IPython kernel.
df_india_test["p2t_ratio"]= np.round(100*df_india_test["c_positive"]/df_india_test["c_tests"],2)
df_india_test["positive"] = df_india_test["c_positive"].diff()
df_india_test["tests"] = df_india_test["c_tests"].diff()
df_india_test["p2t_ratio"]= np.round(100*df_india_test["positive"]/df_india_test["tests"],2)
df_india_test = df_india_test[1:]
f = plt.figure(figsize=(10,5))
f.add_subplot(111)
plt.axes(axisbelow=True)
plt.bar(df_india_test["day"],df_india_test["tests"].values[:],color="darkcyan",label="Test (Negetive)"+st
plt.bar(df_india_test["day"],df_india_test["positive"].values[:],bottom=df_india_test["tests"].values[:].
plt.tick_params(size=5,labelsize = 13)
plt.tick params(axis="x",size=5,labelsize = 13,labelrotation=60 )
plt.xlabel("Date", fontsize=18)
plt.ylabel("Test Results", fontsize=18)
plt.title("Tests in India", fontsize=20)
plt.grid(alpha=0.3)
plt.legend()
```



```
india_data_json = requests.get('https://api.rootnet.in/covid19-in/unofficial/covid19india.org/statewise'
df_india = pd.io.json.json_normalize(india_data_json['data']['statewise'])
df_india = df_india.set_index("state")
```

ry /usr/local/lib/python3.6/dist-packages/ipykernel\_launcher.py:2: FutureWarning: pandas.io.json\_n

```
total = df_india.sum()
total.name = "Total"
pd.DataFrame(total).transpose().style.background_gradient(cmap='Wistia',axis=1)
```

C→ confirmed recovered deaths active
Total 9211 1086 331 7794

df\_india.style.background\_gradient(cmap='Wistia')

₽

### confirmed recovered deaths active

state				
Maharashtra	1982	217	149	1616
Delhi	1154	28	24	1102
Tamil Nadu	1075	50	11	1014
Rajasthan	804	121	11	672
Madhya Pradesh	562	41	43	478
Telangana	531	103	16	412
Gujarat	516	44	24	448
Uttar Pradesh	483	45	5	433
Andhra Pradesh	420	12	7	401
Kerala	375	179	2	194
Jammu and Kashmir	245	6	4	235
Karnataka	232	54	6	172
Haryana	195	44	3	148
Punjab	170	23	12	135
West Bengal	134	19	7	108
Bihar	64	26	1	37
Odisha	54	12	1	41
Uttarakhand	35	5	0	30
Himachal Pradesh	32	12	2	18
Assam	29	0	1	28
Chhattisgarh	31	10	0	21
Chandigarh	21	7	0	14
Jharkhand	19	0	2	17
Ladakh	15	11	0	4
Andaman and Nicobar Islands	11	10	0	1
Goa	7	5	0	2
Puducherry	7	1	0	6
Manipur	2	1	0	1
Tripura	2	0	0	2
Mizoram	1	0	0	1
Arunachal Pradesh	1	0	0	1
Dadra and Nagar Haveli	1	0	0	1
Nagaland	1	0	0	1
Daman and Diu	0	0	0	0
Lakshadweep	0	0	0	0
Meghalaya	0	0	0	0
Sikkim	0	0	0	0

df\_india[df\_india['deaths'] > 0].style.background\_gradient(cmap='Wistia')

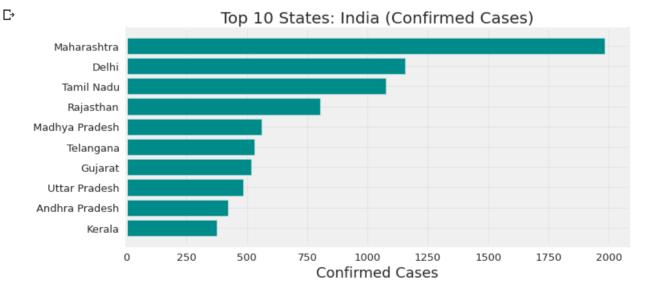
₽

#### confirmed recovered deaths active

state				
Maharashtra	1982	217	149	1616
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West Bengal	134	19	7	108
Bihar	64	26	1	37
Odisha	54	12	1	41
Himachal Pradesh	32	12	2	18
Assam	29	0	1	28
Jharkhand	19	0	2	17

```
f = plt.figure(figsize=(10,5))
f.add_subplot(111)

plt.axes(axisbelow=True)
plt.barh(df_india.sort_values('confirmed')["confirmed"].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.sort_values('confirmed')["confirmed'].index[-10:],df_india.
```

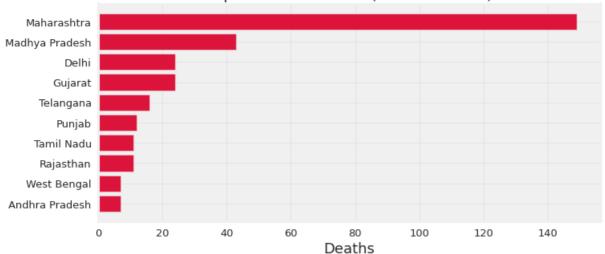


```
f = plt.figure(figsize=(10,5))
f.add_subplot(111)

plt.axes(axisbelow=True)
plt.barh(df_india.sort_values('deaths')["deaths"].index[-10:],df_india.sort_values('deaths')["deaths"].vaplt.tick_params(size=5,labelsize = 13)
plt.xlabel("Deaths",fontsize=18)
plt.title("Top 10 States: India (Deaths Cases)",fontsize=20)
plt.grid(alpha=0.3)
#plt.savefig(out+'Top 10 States_India (Deaths Cases).png')
```







## ▼ Merging Data (States, Hospitals, Police officers, Population, Gener Ratio):

```
police = pd.read_excel('/content/datafile.xls')

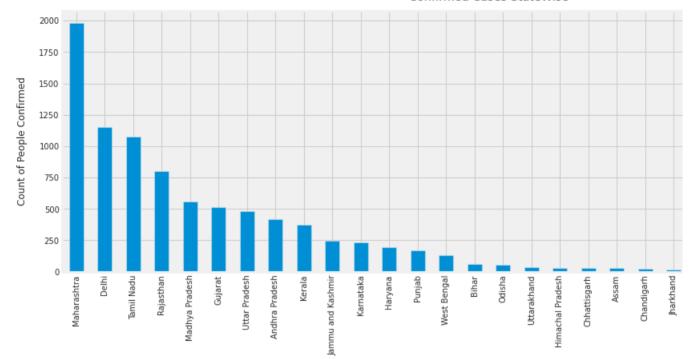
    ₩ARNING *** OLE2 inconsistency: SSCS size is 0 but SSAT size is non-zero

merged_df = population.merge(police, how = 'inner', right_on='State/UT', left_on='State / Union Territory
merged df.drop(['Sno','State / Union Territory','S1.No'], axis = 1, inplace=True)
merged df = merged df.fillna(0)
merged_df['Total Police'] = merged_df['Sanctioned strength of Police Personnel']
cols = ['Total Police per lakh of population - Sanctioned','Total Police per lakh of population - Actual
state_features = merged_df
india_data_unofficial = requests.get('https://api.rootnet.in/covid19-in/unofficial/covid19india.org').jsc
raw_patient_data = pd.io.json.json_normalize(india_data_unofficial['data']['rawPatientData'])
   /usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:2: FutureWarning: pandas.io.json.json_n
raw patient data = raw patient data.drop(['contractedFrom', 'nationality', 'place_attributes', 'relationship
hospitals_unofficial = requests.get('https://api.rootnet.in/covid19-in/stats/hospitals').json()
hospital data = pd.io.json.json normalize(hospitals_unofficial['data']['regional'])
/y /usr/local/lib/python3.6/dist-packages/ipykernel launcher.py:2: FutureWarning: pandas.io.json.json n
hospital_data['state'] = hospital_data['state'].str.replace('&', 'and')
merged df_individual = merged df.merge(raw_patient_data, how = 'inner', left_on='State/UT', right_on='state/UT',
merged_df_individual = merged_df_individual.merge(hospital_data, how = 'inner', left_on = 'State/UT', riq
```

```
merged_dr_individual['State/UT'].value_counts().plot(kind='bar', figsize=(17, 6))
plt.xlabel("States", labelpad=14)
plt.ylabel("Count of People Confirmed", labelpad=14)
plt.title("Confirmed Cases StateWise", y=1.02)
plt.show()
```



#### Confirmed Cases StateWise



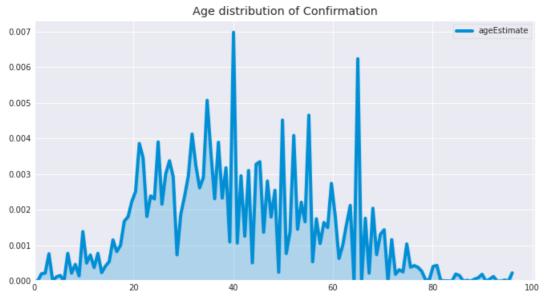
States

```
merged_df_individual.drop(cols, axis =1 , inplace=True)

merged_df_individual.drop(['asOn','notes','state_x','state_y'], axis=1, inplace=True)

merged_df_individual['ageEstimate'].fillna(0)
merged_df_individual['ageEstimate'] = merged_df_individual['ageEstimate'].replace('28-35','30')
merged_df_individual['ageEstimate'] = merged_df_individual['ageEstimate'].replace('','0')
#merged_df_individual['ageEstimate'].astype(int)
plt.figure(figsize=(10,6))
sns.set_style("darkgrid")
plt.title("Age distribution of Confirmation")
sns.kdeplot(data=merged_df_individual['ageEstimate'].astype(float), shade=True).set(xlim=(0))
```

₽



```
merged_df_individual['reportedOn'] = pd.to_datetime(merged_df_individual['reportedOn'], dayfirst=True)
merged_df_individual = merged_df_individual.sort_values(by='reportedOn')
state_features = state_features.merge(hospital_data, how = 'inner', left_on='State/UT', right_on='state'
statewise_unofficial = requests.get('https://api.rootnet.in/covid19-in/unofficial/covid19india.org/statesstatewise_patient_data = pd.io.json.json_normalize(statewise_unofficial['data']['history'])
```

\_\_\_\_ /usr/local/lib/python3.6/dist-packages/ipykernel\_launcher.py:2: FutureWarning: pandas.io.json\_n

```
cols_date_state = statewise_patient_data['day'],statewise_patient_data['statewise']
date_state = pd.DataFrame(cols_date_state)

date_state = date_state.T
date_state
```

 $\Box$ 

```
2020-03-14
                            [{'state': 'Kerala', 'confirmed': 19, 'recover...
        0
            2020-03-15
                           [{'state': 'Maharashtra', 'confirmed': 31, 're...
        1
        2
            2020-03-16
                           [{'state': 'Maharashtra', 'confirmed': 38, 're...
            2020-03-17
                           [{'state': 'Maharashtra', 'confirmed': 41, 're...
        3
        4
            2020-03-18
                           [{'state': 'Maharashtra', 'confirmed': 43, 're...
        5
            2020-03-19
                           [{'state': 'Maharashtra', 'confirmed': 49, 're...
        6
            2020-03-20
                           [{'state': 'Maharashtra', 'confirmed': 52, 're...
        7
            2020-03-21
                           [{'state': 'Maharashtra', 'confirmed': 64, 're...
        8
            2020-03-22
                           [{'state': 'Maharashtra', 'confirmed': 74, 're...
            2020-03-23
                           [{'state': 'Maharashtra', 'confirmed': 97, 're...
        9
        10
            2020-03-24
                           [{'state': 'Kerala', 'confirmed': 109, 'recove...
        11
            2020-03-25
                           [{'state': 'Kerala', 'confirmed': 118, 'recove...
       12
            2020-03-26
                           [{'state': 'Kerala', 'confirmed': 137, 'recove...
       13
            2020-03-27
                           [{'state': 'Kerala', 'confirmed': 176, 'recove...
       14
            2020-03-28
                           [{'state': 'Kerala', 'confirmed': 182, 'recove...
       15
            2020-03-29
                           [{'state': 'Kerala', 'confirmed': 202, 'recove...
        16
            2020-03-30
                           [{'state': 'Maharashtra', 'confirmed': 238, 'r...
            2020-03-31
                           [{'state': 'Maharashtra', 'confirmed': 302, 'r...
       17
            2020-04-01
        18
                           [{'state': 'Maharashtra', 'confirmed': 335, 'r...
       19
            2020-04-02
                           [{'state': 'Maharashtra', 'confirmed': 423, 'r...
            2020-04-03
       20
                           [{'state': 'Maharashtra', 'confirmed': 490, 'r...
            2020-04-04
                           [{'state': 'Maharashtra', 'confirmed': 635, 'r...
       21
                           [{'state': 'Maharashtra', 'confirmed': 748, 'r...
       22
            2020-04-05
       23
            2020-04-06
                           [{'state': 'Maharashtra', 'confirmed': 868, 'r...
       24
            2020-04-07
                          [{'state': 'Maharashtra', 'confirmed': 1018, '...
                          [{'state': 'Maharashtra', 'confirmed': 1135, '...
       25
            2020-04-08
       26
            2020-04-09
                          [{'state': 'Maharashtra', 'confirmed': 1364, '...
       27
            2020-04-10
                          [{'state': 'Maharashtra', 'confirmed': 1574, '...
       28
            2020-04-11
                          [{'state': 'Maharashtra', 'confirmed': 1761, '...
            2020-04-12 [{'state': 'Maharashtra', 'confirmed': 1982, '...
date_state['state'] = date_state['statewise'].to_dict()
output = pd.DataFrame()
i = 0
for i in range(1,len(date_state['day'])):
  output = output.append(date_state['statewise'][i], ignore_index= False)
print(output.head())
                     state
                               confirmed
                                              recovered
                                                             deaths
                                                                         active
 C
      0
             Maharashtra
                                         31
                                                         0
                                                                     0
                                                                              31
                                                         3
                                                                     0
                                                                              21
      1
                    Kerala
                                         2.4
      2
                                         14
                                                         0
                                                                     0
                                                                              14
                  Haryana
      3
          Uttar Pradesh
                                         12
                                                         4
                                                                     0
                                                                                8
                                                         2
                                          7
                                                                                4
                     Delhi
                                                                     1
```

statewise

day

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

```
output['diff'] = output['index'].diff()
```

output

₽		index		confirmed	recovered	deaths	active	diff
	0	<b>0</b> 0 Maharas		31	0	0	31	NaN
	<b>1</b> 1 Kerala		24	3	0	21	1.0	
	2	2 Haryana		14	0	0	14	1.0
	3	3	Uttar Pradesh	12	4	0	8	1.0
	4	4	Delhi	7	2	1	4	1.0
	1052	32	Nagaland	1	0	0	1	1.0
	1053	33	Daman and Diu	0	0	0	0	1.0
	1054	34	Lakshadweep	0	0	0	0	1.0
	1055	35	Meghalaya	0	0	0	0	1.0
	1056	36	Sikkim	0	0	0	0	1.0

1057 rows × 7 columns

```
output['day'] = [0]*len(output)
```

```
j=0
for i,v in output.iterrows():
  output['day'][i] = date_state['day'][j]
  if output['diff'][i] <= 0:
    j += 1
    output['day'][i] = date_state['day'][j]</pre>
```

/usr/local/lib/python3.6/dist-packages/ipykernel\_launcher.py:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexi">https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexi</a>
This is separate from the ipykernel package so we can avoid doing imports until
/usr/local/lib/python3.6/dist-packages/pandas/core/indexing.py:671: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexi">https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexi</a> self.\_setitem\_with\_indexer(indexer, value)
/usr/local/lib/python3.6/dist-packages/ipykernel\_launcher.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexi">https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexi</a>

```
date_time_final = output
date_time_final['day'] = pd.to_datetime(date_time_final['day'], dayfirst=True)
date_time_final
```

С⇒

	index	state	confirmed	recovered	deaths	active	diff	day
0	0	Maharashtra	31	0	0	31	NaN	2020-03-14
1	1	Kerala	24	3	0	21	1.0	2020-03-14
2	2	Haryana	14	0	0	14	1.0	2020-03-14
3	3	Uttar Pradesh	12	4	0	8	1.0	2020-03-14
4	4	Delhi	7	2	1	4	1.0	2020-03-14
1052	32	Nagaland	1	0	0	1	1.0	2020-04-11
1053	33	Daman and Diu	0	0	0	0	1.0	2020-04-11
1054	34	Lakshadweep	0	0	0	0	1.0	2020-04-11
1055	35	Meghalaya	0	0	0	0	1.0	2020-04-11
1056	36	Sikkim	0	0	0	0	1.0	2020-04-11

1057 rows × 8 columns

date\_time\_final\_try = date\_time\_final.merge(state\_features, how = 'left', left\_on='state', right\_on='State
date\_time\_final\_try

 $\Box$ 

	index	state_x	confirmed	recovered	deaths	active	diff	day	Population	Rural population	poj
0	0	Maharashtra	31	0	0	31	NaN	2020- 03-14	112374333.0	61556074.0	51
1	1	Kerala	24	3	0	21	1.0	2020- 03-14	33406061.0	17471135.0	1!
2	2	Haryana	14	0	0	14	1.0	2020- 03-14	25351462.0	16509359.0	1
3	3	Uttar Pradesh	12	4	0	8	1.0	2020- 03-14	199812341.0	155317278.0	4
4	4	Delhi	7	2	1	4	1.0	2020- 03-14	16787941.0	419042.0	1(
1052	32	Nagaland	1	0	0	1	1.0	2020- 04-11	1978502.0	1407536.0	
1053	33	Daman and Diu	0	0	0	0	1.0	2020- 04-11	NaN	NaN	
1054	34	Lakshadweep	0	0	0	0	1.0	2020- 04-11	64473.0	14141.0	
1055	35	Meghalaya	0	0	0	0	1.0	2020- 04-11	2966889.0	2371439.0	
1056	36	Sikkim	0	0	0	0	1.0	2020- 04-11	610577.0	456999.0	

1057 rows  $\times$  27 columns

```
date time final try['confirm time diff statewise'].sum()
 [→ 9211.0
date time final try = date time final try.fillna(0)
date_time_final_try[date_time_final_try['State/UT'] == 'Rajasthan']['confirm_time_diff_statewise'].sum()
 □→ 804.0
ls_final_drop = ['index','state x','active','diff','Total Police per lakh of population - Sanctioned','Total Police per lakh of police per lakh of police per l
te_time_final_try.drop(cols_final_drop, inplace=True, axis = 1)
date_time_final_try['Area'] = date_time_final_try['Area'].str.split(n=1).str[0]
date time final_try['Area'] = date_time final_try['Area'].str.replace(',','')
date time final try['Density'] = date time final try['Density'].str.split(n=1).str[0]
date time final try['Density'] = date time final try['Density'].str.replace(',','')
date_time_final_try['Density'] = date_time_final_try['Density'].str.split('/').str[0]
date_time_final_try.fillna(0, inplace=True)
date time final try['Area'] = date time final try['Area'].astype(int)
date time final try['Density'] = date time final try['Density'].astype(int)
date time final try.info()
 <class 'pandas.core.frame.DataFrame'>
        Int64Index: 1057 entries, 0 to 1056
        Data columns (total 21 columns):
         # Column
                                                                     Non-Null Count Dtype
        ---
                                                                    1057 non-null int64
1057 non-null int64
                confirmed
         1
                recovered
                                                                    1057 non-null int64
         2
               deaths
               day
                                                                   1057 non-null datetime64[ns]
               Population
                                                                   1057 non-null float64
          4
                                                                   1057 non-null float64
1057 non-null float64
1057 non-null int64
          5
               Rural population
                Urban population
          6
               Area
               Density
                                                                   1057 non-null int64
                                                                   1057 non-null float64
          9
               Gender Ratio
          10 State/UT
                                                                   1057 non-null object
                                                                   1057 non-null
1057 non-null
          11
                Total Police
                                                                                                 float64
                ruralHospitals
          12
                                                                                                 float64
          13 ruralBeds
                                                                   1057 non-null float64
          14 urbanHospitals
                                                                   1057 non-null float64
                                                                   1057 non-null float64
          15 urbanBeds
          16 totalHospitals
                                                                    1057 non-null float64
          17
                totalBeds
                                                                     1057 non-null
                                                                                                  float64
         18 asOn
                                                                     1057 non-null object
         19 confirm time diff statewise 1057 non-null
                                                                                                 float64
         20 death_time_diff_statewise 1057 non-null
                                                                                                float64
        dtypes: datetime64[ns](1), float64(13), int64(5), object(2)
        memory usage: 181.7+ KB
date = '2020-03-14'
date = datetime.strptime(date, '%Y-%m-%d')
date_time_final_try['Day'] = date_time_final_try['day'] - date
date_time_final_try['Day'] = date_time_final_try['Day'].astype(str).str.split('').str[0]
date_time_final_try['Day'] = date_time_final_try['Day'].astype(int)
data tima final two - data tima final twott/Danulatian/ (Danul acoulation) (Timban manulation) (Aman)
```

```
date_time_rinal_try = date_time_rinal_try[[ ropulation , kural population , orden population , Area ,
date_time_final_try.sort_values(by='Day', inplace=True)
date_time_final_try.reset_index(inplace=True)
```

### Predicting cases and deaths in India using all the statistics

date\_time\_final\_try.fillna(0, inplace= True)

```
X = date_time_final_try.iloc[:,:-2]
Y_death = date_time_final_try.iloc[:,-1]
Y = date_time_final_try.iloc[:,-2]
X['State/UT'] = X['State/UT'].astype(str)

from sklearn.preprocessing import OneHotEncoder
# creating instance of one-hot-encoder
enc = OneHotEncoder(handle_unknown='ignore')
# passing bridge-types-cat column (label encoded values of bridge_types)
enc_df = pd.DataFrame(enc.fit_transform(X[['State/UT']]).toarray())
# merge with main df bridge_df on key values
X = X.join(enc_df)

X.drop(['State/UT','index'], inplace=True, axis = 1)
```

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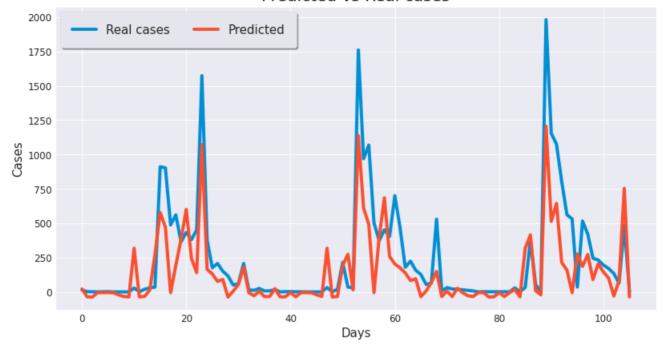
₽		Population	Rural population	Urban population	Area	Density	Gender Ratio	Total Police	ruralHospitals	ruralBeds
	0	112374333.0	61556074.0	50818259.0	307713	365	929.0	220126.0	273.0	12398.0
	1	3673917.0	2712464.0	961453.0	10486	350	960.0	12537.0	99.0	1140.0
	2	91276115.0	62183113.0	29093002.0	88752	1029	953.0	107777.0	1272.0	19684.0
	3	1458545.0	551731.0	906814.0	3702	394	973.0	5630.0	17.0	1405.0
	4	1383727.0	1066358.0	317369.0	83743	17	938.0	8538.0	208.0	2136.0
	1052	27743338.0	17344192.0	10399146.0	50362	550	895.0	68902.0	510.0	5805.0
	1053	91276115.0	62183113.0	29093002.0	88752	1029	953.0	107777.0	1272.0	19684.0
	1054	104099452.0	92341436.0	11758016.0	94163	1102	918.0	92422.0	930.0	6083.0
	1055	199812341.0	155317278.0	44495063.0	240928	828	912.0	377009.0	4442.0	39104.0
	1056	610577.0	456999.0	153578.0	7096	86	890.0	2482.0	24.0	260.0

1057 rows × 46 columns

```
X test = X test[X test['Population'] == 49577103.0]
y_test = date_time_final_try['confirmed'].iloc[X_test.index]
poly = PolynomialFeatures(degree= 4)
poly X train = poly.fit transform(X train)
poly X test = poly.fit transform(X test)
# Transform our death data for polynomial regression
poly_death = PolynomialFeatures(degree= 4)
poly_X_train_death = poly_death.fit_transform(X_train_death)
poly_X_test_death = poly_death.fit_transform(X_test_death)
# polynomial regression cases
linear model = LinearRegression()
linear_model.fit(poly_X_train, y_train)
test_linear_pred = linear_model.predict(poly_X_test)
\# evaluating with MAE and MSE
print('MAE:', mean_absolute_error(test_linear_pred, y_test))
print('MSE:',mean_squared_error(test_linear_pred, y_test))
MAE: 142.1817589889108
    MSE: 52873.184687062814
plt.figure(figsize=(12,7))
plt.plot(y_test, label = "Real cases")
plt.plot(test_linear_pred, label = "Predicted")
plt.title("Predicted vs Real cases", size = 20)
plt.xlabel('Days', size = 15)
plt.ylabel('Cases', size = 15)
plt.xticks(size=12)
plt.yticks(size=12)
# defyning legend config
plt.legend(loc = "upper left"
           , frameon = True
           , ncol = 2
           , fancybox = True
           , framealpha = 0.95
           , shadow = True
           , borderpad = 1
           , prop={'size': 15});
```

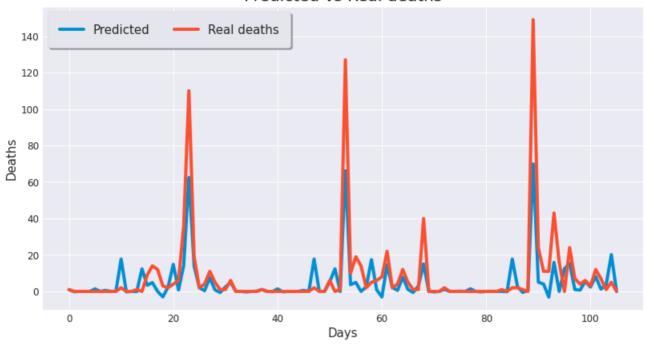
С→

### Predicted vs Real cases



```
linear_model_death = LinearRegression(fit_intercept=False)
linear_model_death.fit(poly_X_train_death, y_train_death)
test linear pred death = linear model death.predict(poly X test death)
# evaluating with MAE and MSE
print('MAE:', mean_absolute_error(test_linear_pred_death, y_test_death))
print('MSE:',mean_squared_error(test_linear_pred_death, y_test_death))
    MAE: 5.835305523828821
    MSE: 168.76949582190258
plt.figure(figsize=(12,7))
plt.plot(test_linear_pred_death, label = "Predicted")
plt.plot(y_test_death, label = "Real deaths")
plt.title("Predicted vs Real deaths", size = 20)
plt.xlabel('Days', size = 15)
plt.ylabel('Deaths', size = 15)
plt.xticks(size=12)
plt.yticks(size=12)
# defyning legend config
plt.legend(loc = "upper left"
           , frameon = True
           , ncol = 2
           , fancybox = True
           , framealpha = 0.95
           , shadow = True
           , borderpad = 1
           , prop={'size': 15});
```

#### Predicted vs Real deaths

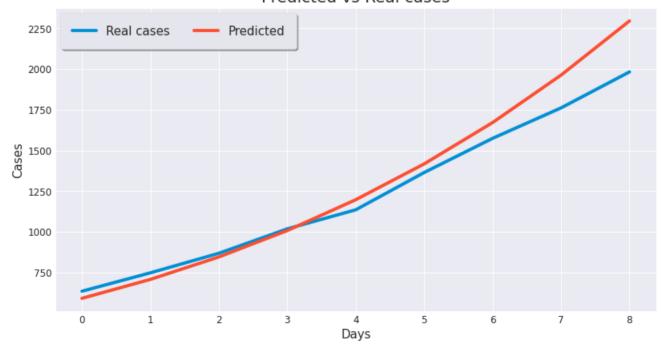


Predicting Cases and Deaths in Maharashtra using only the number of cases and deatl

```
X = date time final try.iloc[:,:-2]
Y death = date time final try.iloc[:,-1]
Y = date_time_final_try.iloc[:,-2]
date time final try m = date time final try[date time final try['State/UT'] == 'Maharashtra']
cases_m = date_time_final_try_m['confirmed'].groupby(date_time_final_try['Day']).sum().sort_values(ascence)
deaths_m = date_time_final_try_m['deaths'].groupby(date_time_final_try['Day']).sum().sort_values(ascendi
maharasthra_cases = np.array(cases_m).reshape(-1, 1)
maharashtra_deaths = np.array(deaths_m).reshape(-1, 1)
days since first case m = np.array([i for i in range(len(cases m.index))]).reshape(-1, 1)
days_since_first_death_m = np.array([i for i in range(len(deaths_m.index))]).reshape(-1, 1)
days in future = 15
future_forcast m = np.array([i for i in range(len(cases m.index)+days in future)]).reshape(-1, 1)
adjusted_dates = future_forcast_m[:-15]
future_forcast_deaths m = np.array([i for i in range(len(deaths m.index)+days in_future)]).reshape(-1, 1
adjusted_dates_deaths = future_forcast_deaths_m[:-15]
X train, X test, y train, y test = train_test_split(days_since_first_case_m
                                                     , maharasthra cases
                                                    , test size=0.30
                                                    , shuffle=False)
X_train_death, X_test_death, y_train_death, y_test_death = train_test_split(days_since_first_death_m
                                                    , maharashtra_deaths
                                                    , test_size=0.3
                                                    , shuffle=False
                                                    , random_state = 42)
poly = PolynomialFeatures(degree= 4)
poly_X_train = poly.fit_transform(X_train)
poly X_test = poly.fit_transform(X_test)
poly_future_forcast = poly.fit_transform(future_forcast_m)
# Transform our death data for polynomial regression
```

```
poly_death = PolynomialFeatures(degree= 4)
poly X train death = poly death.fit transform(X train death)
poly_X_test_death = poly_death.fit_transform(X_test_death)
poly future forcast death = poly death.fit transform(future forcast deaths_m)
# polynomial regression cases
linear model = LinearRegression()
linear_model.fit(poly_X_train, y_train)
test_linear_pred = linear_model.predict(poly_X_test)
linear_pred = linear_model.predict(poly_future_forcast)
\# evaluating with MAE and MSE
print('MAE:', mean_absolute_error(test_linear_pred, y_test))
print('MSE:',mean squared_error(test_linear pred, y test))
    MAE: 93.74065935496594
    MSE: 17646.22555561457
plt.figure(figsize=(12,7))
plt.plot(y_test, label = "Real cases")
plt.plot(test_linear_pred, label = "Predicted")
plt.title("Predicted vs Real cases", size = 20)
plt.xlabel('Days', size = 15)
plt.ylabel('Cases', size = 15)
plt.xticks(size=12)
plt.yticks(size=12)
# defyning legend config
plt.legend(loc = "upper left"
           , frameon = True
           , ncol = 2
           , fancybox = True
           , framealpha = 0.95
           , shadow = True
           , borderpad = 1
           , prop={'size': 15});
```

### Predicted vs Real cases



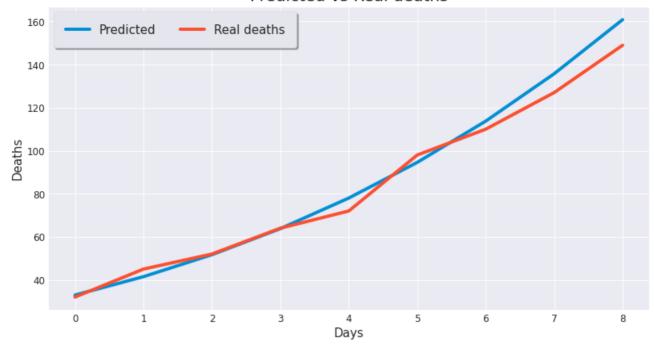
```
plt.figure(figsize=(16, 9))
plt.plot(adjusted_dates
    , maharasthra_cases
    , label = "Real cases")
```

```
plt.plot(future_forcast_m
         , linear pred
         , label = "Polynomial Regression Predictions"
         , linestyle='dashed'
         , color='orange')
plt.title('Cases in Maharashtra over the time: Predicting Next 2 Weeks', size=30)
plt.xlabel('Days Since 03/14/20', size=30)
plt.ylabel('Cases', size=30)
plt.xticks(size=20)
plt.yticks(size=20)
plt.axvline(31, color='black'
            , linestyle="--"
            , linewidth=1)
plt.text(18, 5000
         , "model training"
         , size = 15
         , color = "black")
plt.text(31.2, 15000
        , "prediction"
         , size = 15
         , color = "black")
# defyning legend config
plt.legend(loc = "upper left"
          , frameon = True
           , ncol = 2
           , fancybox = True
           , framealpha = 0.95
           , shadow = True
           , borderpad = 1
           , prop={'size': 15})
plt.show();
₽
```

## Cases in Maharashtra over the time: Predictin 16000 - Polynomial Regression Predictions pre Real cases 14000 12000 10000 8000 6000 model training 4000 2000 0 30 0 10 Days Since 03/14/20

```
linear model death = LinearRegression(fit intercept=False)
linear model death.fit(poly X train death, y train death)
test linear pred death = linear model death.predict(poly X test death)
linear pred death = linear model death.predict(poly future forcast death)
# evaluating with MAE and MSE
print('MAE:', mean_absolute_error(test_linear_pred_death, y_test_death))
print('MSE:',mean_squared_error(test_linear_pred_death, y_test_death))
MAE: 4.312663698953685
    MSE: 32.603175641518675
plt.figure(figsize=(12,7))
plt.plot(test_linear_pred_death, label = "Predicted")
plt.plot(y test death, label = "Real deaths")
plt.title("Predicted vs Real deaths", size = 20)
plt.xlabel('Days', size = 15)
plt.ylabel('Deaths', size = 15)
plt.xticks(size=12)
plt.yticks(size=12)
# defyning legend config
plt.legend(loc = "upper left"
           , frameon = True
           , ncol = 2
           , fancybox = True
           , framealpha = 0.95
           , shadow = True
           , borderpad = 1
           , prop={'size': 15});
```

### Predicted vs Real deaths

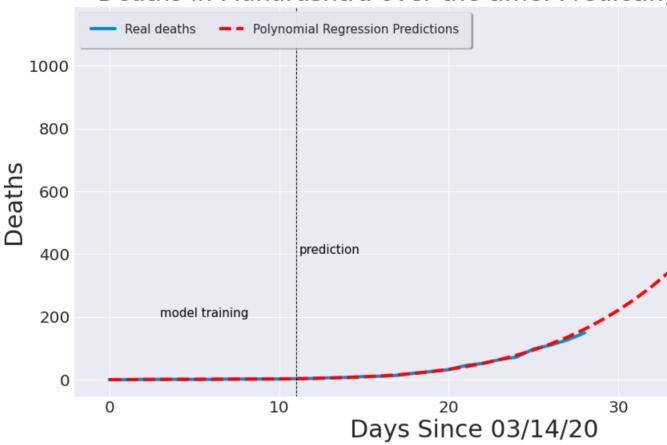


```
plt.figure(figsize=(16, 9))
plt.plot(adjusted_dates_deaths
         , maharashtra_deaths
         , label = "Real deaths")
plt.plot(future_forcast_deaths_m
         , linear_pred_death
         , label = "Polynomial Regression Predictions"
         , linestyle='dashed'
         , color='red')
plt.title('Deaths in Maharashtra over the time: Predicting Next 2 Weeks', size=30)
plt.xlabel('Days Since 03/14/20', size=30)
plt.ylabel('Deaths', size=30)
plt.xticks(size=20)
plt.yticks(size=20)
plt.axvline(11, color='black'
            , linestyle="--"
            , linewidth=1)
plt.text(3, 200
         , "model training"
         , size = 15
         , color = "black")
plt.text(11.2, 400
         , "prediction"
         , size = 15
         , color = "black")
# defyning legend config
plt.legend(loc = "upper left"
           , frameon = True
           , ncol = 2
           , fancybox = True
           , framealpha = 0.95
           , shadow = True
           , borderpad = 1
           , prop={'size': 15})
```

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## Deaths in Maharashtra over the time: Predicting



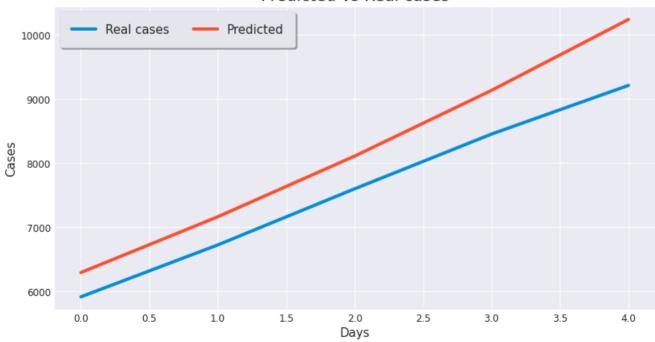
▼ Predicting Cases and Deaths in India using only the number of cases and deaths per d

```
cases = date_time_final_try['confirmed'].groupby(date_time_final_try['Day']).sum().sort_values(ascending-
deaths = date_time_final_try['deaths'].groupby(date_time_final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sum().sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(ascending=final_try['Day']).sort_values(
India_cases = np.array(cases).reshape(-1, 1)
India_deaths = np.array(deaths).reshape(-1, 1)
days_since_first_case = np.array([i for i in range(len(cases.index))]).reshape(-1, 1)
days_since_first_death = np.array([i for i in range(len(deaths.index))]).reshape(-1, 1)
days_in_future = 15
future_forcast = np.array([i for i in range(len(cases.index)+days_in_future)]).reshape(-1, 1)
adjusted dates = future forcast[:-15]
future_forcast_deaths = np.array([i for i in range(len(deaths.index)+days_in_future)]).reshape(-1, 1)
adjusted dates_deaths = future_forcast_deaths[:-15]
X_train, X_test, y_train, y_test = train_test_split(days_since_first_case
                                                                                                                                      , India cases
                                                                                                                                       , test_size=0.15
                                                                                                                                      , shuffle=False)
X_train_death, X_test_death, y_train_death, y_test_death = train_test_split(days_since_first_death
                                                                                                                                      , India_deaths
                                                                                                                                      , test_size=0.15
                                                                                                                                      , shuffle=False
                                                                                                                                      , random_state = 42)
poly = PolynomialFeatures(degree= 4)
poly X train = poly.fit transform(X train)
poly X_test = poly.fit_transform(X_test)
poly future forcast = poly.fit transform(future forcast)
```

```
# Transform our death data for polynomial regression
poly_death = PolynomialFeatures(degree= 4)
poly X train death = poly death.fit transform(X train death)
poly X test death = poly death.fit transform(X test death)
poly_future_forcast_death = poly_death.fit_transform(future_forcast_deaths)
# polynomial regression cases
linear_model = LinearRegression()
linear_model.fit(poly_X_train, y_train)
test_linear_pred = linear_model.predict(poly_X_test)
linear_pred = linear_model.predict(poly_future_forcast)
# evaluating with MAE and MSE
print('MAE:', mean absolute error(test linear pred, y test))
print('MSE:',mean_squared_error(test_linear_pred, y_test))
    MAE: 607.1298125673926
    MSE: 423154.524667917
plt.figure(figsize=(12,7))
plt.plot(y_test, label = "Real cases")
plt.plot(test_linear_pred, label = "Predicted")
plt.title("Predicted vs Real cases", size = 20)
plt.xlabel('Days', size = 15)
plt.ylabel('Cases', size = 15)
plt.xticks(size=12)
plt.yticks(size=12)
# defyning legend config
plt.legend(loc = "upper left"
           , frameon = True
           , ncol = 2
           , fancybox = True
           , framealpha = 0.95
           , shadow = True
           , borderpad = 1
           , prop={'size': 15});
```

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### Predicted vs Real cases

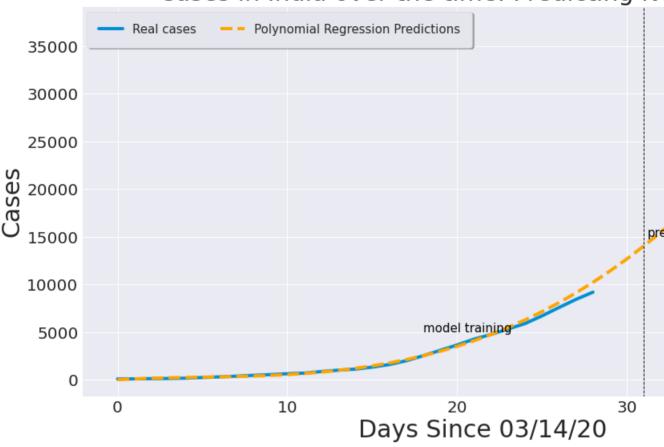


plt.figure(figsize=(16, 9))

plt.plot(adjusted dates

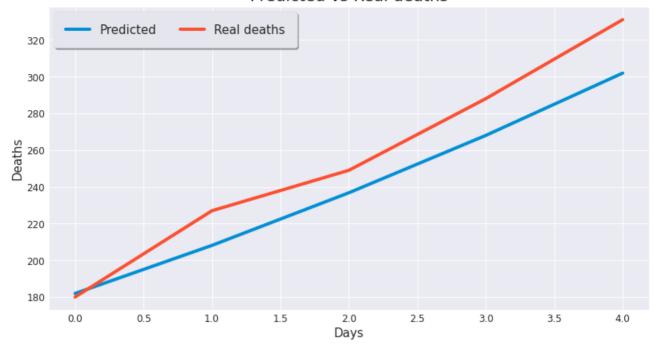
```
Judia_cases
         , label = "Real cases")
plt.plot(future_forcast
        , linear_pred
         , label = "Polynomial Regression Predictions"
         , linestyle='dashed'
         , color='orange')
plt.title('Cases in India over the time: Predicting Next 2 Weeks', size=30)
plt.xlabel('Days Since 03/14/20', size=30)
plt.ylabel('Cases', size=30)
plt.xticks(size=20)
plt.yticks(size=20)
plt.axvline(31, color='black'
            , linestyle="--"
            , linewidth=1)
plt.text(18, 5000
         , "model training"
         , size = 15
         , color = "black")
plt.text(31.2, 15000
        , "prediction"
         , size = 15
         , color = "black")
# defyning legend config
plt.legend(loc = "upper left"
          , frameon = True
          , ncol = 2
          , fancybox = True
          , framealpha = 0.95
          , shadow = True
          , borderpad = 1
           , prop={'size': 15})
plt.show();
С⇒
```

## Cases in India over the time: Predicting Ne



```
linear model death = LinearRegression(fit intercept=False)
linear model death.fit(poly X train death, y train death)
test linear pred death = linear model death.predict(poly X test death)
linear pred death = linear model death.predict(poly future forcast death)
# evaluating with MAE and MSE
print('MAE:', mean_absolute_error(test_linear_pred_death, y_test_death))
print('MSE:',mean_squared_error(test_linear_pred_death, y_test_death))
MAE: 16.430685843913515
    MSE: 351.21935863785046
plt.figure(figsize=(12,7))
plt.plot(test_linear_pred_death, label = "Predicted")
plt.plot(y test death, label = "Real deaths")
plt.title("Predicted vs Real deaths", size = 20)
plt.xlabel('Days', size = 15)
plt.ylabel('Deaths', size = 15)
plt.xticks(size=12)
plt.yticks(size=12)
# defyning legend config
plt.legend(loc = "upper left"
           , frameon = True
           , ncol = 2
           , fancybox = True
           , framealpha = 0.95
           , shadow = True
           , borderpad = 1
           , prop={'size': 15});
```

### Predicted vs Real deaths



```
plt.figure(figsize=(16, 9))
plt.plot(adjusted_dates_deaths
         , India_deaths
         , label = "Real deaths")
plt.plot(future_forcast_deaths
         , linear_pred_death
         , label = "Polynomial Regression Predictions"
         , linestyle='dashed'
         , color='red')
plt.title('Deaths in India over the time: Predicting Next 2 Weeks', size=30)
plt.xlabel('Days Since 03/14/20', size=30)
plt.ylabel('Deaths', size=30)
plt.xticks(size=20)
plt.yticks(size=20)
plt.axvline(11, color='black'
            , linestyle="--"
            , linewidth=1)
plt.text(3, 200
         , "model training"
         , size = 15
         , color = "black")
plt.text(11.2, 400
         , "prediction"
         , size = 15
         , color = "black")
# defyning legend config
plt.legend(loc = "upper left"
           , frameon = True
           , ncol = 2
           , fancybox = True
           , framealpha = 0.95
           , shadow = True
           , borderpad = 1
           , prop={'size': 15})
```

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# Deaths in India over the time: Predicting $N\epsilon$

