A linear regression model describes the relationship between a dependent variable, y, and one or more independent variables, X. The dependent variable is also called the response variable. Independent variables are also called explanatory or predictor variables.

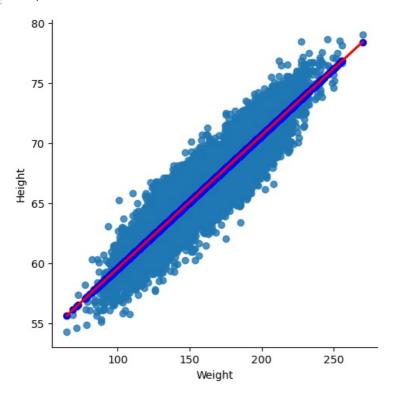
## Example

```
import pandas as pd
 In [1]:
          import numpy as np
          import matplotlib.pyplot as plt
 In [2]: df=pd.read csv(r"C:\Users\USER\Downloads\archive\weight-height.csv")
 In [3]:
         df.head()
            Gender
                      Height
                                Weight
         0
              Male
                   73.847017 241.893563
         1
              Male
                   68.781904 162.310473
                   74.110105 212.740856
                   71.730978 220.042470
              Male
                   69.881796 206.349801
              Male
In [10]:
         import seaborn as sns
          import warnings
         warnings.filterwarnings("ignore")
In [11]: sns.lmplot(x='Weight', y='Height', data=df, line_kws={'color': 'red'})
         <seaborn.axisgrid.FacetGrid at 0x18ed3e3fad0>
Out[11]:
             80
             75
             70
          Height
             65
             60
             55
                          100
                                       150
                                                   200
                                                                250
                                          Weight
         from sklearn.linear_model import LinearRegression
In [12]:
         X=df[['Weight']]
In [15]:
         y=df['Height']
In [16]:
         # Fit the linear regression model
          model = LinearRegression()
         model.fit(X, y)
Out[16]: ▼ LinearRegression
         LinearRegression()
In [17]: # Make predictions
         df['predicted_height'] = model.predict(X)
In [19]:
         # Calculate residuals
         df['residuals'] = df['Height'] - df['predicted_height']
```

```
In [20]: df.head()
              Gender
                                     Weight predicted_height residuals
Out[20]:
                         Height
                Male
                      73.847017
                                 241.893563
                                                   75.282804
                                                             -1.435787
                      68.781904
                                 162.310473
                                                   66.463980
                                                              2.317924
                Male
                 Male
                      74.110105
                                 212.740856
                                                   72.052311
                                                              2.057794
                      71.730978
                                 220.042470
                                                   72.861424
                                                             -1.130445
                 Male
                      69.881796
                                 206.349801
                                                   71.344101 -1.462305
```

```
In [23]: sns.lmplot(x='Weight', y='Height', data=df, line_kws={'color': 'red'})
plt.scatter(df['Weight'], df['predicted_height'], color='blue')
```

~matplotlib.collections.PathCollection at 0x18edba5cd90>



So, here predicted\_height is best fit line.

The best fit line (or line of best fit) is a straight line that best represents the data on a scatter plot. This line minimizes the sum of the squared differences (residuals) between the observed values and the values predicted by the line

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