# DATA ANALYSIS: HAPPINESS DETERMINANTS PER COUNTRY - 2023

The 2023 data resulted from the merger of the 3 csv files below:

- 1. Crime.csv (Crime Rate by Country 2023 retrieved from <a href="https://worldpopulationreview.com/country-rankings/crime-rate-by-country">https://worldpopulationreview.com/country-rankings/crime-rate-by-country</a>),
- 2. Happiness.csv (World Happiness Report 2023 retrieved from <a href="https://www.kaggle.com/datasets/ajaypalsinghlo/world-happiness-report-2023">https://www.kaggle.com/datasets/ajaypalsinghlo/world-happiness-report-2023</a>)), and
- 3. Freedom.csv (2023 Heritage Foundation Economic Freedom data retrieved from <a href="https://www.heritage.org/index/explore?view=by-region-country-year&countryids=&regionids=&yearids=24">https://www.heritage.org/index/explore?view=by-region-country-year&countryids=&regionids=&yearids=24</a>)).

The .csv files were imported into SQL where the data was housed and merged to produce one database and a corresponding.csv file with 111 unique records: Happiness\_Factors.sql and Happiness\_Factors.csv

The analysis consisted of Scatterplots, Multivariate Regressions, and Interactive Bar Charts using the following variables:

- · y variable: Happiness Score
- x variables: GDP per Capita, Social Support, Healthy Life Expectancy, Corruption, Freedom\_Score, Property\_Rights, Government Spending, Business, Trade, and Crime Index

#### **DATA SETUP**

```
In [1]: # Import Dependencies
    import csv
    import pandas as pd
    import statsmodels.api as sm
    import matplotlib.pyplot as plt

In [2]: # Read the CSV file and create pandas dataframe
    df = pd.read_csv('Happiness_Factors.csv')
        # Print the dataframe
    df.head()
Out[2]:

Country Happiness Score GDP per capita Social support Healthy life expectancy Corruption Id Freedom Score Prov
```

	Country	Happiness_Score	GDP_per_capita	Social_support	Healthy_life_expectancy	Corruption	ld	Freedom_Score	Proj
0	Afghanistan	1.859	7.324	0.341	54,712	0.847	1	0,0	
1	Albania	5.277	9.567	0.718	69.150	0.878	2	65.3	
2	Algeria	5.329	9.300	0.855	66.549	0.717	3	43.2	
3	Argentina	6.024	9.959	0.891	67.200	0.814	5	51.0	
4	Armenia	5.342	9.615	0.790	67.789	0.705	6	65.1	
4		_	_	_					

The link below leads to a Github page with a Line Graph of Happiness Scores by Country:

https://jcndongo.github.io/happinessline.github.io/Happiness\_LINE\_index.html (https://jcndongo.github.io/happinessline.github.io/Happiness\_LINE\_index.html)

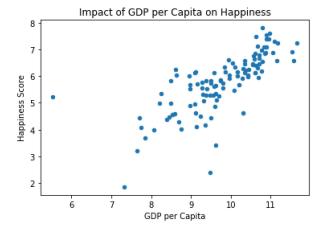
## **SCATTERPLOTS**

The Section Below uses Scatterplots to Illustrate the Impact of GDP per Capita, Social Support, Healthy Life Expectancy, Corruption, Freedom\_Score, Property\_Rights, Government\_Spending, Business, Trade, and Crime\_Index on Happiness Scores

```
In [3]: # Create a scatter plot: Happiness & GDP per Capita
df.plot.scatter(x='GDP_per_capita', y='Happiness_Score')

# Set the axis labels and plot title
plt.xlabel('GDP per Capita')
plt.ylabel('Happiness Score')
plt.title('Impact of GDP per Capita on Happiness')

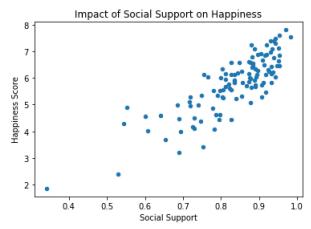
# Display the plot
plt.show()
```



```
In [4]: # Create a scatter plot: Happiness & Social Support
    df.plot.scatter(x='Social_support', y='Happiness_Score')

# Set the axis Labels and plot title
    plt.xlabel('Social Support')
    plt.ylabel('Happiness Score')
    plt.title('Impact of Social Support on Happiness')

# Display the plot
    plt.show()
```



```
In [5]: # Create a scatter plot: Happiness & Healthy Life Expectancy
df.plot.scatter(x='Healthy_life_expectancy', y='Happiness_Score')

# Set the axis labels and plot title
plt.xlabel('Healthy Life Expectancy')
plt.ylabel('Happiness Score')
plt.title('Impact of Healthy Life Expectancy on Happiness')

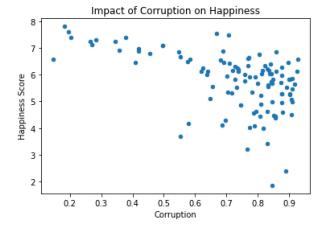
# Display the plot
plt.show()
```



```
In [6]: # Create a scatter plot: Happiness & Corruption
    df.plot.scatter(x='Corruption', y='Happiness_Score')

# Set the axis Labels and plot title
    plt.xlabel('Corruption')
    plt.ylabel('Happiness Score')
    plt.title('Impact of Corruption on Happiness')

# Display the plot
    plt.show()
```

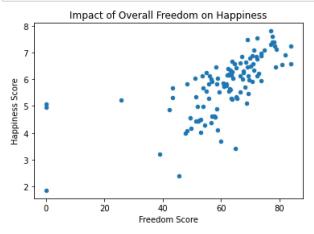


```
In [7]: # Create a scatter plot: Happiness & Overall Freedom

df.plot.scatter(x='Freedom_Score', y='Happiness_Score')

# Set the axis Labels and plot title
plt.xlabel('Freedom Score')
plt.ylabel('Happiness Score')
plt.title('Impact of Overall Freedom on Happiness')

# Display the plot
plt.show()
```

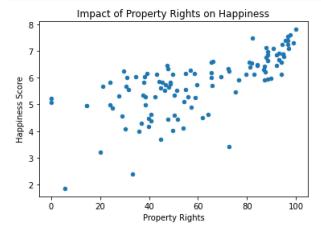


```
In [8]: # Create a scatter plot: Happiness & Property Rights

df.plot.scatter(x='Property_Rights', y='Happiness_Score')

# Set the axis labels and plot title
plt.xlabel('Property Rights')
plt.ylabel('Happiness Score')
plt.title('Impact of Property Rights on Happiness')

# Display the plot
plt.show()
```

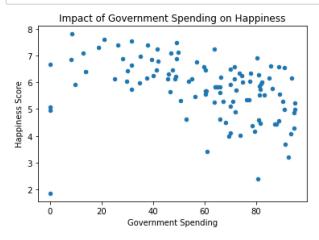


```
In [9]: # Create a scatter plot: Happiness & Government Spending

df.plot.scatter(x='Government_Spending', y='Happiness_Score')

# Set the axis Labels and plot title
plt.xlabel('Government Spending')
plt.ylabel('Happiness Score')
plt.title('Impact of Government Spending on Happiness')

# Display the plot
plt.show()
```

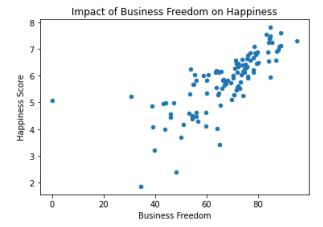


```
In [10]: # Create a scatter plot: Happiness & Business Freedom

df.plot.scatter(x='Business', y='Happiness_Score')

# Set the axis labels and plot title
plt.xlabel('Business Freedom')
plt.ylabel('Happiness Score')
plt.title('Impact of Business Freedom on Happiness')

# Display the plot
plt.show()
```

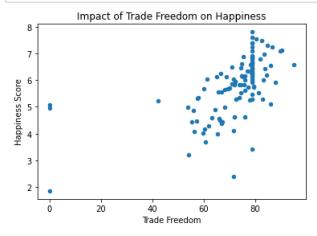


```
In [11]: # Create a scatter plot: Happiness & Trade Freedom

df.plot.scatter(x='Trade', y='Happiness_Score')

# Set the axis Labels and plot title
plt.xlabel('Trade Freedom')
plt.ylabel('Happiness Score')
plt.title('Impact of Trade Freedom on Happiness')

# Display the plot
plt.show()
```

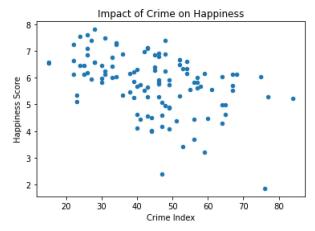


```
In [12]: # Create a scatter plot: Happiness & Crime

df.plot.scatter(x='Crime_Index', y='Happiness_Score')

# Set the axis labels and plot title
plt.xlabel('Crime Index')
plt.ylabel('Happiness Score')
plt.title('Impact of Crime on Happiness')

# Display the plot
plt.show()
```



# **MULTIVARIATE REGRESSION ANALYSIS**

The Section below represents a multivariate regression analysis of the Impact of GDP per Capita, Social Support, Healthy Life Expectancy, Corruption, Freedom\_Score, Property\_Rights, Government\_Spending, Business, Trade, Investment, Crime\_Index, and Population on Happiness Scores

```
In [13]: # Regression analysis to test model

# Define the dependent variable (y) and independent variables (x)
y = df['Happiness_Score']
x = df[['GDP_per_capita', 'Social_support', 'Healthy_life_expectancy', 'Corruption', 'Freedom_Score',
'Property_Rights', 'Government_Spending', 'Business', 'Trade', 'Investment', 'Crime_Index', 'Populatio
n']]

# Add a constant term to the independent variables
x = sm.add_constant(x)

# Perform the multivariate regression analysis
model = sm.OLS(y, x).fit()

# Print the summary statistics of the regression model
print(model.summary())
```

# OLS Regression Results

Dep. Variable: H	appiness_Score	R-squar	R-squared:		0.850			
Model:	OLS	Adj. R-squared:		0.831				
Method:	Least Squares	F-stati	Prob (F-statistic): Log-Likelihood:		46.23			
Date: Tu	e, 30 May 2023	B Prob (F			6.33e-35			
Time:	16:42:50	Dog-Lik			-61.850 149.7			
No. Observations:	111	L AIC:						
Df Residuals:	98	BIC:			184.9			
Df Model:	12	2						
Covariance Type:	nonrobust							
	coef		t	P> t	[0.025	0.975]		
const	-2.2913	1.102	-2.080	0.040	-4.478	-0.105		
GDP_per_capita	0.0597	0.094	0.637	0.526	-0.126	0.246		
Social_support	5.6560	0.596	9.489	0.000	4.473	6.839		
<pre>Healthy_life_expectancy</pre>	0.0325	0.015	2.168	0.033	0.003	0.062		
Corruption	-0.7455	0.292	-2.549	0.012	-1.326	-0.165		
Freedom_Score	0.0124	0.015	0.811	0.419	-0.018	0.043		
Property_Rights	-0.0101	0.005	-1.988	0.050	-0.020	-1.87e-05		
Government_Spending	-0.0018	0.003	-0.563	0.575	-0.008	0.005		
Business	0.0291	0.008	3.431	0.001	0.012	0.046		
Trade	-0.0167	0.008	-2.052	0.043	-0.033	-0.001		
Investment	0.0044	0.004	1.067	0.289	-0.004	0.013		
Crime_Index	0.0018	0.004	0.437	0.663	-0.006	0.010		
Population	9.045e-11	2.41e-10	0.375	0.708	-3.88e-10	5.69e-10		

Omnibus:	0.060	Durbin-Watson:	1.990					
Prob(Omnibus):	0.970	Jarque-Bera (JB):	0.040					
Skew:	-0.038	Prob(JB):	0.980					
Kurtosis:	2.944	Cond. No.	5.31e+09					

Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 5.31e+09. This might indicate that there are strong multicollinearity or other numerical problems.

CONCLUSION: based on the model above, social support, corruption, business freedom, and trade freedom are the only statistically significant determinants of happiness for all the countries in our dataset.

## TABLES & INTERACTIVE BAR CHARTS: Social Support and Corruption

Bottom 6 - The table illustrates the Bottom 6 Countries based on Happiness Score and the link leads to a Github page with an interactive bar chart of social support and corruption for those countries:

https://jcndongo.github.io/bottom6countries.github.io/BOTTOM6\_index.html (https://jcndongo.github.io/bottom6countries.github.io/BOTTOM6\_index.html)

```
In [14]: # Print of by ascending Happiness Score - Least Happy
           Bottom6 = df.sort values('Happiness Score', ascending=True)
           # Print the sorted dataframe
           Bottom6.head(6)
Out[14]:
                   Country Happiness_Score GDP_per_capita Social_support Healthy_life_expectancy Corruption
                                                                                                                Id Freedom_Score
              0 Afghanistan
                                       1.859
                                                       7.324
                                                                      0.341
                                                                                            54,712
                                                                                                        0.847
             57
                   Lebanon
                                       2.392
                                                       9.478
                                                                      0.530
                                                                                            66.149
                                                                                                        0.891
                                                                                                                93
                                                                                                                              45.6
            110
                  Zimbabwe
                                       3.204
                                                       7.641
                                                                      0.690
                                                                                            54.050
                                                                                                        0.766 183
                                                                                                                              39.0
                                                       9.629
             12
                  Botswana
                                       3.435
                                                                      0.753
                                                                                            54.725
                                                                                                        0.830
                                                                                                                21
                                                                                                                              64.9
                                       3.694
                                                       7.857
                                                                      0.653
                                                                                            59.401
                                                                                                        0.554 162
                                                                                                                              60.0
             96
                   Tanzania
                                                       8.074
                                                                                            55.032
            109
                    Zambia
                                       3.982
                                                                      0.694
                                                                                                        0.818 182
                                                                                                                              47.8
```

Top 6 - The table illustrates the Top 6 Countries based on Happiness Score and the link leads to a Github page with an interactive bar chart of social support and corruption for those countries:

https://jcndongo.github.io/top6countries.github.io/TOP6\_index.html (https://jcndongo.github.io/top6countries.github.io/TOP6\_index.html)

```
In [15]: # Print df by descending Happiness Score - Most Happy
Top6 = df.sort_values('Happiness_Score', ascending=False)
# Print the sorted dataframe
Top6.head(6)
```

### Out[15]:

	Country	Happiness_Score	GDP_per_capita	Social_support	Healthy_life_expectancy	Corruption	ld	Freedom_Score	F
32	Finland	7.804	10.792	0.969	71.150	0.182	57	77.1	
25	Denmark	7.586	10.962	0.954	71.250	0.196	45	77.6	
41	Iceland	7.530	10.896	0.983	72.050	0.668	73	72.2	
47	Israel	7.473	10.639	0.943	72.697	0.708	79	68.9	
71	Netherlands	7.403	10.942	0.930	71.550	0.379	119	78.0	
93	Sweden	7.395	10.883	0.939	72.150	0.202	157	77.5	
4									

In [ ]: