```
Stock_Market = pd.read_csv('Stock market data.csv')
         df = pd.DataFrame(Stock_Market, columns=['Year', 'Month', 'Interest_Rate', 'Unemployment_Rate', 'Stock_Index_Price'])
         print (df)
            Year
                 Month Interest_Rate Unemployment_Rate Stock_Index_Price
            2017
                    12
                                 2.75
                                                    5.3
                                 2.50
                                                                      1394
        1
            2017
                     11
                                                    5.3
                                 2.50
                                                    5.3
                                                                      1357
        2
            2017
                     10
                                 2.50
                                                                      1293
        3
            2017
                     9
                                                    5.3
                                                                      1256
        4
            2017
                     8
                                 2.50
                                                    5.4
            2017
                                 2.50
                                                                      1254
        5
                     7
                                                    5.6
            2017
                                 2.50
                                                                      1234
        6
                     6
                                                    5.5
            2017
                                 2.25
                                                                      1195
        7
                     5
                                                    5.5
            2017
                                 2.25
                                                                      1159
        8
                     4
                                                    5.5
                                 2.25
        9
            2017
                     3
                                                    5.6
                                                                      1167
            2017
                     2
                                 2.00
                                                    5.7
                                                                      1130
        10
            2017
                                 2.00
                                                                      1075
        11
                     1
                                                    5.9
        12
           2016
                     12
                                 2.00
                                                    6.0
                                                                      1047
        13
            2016
                     11
                                 1.75
                                                    5.9
                                                                       965
            2016
                     10
                                 1.75
                                                    5.8
                                                                       943
        14
        15
            2016
                     9
                                 1.75
                                                    6.1
                                                                       958
        16
           2016
                     8
                                 1.75
                                                    6.2
                                                                       971
           2016
                                                                       949
        17
                     7
                                 1.75
                                                    6.1
        18
            2016
                     6
                                 1.75
                                                    6.1
                                                                       884
        19
           2016
                     5
                                 1.75
                                                    6.1
                                                                       866
        20
           2016
                     4
                                 1.75
                                                    5.9
                                                                       876
           2016
                                 1.75
                                                                       822
        21
                     3
                                                    6.2
           2016
                                                                       704
        22
                     2
                                 1.75
                                                    6.2
        23 2016
                                 1.75
                                                                       719
                                                    6.1
                     1
In [3]:
         from pandas import DataFrame
         import matplotlib.pyplot as plt
         plt.scatter(df['Interest_Rate'], df['Stock_Index_Price'], color='red')
         plt.title('Stock Index Price Vs Interest Rate', fontsize=14)
         plt.xlabel('Interest Rate', fontsize=14)
         plt.ylabel('Stock Index Price', fontsize=14)
         plt.grid(True)
         plt.show()
                     Stock Index Price Vs Interest Rate
          1500
          1400
        P1300
1300
        Stock Index
          1100
          1000
           900
            800
           700
                         2.0
                                2.2
                                        2.4
                                               2.6
                                                       2.8
                 1.8
                              Interest Rate
In [4]:
         plt.scatter(df['Unemployment_Rate'], df['Stock_Index_Price'], color='green')
         plt.title('Stock Index Price Vs Unemployment Rate', fontsize=14)
         plt.xlabel('Unemployment Rate', fontsize=14)
         plt.ylabel('Stock Index Price', fontsize=14)
         plt.grid(True)
         plt.show()
                 Stock Index Price Vs Unemployment Rate
          1500
          1400
        Price 1300 -
        Stock Index
          1100
          1000
           900
           800
           700
                    5.4
                                                     6.2
                            5.6
                                    5.8
                                             6.0
                          Unemployment Rate
         from sklearn import linear_model
         import statsmodels.api as sm
         X = df[['Interest_Rate', 'Unemployment_Rate']] # here we have 2 variables for multiple regression. If you just want to use one variable for simple linear regression, then use X = df
         Y = df['Stock_Index_Price']
         # with sklearn
         regr = linear_model.LinearRegression()
         regr.fit(X, Y)
         print('Intercept: \n', regr.intercept_)
         print('Coefficients: \n', regr.coef_)
        Intercept:
         1798.403977625855
        Coefficients:
         [ 345.54008701 -250.14657137]
In [6]:
         # prediction with sklearn
         New_Interest_Rate = 2.75
         New_Unemployment_Rate = 5.3
         print ('Predicted Stock Index Price: \n', regr.predict([[New_Interest_Rate , New_Unemployment_Rate]]))
         # with statsmodels
         X = sm.add_constant(X) # adding a constant
         model = sm.OLS(Y, X).fit()
         predictions = model.predict(X)
         print_model = model.summary()
         print(print_model)
        Predicted Stock Index Price:
         [1422.86238865]
                                   OLS Regression Results
        ______
        Dep. Variable:
                           Stock_Index_Price
                                               R-squared:
                                                                               0.898
        Model:
                                         0LS
                                              Adj. R-squared:
                                                                               0.888
        Method:
                               Least Squares
                                              F-statistic:
                                                                               92.07
                            Sun, 30 Apr 2023
                                              Prob (F-statistic):
        Date:
                                                                            4.04e-11
                                    12:18:42
        Time:
                                              Log-Likelihood:
                                                                             -134.61
        No. Observations:
                                          24
                                              AIC:
                                                                               275.2
        Df Residuals:
                                          21
                                               BIC:
                                                                               278.8
        Df Model:
                                           2
                                   nonrobust
        Covariance Type:
        ______
                                                                                    0.975]
                                                                         [0.025
                               coef
                                       std err
                                                              P>|t|
                          1798.4040
        const
                                       899.248
                                                   2.000
                                                              0.059
                                                                        -71.685
                                                                                  3668.493
        Interest_Rate
                           345.5401
                                      111.367
                                                   3.103
                                                              0.005
                                                                        113.940
                                                                                   577.140
                                      117.950
        Unemployment_Rate -250.1466
                                                   -2.121
                                                              0.046
                                                                       -495.437
                                                                                    -4.856
        ______
        Omnibus:
                                               Durbin-Watson:
                                                                               0.530
                                       2.691
        Prob(Omnibus):
                                       0.260
                                               Jarque-Bera (JB):
                                                                               1.551
        Skew:
                                      -0.612
                                              Prob(JB):
                                                                               0.461
        Kurtosis:
                                       3.226 Cond. No.
                                                                                394.
        ______
        Notes:
        [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
        C:\Users\dnhac\anaconda3\lib\site-packages\statsmodels\tsa\tsatools.py:142: FutureWarning: In a future version of pandas all arguments of concat except for the argument 'objs' will
        be keyword-only
         x = pd.concat(x[::order], 1)
In [7]:
         import tkinter as tk
         import statsmodels.api as sm
         X = df[['Interest_Rate', 'Unemployment_Rate']] # here we have 2 input variables for multiple regression. If you just want to use one variable for simple linear regression, then use
         Y = df['Stock_Index_Price'] # output variable (what we are trying to predict)
         # with sklearn
         regr = linear_model.LinearRegression()
         regr.fit(X, Y)
         print('Intercept: \n', regr.intercept_)
         print('Coefficients: \n', regr.coef_)
         # with statsmodels
         X = sm.add_constant(X) # adding a constant
         model = sm.OLS(Y, X).fit()
         predictions = model.predict(X)
        Intercept:
         1798.403977625855
        Coefficients:
         [ 345.54008701 -250.14657137]
        C:\Users\dnhac\anaconda3\lib\site-packages\statsmodels\tsa\tsatools.py:142: FutureWarning: In a future version of pandas all arguments of concat except for the argument 'objs' will
        be keyword-only
          x = pd.concat(x[::order], 1)
In [ ]:
         # tkinter GUI
         root= tk.Tk()
         canvas1 = tk.Canvas(root, width = 1200, height = 450)
         canvas1.pack()
         # with sklearn
         Intercept_result = ('Intercept: ', regr.intercept_)
         label_Intercept = tk.Label(root, text=Intercept_result, justify = 'center')
         canvas1.create_window(260, 220, window=label_Intercept)
         # with sklearn
         Coefficients_result = ('Coefficients: ', regr.coef_)
         label_Coefficients = tk.Label(root, text=Coefficients_result, justify = 'center')
         canvas1.create_window(260, 240, window=label_Coefficients)
         # with statsmodels
         print_model = model.summary()
         label_model = tk.Label(root, text=print_model, justify = 'center', relief = 'solid', bg='LightSkyBlue1')
         canvas1.create_window(800, 220, window=label_model)
         # New_Interest_Rate label and input box
         label1 = tk.Label(root, text='Type Interest Rate: ')
         canvas1.create_window(100, 100, window=label1)
         entry1 = tk.Entry (root) # create 1st entry box
         canvas1.create_window(270, 100, window=entry1)
         # New_Unemployment_Rate label and input box
         label2 = tk.Label(root, text=' Type Unemployment Rate: ')
         canvas1.create_window(120, 120, window=label2)
         entry2 = tk.Entry (root) # create 2nd entry box
         canvas1.create_window(270, 120, window=entry2)
         def values():
             global New_Interest_Rate #our 1st input variable
             New_Interest_Rate = float(entry1.get())
             global New_Unemployment_Rate #our 2nd input variable
             New_Unemployment_Rate = float(entry2.get())
             Prediction_result = ('Predicted Stock Index Price: ', regr.predict([[New_Interest_Rate ,New_Unemployment_Rate]]))
             label_Prediction = tk.Label(root, text= Prediction_result, bg='orange')
             canvas1.create_window(260, 280, window=label_Prediction)
         button1 = tk.Button (root, text='Predict Stock Index Price', command=values, bg='orange') # button to call the 'values' command above
         canvas1.create_window(270, 150, window=button1)
         root.mainloop()
In [ ]:
```

In [2]:

import pandas as pd