

covid-19

July 6, 2024

```
[1]: import pandas as pd
```

```
[2]: covid = pd.read_csv('covid.csv')
```

```
[3]: #Having a glance at some of the records  
covid.head()
```

```
[3]: iso_code location      date  total_cases  new_cases  total_deaths  \  
0      ABW      Aruba  2020-03-13           2           2           0  
1      ABW      Aruba  2020-03-20           4           2           0  
2      ABW      Aruba  2020-03-24          12           8           0  
3      ABW      Aruba  2020-03-25          17           5           0  
4      ABW      Aruba  2020-03-26          19           2           0  
  
      new_deaths  total_cases_per_million  new_cases_per_million  \  
0              0              18.733              18.733  
1              0              37.465              18.733  
2              0             112.395             74.930  
3              0             159.227             46.831  
4              0             177.959             18.733  
  
      total_deaths_per_million  ...  aged_65_older  aged_70_older  \  
0              0.0  ...             13.085             7.452  
1              0.0  ...             13.085             7.452  
2              0.0  ...             13.085             7.452  
3              0.0  ...             13.085             7.452  
4              0.0  ...             13.085             7.452  
  
      gdp_per_capita  extreme_poverty  cvd_death_rate  diabetes_prevalence  \  
0      35973.781              NaN              NaN              11.62  
1      35973.781              NaN              NaN              11.62  
2      35973.781              NaN              NaN              11.62  
3      35973.781              NaN              NaN              11.62  
4      35973.781              NaN              NaN              11.62  
  
      female_smokers  male_smokers  handwashing_facilities  hospital_beds_per_100k  
0              NaN              NaN              NaN              NaN  
1              NaN              NaN              NaN              NaN
```

2	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN
4	NaN	NaN	NaN	NaN

[5 rows x 32 columns]

```
[4]: #Looking at the shape
covid.shape
```

```
[4]: (19496, 32)
```

```
[5]: covid.columns
```

```
[5]: Index(['iso_code', 'location', 'date', 'total_cases', 'new_cases',
        'total_deaths', 'new_deaths', 'total_cases_per_million',
        'new_cases_per_million', 'total_deaths_per_million',
        'new_deaths_per_million', 'total_tests', 'new_tests',
        'total_tests_per_thousand', 'new_tests_per_thousand',
        'new_tests_smoothed', 'new_tests_smoothed_per_thousand', 'tests_units',
        'stringency_index', 'population', 'population_density', 'median_age',
        'aged_65_older', 'aged_70_older', 'gdp_per_capita', 'extreme_poverty',
        'cvd_death_rate', 'diabetes_prevalence', 'female_smokers',
        'male_smokers', 'handwashing_facilities', 'hospital_beds_per_100k'],
        dtype='object')
```

```
[6]: #Looking at the different locations
covid["location"].value_counts()
```

```
[6]: location
Sweden          146
Canada          146
United States   146
Lithuania       146
Brazil          146
...
Yemen           45
Western Sahara  29
Tajikistan      24
Comoros         23
Lesotho         10
Name: count, Length: 212, dtype: int64
```

```
[7]: #Checking if columns have null values
covid.isna().any()
```

```
[7]: iso_code          True
location             False
date                 False
```

total_cases	False
new_cases	False
total_deaths	False
new_deaths	False
total_cases_per_million	True
new_cases_per_million	True
total_deaths_per_million	True
new_deaths_per_million	True
total_tests	True
new_tests	True
total_tests_per_thousand	True
new_tests_per_thousand	True
new_tests_smoothed	True
new_tests_smoothed_per_thousand	True
tests_units	True
stringency_index	True
population	True
population_density	True
median_age	True
aged_65_older	True
aged_70_older	True
gdp_per_capita	True
extreme_poverty	True
cvd_death_rate	True
diabetes_prevalence	True
female_smokers	True
male_smokers	True
handwashing_facilities	True
hospital_beds_per_100k	True
dtype: bool	

```
[8]: #Getting the sum of null values across each column
      covid.isna().sum()
```

```
[8]: iso_code          64
      location         0
      date             0
      total_cases      0
      new_cases        0
      total_deaths     0
      new_deaths       0
      total_cases_per_million  377
      new_cases_per_million   377
      total_deaths_per_million  377
      new_deaths_per_million   377
      total_tests      14332
      new_tests        14904
```

```

total_tests_per_thousand      14332
new_tests_per_thousand        14904
new_tests_smoothed            13866
new_tests_smoothed_per_thousand 13866
tests_units                   13267
stringency_index              4500
population                     64
population_density            850
median_age                    1743
aged_65_olders                1980
aged_70_olders                1832
gdp_per_capita                1982
extreme_poverty               7878
cvd_death_rate                1817
diabetes_prevalence           1174
female_smokers                 5052
male_smokers                   5206
handwashing_facilities        11822
hospital_beds_per_100k        3160
dtype: int64

```

```

[9]: #Getting the cases in India
india_case=covid[covid["location"]=="India"]

```

```

[10]: india_case.head()

```

```

[10]:   iso_code location      date  total_cases  new_cases  total_deaths  \
8379     IND     India  2019-12-31           0           0           0
8380     IND     India  2020-01-01           0           0           0
8381     IND     India  2020-01-02           0           0           0
8382     IND     India  2020-01-03           0           0           0
8383     IND     India  2020-01-04           0           0           0

      new_deaths  total_cases_per_million  new_cases_per_million  \
8379           0                0.0                0.0
8380           0                0.0                0.0
8381           0                0.0                0.0
8382           0                0.0                0.0
8383           0                0.0                0.0

      total_deaths_per_million  ...  aged_65_olders  aged_70_olders  \
8379                0.0  ...         5.989         3.414
8380                0.0  ...         5.989         3.414
8381                0.0  ...         5.989         3.414
8382                0.0  ...         5.989         3.414
8383                0.0  ...         5.989         3.414

```

	gdp_per_capita	extreme_poverty	cvd_death_rate	diabetes_prevalence	\
8379	6426.674	21.2	282.28	10.39	
8380	6426.674	21.2	282.28	10.39	
8381	6426.674	21.2	282.28	10.39	
8382	6426.674	21.2	282.28	10.39	
8383	6426.674	21.2	282.28	10.39	

	female_smokers	male_smokers	handwashing_facilities	\
8379	1.9	20.6	59.55	
8380	1.9	20.6	59.55	
8381	1.9	20.6	59.55	
8382	1.9	20.6	59.55	
8383	1.9	20.6	59.55	

	hospital_beds_per_100k
8379	0.53
8380	0.53
8381	0.53
8382	0.53
8383	0.53

[5 rows x 32 columns]

```
[11]: india_case.tail()
```

	iso_code	location	date	total_cases	new_cases	total_deaths	\
8519	IND	India	2020-05-20	106750	5611	3303	
8520	IND	India	2020-05-21	112359	5609	3435	
8521	IND	India	2020-05-22	118447	6088	3583	
8522	IND	India	2020-05-23	125101	6654	3720	
8523	IND	India	2020-05-24	131868	6767	3867	

	new_deaths	total_cases_per_million	new_cases_per_million	\
8519	140	77.355	4.066	
8520	132	81.419	4.064	
8521	148	85.831	4.412	
8522	137	90.653	4.822	
8523	147	95.556	4.904	

	total_deaths_per_million	...	aged_65_older	aged_70_older	\
8519	2.393	...	5.989	3.414	
8520	2.489	...	5.989	3.414	
8521	2.596	...	5.989	3.414	
8522	2.696	...	5.989	3.414	
8523	2.802	...	5.989	3.414	

	gdp_per_capita	extreme_poverty	cvd_death_rate	diabetes_prevalence	\
--	----------------	-----------------	----------------	---------------------	---

8519	6426.674	21.2	282.28	10.39
8520	6426.674	21.2	282.28	10.39
8521	6426.674	21.2	282.28	10.39
8522	6426.674	21.2	282.28	10.39
8523	6426.674	21.2	282.28	10.39

	female_smokers	male_smokers	handwashing_facilities	\
8519	1.9	20.6	59.55	
8520	1.9	20.6	59.55	
8521	1.9	20.6	59.55	
8522	1.9	20.6	59.55	
8523	1.9	20.6	59.55	

	hospital_beds_per_100k
8519	0.53
8520	0.53
8521	0.53
8522	0.53
8523	0.53

[5 rows x 32 columns]

```
[12]: import seaborn as sns
      from matplotlib import pyplot as plt
```

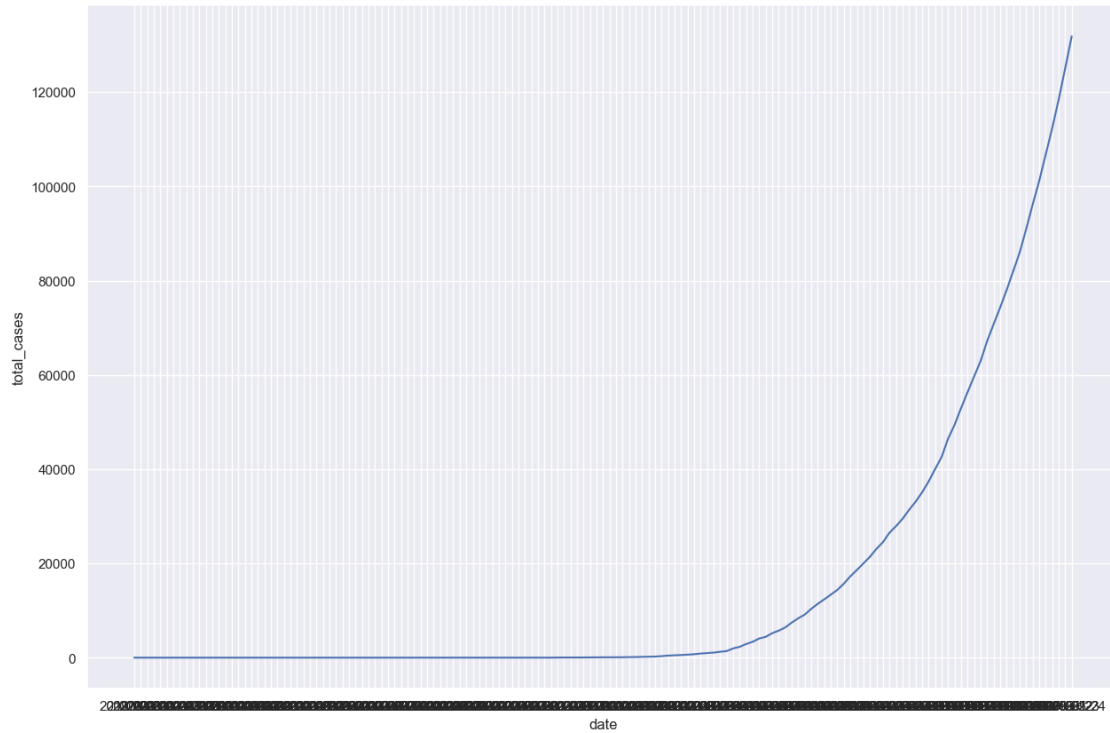
```
[13]: #Total cases per day
      sns.set(rc={'figure.figsize':(15,10)})
      sns.lineplot(x="date",y="total_cases",data=india_case)
      plt.show()
      import warnings
      warnings.filterwarnings('ignore')
```

D:\SOFTWARES\Anaconda\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

```
with pd.option_context('mode.use_inf_as_na', True):
```

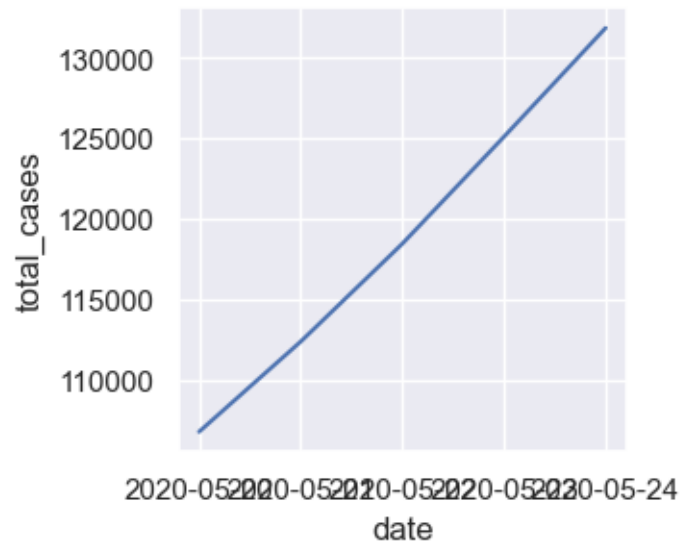
D:\SOFTWARES\Anaconda\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

```
with pd.option_context('mode.use_inf_as_na', True):
```

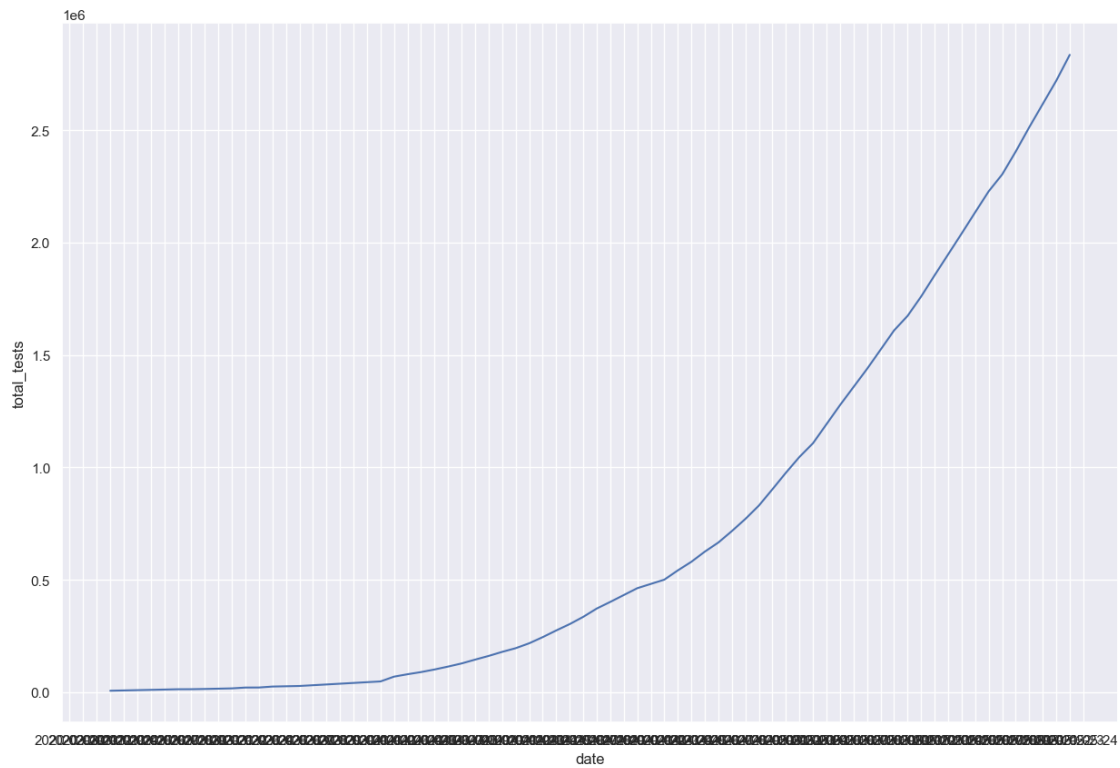


```
[14]: #Making a dataframe for last 5 days
india_last_5_days=india_case.tail()
```

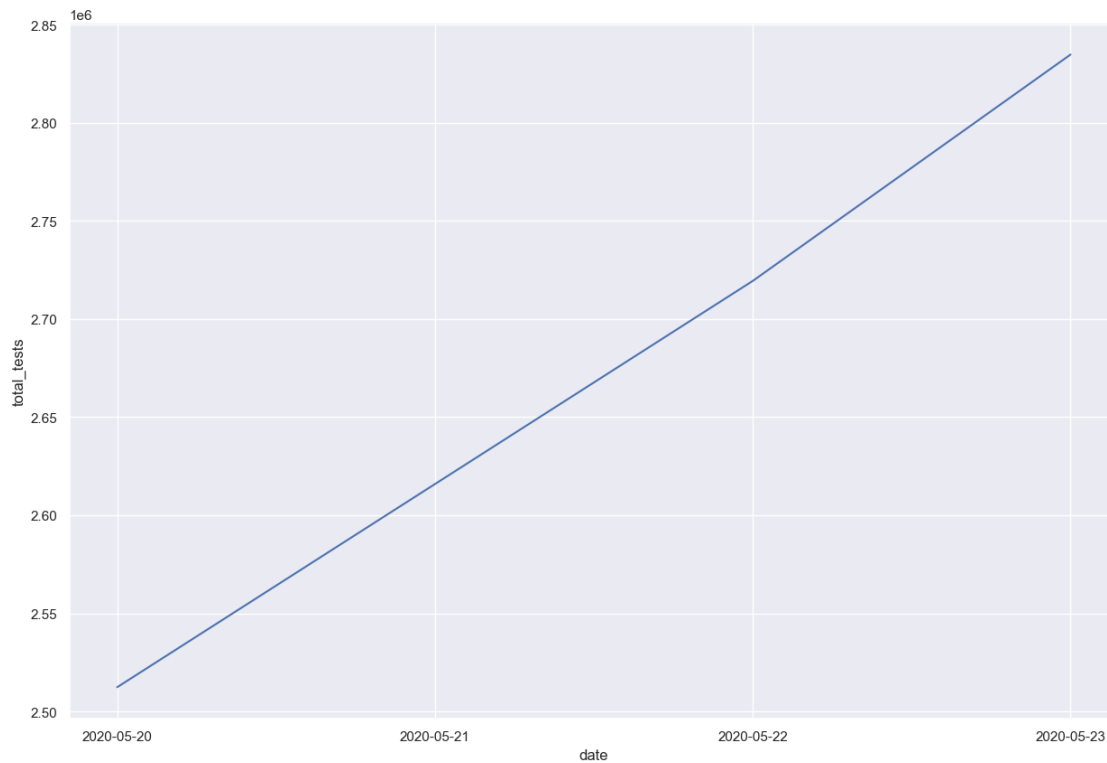
```
[15]: #Total cases in last 5 days
sns.set(rc={'figure.figsize':(3,3)})
sns.lineplot(x="date",y="total_cases",data=india_last_5_days)
plt.show()
```



```
[16]: #Total tests per day
sns.set(rc={'figure.figsize':(15,10)})
sns.lineplot(x="date",y="total_tests",data=india_case)
plt.show()
```



```
[17]: #Total tests in last 5 days
sns.set(rc={'figure.figsize':(15,10)})
sns.lineplot(x="date",y="total_tests",data=india_last_5_days)
plt.show()
```

```
[18]: #Brazil Case
brazil_case=covid[covid["location"]=="Brazil"]
brazil_case.head()
```

```
[18]:
```

	iso_code	location	date	total_cases	new_cases	total_deaths	\
2510	BRA	Brazil	2019-12-31	0	0	0	
2511	BRA	Brazil	2020-01-01	0	0	0	
2512	BRA	Brazil	2020-01-02	0	0	0	
2513	BRA	Brazil	2020-01-03	0	0	0	
2514	BRA	Brazil	2020-01-04	0	0	0	

	new_deaths	total_cases_per_million	new_cases_per_million	\
2510	0	0.0	0.0	
2511	0	0.0	0.0	
2512	0	0.0	0.0	
2513	0	0.0	0.0	
2514	0	0.0	0.0	

	total_deaths_per_million	...	aged_65_older	aged_70_older	\
2510	0.0	...	8.552	5.06	
2511	0.0	...	8.552	5.06	
2512	0.0	...	8.552	5.06	

2513	0.0	...	8.552	5.06
2514	0.0	...	8.552	5.06

	gdp_per_capita	extreme_poverty	cvd_death_rate	diabetes_prevalence	\
2510	14103.452	3.4	177.961	8.11	
2511	14103.452	3.4	177.961	8.11	
2512	14103.452	3.4	177.961	8.11	
2513	14103.452	3.4	177.961	8.11	
2514	14103.452	3.4	177.961	8.11	

	female_smokers	male_smokers	handwashing_facilities	\
2510	10.1	17.9	NaN	
2511	10.1	17.9	NaN	
2512	10.1	17.9	NaN	
2513	10.1	17.9	NaN	
2514	10.1	17.9	NaN	

	hospital_beds_per_100k
2510	2.2
2511	2.2
2512	2.2
2513	2.2
2514	2.2

[5 rows x 32 columns]

```
[19]: brazil_case.tail()
```

```
[19]:
```

	iso_code	location	date	total_cases	new_cases	total_deaths	\
2651	BRA	Brazil	2020-05-20	271628	17408	17971	
2652	BRA	Brazil	2020-05-21	291579	19951	18859	
2653	BRA	Brazil	2020-05-22	310087	18508	20047	
2654	BRA	Brazil	2020-05-23	330890	20803	21048	
2655	BRA	Brazil	2020-05-24	347398	16508	22013	

	new_deaths	total_cases_per_million	new_cases_per_million	\
2651	1179	1277.892	81.897	
2652	888	1371.753	93.861	
2653	1188	1458.825	87.072	
2654	1001	1556.694	97.869	
2655	965	1634.357	77.663	

	total_deaths_per_million	...	aged_65_older	aged_70_older	\
2651	84.546	...	8.552	5.06	
2652	88.723	...	8.552	5.06	
2653	94.312	...	8.552	5.06	
2654	99.022	...	8.552	5.06	

2655	103.562	...	8.552	5.06
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	gdp_per_capita	extreme_poverty	cvd_death_rate	diabetes_prevalence	\
2651	14103.452	3.4	177.961	8.11	
2652	14103.452	3.4	177.961	8.11	
2653	14103.452	3.4	177.961	8.11	
2654	14103.452	3.4	177.961	8.11	
2655	14103.452	3.4	177.961	8.11	

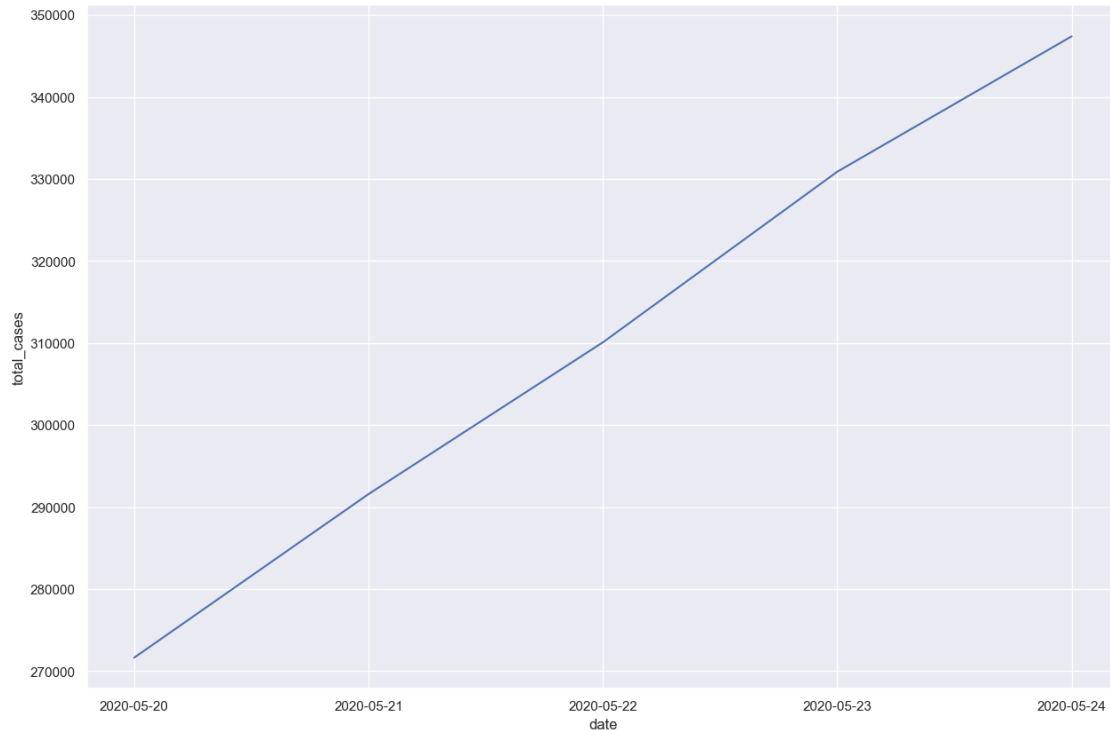
	female_smokers	male_smokers	handwashing_facilities	\
2651	10.1	17.9	NaN	
2652	10.1	17.9	NaN	
2653	10.1	17.9	NaN	
2654	10.1	17.9	NaN	
2655	10.1	17.9	NaN	

	hospital_beds_per_100k
2651	2.2
2652	2.2
2653	2.2
2654	2.2
2655	2.2

[5 rows x 32 columns]

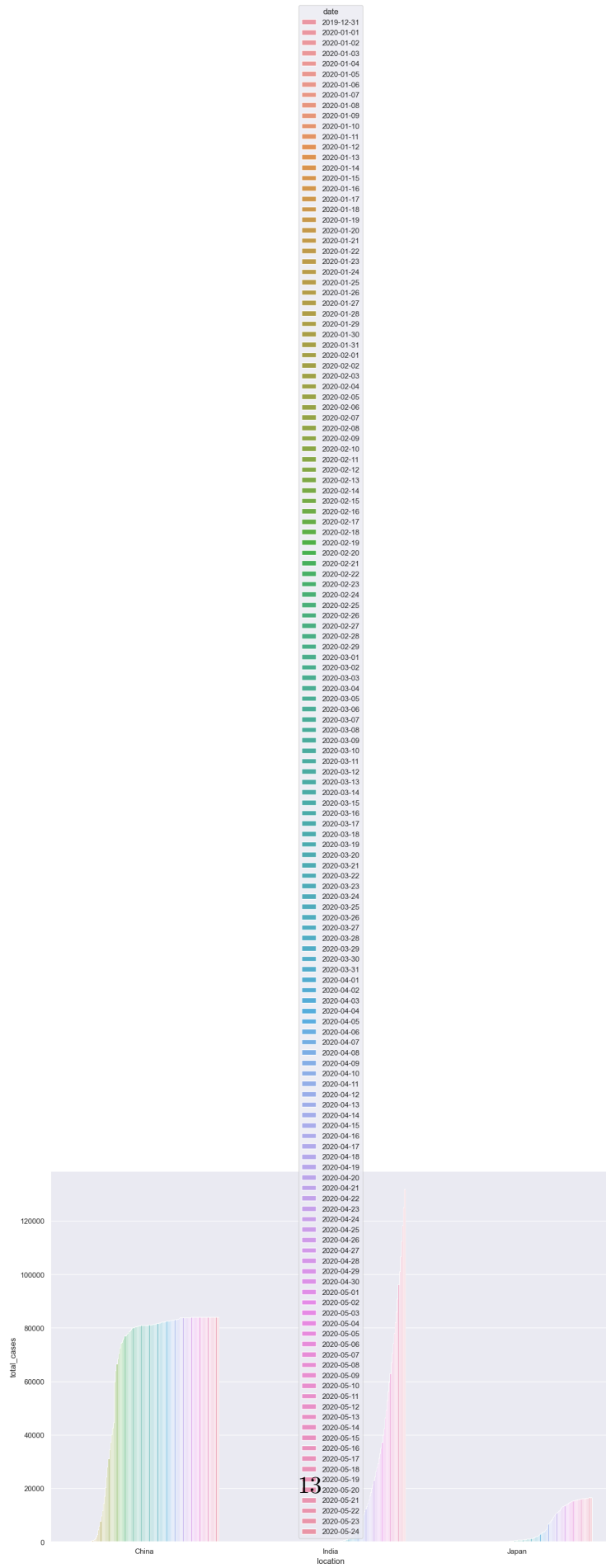
```
[20]: #Making a dataframe for brazil for last 5 days
brazil_last_5_days=brazil_case.tail()
```

```
[21]: #Total cases in last 5 days
sns.set(rc={'figure.figsize':(15,10)})
sns.lineplot(x="date",y="total_cases",data=brazil_last_5_days)
plt.show()
```



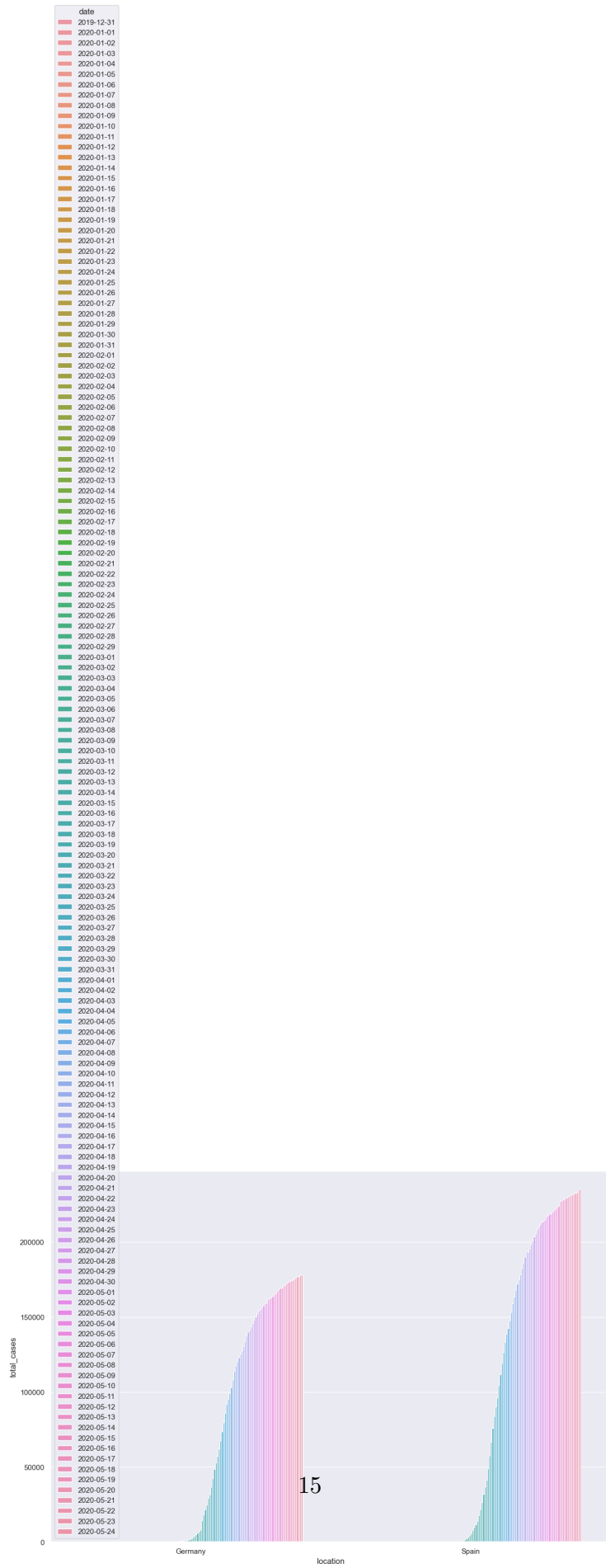
```
[22]: #Understanding cases of India, China and Japan
india_japan_china=covid[(covid["location"] == "India") | (covid["location"]_
↪ == "China") | (covid["location"] == "Japan")]
```

```
[23]: #Plotting growth of cases across China, India and Japan
sns.set(rc={'figure.figsize':(15,10)})
sns.barplot(x="location",y="total_cases",data=india_japan_china,hue="date")
plt.show()
```



```
[24]: #Understanding cases of germany and spain  
germany_spain=covid[(covid["location"] == "Germany") | (covid["location"]  
↪ == "Spain")]
```

```
[25]: #Plotting growth of cases across Germany and Spain  
sns.set(rc={'figure.figsize':(15,10)})  
sns.barplot(x="location",y="total_cases",data=germany_spain,hue="date")  
plt.show()
```



```
[26]: #Getting latest data
last_day_cases=covid[covid["date"]=="2020-05-24"]
last_day_cases
```

```
[26]:
```

	iso_code	location	date	total_cases	new_cases	\
62	ABW	Aruba	2020-05-24	101	0	
198	AFG	Afghanistan	2020-05-24	9998	782	
262	AGO	Angola	2020-05-24	60	0	
321	AIA	Anguilla	2020-05-24	3	0	
398	ALB	Albania	2020-05-24	989	8	
...	
19045	YEM	Yemen	2020-05-24	212	7	
19153	ZAF	South Africa	2020-05-24	21343	1218	
19220	ZMB	Zambia	2020-05-24	920	0	
19285	ZWE	Zimbabwe	2020-05-24	56	0	
19431	OWID_WRL	World	2020-05-24	5273572	97636	

	total_deaths	new_deaths	total_cases_per_million	\
62	3	0	945.994	
198	216	11	256.831	
262	3	0	1.826	
321	0	0	199.973	
398	31	0	343.665	
...	
19045	39	6	7.108	
19153	407	10	359.863	
19220	7	0	50.044	
19285	4	0	3.768	
19431	341722	3633	676.550	

	new_cases_per_million	total_deaths_per_million	...	aged_65_older	\
62	0.000	28.099	...	13.085	
198	20.088	5.549	...	2.581	
262	0.000	0.091	...	2.405	
321	0.000	0.000	...	NaN	
398	2.780	10.772	...	13.188	
...	
19045	0.235	1.308	...	2.922	
19153	20.537	6.862	...	5.344	
19220	0.000	0.381	...	2.480	
19285	0.000	0.269	...	2.822	
19431	12.526	43.840	...	8.696	

	aged_70_older	gdp_per_capita	extreme_poverty	cvd_death_rate	\
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62	7.452	35973.781	NaN	NaN
198	1.337	1803.987	NaN	597.029
262	1.362	5819.495	NaN	276.045
321	NaN	NaN	NaN	NaN
398	8.643	11803.431	1.1	304.195
...
19045	1.583	1479.147	18.8	495.003
19153	3.053	12294.876	18.9	200.380
19220	1.542	3689.251	57.5	234.499
19285	1.882	1899.775	21.4	307.846
19431	5.355	15469.207	10.0	233.070

	diabetes_prevalence	female_smokers	male_smokers	\
62	11.62	NaN	NaN	
198	9.59	NaN	NaN	
262	3.94	NaN	NaN	
321	NaN	NaN	NaN	
398	10.08	7.100	51.200	
...	
19045	5.35	7.600	29.200	
19153	5.52	8.100	33.200	
19220	3.94	3.100	24.700	
19285	1.82	1.600	30.700	
19431	8.51	6.434	34.635	

	handwashing_facilities	hospital_beds_per_100k
62	NaN	NaN
198	37.746	0.500
262	26.664	NaN
321	NaN	NaN
398	NaN	2.890
...
19045	49.542	0.700
19153	43.993	2.320
19220	13.938	2.000
19285	36.791	1.700
19431	60.130	2.705

[207 rows x 32 columns]

```
[27]: #Sorting data w.r.t total_cases
max_cases_country=last_day_cases.sort_values(by="total_cases",ascending=False)
max_cases_country
```

```
[27]:
```

	iso_code	location	date	total_cases	\
19431	OWID_WRL	World	2020-05-24	5273572	
18391	USA	United States	2020-05-24	1622670	

2655	BRA	Brazil	2020-05-24	347398
15569	RUS	Russia	2020-05-24	335882
9396	ITA	Italy	2020-05-24	229327
...
18723	VGB	British Virgin Islands	2020-05-24	8
1645	BES	Bonaire Sint Eustatius and Saba	2020-05-24	6
5543	ESH	Western Sahara	2020-05-24	6
321	AIA	Anguilla	2020-05-24	3
11086	LSO	Lesotho	2020-05-24	2

	new_cases	total_deaths	new_deaths	total_cases_per_million	\
19431	97636	341722	3633	676.550	
18391	21236	97087	1080	4902.287	
2655	16508	22013	965	1634.357	
15569	9434	3388	139	2301.595	
9396	669	32735	119	3792.922	
...	
18723	0	1	0	264.577	
1645	0	0	0	228.824	
5543	0	0	0	10.045	
321	0	0	0	199.973	
11086	1	0	0	0.934	

	new_cases_per_million	total_deaths_per_million	...	aged_65_older	\
19431	12.526	43.840	...	8.696	
18391	64.157	293.312	...	15.413	
2655	77.663	103.562	...	8.552	
15569	64.645	23.216	...	14.178	
9396	11.065	541.416	...	23.021	
...	
18723	0.000	33.072	...	NaN	
1645	0.000	0.000	...	NaN	
5543	0.000	0.000	...	NaN	
321	0.000	0.000	...	NaN	
11086	0.467	0.000	...	4.506	

	aged_70_older	gdp_per_capita	extreme_poverty	cvd_death_rate	\
19431	5.355	15469.207	10.0	233.070	
18391	9.732	54225.446	1.2	151.089	
2655	5.060	14103.452	3.4	177.961	
15569	9.393	24765.954	0.1	431.297	
9396	16.240	35220.084	2.0	113.151	
...	
18723	NaN	NaN	NaN	NaN	
1645	NaN	NaN	NaN	NaN	
5543	1.380	NaN	NaN	NaN	
321	NaN	NaN	NaN	NaN	

11086	2.647	2851.153	59.6	405.126
-------	-------	----------	------	---------

	diabetes_prevalence	female_smokers	male_smokers	\
19431	8.51	6.434	34.635	
18391	10.79	19.100	24.600	
2655	8.11	10.100	17.900	
15569	6.18	23.400	58.300	
9396	4.78	19.800	27.800	
...	
18723	13.67	NaN	NaN	
1645	NaN	NaN	NaN	
5543	NaN	NaN	NaN	
321	NaN	NaN	NaN	
11086	3.94	0.400	53.900	

	handwashing_facilities	hospital_beds_per_100k
19431	60.130	2.705
18391	NaN	2.770
2655	NaN	2.200
15569	NaN	8.050
9396	NaN	3.180
...
18723	NaN	NaN
1645	NaN	NaN
5543	NaN	NaN
321	NaN	NaN
11086	2.117	NaN

[207 rows x 32 columns]

```
[28]: #Top 5 countries with maximum cases
max_cases_country[1:6]
```

```
[28]:
```

	iso_code	location	date	total_cases	new_cases	\
18391	USA	United States	2020-05-24	1622670	21236	
2655	BRA	Brazil	2020-05-24	347398	16508	
15569	RUS	Russia	2020-05-24	335882	9434	
9396	ITA	Italy	2020-05-24	229327	669	
4613	DEU	Germany	2020-05-24	178281	431	

	total_deaths	new_deaths	total_cases_per_million	\
18391	97087	1080	4902.287	
2655	22013	965	1634.357	
15569	3388	139	2301.595	
9396	32735	119	3792.922	
4613	8247	31	2127.866	

	new_cases_per_million	total_deaths_per_million	...	aged_65_older	\
18391	64.157	293.312	...	15.413	
2655	77.663	103.562	...	8.552	
15569	64.645	23.216	...	14.178	
9396	11.065	541.416	...	23.021	
4613	5.144	98.432	...	21.453	

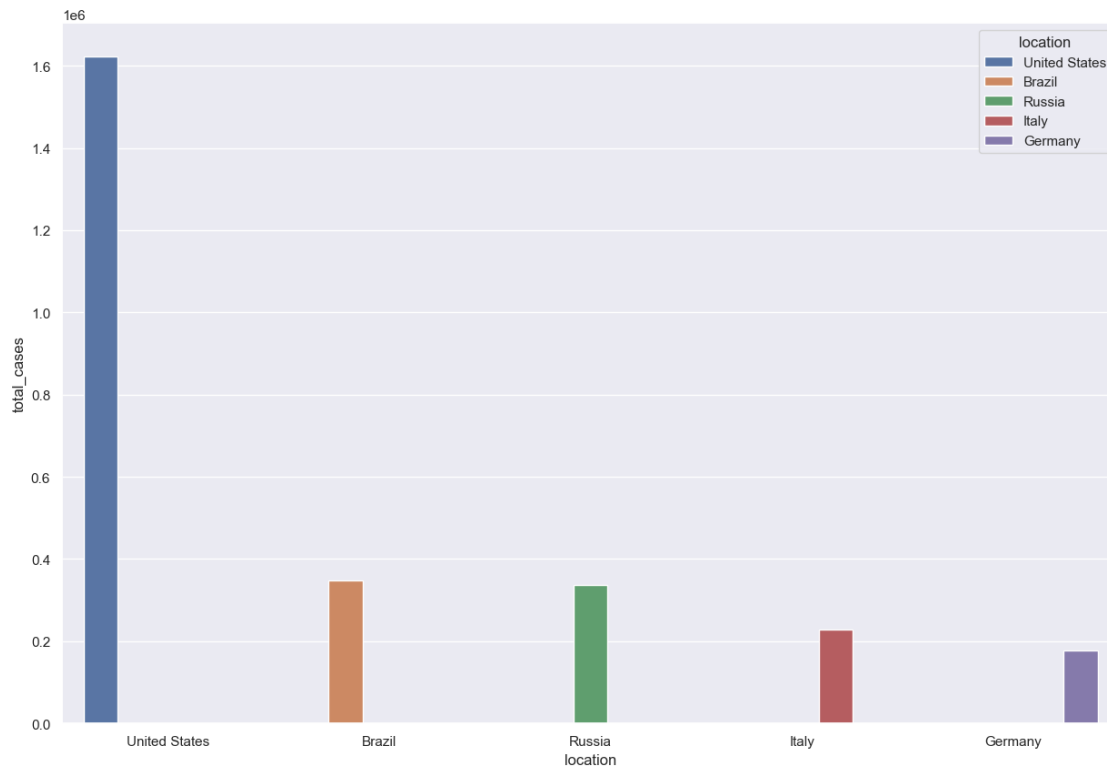
	aged_70_older	gdp_per_capita	extreme_poverty	cvd_death_rate	\
18391	9.732	54225.446	1.2	151.089	
2655	5.060	14103.452	3.4	177.961	
15569	9.393	24765.954	0.1	431.297	
9396	16.240	35220.084	2.0	113.151	
4613	15.957	45229.245	NaN	156.139	

	diabetes_prevalence	female_smokers	male_smokers	\
18391	10.79	19.1	24.6	
2655	8.11	10.1	17.9	
15569	6.18	23.4	58.3	
9396	4.78	19.8	27.8	
4613	8.31	28.2	33.1	

	handwashing_facilities	hospital_beds_per_100k
18391	NaN	2.77
2655	NaN	2.20
15569	NaN	8.05
9396	NaN	3.18
4613	NaN	8.00

[5 rows x 32 columns]

```
[29]: #Making bar-plot for countries with top cases
sns.barplot(x="location",y="total_cases",data=max_cases_country[1:
↪6],hue="location")
plt.show()
```



```
[30]: india_case.head()
```

```
[30]:
```

	iso_code	location	date	total_cases	new_cases	total_deaths	\
8379	IND	India	2019-12-31	0	0	0	
8380	IND	India	2020-01-01	0	0	0	
8381	IND	India	2020-01-02	0	0	0	
8382	IND	India	2020-01-03	0	0	0	
8383	IND	India	2020-01-04	0	0	0	

	new_deaths	total_cases_per_million	new_cases_per_million	\
8379	0	0.0	0.0	
8380	0	0.0	0.0	
8381	0	0.0	0.0	
8382	0	0.0	0.0	
8383	0	0.0	0.0	

	total_deaths_per_million	...	aged_65_older	aged_70_older	\
8379	0.0	...	5.989	3.414	
8380	0.0	...	5.989	3.414	
8381	0.0	...	5.989	3.414	
8382	0.0	...	5.989	3.414	
8383	0.0	...	5.989	3.414	

	gdp_per_capita	extreme_poverty	cvd_death_rate	diabetes_prevalence	\
8379	6426.674	21.2	282.28	10.39	
8380	6426.674	21.2	282.28	10.39	
8381	6426.674	21.2	282.28	10.39	
8382	6426.674	21.2	282.28	10.39	
8383	6426.674	21.2	282.28	10.39	

	female_smokers	male_smokers	handwashing_facilities	\
8379	1.9	20.6	59.55	
8380	1.9	20.6	59.55	
8381	1.9	20.6	59.55	
8382	1.9	20.6	59.55	
8383	1.9	20.6	59.55	

	hospital_beds_per_100k
8379	0.53
8380	0.53
8381	0.53
8382	0.53
8383	0.53

[5 rows x 32 columns]

```
[31]: #Linear regression
from sklearn.model_selection import train_test_split
```

```
[32]: #converting string date to date-time
import datetime as dt
india_case['date'] = pd.to_datetime(india_case['date'])
india_case.head()
```

```
[32]: iso_code location      date  total_cases  new_cases  total_deaths  \
8379      IND      India 2019-12-31           0         0           0
8380      IND      India 2020-01-01           0         0           0
8381      IND      India 2020-01-02           0         0           0
8382      IND      India 2020-01-03           0         0           0
8383      IND      India 2020-01-04           0         0           0
```

	new_deaths	total_cases_per_million	new_cases_per_million	\
8379	0	0.0	0.0	
8380	0	0.0	0.0	
8381	0	0.0	0.0	
8382	0	0.0	0.0	
8383	0	0.0	0.0	

	total_deaths_per_million	...	aged_65_older	aged_70_older	\
--	--------------------------	-----	---------------	---------------	---

8379	0.0	...	5.989	3.414
8380	0.0	...	5.989	3.414
8381	0.0	...	5.989	3.414
8382	0.0	...	5.989	3.414
8383	0.0	...	5.989	3.414

	gdp_per_capita	extreme_poverty	cvd_death_rate	diabetes_prevalence	\
8379	6426.674	21.2	282.28	10.39	
8380	6426.674	21.2	282.28	10.39	
8381	6426.674	21.2	282.28	10.39	
8382	6426.674	21.2	282.28	10.39	
8383	6426.674	21.2	282.28	10.39	

	female_smokers	male_smokers	handwashing_facilities	\
8379	1.9	20.6	59.55	
8380	1.9	20.6	59.55	
8381	1.9	20.6	59.55	
8382	1.9	20.6	59.55	
8383	1.9	20.6	59.55	

	hospital_beds_per_100k
8379	0.53
8380	0.53
8381	0.53
8382	0.53
8383	0.53

[5 rows x 32 columns]

```
[33]: india_case.head()
```

```
[33]:
```

	iso_code	location	date	total_cases	new_cases	total_deaths	\
8379	IND	India	2019-12-31	0	0	0	
8380	IND	India	2020-01-01	0	0	0	
8381	IND	India	2020-01-02	0	0	0	
8382	IND	India	2020-01-03	0	0	0	
8383	IND	India	2020-01-04	0	0	0	

	new_deaths	total_cases_per_million	new_cases_per_million	\
8379	0	0.0	0.0	
8380	0	0.0	0.0	
8381	0	0.0	0.0	
8382	0	0.0	0.0	
8383	0	0.0	0.0	

	total_deaths_per_million	...	aged_65_older	aged_70_older	\
8379	0.0	...	5.989	3.414	

8380	0.0	...	5.989	3.414
8381	0.0	...	5.989	3.414
8382	0.0	...	5.989	3.414
8383	0.0	...	5.989	3.414

	gdp_per_capita	extreme_poverty	cvd_death_rate	diabetes_prevalence	\
8379	6426.674	21.2	282.28	10.39	
8380	6426.674	21.2	282.28	10.39	
8381	6426.674	21.2	282.28	10.39	
8382	6426.674	21.2	282.28	10.39	
8383	6426.674	21.2	282.28	10.39	

	female_smokers	male_smokers	handwashing_facilities	\
8379	1.9	20.6	59.55	
8380	1.9	20.6	59.55	
8381	1.9	20.6	59.55	
8382	1.9	20.6	59.55	
8383	1.9	20.6	59.55	

	hospital_beds_per_100k
8379	0.53
8380	0.53
8381	0.53
8382	0.53
8383	0.53

[5 rows x 32 columns]

```
[34]: #converting date-time to ordinal
india_case['date']=india_case['date'].map(dt.datetime.toordinal)
india_case.head()
```

```
[34]: iso_code location    date  total_cases  new_cases  total_deaths  \
8379      IND      India  737424           0           0           0
8380      IND      India  737425           0           0           0
8381      IND      India  737426           0           0           0
8382      IND      India  737427           0           0           0
8383      IND      India  737428           0           0           0
```

	new_deaths	total_cases_per_million	new_cases_per_million	\
8379	0	0.0	0.0	
8380	0	0.0	0.0	
8381	0	0.0	0.0	
8382	0	0.0	0.0	
8383	0	0.0	0.0	

	total_deaths_per_million	...	aged_65_older	aged_70_older	\
--	--------------------------	-----	---------------	---------------	---

8379	0.0	...	5.989	3.414
8380	0.0	...	5.989	3.414
8381	0.0	...	5.989	3.414
8382	0.0	...	5.989	3.414
8383	0.0	...	5.989	3.414

	gdp_per_capita	extreme_poverty	cvd_death_rate	diabetes_prevalence	\
8379	6426.674	21.2	282.28	10.39	
8380	6426.674	21.2	282.28	10.39	
8381	6426.674	21.2	282.28	10.39	
8382	6426.674	21.2	282.28	10.39	
8383	6426.674	21.2	282.28	10.39	

	female_smokers	male_smokers	handwashing_facilities	\
8379	1.9	20.6	59.55	
8380	1.9	20.6	59.55	
8381	1.9	20.6	59.55	
8382	1.9	20.6	59.55	
8383	1.9	20.6	59.55	

	hospital_beds_per_100k
8379	0.53
8380	0.53
8381	0.53
8382	0.53
8383	0.53

[5 rows x 32 columns]

```
[35]: #getting dependent variable and inpedent variable
x=india_case['date']
y=india_case['total_cases']
```

```
[36]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
```

```
[37]: from sklearn.linear_model import LinearRegression
```

```
[38]: lr = LinearRegression()
```

```
[39]: import numpy as np
lr.fit(np.array(x_train).reshape(-1,1),np.array(y_train).reshape(-1,1))
```

```
[39]: LinearRegression()
```

```
[40]: india_case.tail()
```

```

[40]:      iso_code location      date  total_cases  new_cases  total_deaths  \
8519      IND      India  737565      106750      5611      3303
8520      IND      India  737566      112359      5609      3435
8521      IND      India  737567      118447      6088      3583
8522      IND      India  737568      125101      6654      3720
8523      IND      India  737569      131868      6767      3867

      new_deaths  total_cases_per_million  new_cases_per_million  \
8519          140              77.355              4.066
8520          132              81.419              4.064
8521          148              85.831              4.412
8522          137              90.653              4.822
8523          147              95.556              4.904

      total_deaths_per_million  ...  aged_65_older  aged_70_older  \
8519              2.393  ...      5.989      3.414
8520              2.489  ...      5.989      3.414
8521              2.596  ...      5.989      3.414
8522              2.696  ...      5.989      3.414
8523              2.802  ...      5.989      3.414

      gdp_per_capita  extreme_poverty  cvd_death_rate  diabetes_prevalence  \
8519      6426.674      21.2      282.28      10.39
8520      6426.674      21.2      282.28      10.39
8521      6426.674      21.2      282.28      10.39
8522      6426.674      21.2      282.28      10.39
8523      6426.674      21.2      282.28      10.39

      female_smokers  male_smokers  handwashing_facilities  \
8519          1.9      20.6      59.55
8520          1.9      20.6      59.55
8521          1.9      20.6      59.55
8522          1.9      20.6      59.55
8523          1.9      20.6      59.55

      hospital_beds_per_100k
8519          0.53
8520          0.53
8521          0.53
8522          0.53
8523          0.53

```

[5 rows x 32 columns]

```
[41]: y_pred=lr.predict(np.array(x_test).reshape(-1,1))
```

```
[42]: from sklearn.metrics import mean_squared_error
```

```
[43]: mean_squared_error(x_test,y_pred)
```

```
[43]: 517987319887.33966
```

```
[44]: lr.predict(np.array([[737573]]))
```

```
[44]: array([[54969.10209972]])
```

```
[ ]:
```