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In [4]:	1. Simple Linear Regression import pandas as pd df = pd.read_csv("ml_data_salary.csv") df
Out[4]:	age distance YearsExperience Salary 0 31.1 77.75 1.1 39343 1 31.3 78.25 1.3 46205 2 31.5 78.75 1.5 37731 3 32.0 80.00 2.0 43525
	4 32.2 80.50 2.2 39891 5 32.9 82.25 2.9 56642 6 33.0 82.50 3.0 60150 7 33.2 83.00 3.2 54445 8 33.2 83.00 3.2 64445
	9 33.7 84.25 3.7 57189 10 33.9 84.75 3.9 63218 11 34.0 85.00 4.0 55794 12 34.0 85.00 4.0 56957 13 34.1 85.25 4.1 57081
	14 34.5 86.25 4.5 61111 15 34.9 87.25 4.9 67938 16 35.1 87.75 5.1 66029 17 35.3 88.25 5.3 83088 18 35.9 89.75 5.9 81363
	19 36.0 90.00 6.0 93940 20 36.8 92.00 6.8 91738 21 37.1 92.75 7.1 98273 22 37.9 94.75 7.9 101302 23 38.2 95.50 8.2 113812
	24 38.7 96.75 8.7 109431 25 39.0 97.50 9.0 105582 26 39.5 98.75 9.5 116969 27 39.6 99.00 9.6 112635 28 40.3 100.75 10.3 122391
In [5]: Out[5]:	29 40.5 101.25 10.5 121872 df = df.drop(["age"], axis = 1) df.head() distance YearsExperience Salary
	0 77.75 1.1 39343 1 78.25 1.3 46205 2 78.75 1.5 37731 3 80.00 2.0 43525 4 80.50 2.2 39891
In [6]: Out[6]:	<pre>df = df.drop(["distance"], axis = 1) df.head()</pre>
	1 1.3 46205 2 1.5 37731 3 2.0 43525 4 2.2 39891
In [7]: In [8]:	Step-2 : Splitting dataset into training data and testing data
Out[8]:	
In [9]: Out[9]:	3 2.0 4 2.2 y,head() 0 39343
In [10]:	1 46205 2 37731 3 43525 4 39891 Name: Salary, dtype: int64 # import library and split data
	<pre>from sklearn.model_selection import train_test_split X_train, X_test, y_train, y_test =train_test_split(X,y,test_size=0.2, random_state=0) Step-2: Fit Linear Regression Model</pre>
<pre>In [11]: Out[11]:</pre>	<pre>from sklearn.linear_model import LinearRegression model = LinearRegression() model = model.fit(X_train, y_train) model LinearRegression()</pre>
In [12]:	Step-4: Plotting import seaborn as sns import matplotlib.pyplot as plt sns.set_style("darkgrid")
Out[12]:	<pre>plt.scatter(X_train, y_train) plt.plot(X_train, model.predict(X_train), color ="Green") plt.xlabel("Years of Experience") plt.ylabel("Salary") plt.title("Train Plot")</pre> Text(0.5, 1.0, 'Train Plot')
	Train Plot 120000 100000 80000
	60000 40000 2 4 6 8 10 Years of Experience
<pre>In [13]: Out[13]:</pre>	<pre>plt.scatter(X_test, y_test) plt.plot(X_test, model.predict(X_test), color ="Green") plt.xlabel("Years of Experience") plt.ylabel("Salary") plt.title("Train Plot")</pre> Text(0.5, 1.0, 'Train Plot')
_ ~1'	Train Plot 120000 100000 80000
	80000 40000 2 4 6 8 10 Years of Experience
In [14]:	Step-5: testing or Evaluating the Model # Model Fitness Checking model.score(X_test, y_test)
Out[14]: In [15]: Out[15]:	<pre>model.score(X_train, y_train)</pre>
In [16]:	Step-6: Prediction of Unknown values model.predict([[5]]) C:\Users\syedriaz\AppData\Local\Programs\Python\Python39\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names warnings.warn(
Out[16]:	<pre>array([73342.97478427]) model.predict([[10]]) C:\Users\syedriaz\AppData\Local\Programs\Python\Python39\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names warnings.warn(</pre>
Out[17]: In [18]: Out[18]:	<pre>array([119905.85041792]) model.predict(X_test) array([40748.96184072, 122699.62295594, 64961.65717022, 63099.14214487,</pre>
In [19]:	# How to predict value for multiple data model.predict([[10],[5],[1]]) C:\Users\syedriaz\AppData\Local\Programs\Python\Python39\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names warnings.warn(
Out[19]: In [20]:	<pre>x = ([10],[20],[30],[5]) model.predict(x) C:\Users\syedriaz\AppData\Local\Programs\Python\Python39\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names</pre>
Out[20]: In [21]:	<pre>warnings.warn(array([119905.85041792, 213031.60168521, 306157.3529525 , 73342.97478427]) 2. Multiple Linear Regression df = pd.read_csv("ml_data_salary.csv") df.head()</pre>
Out[21]:	age distance YearsExperience Salary 0 31.1 77.75 1.1 39343 1 31.3 78.25 1.3 46205 2 31.5 78.75 1.5 37731
In [22]:	3 32.0 80.00 2.0 43525 4 32.2 80.50 2.2 39891 X = df[["age", "distance", "YearsExperience"]] X.head()
Out[22]:	age distance YearsExperience 0 31.1 77.75 1.1 1 31.3 78.25 1.3 2 31.5 78.75 1.5 3 32.0 80.00 2.0 4 32.2 80.50 2.2
In [23]: Out[23]:	<pre>y = df["Salary"] y.head() 0 39343 1 46205 2 37731</pre>
In [24]:	<pre>3 43525 4 39891 Name: Salary, dtype: int64 # Create & Fit the model model = LinearRegression().fit(X, y) model</pre>
Out[24]:	<pre># Coeffients model.coef_</pre>
Out[25]: In [26]: Out[26]: In [27]:	array([-2.68055892e+15, 1.06092560e+15, 2.82449143e+13]) model.intercept_ 847347429532075.5 model.predict([[31.1,77.75,1.1]])
Out[27]: In [28]:	C:\Users\syedriaz\AppData\Local\Programs\Python\Python39\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names warnings.warn(array([36217.125]) model.score(X, y)
Out[28]: In [29]:	<pre>0.9569520722791693 df = pd.read_csv("ml_data_salary.csv") df.head()</pre>
Out[29]:	age distance YearsExperience Salary 0 31.1 77.75 1.1 39343 1 31.3 78.25 1.3 46205 2 31.5 78.75 1.5 37731 3 32.0 80.00 2.0 43525 4 32.2 80.50 2.2 39891
In [30]:	<pre>from sklearn.model_selection import train_test_split X_train, X_test, y_train, y_test =train_test_split(X,y,test_size=0.2, random_state=100)</pre>
<pre>In [31]: Out[31]:</pre>	<pre>from sklearn.linear_model import LinearRegression model = LinearRegression() model = model.fit(X_train, y_train) model LinearRegression()</pre>
In [32]: Out[32]: In [33]:	<pre>X_train.shape (24, 3) y_train.shape</pre>
Out[33]: In [34]: Out[34]:	model.score(X_train, y_train) 0.9514648186125461
In [35]: Out[35]: In [36]:	<pre>model.score(X_test, y_test) 0.9726052357502972 sns.regplot(x=X_train["age"], y=y_train, color='blue', marker='+') <axessubplot:xlabel='age', ylabel="Salary"></axessubplot:xlabel='age',></pre>
Out[36]:	<pre><axessubplot:xlabel='age', ylabel="Salary"> 120000 100000 80000</axessubplot:xlabel='age',></pre>
	8000 6000 4000 32 34 36 38 40 age
In [37]: Out[37]:	<pre>sns.regplot(x=X_train["distance"], y=y_train, color='green', marker='+')</pre>
	$\frac{100000}{\frac{1}{8}}$ 80000
In [38]:	sns.regplot(x=X_train["YearsExperience"], y=y_train, color='red', marker='+')
Out[38]:	<pre><axessubplot:xlabel='yearsexperience', ylabel="Salary"></axessubplot:xlabel='yearsexperience',></pre>
	10000
In [39]:	100000 \(\frac{1}{2} \) \(\f
In [39]:	import matplotlib.pyplot as plt import pandas as pd import seaborn as sns
In [39]:	import matplotlib.pyplot as plt import pandas as pd import seaborn as sns from mpl_toolkits.mplot3d import Axes3D fig = plt.figure() ax = fig.add_subplot(iii, projection = '3d') x = X_train["YearsExperience"] y = X_train["distance"]
In [39]:	import matplotlib.pyplot as plt import pandas as pd import seaborn as sms import seaborn as sms from up.tronksts.mplotdd import Axes3D x = fig.add.subplot(111, projection = '3d') x = X.train["Year sExperience"] y = X.train["distance"] z = X.train["distance"] ax. sox.xlabel("Waar sExperience") ax. sex.xlabel("Waar sExperience") ax. sex.xlabel("Waar sExperience") ax. sex.xlabel("Waar sExperience") ax. sex.xlabel("distance")
	import matplotlib.pyplot as plt import pandas as pd import seaborn as sms import seaborn as sms from up.troksts.mplotdd import Axes3D x = fig.add.subplot(111, projection = '3d') x = X.train["Year sExperience"] y = X.train["distance"] z = X.train["distance"] ax. sox.xlabel("Waar sExperience") ax. sex.xlabel("Waar sExperience") ax. sex.xlabel("Waar sExperience") ax. sex.xlabel("Waar sExperience") ax. sex.xlabel("distance")
In [40]:	Import model in reduction as a process of the state of th
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