```
import requests
In [1]:
         import io
         import sys
         import json
         import itertools
         from IPython.display import Image
         from bs4 import BeautifulSoup
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         %matplotlib inline
         import seaborn as sns
         import plotly.express as px
         import sqlite3 as sql
         from sklearn.model_selection import train_test_split, cross_val_score
         import statsmodels.api as sm
         import statsmodels.stats.api as sms
         import scipy.stats as stats
         from scipy.stats import chisquare
         from sklearn.linear_model import LogisticRegression
         from sklearn.metrics import confusion matrix, precision score, recall score, accuracy score, f1 score, precision recall curve, roc curve, auc, mean squared error
         from statsmodels.tsa.stattools import adfuller
         from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
         from imblearn.over sampling import RandomOverSampler
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.ensemble import GradientBoostingClassifier
        import warnings
In [2]:
```

Imports

A class of functions that pull data from sourses and store the dataframes in a json.

warnings.filterwarnings("ignore")

```
class Json():
In [3]:
             global json_storage
             d=\{\}
             s=json.dumps(d)
             json storage = json.loads(s)
             def __init__(self, df_name):
                 self.df name=df name
             def csv(self, url):
                 if url[0]=='r':
                     download = requests.get(url).content
                     df = pd.read csv(io.StringIO(download.decode('utf-8')))
                     df=pd.DataFrame(df)
                     json_storage[self.df_name]=[url,df]
                 else:
                     df=pd.read_csv(url)
                     df=pd.DataFrame(df)
                     json_storage[self.df_name]=[url,df]
                 return(df)
             def excel(self, e, s):
                 df=pd.read_excel(e, s)
                 json_storage[self.df_name]=[e,df]
                 return(df)
             def web scrape(self, u, c):
                 url = requests.get(u).text
                 soup = BeautifulSoup(url, 'lxml')
                 table = soup.find('table')
```

```
table_rows = table.find_all('tr')
l = []
for tr in table_rows:
    td = tr.find_all('td')
    row = [tr.text for tr in td]
    l.append(row)
    df=pd.DataFrame(l, columns=c)
    json_storage[self.df_name]=[u,df]
    return(df)

def file(self, f):
    file = open(f, "r")
    read=file.read()
    json_storage[self.df_name]=read
    return(read)
```

A class of graphing functions.

```
In [ ]: | class graph:
             def choropleth(self, d, location, lm, c, af, s, t):
                 fig = px.choropleth(d , locations = location, locationmode = lm, color = c
                             ,animation frame=af, scope=s)
                 fig.update_layout(title_text = t)
                 fig.show()
             def bar_chart(self, x, y, t, x2, y2):
                 fig, ax = plt.subplots(figsize=(50,150))
                 width=.25
                 ax.barh(x, y, width, color='red')
                 for i, v in enumerate(y):
                     ax.text(v, i, str(v), color='blue')
                 plt.rcParams.update({'font.size': 20})
                 plt.title(t, fontsize=40)
                 plt.xlabel(x2, fontsize=30)
                 plt.ylabel(y2, fontsize=30)
                 plt.show()
```

A function that prints summary statistics.

```
In []:
    def desc(x):
        print(x.name)
        print('Count:', x.count())
        print('Mean:', x.mean())
        print('Standard Deviation:', x.std())
        print('Min:', x.min())
        print("Q1 quantile: ", np.quantile(x, .25))
        print("Q2 quantile: ", np.quantile(x, .5))
        print("Q3 quantile: ", np.quantile(x, .75))
        print('Max:', x.max())
```

A function that creates a dataframe of covid-19 totals and prints the totals.

```
t=dd1.loc[dd1[name]==f'{i}']['Total Tests']
t=int(t)
a=dd1.loc[dd1[name]==f'{i}']['Active_Cases']
a=int(a)
if name=='Country':
   try:
        v=dd2.loc[dd2[name]==f'{i}'].iloc[[-1]]['total_vaccinations']
        v=0
elif name == 'State':
    try:
        v=dd2.loc[dd2[name]==f'{i}']['Vaccines_Administered']
    except:
v=int(v)
try:
    tp=(t/p)*100
    if tp>=100:
       tp=99.99
except:
    tp=0
try:
   ct=(c/t)*100
   if ct>=100:
        ct=99.99
except:
   ct=0
try:
    rc=(r/c)*100
   if rc>=100:
        rc=99.99
except:
    rc=0
try:
    dc = (d/c) * 100
    if dc>=100:
        dc=99.99
except:
    dc=0
try:
    vp = (v/p) * 100
   if vp>=100:
        vp=99.99
except:
    vp=0
print(f'{i}')
print('Population Total:', p)
print('Tested Total:', t)
print('Tested Percentage of Population:', tp, '%')
print('Cases Total:', c)
print('Active Cases:', a)
print('Confirmed Cases Percentage of Tested:', ct, '%')
print('Recoveries Total:', r)
print('Recoveries Percentage of Confirmed Cases:', rc, '%')
print('Deaths Total:', d)
print('Deaths Percentage of Confirmed Cases:', dc, '%')
print('Vaccinations:', v)
print('Vaccinations of Population:', vp, '%')
print('----')
ddd \cdot loc[len(ddd \cdot index)] = [f'\{i\}', p, t, tp, c, a, ct, r, rc, d, dc, v, vp]
```

A function that prints a confusion matrix.

```
def con_mat(y, y_pred):
    print('\nConfusion Matrix')
    print('-----')
    cm=pd.DataFrame(confusion_matrix(y, y_pred))
    print(cm)
    plt.matshow(cm)
    plt.title('Predicted')
    plt.ylabel('Actual')
    plt.colorbar()
    cm
```

**Imports** 

A function that prints a classification metrics.

```
In [4]: def Metrics(labels, preds):
             actual_pos = labels == 1
             actual_neg = labels == 0
             tp = (preds == 1) & (actual_pos)
             fp = (preds == 1) & (actual neg)
             tn = (preds == 0) & (actual_neg)
             fn = (preds == 0) & (actual pos)
             precision_score(labels, preds)
             recall_score(labels, preds)
             f1=f1_score(labels, preds)
             accuracy=accuracy_score(labels, preds)
             specificity = sum(tn) / (sum(tn) + sum(fn))
             print("Precision Score: {}".format(precision))
             print("Recall Score: {}".format(recall))
             print("F1 Score: {}".format(f1))
             print("Accuracy Score: {}".format(accuracy))
             print("Specificity Score: {}".format(specificity))
             lr_precision, lr_recall, _ = precision_recall_curve(labels, preds)
             plt.plot(lr_recall, lr_precision, marker='o')
             plt.title('Precision Recall Tradeoff')
             plt.xlabel('Recall')
             plt.ylabel('Precision')
```

a function that prints roc auc.

```
In [ ]: | def roc(y, y_hat):
             false_positive_rate, true_positive_rate, thresholds = roc_curve(y, y_hat)
             roc_auc = auc(false_positive_rate, true_positive_rate)
             sns.set_style('darkgrid', {'axes.facecolor': '0.9'})
             plt.figure(figsize=(10, 8))
             lw = 2
             plt.plot(false_positive_rate, true_positive_rate, color='darkorange',
                      lw=lw, label='ROC curve')
             plt.plot([0, 1], [0, 1], color='navy', lw=lw, linestyle='--')
             plt.xlim([0.0, 1.0])
             plt.ylim([0.0, 1.05])
             plt.yticks([i/20.0 for i in range(21)])
             plt.xticks([i/20.0 for i in range(21)])
             plt.xlabel('False Positive Rate')
             plt.ylabel('True Positive Rate')
             plt.title('Receiver operating characteristic (ROC) Curve')
             plt.legend(loc='lower right')
```

```
print('AUC: {}'.format(auc(false_positive_rate, true_positive_rate)))
plt.show()
```

A function that prints muticlass metrics.

```
def multiclass(model,x,y,y_hat,classes):
    sort=sorted(set(classes))
    for i,l,c in zip(y,y_hat,sort):
        print(f'Class:{c}')
        Metrics(i,l)
        roc(i,l)
        cv_score = cross_val_score(model, x, l, cv=5, scoring='roc_auc')
        mean_cv_score = np.mean(cv_score)
        print(f"Cross Validated ROC AUC score: {mean_cv_score}")
        print('_______')
```

A function that prints time series analysis.

```
In [ ]: def forecast(dd, w, start, end, s, l, date1, date2, title):
                     date3=dd.index[0]
                     roll mean = dd.rolling(window=w, center=False).mean()
                     roll_std = dd.rolling(window=w, center=False).std()
                     fig = plt.figure(figsize=(12,7))
                     plt.plot(dd, color='blue', label='Original')
                     plt.plot(roll mean, color='red', label='Rolling Mean')
                     plt.plot(roll_std, color='black', label = 'Rolling Std')
                     plt.legend(loc='best')
                     plt.title(f'{title} Trend')
                     plt.show(block=False)
                     dftest = adfuller(dd)
                     dfoutput = pd.Series(dftest[0:4], index=['Test Statistic', 'p-value', '#Lags Used', 'Number of Observations Used'])
                     for key,value in dftest[4].items():
                         dfoutput['Critical Value (%s)'%key] = value
                     print (f'{title} Dickey-Fuller test results: \n')
                     print(dfoutput)
                     fig, ax = plt.subplots(figsize=(16,3))
                     plot_acf(dd, ax=ax, lags=1)
                     plt.title(f'{title} autocorrelation')
                     plt.show()
                     fig, ax = plt.subplots(figsize=(16,3))
                     plot_pacf(dd, ax=ax, lags=1)
                     plt.title(f'{title} partial autocorrelation')
                     plt.show()
                     print(f'{title} AIC Scores:')
                     # Define the p, d and q parameters to take any value between 0 and 2
                     p = d = q = range(0, 2)
                     # Generate all different combinations of p, d and q triplets
                     pdq = list(itertools.product(p, d, q))
                     # Generate all different combinations of seasonal p, d and q triplets
                     pdgs = [(x[0], x[1], x[2], 7) for x in list(itertools.product(p, d, q))]
                     # Run a grid with pdg and seasonal pdg parameters calculated above and get the best AIC value
                     ans = []
                     for comb in pdq:
                         for combs in pdqs:
```

```
try:
            mod = sm.tsa.statespace.SARIMAX(dd,
                                            order=comb,
                                            seasonal_order=combs,
                                            enforce_stationarity=False,
                                            enforce_invertibility=False)
            output = mod.fit()
            ans.append([comb, combs, output.aic])
            print('ARIMA {} x {}12 : AIC Calculated ={}'.format(comb, combs, output.aic))
        except:
# Find the parameters with minimal AIC value
ans_df = pd.DataFrame(ans, columns=['pdq', 'pdqs', 'aic'])
values=ans_df.loc[ans_df['aic'].idxmin()]
# Plug the optimal parameter values into a new SARIMAX model
ARIMA MODEL = sm.tsa.statespace.SARIMAX(dd,
                                        order=values['pdq'],
                                        seasonal order=values['pdgs'],
                                        enforce_stationarity=False,
                                        enforce_invertibility=False)
# Fit the model and print results
output = ARIMA_MODEL.fit()
print(f'{title} ARIMA:', output.summary())
output.plot_diagnostics(figsize=(15, 18))
plt.show()
# Get predictions and calculate confidence intervals
pred = output.get_prediction(start=pd.to_datetime(date3), dynamic=False)
pred_conf = pred.conf_int()
# Plot real vs predicted values along with confidence interval
# Plot observed values
ax = dd.plot(label='observed', figsize=(15, 18))
# Plot predicted values
pred.predicted mean.plot(ax=ax, label='One-step ahead Forecast', alpha=0.9)
# Plot the range for confidence intervals
ax.fill_between(pred_conf.index,
                pred_conf.iloc[:, 0],
                pred_conf.iloc[:, 1], color='g', alpha=0.5)
# Set axes labels
plt.title(f'{title} actual and predicted values with confidence interval')
ax.set_xlabel('Date')
ax.set_ylabel(title)
plt.legend()
plt.show()
# Get the real and predicted values
forecasted = pred.predicted_mean
truth = dd
# Compute the mean square error
rmse=np.sqrt(mean_squared_error(truth, forecasted))
```

```
print(f'{title} Root Mean Squared Error:{round(rmse, 2)}')
f=output.forecast(steps=s)
prediction = output.get_forecast(steps=s)
pred_conf_f = prediction.conf_int()
print(f'{title} Forecast:')
print(f)
start[title]=dd.iloc[-1]
end[title]=f.iloc[-1]
# Plot future predictions with confidence intervals
ax = dd.plot(label='observed', figsize=(20, 15))
prediction.predicted_mean.plot(ax=ax, label='Forecast')
ax.fill_between(pred_conf_f.index,
               pred_conf_f.iloc[:, 0],
               pred_conf_f.iloc[:, 1], color='k', alpha=0.25)
ax.set_xlabel('Date')
ax.set_ylabel(title)
plt.title(f'{title} trend and future predictions with confidence interval')
plt.legend()
plt.show()
```

A function that returns percentage change.