```
In [ ]: import pandas as pd
        import numpy as np
        import seaborn as s
        from sklearn.model_selection import train_test_split
        from sklearn.linear_model import LinearRegression
        from sklearn.impute import SimpleImputer
        from datascience import *
        # These lines do some fancy plotting magic.
        import matplotlib
        %matplotlib inline
        import matplotlib.pyplot as plt
        plt.style.use('fivethirtyeight')
        import warnings
        warnings.simplefilter('ignore', FutureWarning)
In [ ]: #reading in the csv file as a dataframe
        data=pd.read_csv("Life_Expectancy_Data.csv")
        ###replacing null cells with the means of that column
        for col in data.columns:
            if data[col].isnull().any():
                data[col] = data[col].fillna(data[col].mean())
        data.isnull().sum()
Out[]: Country
                                            0
        Continent
                                            0
        Year
                                            0
                                            0
        Status
                                            0
        Life_expectancy
        Adult_Mortality
                                            0
        infant_deaths
                                            0
        Alcohol
                                            0
                                            0
        percentage_expenditure
        Hepatitis_B
                                            0
        Measles
                                            0
         BMT
                                            0
        under_five_deaths
        Polio
        Total_expenditure
                                            0
        Diphtheria
                                            0
         HIV/AIDS
                                            0
                                            0
        Population
         thinness 1-19 years
         thinness 5-9 years
        Income_composition_of_resources
        Schooling
        dtype: int64
In [ ]: data.groupby('Country').count()
```

Continent Year Status Life_expectancy Adult_Mortality infant_deaths Alcohol per

Out[]:

Country							
Afghanistan	16	16	16	16	16	16	16
Albania	16	16	16	16	16	16	16
Algeria	16	16	16	16	16	16	16
Angola	16	16	16	16	16	16	16
Antigua and Barbuda	16	16	16	16	16	16	16
•••							
Uruguay	16	16	16	16	16	16	16
Uzbekistan	16	16	16	16	16	16	16
Vanuatu	16	16	16	16	16	16	16
Zambia	16	16	16	16	16	16	16
Zimbabwe	16	16	16	16	16	16	16

156 rows × 22 columns

In []: data['Country'].unique()

```
Out[]: array(['Afghanistan', 'Albania', 'Algeria', 'Angola',
                'Antigua and Barbuda', 'Argentina', 'Armenia', 'Australia',
                'Austria', 'Azerbaijan', 'Bahrain', 'Bangladesh', 'Barbados',
                'Belarus', 'Belgium', 'Belize', 'Benin', 'Bhutan',
                'Bosnia and Herzegovina', 'Botswana', 'Brazil', 'Brunei Darussalam',
                'Bulgaria', 'Burkina Faso', 'Burundi', 'Cabo Verde', 'Cambodia',
                'Cameroon', 'Canada', 'Central African Republic', 'Chad', 'Chile',
                'China', 'Colombia', 'Comoros', 'Costa Rica', 'Croatia', 'Cuba',
                'Cyprus', 'Denmark', 'Djibouti', 'Dominican Republic', 'Ecuador',
                'El Salvador', 'Equatorial Guinea', 'Eritrea', 'Estonia',
                'Ethiopia', 'Fiji', 'Finland', 'France', 'Gabon', 'Georgia',
                'Germany', 'Ghana', 'Greece', 'Grenada', 'Guatemala', 'Guinea',
                'Guinea-Bissau', 'Guyana', 'Haiti', 'Honduras', 'Hungary',
                'Iceland', 'India', 'Indonesia', 'Iraq', 'Ireland', 'Israel',
                'Italy', 'Jamaica', 'Japan', 'Jordan', 'Kazakhstan', 'Kenya',
                'Kiribati', 'Kuwait', 'Latvia', 'Lebanon', 'Lesotho', 'Liberia',
                'Libya', 'Lithuania', 'Luxembourg', 'Madagascar', 'Malawi',
                'Malaysia', 'Maldives', 'Mali', 'Malta', 'Mauritania', 'Mauritius',
                'Mexico', 'Mongolia', 'Montenegro', 'Morocco', 'Mozambique',
                'Myanmar', 'Namibia', 'Nepal', 'Netherlands', 'New Zealand',
                'Nicaragua', 'Niger', 'Nigeria', 'Norway', 'Oman', 'Pakistan',
                'Panama', 'Papua New Guinea', 'Paraguay', 'Peru', 'Philippines',
                'Poland', 'Portugal', 'Qatar', 'Romania', 'Russian Federation',
                'Rwanda', 'Samoa', 'Sao Tome and Principe', 'Saudi Arabia',
                'Senegal', 'Serbia', 'Seychelles', 'Sierra Leone', 'Singapore',
                'Slovenia', 'Solomon Islands', 'Somalia', 'South Africa', 'Spain',
                'Sri Lanka', 'Suriname', 'Swaziland', 'Sweden', 'Switzerland',
                'Syrian Arab Republic', 'Tajikistan', 'Thailand', 'Timor-Leste',
                'Togo', 'Tonga', 'Trinidad and Tobago', 'Tunisia', 'Turkey',
                'Turkmenistan', 'Uganda', 'Ukraine', 'United Arab Emirates',
                'Uruguay', 'Uzbekistan', 'Vanuatu', 'Zambia', 'Zimbabwe'], dtype=object)
In [ ]: morocco = data[data['Country'] == 'Morocco']
        predictVar=morocco[['GDP','Schooling','Income composition of resources']]
        target=morocco['Life_expectancy ']
In [ ]: predictVar.head(8)
                    GDP Schooling Income_composition_of_resources
Out[]:
        1524 2847.285569
                              12.1
                                                          0.645
        1525 3154.513484
                              12.1
                                                           0.640
        1526 3111.762887
                              12.1
                                                          0.634
        1527
              294.746728
                              11.6
                                                           0.623
```

0.612

0.603

0.596

0.589

1528

339.916160

1529 2834.247200

1530 2861.554500

1531 2884.947760

11.2

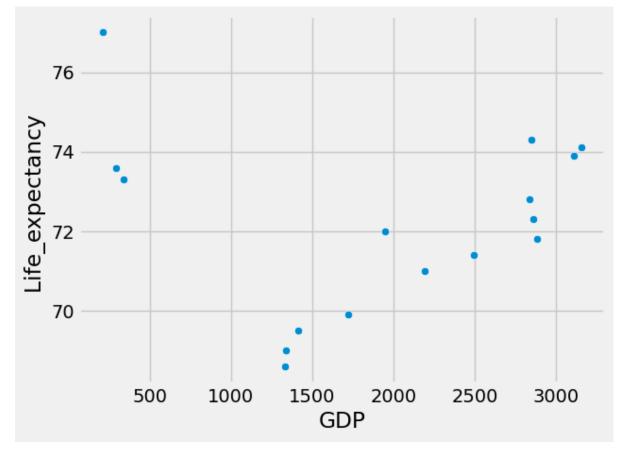
10.7

10.5

10.3

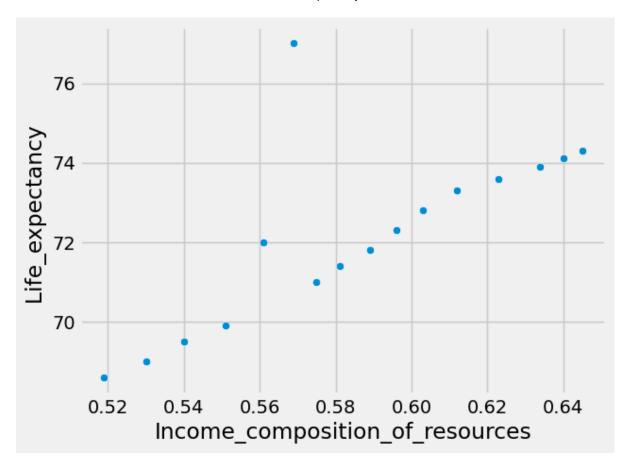
```
In [ ]: s.scatterplot(data=morocco,x='GDP',y='Life_expectancy ')
```

Out[]: <Axes: xlabel='GDP', ylabel='Life_expectancy '>



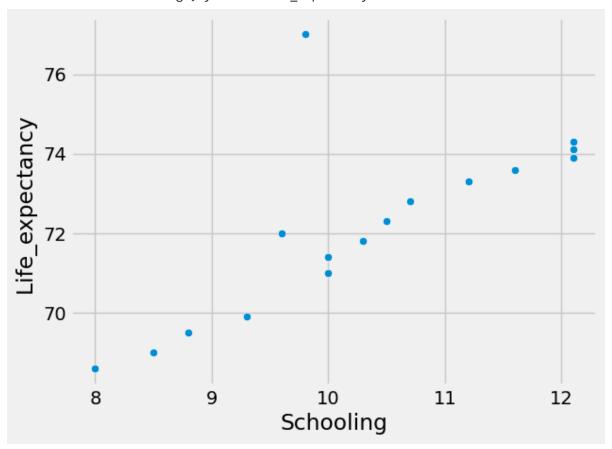
```
In [ ]: s.scatterplot(data=morocco,x='Income_composition_of_resources',y='Life_expectancy '
```

Out[]: <Axes: xlabel='Income_composition_of_resources', ylabel='Life_expectancy '>



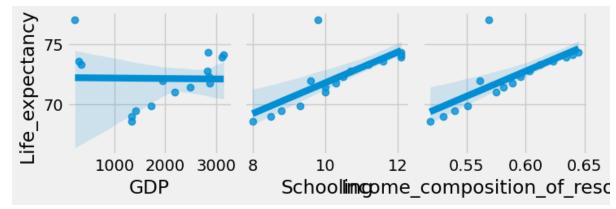
```
In [ ]: s.scatterplot(data=morocco,x='Schooling',y='Life_expectancy ')
```

Out[]: <Axes: xlabel='Schooling', ylabel='Life_expectancy '>



```
In [ ]: s.pairplot(morocco,x_vars=predictVar,y_vars='Life_expectancy ',kind='reg')
```

```
Out[]: <seaborn.axisgrid.PairGrid at 0x202ef126790>
```



```
In [ ]: # metrics
    from sklearn.metrics import mean_squared_error
    from sklearn.metrics import r2_score
```

```
In [ ]: X = data.drop(["Life_expectancy "],axis=1)
        y = data["Life_expectancy "]
        #X = pd.get_dummies(X, dummy_na=True)
        from sklearn.preprocessing import OneHotEncoder
        from sklearn.pipeline import Pipeline
        from sklearn.compose import ColumnTransformer
        #using the oneHotEncoder to transform the categorical columns
        categorical_features = ['Country', 'Status','Continent','Diphtheria ','Population'
        preprocessor = ColumnTransformer(
            transformers=[
                ('cat', OneHotEncoder(handle_unknown='ignore'), categorical_features)
            ],
            remainder='passthrough'
        X encoded = preprocessor.fit transform(X)
        X_train, X_test, y_train, y_test = train_test_split(predictVar, target, test_size=0
        #establishes linear regression
        lr=LinearRegression()
        #fits the model onto training set
        lr.fit(X_train,y_train)
        prediction=lr.predict(X_test)
        #checks to see models efficenecy
        RMSE=(str(mean_squared_error( prediction , y_test, squared=False )))
        R2_score=(str(r2_score( y_test , prediction ) * 100 ) + " %")
```

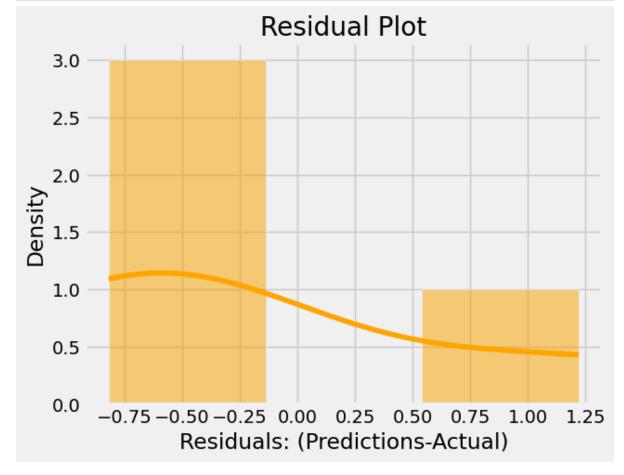
```
In [ ]: print("Model Efficenecy")
    print("RMSE:",RMSE)
    print("r2:",R2_score)
    print("Intercept: ",lr.intercept_)
```

```
print(list(zip(predictVar, lr.coef_)))

Model Efficenecy
RMSE: 0.82194832623
r2: 85.0943397465 %
Intercept: 58.6434263205
Coefficent: [ -7.87883564e-04    1.16447122e+00    5.15689058e+00]
[('GDP', -0.00078788356391796069), ('Schooling', 1.1644712166367723), ('Income_com position_of_resources', 5.1568905812040358)]
```

print("Coefficent: ",lr.coef_)

```
In []: predictions=lr.predict(X_test)
    residuals=(predictions-y_test)
    s.histplot(residuals, kde=True,color="orange")
    plt.title('Residual Plot')
    plt.xlabel('Residuals: (Predictions-Actual)')
    plt.ylabel('Density')
    plt.show()
    residuals.head(20)
```



```
Out[]: 1524 -0.483607

1525 -0.551451

1529 -0.820183

1538 1.221360

Name: Life_expectancy , dtype: float64
```

```
In [ ]: #Y=58.64-.00078(GDP)+1.16(Schooling)+5.16(ICOR)
```