	About This document contains basic exploaratory data analysis of COVID-19 Disease in India. This notebook serves to analyze and visualize progress of the pandemic from various perspectives.
,	Introduction The first signs of COVID-19 in India was reported in some towns of Kerala, among three Indian medical students who had returned from Wuhan. After that, the Government of India had announced lockdown on 25 March 2020. India faced its first wave from May 2020 to January 2020 with an Amplitude of around 90,000 new infections a day. As of now India is going under second wave which has proved be more deadlier than previous one. 1. Cases, Deaths and Recovery import pandas as pd from matplotlib import pyplot as plt from matplotlib import dates as mpl_dates
2]: 3]:	ind_covid_df = pd.read_csv('https://api.covid19india.org/csv/latest/case_time_series.csv') Date Date_YMD Daily Confirmed Total Confirmed Daily Recovered Total Recovered Daily Deceased Total Deceased
	4 3 February 2020 2020-02-03 1 3 0 0 0 0 0 <t< td=""></t<>
4]:	<pre>ind_covid_df.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 504 entries, 0 to 503 Data columns (total 8 columns): # Column Non-Null Count Dtype</class></pre>
5]: 5]:	6 Daily Deceased 504 non-null int64 7 Total Deceased 504 non-null int64 dtypes: int64(6), object(2) memory usage: 31.6+ KB ind_covid_df.isnull().sum() Date 0 Date_YMD 0 Daily Confirmed 0 Total Confirmed 0 Daily Recovered 0 Total Recovered 0
6]: 7]: 7]:	Daily Deceased Total Deceased dtype: int64 ind_covid_df['Date_YMD'] = pd.to_datetime(ind_covid_df['Date_YMD']) ind_covid_df.tail(1) Date Date_YMD Daily Confirmed Total Confirmed Daily Recovered Total Recovered Daily Deceased Total Deceased 503 16 June 2021 2021-06-16 67289 29699371 103900 28484360 2329 381340
8]: 9]: 9]:	<pre>total_cases = ind_covid_df['Total Confirmed'] dates = ind_covid_df['Date_YMD'] curr_date = dates.max() curr_total_cases = int(total_cases.tail(1)) dates.max() Timestamp('2021-06-16 00:00:00')</pre>
1]:	
	Total Cases by Time 30 29699371 25 15 10 10
	The logarithmic rise of total cases was observed from July end 2020 till December 2020 which seemed to saturate in january 2020. But April 2020 onwards, cases started to increase at much higher rate than before
3]:	<pre>daily_cases = ind_covid_df['Daily Confirmed'] plt.figure(figsize=(10,6)) plt.plot(dates, daily_cases,'-', linewidth=2) plt.gcf().autofmt_xdate() date_format = mpl_dates.DateFormatter('%d %b, %Y') plt.gca().xaxis.set_major_formatter(date_format) plt.ylabel('Count') plt.suptitle('Daily New Cases by time',fontsize=22) plt.annotate(text=str(today_cases), xy=(curr_date, today_cases),</pre>
	300000 100000 100000 67289\
	From July 2020 Onwards infection rate started to increase and reached its first peak at September 2020 with over 90,000 cases report per-day.\ Cases began to decline from October 2020 and were reported below 15,000 in January 2021 which was a good sign. A second wave beginning in March 2021 was much larger than first, with shortages of vaccines, hospital beds, oxygen cylinders and o medicines such as remdesivir in parts of the country. By April end daily infection count reached over 400,000 which was new record
5]: 6]:	<pre>total_deaths = ind_covid_df['Total Deceased'] plt.figure(figsize=(10,6)) plt.plot(dates, total_deaths, color='red') plt.gcf().autofmt_xdate() date_format = mpl_dates.DateFormatter('%d %b, %Y') plt.gca().xaxis.set_major_formatter(date_format) plt.ylabel('Count')</pre>
6]:	plt.suptitle('Total Deaths by Time', fontsize=22) plt.annotate(text=str(curr_total_deaths), xy=(curr_date, curr_total_deaths), xycoords='data', xytext=(-56,1), textcoords='offset points', fontsize=14) Text(-56, 1, '381340') Total Deaths by Time 400000 381340
	300000 250000 150000 100000 50000
7]: 8]:	<pre>daily_deaths = ind_covid_df['Daily Deceased'] plt.figure(figsize=(10,6)) plt.plot(dates, daily_deaths,'-r', linewidth=1) plt.gcf().autofmt_xdate()</pre>
8]:	<pre>date_format = mpl_dates.DateFormatter('%d %b, %Y') plt.gca().xaxis.set_major_formatter(date_format) plt.ylabel('Count') plt.suptitle('Daily New Deaths by time', fontsize=22) plt.annotate(text=str(today_deaths), xy=(curr_date, today_deaths),</pre>
	6000 5000 4000 3000 2000
	Above plot depicts that their were large no. of deaths in August, September and October months of year 2020. Sudden spike of deaths was seen in mid-June month. In Second wave the deaths are 4 to 5 times more than the previous wave
9]: 9]:	Let us see if their is any correlation between new cases and new deaths on daily basis plt.figure(figsize=(8,5)) plt.scatter(daily_cases, daily_deaths, edgecolor='black', alpha=.3) plt.xlabel('New Cases') plt.ylabel('New Deaths') Text(0, 0.5, 'New Deaths')
	5000
	0 100000 200000 300000 400000 The Scatterplot shows that Daily New deaths are linearly correlated with new cases on daily basis. Their is positive, strong relation between the two, as more points overlapp to form a line i.e. Deaths occuring each day depends on the fresh Covid cases on that day. More the no. of cases are found more deaths will occur. From above plots we can conclude, that if we could stop or supress the fresh Covid cases, then their would be less deaths.
9]: 1]:	<pre>if we could prevent new cases from happening, deaths would reduce total_recovered = ind_covid_df['Total Recovered'] plt.figure(figsize=(10,6)) plt.plot(dates, total_recovered/10**6, color='green') plt.gcf().autofmt_xdate() date_format = mpl_dates.DateFormatter('%d %b, %Y') plt.gca().xaxis.set_major_formatter(date_format)</pre>
	<pre>plt.ylabel('Count (in Million)') plt.suptitle('Total Recovery by Time', fontsize=22) plt.annotate(text=str(curr_total_recovered), xy=(curr_date, curr_total_recovered/10**6),</pre>
	(C) 20 15 15 10 5
2]:	0
3]:	<pre>plt.plot(dates, daily_recovered,'-g', linewidth=1) plt.gcf().autofmt_xdate() date_format = mpl_dates.DateFormatter('%d %b, %Y') plt.gca().xaxis.set_major_formatter(date_format) plt.ylabel('Count') plt.suptitle('Daily Recovered by time',fontsize=22) plt.annotate(text=str(today_recovered), xy=(curr_date, today_recovered),</pre>
	Daily Recovered by time 400000 300000 2000000
4]:	100000 100000 10039
5]:	<pre>curr_active_cases = curr_total_cases - curr_total_deaths - curr_total_recovered plt.figure(figsize=(10,6)) plt.plot(dates, active_cases/10**6, color='#483096', linewidth=2) plt.gcf().autofmt_xdate() date_format = mpl_dates.DateFormatter('%d %b, %Y') plt.gca().xaxis.set_major_formatter(date_format) plt.ylabel('Count (in Millions)')</pre>
6]:	plt.suptitle('Active Cases over Time', fontsize=22) plt.annotate(text=str(curr_active_cases), xy=(curr_date, curr_active_cases/10**6),
	3.0 2.5 2.0 1.5 1.0 0.5 0.0
	Summary 1. Case Fatality Ratio (CFR) Case fatality ratio(CFR) is ratio to measure risk of death when person is infected with a disease. The actual probability of death of person diagonsed with a disease is generally less since everybody is not tested to have a disease or not. Hence their would be a scenario when the content of the c
	their are people who have the disease but are not diagonsed. CFR can increase or decrease, or could vary by location and characterist of the infected person. $CFR \text{ gives rough chances of death if person is infected with COVID-19}$ $CFR = \frac{Number\ of\ deaths\ from\ disease}{Number\ of\ diagonsed\ case\ of\ disease} X\ 100$ $\text{inf_fatality_ratio} = (\text{total_deaths/total_cases})*100$
9]:	<pre>curr_fat_ratio = (curr_total_deaths/curr_total_cases)*100 plt.figure(figsize=(10,6)) plt.plot(dates, inf_fatality_ratio,'-m', linewidth=1) plt.gcf().autofmt_xdate() date_format = mpl_dates.DateFormatter('%d %b, %Y') plt.gca().xaxis.set_major_formatter(date_format) plt.ylabel('Percent') plt.suptitle('Infection Fatality Ratio over Time', fontsize=22) # plt.title('(Chances of Death)')</pre>
9]:	plt.annotate(text=str(round(curr_fat_ratio, 3)), xy=(curr_date, curr_fat_ratio), xycoords='data', xytext=(-48,-10), textcoords='offset points', fontsize=14) Text(-48, -10, '1.284') Infection Fatality Ratio over Time 3.5 3.0
	2.5 2.0 1.5 1.0 0.5 0.0
	$ 2. \text{ Rate of Recovery} $ $ Recovery Rate = \frac{Number\ of\ recovries\ from\ disease}{Number\ of\ diagonsed\ case\ of\ disease} X\ 100 $ During the rise of second wave, recovery rate started falling from March 2021 and settled at 80% after which has started to grow again
0]: 1]: 2]:	<pre>recovery_rate = (total_recovered/total_cases)*100 curr_rec_ratio = (curr_total_recovered/curr_total_cases)*100 plt.figure(figsize=(10,6)) plt.plot(dates, recovery_rate, color='#3b7d24', linewidth=1) plt.gcf().autofmt_xdate() date_format = mpl_dates.DateFormatter('%d %b, %Y') plt.gca().xaxis.set_major_formatter(date_format)</pre>
2]:	plt.ylabel('Percent') plt.suptitle('Recovery rate over Time', fontsize=22) plt.annotate(text=str(round(curr_rec_ratio, 2)), xy=(curr_date, curr_rec_ratio),
	80 60 40 20 20
3]:	o
5]:	<pre>plt.figure(figsize=(10,6)) plt.plot(dates, per_act_cases, color='#8c2730', linewidth=1) plt.gcf().autofmt_xdate() date_format = mpl_dates.DateFormatter('%d %b, %Y') plt.gca().xaxis.set_major_formatter(date_format) plt.ylabel('Percent') plt.suptitle('Percent of Active Cases over Time',fontsize=22) plt.annotate(text=str(round(curr_per_act_cases, 2)), xy=(curr_date, curr_per_act_cases),</pre>
5]:	Percent of Active Cases over Time
	20 01 Mar. 2020 01 May. 2020 01 May. 2020 01 Mov. 2020 01 Mov. 2020 01 Mar. 2021 01 May. 2021 01 May. 2021 01 May. 2021
6]:	<pre>fig, ax = plt.subplots() fig.set_figheight(8) fig.set_figwidth(10) labels = ['Deaths', 'Recovered', 'Active'] ax.stackplot(dates, total_deaths/10**6, total_recovered/10**6, active_cases/10**6, alpha=.8, labels=labels) ax.set_ylabel('Count (in Million)') ax.legend(loc='upper left') fig.autofmt_xdate() date_format = mpl_dates.DateFormatter('%d %b, %Y')</pre>
67	<pre>ax.xaxis.set_major_locator(plt.MaxNLocator(18)) ax.xaxis.set_major_formatter(date_format) plt.annotate(text=str(curr_total_deaths), xy=(curr_date, curr_total_deaths/10**6),</pre>
<i>,</i> :	Deaths Recovered Active 28386360
	(i) tuno 10 10 15 15 15 15 16 16 16 16 16 16 16 16 16 16 16 16 16
7]: 7]: 8]:	total_deaths.max() 19 ax = plt.subplots(figsize=(10,6)) labels = ['Deceased', 'Active', 'Recovered', 'Confirmed'] values = [total_deaths.max(), active_cases.max(), total_recovered.max(), total_cases.max()]
	<pre>values = [total_deaths.max(), active_cases.max(), total_recovered.max(), total_cases.max()] ax.bar(labels, values, color=['#a83232','#3267a8','#67a832','#5d32a8']) ax.set_ylabel('Count') # create a list to collect the plt.patches data totals = [] # find the values and append to list for i in ax.patches: totals.append(i.get_height()) # set individual bar lables using above list for i in ax.patches:</pre>
	2.0 1.5 1.0
	2. Vaccination India began its vaccination program on 16 January 2021. India has approved two vaccines for emergency use, including Oxford-AstraZeneca vaccine also known as Covisheld manufactured by the Serum Institue of India, and Covaxin developed by Biotech. In A 2021 ,Sputnik V was approved as a third vaccine.
	India first started with vaccinating Health care workers being first to receive the vaccine. On April 1 2021 vaccination of people above 45 was started. Followed by vaccination of age group 18-44 from 1 May onwards. vac_df = pd.read_csv('http://api.covid19india.org/csv/latest/cowin_vaccine_data_statewise.csv') pd.set_option('display.max_rows', 10) filt = vac_df.State=='India' ind_vac_df = vac_df.loc[filt].copy()
0]: 1]:	ind_vac_df
	Updated On State Total Individuals Vaccinated Total Sessions Conducted First Dose Administered Second Dose Administered Male(Individuals Vaccinated) Female(Individuals Vaccinated) Transgen Vaccinated 0 16/01/2021 India 48276.0 2957.0 48276.0 0.0 23757.0 24517.0 1 17/01/2021 India 58604.0 8532.0 4954.0 58604.0 0.0 27348.0 31252.0 2 18/01/2021 India 99449.0 13611.0 6583.0 99449.0 0.0 41361.0 58083.0 3 19/01/2021 India 195525.0 7951.0 195525.0 0.0 81901.0 113613.0 4 20/01/2021 India 251280.0 25472.0 10504.0 251280.0 0.0 98111.0 153145.0

