Final Project: Covid-19

```
In [34]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as seabornInstance
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn import metrics
%matplotlib inline
import plotly as py

import seaborn as sns
import datetime
```

Import Datasets

```
In [35]: countries = pd.read_csv("4.18states.csv")
    us_counties = pd.read_csv("abridged_couties.csv")
    us_confirmed_dates = pd.read_csv("time_series_covid19_confirmed_US.csv")
    us_confirmed_deaths = pd.read_csv("time_series_covid19_deaths_US.csv")
```

In [36]: countries.head()

Out[36]:

	Province_State	Country_Region	Last_Update	Lat	Long_	Confirmed	Deaths	Recovered	Active	FIPS	Incident_Rate	People_Tested	F
0	Alabama	US	2020-04-18 22:32:47	32.3182	-86.9023	4712	153	NaN	4559.0	1.0	100.492717	42538.0	_
1	Alaska	US	2020-04-18 22:32:47	61.3707	-152.4044	314	9	147.0	305.0	2.0	52.530410	9655.0	
2	American Samoa	US	NaN	-14.2710	-170.1320	0	0	NaN	NaN	60.0	0.000000	3.0	
3	Arizona	US	2020-04-18 22:32:47	33.7298	-111.4312	4724	180	539.0	4544.0	4.0	64.901548	51045.0	
4	Arkansas	US	2020-04-18 22:32:47	34.9697	-92.3731	1744	38	703.0	1706.0	5.0	67.361213	24141.0	

In [37]: us_counties.head(5)

Out[37]:

	countyFIPS	STATEFP	COUNTYFP	CountyName	StateName	State	lat	lon	POP_LATITUDE	POP_LONGITUDE	 >500 gatherings
0	01001	1.0	1.0	Autauga	AL	Alabama	32.540091	-86.645649	32.500389	-86.494165	 737497.0
1	01003	1.0	3.0	Baldwin	AL	Alabama	30.738314	-87.726272	30.548923	-87.762381	 737497.0
2	01005	1.0	5.0	Barbour	AL	Alabama	31.874030	-85.397327	31.844036	-85.310038	 737497.0
3	01007	1.0	7.0	Bibb	AL	Alabama	32.999024	-87.125260	33.030921	-87.127659	 737497.0
4	01009	1.0	9.0	Blount	AL	Alabama	33.990440	-86.562711	33.955243	-86.591491	 737497.0

5 rows × 87 columns

Convert null values in State by examining their StateName and using a dictionary with key, value pairs to apply the transformation.

```
In [38]:
         abbrev to state = {
                  'AK': 'Alaska',
                  'AL': 'Alabama',
                  'AR': 'Arkansas',
                  'AS': 'American Samoa',
                  'AZ': 'Arizona',
                  'CA': 'California',
                  'CO': 'Colorado',
                  'CT': 'Connecticut',
                  'DC': 'District of Columbia',
                  'DE': 'Delaware',
                  'FL': 'Florida',
                  'GA': 'Georgia',
                  'GU': 'Guam',
                  'HI': 'Hawaii',
                  'IA': 'Iowa',
                  'ID': 'Idaho',
                  'IL': 'Illinois',
                  'IN': 'Indiana',
                  'KS': 'Kansas',
                  'KY': 'Kentucky',
                  'LA': 'Louisiana',
                  'MA': 'Massachusetts',
                  'MD': 'Maryland',
                  'ME': 'Maine',
                  'MI': 'Michigan',
                  'MN': 'Minnesota',
                  'MO': 'Missouri',
                  'MP': 'Northern Mariana Islands',
                  'MS': 'Mississippi',
                  'MT': 'Montana',
                  'NA': 'National',
                  'NC': 'North Carolina',
                  'ND': 'North Dakota',
                  'NE': 'Nebraska',
                  'NH': 'New Hampshire',
                  'NJ': 'New Jersey',
                  'NM': 'New Mexico',
                  'NV': 'Nevada',
                  'NY': 'New York',
```

```
'OH': 'Ohio',
        'OK': 'Oklahoma',
        'OR': 'Oregon',
        'PA': 'Pennsylvania',
        'PR': 'Puerto Rico',
        'RI': 'Rhode Island',
        'SC': 'South Carolina',
        'SD': 'South Dakota',
        'TN': 'Tennessee',
        'TX': 'Texas',
        'UT': 'Utah',
        'VA': 'Virginia',
        'VI': 'Virgin Islands',
        'VT': 'Vermont',
        'WA': 'Washington',
        'WI': 'Wisconsin',
        'WV': 'West Virginia',
        'WY': 'Wyoming'
us_counties['State']=us_counties['StateName'].map(abbrev_to_state)
```

In [39]: us_confirmed_dates.head()

Out[39]:

	UID	iso2	iso3	code3	FIPS	Admin2	Province_State	Country_Region	Lat	Long_	 4/9/20	4/10/20	4/11/20	4/12/20	4/13/20	4/1
0	16	AS	ASM	16	60.0	NaN	American Samoa	US	-14.2710	-170.1320	 0	0	0	0	0	
1	316	GU	GUM	316	66.0	NaN	Guam	US	13.4443	144.7937	 128	130	133	133	133	
2	580	MP	MNP	580	69.0	NaN	Northern Mariana Islands	US	15.0979	145.6739	 11	11	11	11	11	
3	630	PR	PRI	630	72.0	NaN	Puerto Rico	US	18.2208	-66.5901	 683	725	788	897	903	
4	850	VI	VIR	850	78.0	NaN	Virgin Islands	US	18.3358	-64.8963	 45	50	51	51	51	

5 rows × 99 columns

In [40]: us_confirmed_deaths.head()

Out[40]:

	UID	iso2	iso3	code3	FIPS	Admin2	Province_State	Country_Region	Lat	Long_	 4/9/20	4/10/20	4/11/20	4/12/20	4/13/20	4/1
0	16	AS	ASM	16	60.0	NaN	American Samoa	US	-14.2710	-170.1320	 0	0	0	0	0	
1	316	GU	GUM	316	66.0	NaN	Guam	US	13.4443	144.7937	 4	4	5	5	5	
2	580	MP	MNP	580	69.0	NaN	Northern Mariana Islands	US	15.0979	145.6739	 2	2	2	2	2	
3	630	PR	PRI	630	72.0	NaN	Puerto Rico	US	18.2208	-66.5901	 33	39	42	44	45	
4	850	VI	VIR	850	78.0	NaN	Virgin Islands	US	18.3358	-64.8963	 1	1	1	1	1	

5 rows × 100 columns

Objective

What factors affect the transmission rate of COVID-19? How can we predict recovery rates based on these factors across the U.S.?

Train a machine learning model to predict Recovery_Rate using Linear Regression.

Data Cleaning

```
drop_us_county_columns = ['PopMale<52010',</pre>
In [41]:
                           'PopFmle<52010',
                            'PopMale5-92010',
                            'PopFmle5-92010',
                            'PopMale10-142010',
                            'PopFmle10-142010',
                            'PopMale15-192010',
                            'PopFmle15-192010',
                            'PopMale20-242010',
                            'PopFmle20-242010',
                            'PopMale25-292010',
                            'PopFmle25-292010',
                            'PopMale30-342010',
                            'PopFmle30-342010',
                            'PopMale35-442010',
                            'PopFmle35-442010',
                            'PopMale45-542010',
                            'PopFmle45-542010',
                            'PopMale55-592010',
                            'PopFmle55-592010',
                            'PopMale60-642010',
                            'PopFmle60-642010',
                            'PopMale65-742010',
                            'PopFmle65-742010',
                            'PopMale75-842010',
                            'PopFmle75-842010',
                            'PopMale>842010',
                            'PopFmle>842010',
                           'countyFIPS',
                           'StateName',
                           'STATEFP',
                           'COUNTYFP',
                           'CensusRegionName',
                           'CensusDivisionName',
                           'Rural-UrbanContinuumCode2013',
                            'PopTotalMale2017', 'PopTotalFemale2017', 'FracMale2017',
                          'dem to rep ratio',
                           '3-YrMortalityAge<1Year2015-17',
                            '3-YrMortalityAge1-4Years2015-17',
                            '3-YrMortalityAge5-14Years2015-17',
```

```
'3-YrMortalityAge25-34Years2015-17',
'3-YrMortalityAge25-34Years2015-17',
'3-YrMortalityAge35-44Years2015-17',
'3-YrMortalityAge45-54Years2015-17',
'3-YrMortalityAge55-64Years2015-17',
'3-YrMortalityAge65-74Years2015-17',
'3-YrMortalityAge65-74Years2015-17',
'3-YrMortalityAge75-84Years2015-17',
'3-YrMortalityAge85+Years2015-17',
'#EligibleforMedicare2018',
'MedicareEnrollment,AgedTot2017']
us_counties = us_counties.drop(columns = drop_us_county_columns)
```

Convert ordinal dates in columns stay at home, >50 gatherings, >500 gatherings, public schools, restaurant dine-in, entertainment/gym, federal guidelines, foreign travel ban into timestamps.

```
In [42]:
         us counties.iloc[:, 23:31] = us counties.iloc[:, 23:31].fillna(0).astype(int)
         def convert timestamps(col):
             for i in range(len(us counties[col])):
                 if (us counties[col][i] == 0):
                     us counties[col][i] = 0
                 else:
                     us counties[col][i] = pd.Timestamp.fromordinal(us counties[col][i]).date()
         convert timestamps('stay at home')
         convert timestamps('>50 gatherings')
         convert timestamps('>500 gatherings')
         convert timestamps('public schools')
         convert timestamps('restaurant dine-in')
         convert timestamps('entertainment/gym')
         convert timestamps('federal guidelines')
         convert timestamps('foreign travel ban')
```

/srv/conda/envs/data100/lib/python3.7/site-packages/ipykernel_launcher.py:8: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#r eturning-a-view-versus-a-copy

/srv/conda/envs/data100/lib/python3.7/site-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: http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

final proj 6/15/2020

```
guidelines = us_counties.iloc[:, 23:31]
In [43]:
         guidelines.head(5)
```

Out[43]:

	stay at home	>50 gatherings	>500 gatherings	public schools	restaurant dine-in	entertainment/gym	federal guidelines	foreign travel ban
0	2020-04-04	2020-03-20	2020-03-13	2020-03-16	2020-03-19	2020-03-28	2020-03-16	2020-03-11
1	2020-04-04	2020-03-20	2020-03-13	2020-03-16	2020-03-19	2020-03-28	2020-03-16	2020-03-11
2	2020-04-04	2020-03-20	2020-03-13	2020-03-16	2020-03-19	2020-03-28	2020-03-16	2020-03-11
3	2020-04-04	2020-03-20	2020-03-13	2020-03-16	2020-03-19	2020-03-28	2020-03-16	2020-03-11
4	2020-04-04	2020-03-20	2020-03-13	2020-03-16	2020-03-19	2020-03-28	2020-03-16	2020-03-11

In [44]: grouped_states = us_counties.groupby(['State']).mean() grouped_states.head()

Out[44]:

	lat	lon	POP_LATITUDE	POP_LONGITUDE	PopulationEstimate2018	PopulationEstimate65+2017	PopulationDensityperSqMile
State							
Alabama	32.887935	-86.709300	32.878325	-86.700935	72953.298507	11996.582090	90.2
Alaska	NaN	NaN	60.190155	-148.255559	25428.896552	2847.586207	7.7
American Samoa	NaN	NaN	NaN	NaN	NaN	NaN	
Arizona	33.678351	-111.467022	33.581097	-111.477913	478109.733333	80116.400000	52.0
Arkansas	34.911163	-92.437589	34.924027	-92.428029	40184.333333	6655.253333	54.3

5 rows × 25 columns

Out[45]:

	Province_State	Country_Region	Lat	Long_	Confirmed	Deaths	Recovered	Active	Incident_Rate	People_Tested	People_Hospitalized
0	Alabama	US	32.3182	-86.9023	4712	153	NaN	4559.0	100.492717	42538.0	620.0
1	Alaska	US	61.3707	-152.4044	314	9	147.0	305.0	52.530410	9655.0	39.0
2	American Samoa	US	-14.2710	-170.1320	0	0	NaN	NaN	0.000000	3.0	NaN
3	Arizona	US	33.7298	-111.4312	4724	180	539.0	4544.0	64.901548	51045.0	566.0
4	Arkansas	US	34.9697	-92.3731	1744	38	703.0	1706.0	67.361213	24141.0	291.0

In [46]: grouped_countries = countries.groupby(['Province_State']).mean()
 grouped_countries.head()

Out[46]:

	Lat	Long_	Confirmed	Deaths	Recovered	Active	Incident_Rate	People_Tested	People_Hospitalized	Mortality_Rate	Testi
Province_State											
Alabama	32.3182	-86.9023	4712	153	NaN	4559.0	100.492717	42538.0	620.0	3.247029	907
Alaska	61.3707	-152.4044	314	9	147.0	305.0	52.530410	9655.0	39.0	2.866242	1615
Alberta	53.9333	-116.5765	2562	51	0.0	2511.0	58.053824	NaN	NaN	1.990632	
American Samoa	-14.2710	-170.1320	0	0	NaN	NaN	0.000000	3.0	NaN	NaN	Ę
Anguilla	18.2206	-63.0686	3	0	1.0	2.0	19.997334	NaN	NaN	0.000000	

In [47]: us_confirmed_dates.head()

Out[47]:

	UID	iso2	iso3	code3	FIPS	Admin2	Province_State	Country_Region	Lat	Long_	 4/9/20	4/10/20	4/11/20	4/12/20	4/13/20	4/1
0	16	AS	ASM	16	60.0	NaN	American Samoa	US	-14.2710	-170.1320	 0	0	0	0	0	
1	316	GU	GUM	316	66.0	NaN	Guam	US	13.4443	144.7937	 128	130	133	133	133	
2	580	MP	MNP	580	69.0	NaN	Northern Mariana Islands	US	15.0979	145.6739	 11	11	11	11	11	
3	630	PR	PRI	630	72.0	NaN	Puerto Rico	US	18.2208	-66.5901	 683	725	788	897	903	
4	850	VI	VIR	850	78.0	NaN	Virgin Islands	US	18.3358	-64.8963	 45	50	51	51	51	

5 rows × 99 columns

In [48]: us_confirmed_dates = us_confirmed_dates.drop(columns=['UID', 'iso2', 'iso3', 'code3', 'Admin2', 'FIPS'])

Out[49]:

	1/22/20	1/23/20	1/24/20	1/25/20	1/26/20	1/27/20	1/28/20	1/29/20	1/30/20	1/31/20	 4/9/20	4/10/20	4/11/20	4/12/20	4/13/
Province_State															
Alabama	0	0	0	0	0	0	0	0	0	0	 2703	2947	3217	3563	37
Alaska	0	0	0	0	0	0	0	0	0	0	 235	246	257	272	2
Arizona	0	0	0	0	1	1	1	1	1	1	 3018	3112	3393	3542	37
Arkansas	0	0	0	0	0	0	0	0	0	0	 1119	1171	1228	1280	14
California	0	0	0	0	2	2	2	2	2	3	 19710	21081	21706	22795	239

5 rows × 88 columns

In [50]: transposed_dates1 = us_confirmed_dates_groupedby_states.T
 transposed_dates1.head()

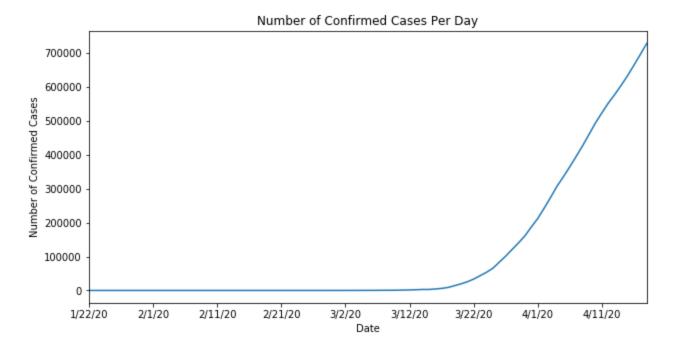
Out[50]:

Province_State	Alabama	Alaska	Arizona	Arkansas	California	Colorado	Connecticut	Delaware	Florida	Georgia	 South Dakota	Tennessee	Texas	
1/22/20	0	0	0	0	0	0	0	0	0	0	 0	0	0	_
1/23/20	0	0	0	0	0	0	0	0	0	0	 0	0	0	
1/24/20	0	0	0	0	0	0	0	0	0	0	 0	0	0	
1/25/20	0	0	0	0	0	0	0	0	0	0	 0	0	0	
1/26/20	0	0	1	0	2	0	0	0	0	0	 0	0	0	

5 rows × 50 columns

```
In [51]: transposed_dates1['Total per Day'] = transposed_dates1.sum(axis=1)
    transposed_dates1['Total per Day'].plot(figsize=(10, 5))
    plt.xlabel('Date')
    plt.ylabel('Number of Confirmed Cases')
    plt.title('Number of Confirmed Cases Per Day')
```

Out[51]: Text(0.5, 1.0, 'Number of Confirmed Cases Per Day')



Out[52]:

	Province_State	1/22/20	1/23/20	1/24/20	1/25/20	1/26/20	1/27/20	1/28/20	1/29/20	1/30/20	 4/9/20	4/10/20	4/11/20	4/12/20	4/13/20	4/
0	American Samoa	0	0	0	0	0	0	0	0	0	 0	0	0	0	0	
1	Guam	0	0	0	0	0	0	0	0	0	 4	4	5	5	5	
2	Northern Mariana Islands	0	0	0	0	0	0	0	0	0	 2	2	2	2	2	
3	Puerto Rico	0	0	0	0	0	0	0	0	0	 33	39	42	44	45	
4	Virgin Islands	0	0	0	0	0	0	0	0	0	 1	1	1	1	1	

5 rows × 89 columns

```
In [54]: us_confirmed_deaths_groupedby_state = us_confirmed_deaths.groupby('Province_State').sum()
    transposed_dates2 = us_confirmed_deaths_groupedby_state.T
    transposed_dates2.head(5)
```

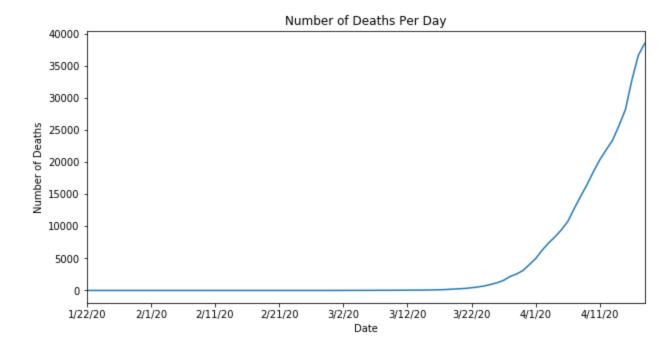
Out[54]:

Province_State	Alabama	Alaska	Arizona	Arkansas	California	Colorado	Connecticut	Delaware	Florida	Georgia	 South Dakota	Tennessee	Texas	
1/22/20	0	0	0	0	0	0	0	0	0	0	 0	0	0	
1/23/20	0	0	0	0	0	0	0	0	0	0	 0	0	0	
1/24/20	0	0	0	0	0	0	0	0	0	0	 0	0	0	
1/25/20	0	0	0	0	0	0	0	0	0	0	 0	0	0	
1/26/20	0	0	0	0	0	0	0	0	0	0	 0	0	0	

5 rows × 50 columns

```
In [55]: transposed_dates2['Total per Day'] = transposed_dates2.sum(axis=1)
    transposed_dates2['Total per Day'].plot(figsize=(10,5))
    plt.xlabel('Date')
    plt.ylabel('Number of Deaths')
    plt.title('Number of Deaths Per Day')
```

Out[55]: Text(0.5, 1.0, 'Number of Deaths Per Day')



GROUPED TABLES

In [56]: grouped_states.head(5)

lat

Out[56]:

State							
Alabama	32.887935	-86.709300	32.878325	-86.700935	72953.298507	11996.582090	90.2
Alaska	NaN	NaN	60.190155	-148.255559	25428.896552	2847.586207	7.7
American Samoa	NaN	NaN	NaN	NaN	NaN	NaN	
Arizona	33.678351	-111.467022	33.581097	-111.477913	478109.733333	80116.400000	52.0
Arkansas	34.911163	-92.437589	34.924027	-92.428029	40184.333333	6655.253333	54.3

Ion POP_LATITUDE POP_LONGITUDE PopulationEstimate2018 PopulationEstimate65+2017 PopulationDensityperSqMile

5 rows × 25 columns

In [57]: grouped_countries.head(5)

Out[57]:

	Lat	Long_	Confirmed	Deaths	Recovered	Active	Incident_Rate	People_Tested	People_Hospitalized	Mortality_Rate	Testi
Province_State											
Alabama	32.3182	-86.9023	4712	153	NaN	4559.0	100.492717	42538.0	620.0	3.247029	907
Alaska	61.3707	-152.4044	314	9	147.0	305.0	52.530410	9655.0	39.0	2.866242	1618
Alberta	53.9333	-116.5765	2562	51	0.0	2511.0	58.053824	NaN	NaN	1.990632	
American Samoa	-14.2710	-170.1320	0	0	NaN	NaN	0.000000	3.0	NaN	NaN	Ę
Anguilla	18.2206	-63.0686	3	0	1.0	2.0	19.997334	NaN	NaN	0.000000	

In [58]:]: us_confirmed_dates_groupedby_states.head(5)															
Out[58]:		1/22/20	1/23/20	1/24/20	1/25/20	1/26/20	1/27/20	1/28/20	1/29/20	1/30/20	1/31/20	 4/9/20	4/10/20	4/11/20	4/12/20	4/13/
	Province_State															
	Alabama	0	0	0	0	0	0	0	0	0	0	 2703	2947	3217	3563	37
	Alaska	0	0	0	0	0	0	0	0	0	0	 235	246	257	272	2
	Arizona	0	0	0	0	1	1	1	1	1	1	 3018	3112	3393	3542	37
	Arkansas	0	0	0	0	0	0	0	0	0	0	 1119	1171	1228	1280	14
	California	0	0	0	0	2	2	2	2	2	3	 19710	21081	21706	22795	239
	5 rows × 88 colu	ımno														
	3 10W3 × 00 COR	JIIIIS														
In [59]:	us_confirmed		s_group	edby_st	tate.he	ad(5)										
In [59]: Out[59]:			s_group	edby_st	ate.he	ad(5)	1/27/20	1/28/20	1/29/20	1/30/20	1/31/20	 4/9/20	4/10/20	4/11/20	4/12/20	4/13/
		l_death					1/27/20	1/28/20	1/29/20	1/30/20	1/31/20	 4/9/20	4/10/20	4/11/20	4/12/20	4/13/
	us_confirmed	l_death					1/27/20	1/28/20	1/29/20	1/30/20		 4/9/20	4/10/20	4/11/20 92	4/12/20 93	4/13/
	us_confirmed	l_death:	1/23/20	1/24/20	1/25/20	1/26/20					0					4/13/
	us_confirmed Province_State Alabama	1_death:	1/23/20	1/24/20	1/25/20	1/26/20	0	0	0	0	0	 70	80	92	93	4/13/
	Province_State Alabama Alaska	1/22/20 0	1/23/20 0 0	1/24/20 0 0	1/25/20 0 0	1/26/20 0 0	0	0	0	0	0	 70 7	80 7	92	93	

5 rows × 88 columns

BAR PLOTS BY REGIONS

Out[61]:

lat	Ion POP_LATITUDE	POP_LONGITUDE	PopulationEstimate2018	PopulationEstimate65+2017	PopulationDensityperSqMil-

_	State							
_	Alabama	32.887935	-86.709300	32.878325	-86.700935	72953.298507	11996.582090	90.2
	Alaska	NaN	NaN	60.190155	-148.255559	25428.896552	2847.586207	7.7
	Arizona	33.678351	-111.467022	33.581097	-111.477913	478109.733333	80116.400000	52.0
	Arkansas	34.911163	-92.437589	34.924027	-92.428029	40184.333333	6655.253333	54.3
	California	37.851530	-120.724312	37.821320	-120.857914	682018.017241	94919.965517	663.2

5 rows × 38 columns

Let's split our data into regions so we can examine the data more closely before choosing features to train our model.

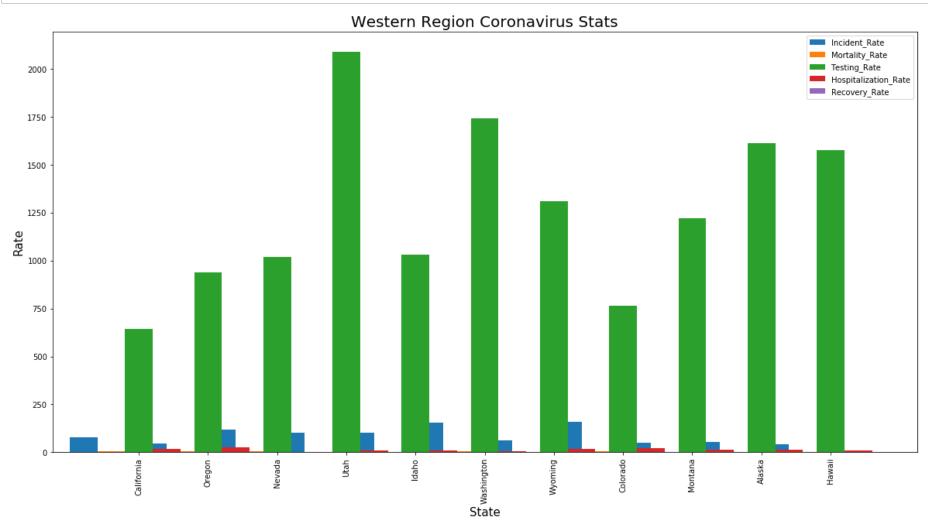
```
In [62]: west_region = ['California', 'Oregon', 'Nevada', 'Utah', 'Idaho', 'Washington', 'Wyoming', 'Colorado', 'Montan
a', 'Alaska', 'Hawaii']
southwest_region = ['Arizona', 'New Mexico', 'Oklahoma', 'Texas']
midwest_region = ['North Dakota', 'South Dakota', 'Nebraska', 'Kansas', 'Minnesota', 'Iowa', 'Missouri', 'Wisc
onsin', 'Illinois', 'Michigan', 'Indiana', 'Ohio']
southeast_region = ['Arkansas', 'Louisiana', 'Mississippi', 'Tennessee', 'Alabama', 'West Virginia', 'Virgini
a', 'North Carolina', 'South Carolina', 'Georgia', 'Florida']
northeast_region = ['Pennsylvania', 'New York', 'Vermont', 'New Hampshire', 'Massachusetts', 'New Jersey', 'Ma
ryland', 'Delaware', 'Connecticut', 'Rhole Island']
```

/srv/conda/envs/data100/lib/python3.7/site-packages/pandas/core/indexing.py:1418: FutureWarning:

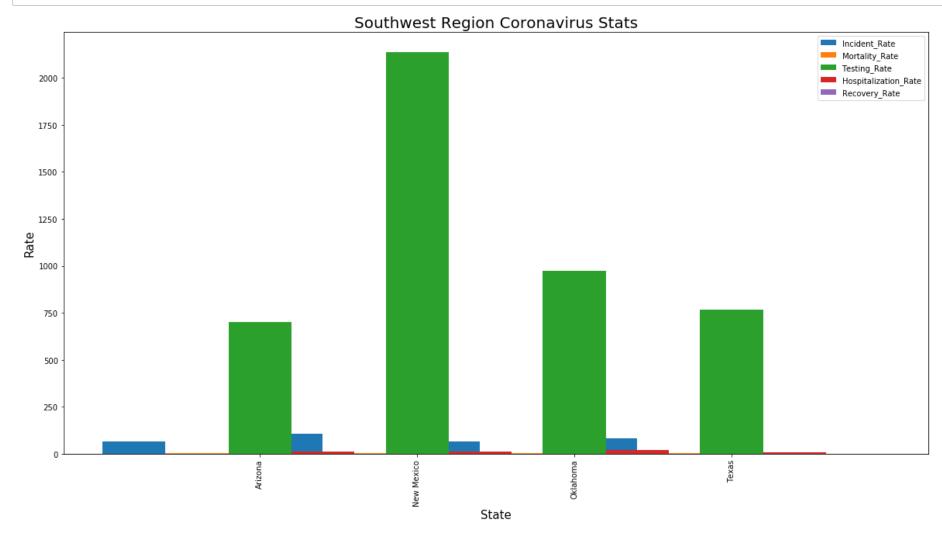
Passing list-likes to .loc or [] with any missing label will raise KeyError in the future, you can use .reindex() as an alternative.

See the documentation here: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#deprecate-loc-reindex-listlike

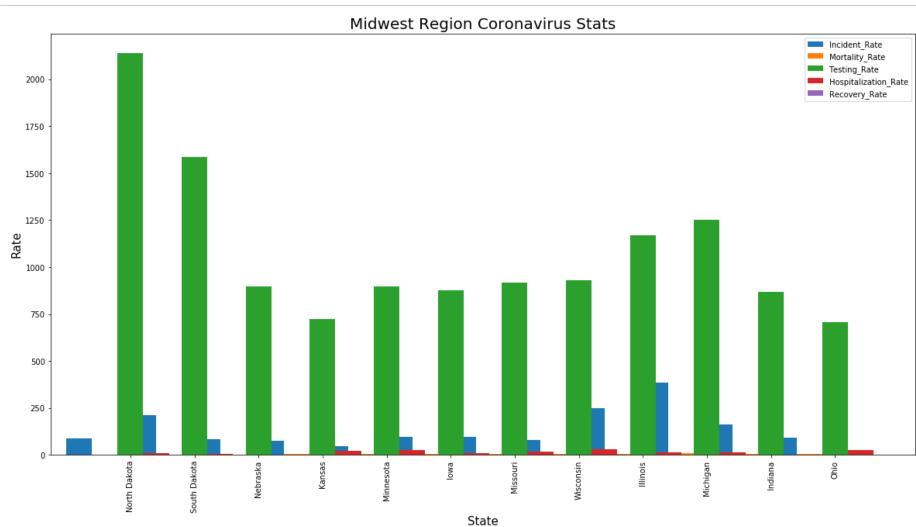
```
In [64]: west_df.loc[:, ['State', 'Incident_Rate', 'Mortality_Rate', 'Testing_Rate', 'Hospitalization_Rate', 'Recovery_
Rate']].plot.bar(x='State', figsize=(20,10), width=2)
plt.title('Western Region Coronavirus Stats').set_size(20)
plt.xlabel('State').set_size(15)
plt.ylabel('Rate').set_size(15)
```

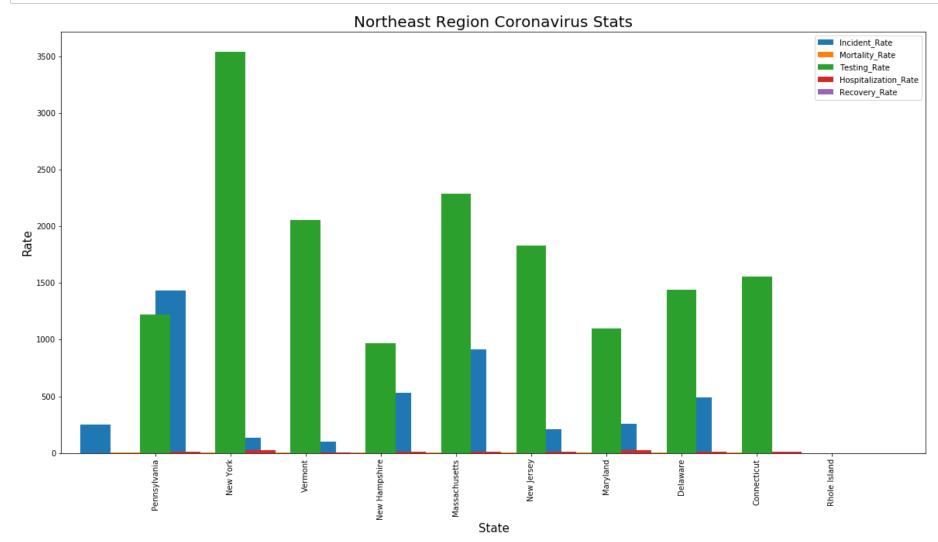


```
In [65]: southwest_df.loc[:, ['State', 'Incident_Rate', 'Mortality_Rate', 'Testing_Rate', 'Hospitalization_Rate', 'Reco
    very_Rate']].plot.bar(x='State', figsize=(20,10), width=2)
    plt.title('Southwest Region Coronavirus Stats').set_size(20)
    plt.xlabel('State').set_size(15)
    plt.ylabel('Rate').set_size(15)
```

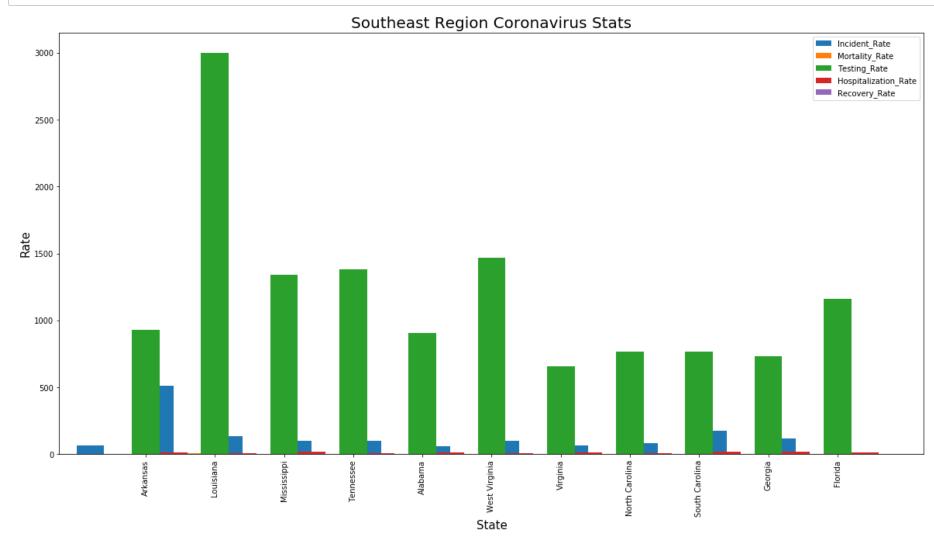


```
In [66]: midwest_df.loc[:, ['State', 'Incident_Rate', 'Mortality_Rate', 'Testing_Rate', 'Hospitalization_Rate', 'Recove ry_Rate']].plot.bar(x='State', figsize=(20,10), width=2)
    plt.title('Midwest Region Coronavirus Stats').set_size(20)
    plt.xlabel('State').set_size(15)
    plt.ylabel('Rate').set_size(15)
```





```
In [68]: southeast_df.loc[:, ['State', 'Incident_Rate', 'Mortality_Rate', 'Testing_Rate', 'Hospitalization_Rate', 'Reco
    very_Rate']].plot.bar(x='State', figsize=(20,10), width=2)
    plt.title('Southeast Region Coronavirus Stats').set_size(20)
    plt.xlabel('State').set_size(15)
    plt.ylabel('Rate').set_size(15)
```



EXPLORE RELATIONSHIPS

We must divide the data into attributes and labels.

Labels: 3-YrDiabetes2015-17, DiabetesPercentage, HeartDiseaseMortality, StrokeMortality, Smokers_Percentage, RespMortalityRate2014, Incident_Rate, Testing_Rate, Hospitalization_Rate

Attribute: Recovery_Rate

```
In [69]: relationships = join countries and states.loc[:, ['3-YrDiabetes2015-17',
                                                            'DiabetesPercentage',
                                                            'HeartDiseaseMortality',
                                                            'StrokeMortality',
                                                            'Smokers Percentage',
                                                            'RespMortalityRate2014',
                                                            'Incident Rate',
                                                            'Testing Rate',
                                                            'Hospitalization Rate',
                                                            'Recovery Rate']]
         relationships['Hospitalization_Rate']=relationships['Hospitalization_Rate'].fillna(np.mean(relationships['Hosp
         italization Rate']))
         corr = relationships.corr()
         plt.figure(figsize=(15,10))
         sns.heatmap(corr, annot=True)
         plt.title('Correlation between Labels and Attribute').set size(20)
```

Correlation between Labels and Attribute

- 1.0

- 0.8

- 0.6

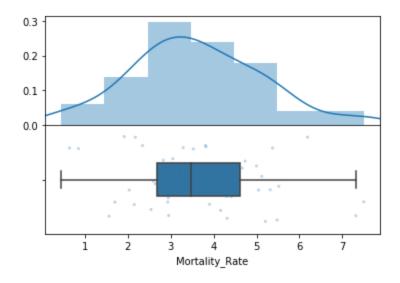
- 0.2

		•	COLLCIA	CIOII DC	CIICCII	Labell	ana /	cciibac	_	
3-YrDiabetes2015-17 -	1	-0.21	-0.27	-0.27	-0.41	-0.39	0.25	-0.0092	-0.086	-0.22
DiabetesPercentage -	-0.21	1	0.86	0.89	0.76	0.65	-0.1	-0.2	0.19	0.092
HeartDiseaseMortality -	-0.27	0.86	1	0.8	0.74	0.73	0.013	-0.087	0.21	0.074
StrokeMortality -	-0.27	0.89	0.8	1	0.66	0.65	-0.3	-0.32	0.17	0.19
Smokers_Percentage -	-0.41	0.76	0.74	0.66	1	0.76	-0.19	-0.16	0.23	0.21
RespMortalityRate2014 -	-0.39	0.65	0.73	0.65	0.76	1	-0.28	-0.34	0.36	0.17
Incident_Rate -	0.25	-0.1	0.013	-0.3	-0.19	-0.28	1	0.67	-0.069	-0.35
Testing_Rate -	-0.0092	-0.2	-0.087	-0.32	-0.16	-0.34	0.67	1	-0.34	-0.3
Hospitalization_Rate -	-0.086	0.19	0.21	0.17	0.23	0.36	-0.069	-0.34	1	0.13
Recovery_Rate -	-0.22	0.092	0.074	0.19	0.21	0.17	-0.35	-0.3	0.13	1
	3-YrDiabetes2015-17 -	DiabetesPercentage -	HeartDiseaseMortality -	StrokeMortality -	Smokers_Percentage -	RespMortalityRate2014 -	Incident_Rate	Testing_Rate	Hospitalization_Rate -	Recovery_Rate

6/15/2020	final proj
-----------	------------

Let's further explore the relationships between the labels and attribute using a **raincloud plot** (combination of a KDE, histogram, strip plot, and box plot.) target variable: Recovery_Rate.

```
In [70]: fig, axs = plt.subplots(nrows=2)
         sns.distplot(
             join_countries_and_states['Mortality_Rate'],
             ax=axs[0]
         sns.stripplot(
             join_countries_and_states['Mortality_Rate'],
             jitter=0.4,
             size=3,
             ax=axs[1],
             alpha=0.3
         sns.boxplot(
             join_countries_and_states['Mortality_Rate'],
             width=0.3,
             ax=axs[1],
             showfliers=False,
         spacer = np.max(join countries and states['Mortality Rate']) * 0.05
         xmin = np.min(join countries and states['Mortality Rate']) - spacer
         xmax = np.max(join countries and states['Mortality Rate']) + spacer
         axs[0].set xlim((xmin, xmax))
         axs[1].set xlim((xmin, xmax))
         plt.subplots adjust(hspace=0)
```

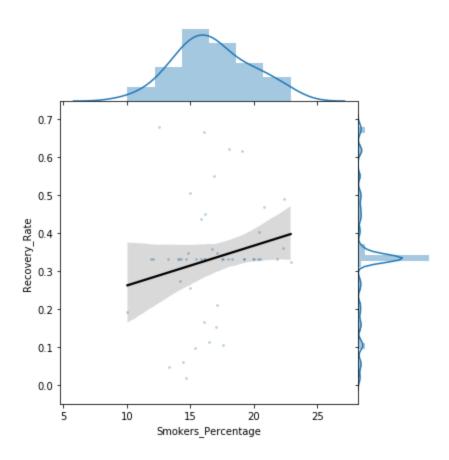


In [71]: join_countries_and_states['Mortality_Rate'].describe()

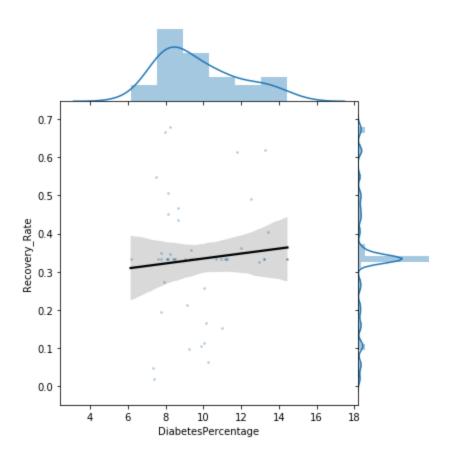
Out[71]: count 50.000000 3.601043 mean 1.529845 std min 0.453956 25% 2.680235 50% 3.473322 75% 4.603837 7.495697 max

Name: Mortality_Rate, dtype: float64

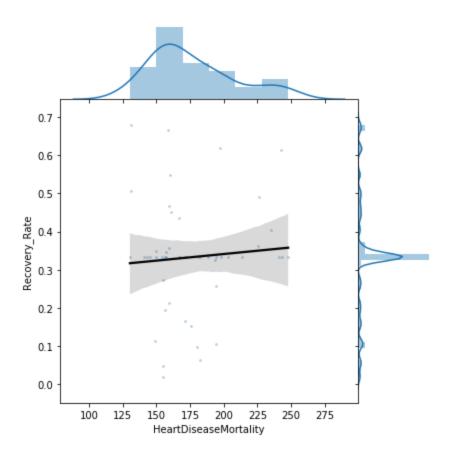
```
In [72]: sns.jointplot(
    x='Smokers_Percentage',
    y='Recovery_Rate',
    data=join_countries_and_states,
    stat_func=None,
    kind="reg",
    ratio=4,
    space=0,
    scatter_kws={
        's': 3,
        'alpha': 0.25
    },
    line_kws={
        'color': 'black'
    });
```



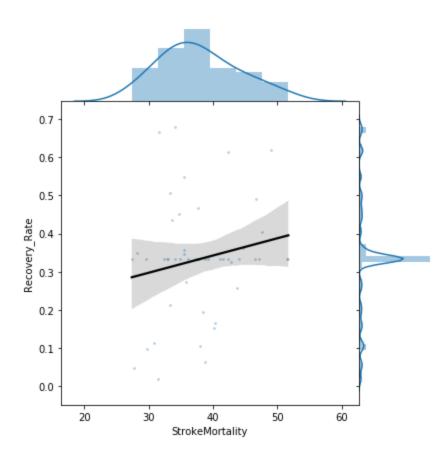
```
In [73]: sns.jointplot(
    x='DiabetesPercentage',
    y='Recovery_Rate',
    data=join_countries_and_states,
    stat_func=None,
    kind="reg",
    ratio=4,
    space=0,
    scatter_kws={
        's': 3,
        'alpha': 0.25
    },
    line_kws={
        'color': 'black'
    });
```



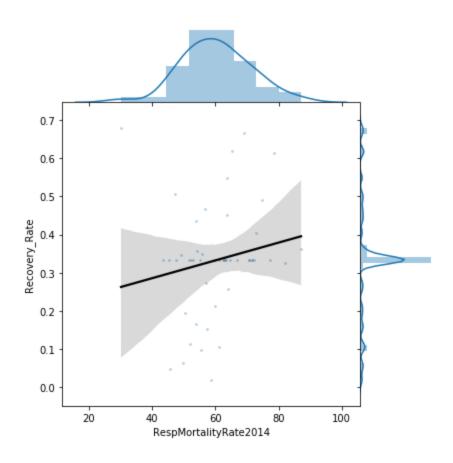
```
In [74]: sns.jointplot(
    x='HeartDiseaseMortality',
    y='Recovery_Rate',
    data=join_countries_and_states,
    stat_func=None,
    kind="reg",
    ratio=4,
    space=0,
    scatter_kws={
        's': 3,
        'alpha': 0.25
    },
    line_kws={
        'color': 'black'
    });
```



```
In [75]: sns.jointplot(
    x='StrokeMortality',
    y='Recovery_Rate',
    data=join_countries_and_states,
    stat_func=None,
    kind="reg",
    ratio=4,
    space=0,
    scatter_kws={
        's': 3,
        'alpha': 0.25
    },
    line_kws={
        'color': 'black'
    });
```



```
In [76]: sns.jointplot(
    x='RespMortalityRate2014',
    y='Recovery_Rate',
    data=join_countries_and_states,
    stat_func=None,
    kind="reg",
    ratio=4,
    space=0,
    scatter_kws={
        's': 3,
        'alpha': 0.25
    },
    line_kws={
        'color': 'black'
    });
```



MODEL

In [78]: X.head()

Out[78]:

3YrDiabetes2015- DiabetesPercentage HeartDiseaseMortality StrokeMortality Smokers_Percentage RespMortalityRate2014 Incident_Rate
17

State							
Alabama	31.437500	14.407463	243.595522	51.450746	19.989231	77.282985	100.492717
Alaska	26.250000	8.667857	159.250000	37.659091	20.806449	56.793448	52.530410
Arizona	146.428571	10.060000	148.826667	30.900000	16.483911	51.968667	64.901548
Arkansas	29.833333	13.432000	235.172000	47.681333	20.388849	72.727067	67.361213
California	207.318182	8.505172	153.908621	37.891379	12.091600	52.153621	77.766063

```
In [79]: train, test = train_test_split(join_countries_and_states, test_size=0.2, random_state=42)
```

```
In [80]: X_train = train.loc[:, ['3-YrDiabetes2015-17', 'DiabetesPercentage', 'HeartDiseaseMortality', 'StrokeMortalit
y', 'Smokers_Percentage', 'RespMortalityRate2014', 'Incident_Rate', 'Testing_Rate', 'Hospitalization_Rate']]
X_train['Hospitalization_Rate']=X_train['Hospitalization_Rate'].fillna(np.mean(X_train['Hospitalization_Rate']))
```

Y_train = train['Recovery_Rate']

```
In [81]:
         def normalize(data):
             Args:
                 data: a dataframe
             Returns:
                 the normalized version of input data with NAN values filled with 0's
             new df = data.copy()
             for i in range(len(data.columns)):
                 std = np.std(data[data.columns[i]])
                 mean = np.mean(data[data.columns[i]])
                 for j in range(len(data[data.columns[i]])):
                     x = data[data.columns[i]][j]
                     if (std == 0):
                         new df[new df.columns[i]] = 0
                     else:
                          new df[new df.columns[i]][j] = (x - mean) / std
                     j+=1
                 i += 1
             return new df
In [82]:
         X train df = pd.DataFrame(X train)
         X train = normalize(X_train_df)
In [83]:
         X test = test.loc[:, ['3-YrDiabetes2015-17', 'DiabetesPercentage', 'HeartDiseaseMortality', 'StrokeMortality',
         'Smokers Percentage', 'RespMortalityRate2014', 'Incident Rate', 'Testing Rate', 'Hospitalization Rate']]
         X test['Hospitalization Rate']=X test['Hospitalization Rate'].fillna(np.mean(X test['Hospitalization Rate']))
         Y test = test['Recovery Rate']
```

3-

In [84]: X_test.head()

Out[84]:

Indiana	33.537037	11.191304	188.691304	41.625000	19.932077	70.725109	162.607117
South Carolina	39.235294	13.267391	197.263043	49.047826	18.045662	65.362391	84.286132
New Mexico	33.222222	9.157576	159.251515	33.366667	17.066621	59.979697	107.823953
Virginia	a 28.473684 10.983459		175.609023	40.103008	16.991193	57.262180	101.843503
.ouisiana	isiana 34.406250 13.232812		241.418750	47.168750	21.792924	62.449688	512.913545

YrDiabetes2015- DiabetesPercentage HeartDiseaseMortality StrokeMortality Smokers_Percentage RespMortalityRate2014 Incident_Rate

```
In [85]: X_test_df = pd.DataFrame(X_test)
X_test = normalize(X_test_df)
```

In [86]: model = LinearRegression()
 model.fit(X_train, Y_train)
 Y_pred = model.predict(X_test)

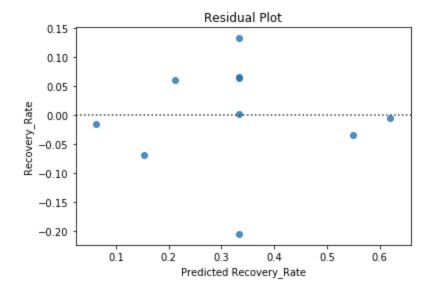
```
In [87]:
         Y_test
Out[87]: State
                            0.333147
         Indiana
         South Carolina
                            0.619821
                            0.212458
         New Mexico
                            0.152490
         Virginia
                            0.333147
         Louisiana
                            0.333147
         Wisconsin
         Nebraska
                            0.333147
                            0.549296
         Montana
         North Carolina
                            0.333147
         Maryland
                            0.062551
         Name: Recovery_Rate, dtype: float64
         Y pred.tolist()
In [88]:
Out[88]: [0.2708828245566953,
          0.31578603195037586,
          0.2827747159922025,
          0.4172656091739701,
          0.20151366215404656,
          0.5389913079338842,
          0.331777186812271,
          0.3497514815884232,
          0.2685228490788606,
          0.37148445162306354]
In [89]: metrics.mean_absolute_error(Y_test, Y_pred)
Out[89]: 0.16133412405652992
In [90]: metrics.mean_squared_error(Y_test, Y_pred)
Out[90]: 0.037050015723393295
In [91]:
         np.sqrt(metrics.mean_squared_error(Y_test, Y_pred))
Out[91]: 0.19248380639262436
```

```
In [93]: sns.residplot(Y_test, Y_test-Y_pred, lowess=True)
    plt.title('Residual Plot')
    plt.xlabel('Actual Recovery_Rate')
    plt.xlabel('Predicted Recovery_Rate')
```

/srv/conda/envs/data100/lib/python3.7/site-packages/statsmodels/nonparametric/smoothers_lowess.py:165: Runti meWarning:

invalid value encountered in greater_equal

Out[93]: Text(0.5, 0, 'Predicted Recovery_Rate')



```
In [94]: from sklearn.model_selection import KFold

kf = KFold(n_splits=4)
   kf.get_n_splits(X_train)
   KFold(n_splits = 4, random_state = 42, shuffle = False)

from sklearn.model_selection import cross_val_score, cross_val_predict
   from sklearn import metrics
   scores = cross_val_score(model, X_train, Y_train, cv=6, scoring='neg_mean_squared_error')
   scores
```

/srv/conda/envs/data100/lib/python3.7/site-packages/sklearn/model selection/ split.py:296: FutureWarning:

Setting a random_state has no effect since shuffle is False. This will raise an error in 0.24. You should le ave random_state to its default (None), or set shuffle=True.

```
Out[94]: array([-0.01103514, -0.02494853, -0.05712099, -0.04246981, -0.06255325, -0.01658819])
```

```
In [95]: X['Hospitalization_Rate'] = X['Hospitalization_Rate'].fillna(np.mean(X['Hospitalization_Rate']))
join_countries_and_states['Predicted Recovery Rates'] = model.predict(normalize(X))
join_countries_and_states.loc[:, ['Recovery_Rate', 'Predicted Recovery Rates']]
```

Out[95]:

	Recovery_Rate	Predicted Recovery Rates
State		
Alabama	0.333147	0.317520
Alaska	0.468153	0.422385
Arizona	0.114098	0.316995
Arkansas	0.403096	0.382509
California	0.333147	0.255202
Colorado	0.333147	0.343062
Connecticut	0.333147	0.251295
Delaware	0.166667	0.292256
Florida	0.333147	0.302038
Georgia	0.333147	0.330098
Hawaii	0.679443	0.419838
Idaho	0.273716	0.341614
Illinois	0.333147	0.299163
Indiana	0.333147	0.340839
Iowa	0.435734	0.395865
Kansas	0.333147	0.390900
Kentucky	0.361655	0.426445
Louisiana	0.333147	0.246747
Maine	0.451004	0.351938
Maryland	0.062551	0.412817
Massachusetts	0.333147	0.202540
Michigan	0.105128	0.316050
Minnesota	0.506111	0.493486

Recovery_Rate Predicted Recovery Rates

State		
Mississippi	0.333147	0.376724
Missouri	0.333147	0.388777
Montana	0.549296	0.368553
Nebraska	0.333147	0.362522
Nevada	0.333147	0.319818
New Hampshire	0.348733	0.385368
New Jersey	0.333147	0.228383
New Mexico	0.212458	0.305412
New York	0.098824	0.068168
North Carolina	0.333147	0.315254
North Dakota	0.346591	0.385804
Ohio	0.333147	0.430326
Oklahoma	0.614604	0.379442
Oregon	0.333147	0.387219
Pennsylvania	0.333147	0.364085
Rhode Island	0.048319	0.225341
South Carolina	0.619821	0.367216
South Dakota	0.357977	0.317775
Tennessee	0.490818	0.334176
Texas	0.256950	0.304318
Utah	0.193692	0.257791
Vermont	0.018680	0.283603
V irginia	0.152490	0.419200
Washington	0.333147	0.229023

Recovery_Rate Predicted Recovery Rates

State

West Virginia	0.324841	0.251634
Wisconsin	0.333147	0.512432
Wyoming	0.666667	0.323784

Line Plot Overtime From Dates 4-19-2020 to 5-10-2020

```
In [96]: april_19 = pd.read_csv("04-19-2020.csv")
    april_20 = pd.read_csv("04-20-2020.csv")
    april_21 = pd.read_csv("04-21-2020.csv")
    april_22 = pd.read_csv("04-22-2020.csv")
    april_23 = pd.read_csv("04-23-2020.csv")
    april_24 = pd.read_csv("04-24-2020.csv")
    april_25 = pd.read_csv("04-25-2020.csv")
    april_26 = pd.read_csv("04-26-2020.csv")
    april_27 = pd.read_csv("04-27-2020.csv")
    april_28 = pd.read_csv("04-28-2020.csv")
    april_29 = pd.read_csv("04-29-2020.csv")
    april_30 = pd.read_csv("04-29-2020.csv")
```

Out[98]:

	Province_State	Country_Region	Last_Update	Lat	Long_	Confirmed	Deaths	Recovered	Active	FIPS	Incident_Rate	People_Tested
0	Alabama	US	2020-04-19 23:41:01	32.3182	-86.9023	4888	157	NaN	4731.0	1.0	104.246265	45712.0
1	Alaska	US	2020-04-19 23:41:01	61.3707	-152.4044	319	9	153.0	310.0	2.0	53.366881	9895.0
2	American Samoa	US	NaN	-14.2710	-170.1320	0	0	NaN	NaN	60.0	0.000000	3.0
3	Arizona	US	2020-04-19 23:41:01	33.7298	-111.4312	4933	184	994.0	4749.0	4.0	67.772933	52990.0
4	Arkansas	US	2020-04-19 23:41:01	34.9697	-92.3731	1781	39	721.0	1742.0	5.0	68.790322	24209.0
53	Virginia	US	2020-05-11 02:32:34	37.7693	-78.1700	24081	839	3201.0	20041.0	51.0	304.544069	143055.0
54	Washington	US	2020-05-11 02:32:34	47.4009	-121.4905	16891	931	NaN	15960.0	53.0	223.739546	242989.0
55	West Virginia	US	2020-05-11 02:32:34	38.4912	-80.9545	1360	54	775.0	531.0	54.0	102.796830	62644.0
56	Wisconsin	US	2020-05-11 02:32:34	44.2685	-89.6165	10219	400	5014.0	4805.0	55.0	197.484970	115382.0
57	Wyoming	US	2020-05-11 02:32:34	42.7560	-107.3025	662	7	443.0	212.0	56.0	133.134905	12064.0

1368 rows × 18 columns

```
In [99]: grouped_by_date = all_dates.groupby('Last_Update').sum()
grouped_by_date = grouped_by_date.loc[:, ['Confirmed', 'Deaths']]
```

```
In [100]: grouped_by_date.plot(figsize = (10, 5))
    plt.xticks(rotation=45)
    plt.ylabel('Number of Cases')
    plt.title('Latest Number of Confirmed Cases and Deaths')
```

Out[100]: Text(0.5, 1.0, 'Latest Number of Confirmed Cases and Deaths')



