Tn [4]:	EXAMPLE import numpy as np
	<pre>import pandas as pd from sklearn.linear_model import LinearRegression data=pd.read_csv('insurance.csv') data</pre>
Out[6]:	
	2 35 167.5 78 38000 3 40 NaN 110 60000 4 45 157.4 85 70000
In [11]:	<pre># Extract the features (independent variables) as a DataFrame X = data[['age', 'weight']] # Note the use of double square brackets [[]] # Extract the target variable (dependent variable) as a Series y = data['premium']</pre>
	<pre># Create a Linear Regression model model = LinearRegression() # Fit the model to the data model.fit(X, y)</pre>
Out[11]:	LinearRegression()
In [15]:	<pre>print("Coefficient (Slope):", model.coef_) print("Intercept:", model.intercept_) print("Predicted Value:", y_pred) Coefficient (Slope): [2290.36842795 255.14619116] Intercept: -57713.70132424285 Predicted Value: [5.8]</pre>
In [17]:	EXAMPLE import pandas as pd
In [18]:	<pre>import numpy as np from sklearn.linear_model import LinearRegression data=pd.read_csv('insurance.csv') data</pre>
Out[18]:	age height weight premium 0 25 162.5 70 18000 1 30 172.7 95 38000
	2 35 167.5 78 38000 3 40 NaN 110 60000 4 45 157.4 85 70000
In [21]: Out[21]:	<pre>data['height'] = data['height'].fillna(mean_height) data['height'] 0 162.500 1 172.700</pre>
In [25]:	<pre>2 167.500 3 165.025 4 157.400 Name: height, dtype: float64 X=data[['age', 'height', 'weight']]</pre>
In [26]:	<pre>y=data['premium'] model = LinearRegression() # Fit the model to the data model.fit(X, y)</pre>
Out[26]:	LinearRegression()
In [38]: In [39]: Out[39]:	orroy/[19200_094690924490_029967741207_70269272_62646_26747026
In [40]:	<pre>print("Coefficient (Slope):", model.coef_) print("Intercept:", model.intercept_) Coefficient (Slope): [2152.80168625 -242.52223936 311.66290353] Intercept: -17827.496816556588</pre>
In [70]:	<pre>import pandas as pd import numpy as np</pre>
In [59]:	<pre>import matplotlib.pyplot as plt from sklearn .linear_model import LinearRegression data=pd.read_csv('Salary_Data[1].csv') data</pre>
Out[59]:	YearsExperience Salary 0 1.1 39343.0 1 1.3 46205.0 2 1.5 37731.0
	3 2.0 43525.0 4 2.2 39891.0 5 2.9 56642.0
	 6 3.0 60150.0 7 3.2 54445.0 8 3.2 64445.0 9 3.7 57189.0
	10 3.9 63218.0 11 4.0 55794.0 12 4.0 56957.0 13 4.1 57081.0
	14 4.5 61111.0 15 4.9 67938.0 16 5.1 66029.0 17 5.3 83088.0
	18 5.9 81363.0 19 6.0 93940.0 20 6.8 91738.0
	21 7.1 98273.0 22 7.9 101302.0 23 8.2 113812.0 24 8.7 109431.0
	25 9.0 105582.0 26 9.5 116969.0 27 9.6 112635.0 28 10.3 122391.0
In [60]:	<pre>X=data[['YearsExperience']] y=data['Salary']</pre>
In [61]:	# Fit the model to the data model.fit(X, y)
Out[61]: In [62]:	<pre>v LinearRegression LinearRegression() y_pred = model.predict(X)</pre>
In [66]:	coffeficent(slope): [9449.96232146]
In [68]:	<pre>Intercept: 25792.20019866871 print("Predicted Value(s):", y_pred) Predicted Value(s): [36187.15875227 38077.15121656 39967.14368085 44692.12484158 46582.11730587 53197.09093089 54142.08716303 56032.07962732</pre>
	56032.07962732 60757.06078805 62647.05325234 63592.04948449 63592.04948449 64537.04571663 68317.03064522 72097.0155738 73987.00803809 75877.00050238 81546.97789525 82491.9741274 90051.94398456 92886.932681 100446.90253816 103281.8912346 108006.87239533 110841.86109176 115566.84225249 116511.83848464 123126.81210966 125016.80457395]
In [78]:	<pre>plt.scatter(X, y, color='yellow', label='Actual Data') plt.plot(X, y_pred, color='black', linewidth=2, label='Regression Line') plt.xlabel('X (Independent Variable)') plt.ylabel('y (Dependent Variable)') plt.title('multiple-Variable Linear Regression')</pre>
	plt.legend() plt.show() multiple-Variable Linear Regression Actual Data
	120000 - Regression Line 100000 - Regression Line
	(Dependent Variable) × 60000 -
	© > 60000 -
	2 4 6 8 10 X (Independent Variable)
In [106…	<pre>import numpy as np</pre>
	<pre>from sklearn.model_selection import train_test_split from sklearn.linear_model import LinearRegression from sklearn.metrics import mean_squared_error, r2_score # Step 1: Load the dataset data = pd.read_csv('Ecommerce_Customers[1].csv')</pre>
	<pre># Step 2: Explore and preprocess the data (cleaning, feature selection, etc.) # For this example, let's assume you're interested in 'Yearly Amount Spent' as the target variable and 'Length of Membership' as a feature. X = data[['Length of Membership']] # Feature(s) y = data['Yearly Amount Spent'] # Target variable</pre>
	<pre># Step 3: Split the data into training and testing sets X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42) # Step 4: Create a Linear Regression model model = LinearRegression()</pre>
	<pre># Step 5: Train the model on the training data model.fit(X_train, y_train) # Step 6: Make predictions on the testing data y_pred = model.predict(X_test)</pre>
	<pre># Step 7: Evaluate the model's performance mse = mean_squared_error(y_test, y_pred) r2 = r2_score(y_test, y_pred) # Step 8: Print model coefficients and evaluation metrics print("Coefficient (Slope):", model.coef_[0]) print("Intercept:", model.intercept_)</pre>
	print("Mean Squared Error (MSE):", mse) print("R-squared (R2):", r2) Coefficient (Slope): 64.64010065386711 Intercept: 271.35211280339297 Mean Squared Error (MSE): 2162.1116327764234
In [2]·	R-squared (R2): 0.5633624502240148 EXAMPLE import pandas as pd
	<pre>import numpy as np from sklearn.model_selection import train_test_split from sklearn.linear_model import LinearRegression from sklearn.metrics import mean_squared_error, r2_score import matplotlib.pyplot as plt</pre>
In [3]: Out[3]:	data
	0 230.1 37.8 69.2 22.1 1 44.5 39.3 45.1 10.4 2 17.2 45.9 69.3 12.0 3 151.5 41.3 58.5 16.5
	4 180.8 10.8 58.4 17.9 195 38.2 3.7 13.8 7.6
	196 94.2 4.9 8.1 14.0 197 177.0 9.3 6.4 14.8 198 283.6 42.0 66.2 25.5 199 232.1 8.6 8.7 18.4
	<pre>200 rows × 4 columns X = data[['TV']] # Feature(s) y = data['Sales'] # Target variable</pre>
	<pre># Step 3: Split the data into training and testing sets X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42) # Step 4: Create a Linear Regression model model = LinearRegression()</pre>
<pre>In [7]: Out[7]:</pre>	# Step 5: Train the model on the training data model.fit(X_train, y_train) * LinearRegression
In [8]:	<pre># Step 6: Make predictions on the testing data y_pred = model.predict(X_test)</pre>
In [9]:	<pre># Step 8: Print model coefficients and evaluation metrics print("Coefficient (Slope):", model.coef_[0]) print("Intercept:", model.intercept_) Coefficient (Slope): 0.05548294393146318 Intercept: 7.007108428241851</pre>
In [10]:	<pre># Step 9: Visualize the regression line and data plt.scatter(X_test, y_test, color='blue', label='Actual Data') plt.plot(X_test, y_pred, color='red', linewidth=2, label='Regression Line') plt.xlabel('TV Advertising Spending') plt.ylabel('Sales')</pre>
	plt.title('Linear Regression: TV vs. Sales') plt.legend() plt.show() Linear Regression: TV vs. Sales
	25.0 - Actual Data Regression Line 20.0 -
	17.5 - \frac{\omega}{\omega} 15.0 -
	12.5 -
	7.5 - 5.0 - 50 100 150 200 250 300 TV Advertising Spending
In []:	