

Ex. No: 1

A PYTHON PROGRAM

Date: 26/7/25. IMPLEMENT UNIVARIATE, BIVARIATE

AND MULTIVARIATE REGRESSION.

Aim: $\text{Polarisiv} = \beta_0 + \beta_1 \text{Salary}$

To implement a Python program using univariate, bivariate and multivariate regression techniques for a given Iris dataset.

Algorithm:

Step 1: Import necessary libraries:

* pandas for data manipulation, numpy for numerical operations and matplotlib.pyplot for plotting.

Step 2: Read the dataset:

* Use the pandas 'read_csv' function to read the dataset.

* Store the dataset in a variable.

Step 3: Prepare the data:

* Extract the independent variables (X) and dependent variable (y) from the dataset.

* Perhaps X and y to be 2D arrays if needed.

Step 4: Univariate Regression:

* For univariate regression, we only one independent variable.

* Make predictions using the model.

* Calculate the R-squared value to evaluate the model's performance.

program:

```
import pandas as pd  
import matplotlib.pyplot as plt  
import seaborn as sns  
import numpy as np  
df = pd.read_csv('./input/iris-dataset/iris.csv')  
df.head(150)  
df.shape  
  
# univariate for sepal width.  
df.loc[df['variety'] == 'Setosa']  
df_Setosa = df.loc[df['variety'] == 'setosa']  
df_Virginica = df.loc[df['variety'] == 'Virginica']  
df_Versicolor = df.loc[df['variety'] == 'Versicolor']  
plt.scatter(df_Setosa['sepal.width'], np.zeros_like(df_Setosa['sepal.  
width']))  
plt.scatter(df_Virginica['sepal.width'], np.zeros_like(df_Virginica['sepal.  
width']))  
plt.scatter(df_Versicolor['sepal.width'], np.zeros_like(df_Versicolor['sepal.  
width']))  
plt.xlabel('sepal.width')  
plt.show()
```

univariate for sepal length.

```
df.loc[df['variety'] == 'Setosa']  
df_Setosa = df.loc[df['variety'] == 'setosa']  
df_Virginica = df.loc[df['variety'] == "virginica"]  
df_Versicolor = df.loc[df['variety'] == 'Versicolor']  
plt.scatter(df_Setosa['sepal.length'], np.zeros_like(df_Setosa['sepal.length']))  
plt.scatter(df_Virginica['sepal.length'], np.zeros_like(df_Virginica['sepal.length']))  
plt.scatter(df_Versicolor['sepal.length'], np.zeros_like(df_Versicolor['sepal.length']))  
plt.xlabel('sepal.length')  
plt.show()
```

Step 5 : Bivariate Regression. log22 - Data + idf

* For bivariate regression, use two independent variables.

* Make predictions using the model, "diprob_log22".

* Calculate the R-squared values to evaluate the model's performance.

Step 6 : Multivariate Regression:

* For multivariate regression, use more than two independent variables. "glaucov" = out, $\{b\}$ following.

* Make predictions using the model.

* Calculate the R-squared value to evaluate the model's performance.

Step 7 : Plot The results!

* For univariate regression, plot the original data pairs (x_1, y) as a scatter plot and the regression line as a line plot.

* For bivariate regression, plot the original data points (x_1, x_2) as a 3D scatter plot and the regression plane.

Step 8 : Display The results :

* Print the coefficients (slope) and intercept for each regression model.

* Print the R-squared value for each regression model.

Step 9 : Complete the program:

* Combine all the steps into a python program.

* Run the program to perform univariate, bivariate, and multivariate regression on the data set.

univariate

df = loc [df['variety']] df['Setosa']

df - Setosa = df - loc [df['variety']] == 'Setosa'

df - virginica = df - loc [df['variety']] == 'virginica'

df - versicolor = df - loc [df['variety']] == 'versicolor'.

plt.scatter (df - Setosa['petal.width'], np.zeros_like (df - Setosa['petal.width']))

plt.scatter (df - virginica['petal.width'], np.zeros_like (df - virginica['petal.width']))

plt.scatter (df - versicolor['petal.width'], np.zeros_like (df - versicolor['petal.width']))

plt.xlabel ('petal. width')

plt.show()

univariate for petal length.

df = loc [df['variety']] == 'Setosa'

df - Setosa = df - loc [df['variety']] == 'Setosa'

df - virginica = df - loc [df['variety']] == 'virginica'

df - versicolor = df - loc [df['variety']] == 'versicolor'

plt.scatter (df - Setosa['petal.length'], np.zeros_like (df - Setosa['petal.length']))

plt.scatter (df - virginica['petal.length'], np.zeros_like (df - virginica['petal.length']))

plt.scatter (df - versicolor['petal.length'], np.zeros_like (df - versicolor['petal.length']))

plt.xlabel ('petal. length')

plt.show()

• below fit species GB to fit of y axis x axis post x

bivariate Sepal. width vs petal. width.

sns.FacetGrid (df, hue = 'variety', size = 5). map(plt.scatter, "Sepal. width",

"Petal. width")

• add - legend (e.)

• plt.show()

bivariate: sepal-length vs. petal-length: 2 73

sns.FacetGrid (df, hue='variety', size=5) .map(plt.scatter,
"sepal_length", "petal_length") : odd (legend) 3 70
or iteration of scatter because of add background 4

plt.show()

multivariate all the features together: 4 70

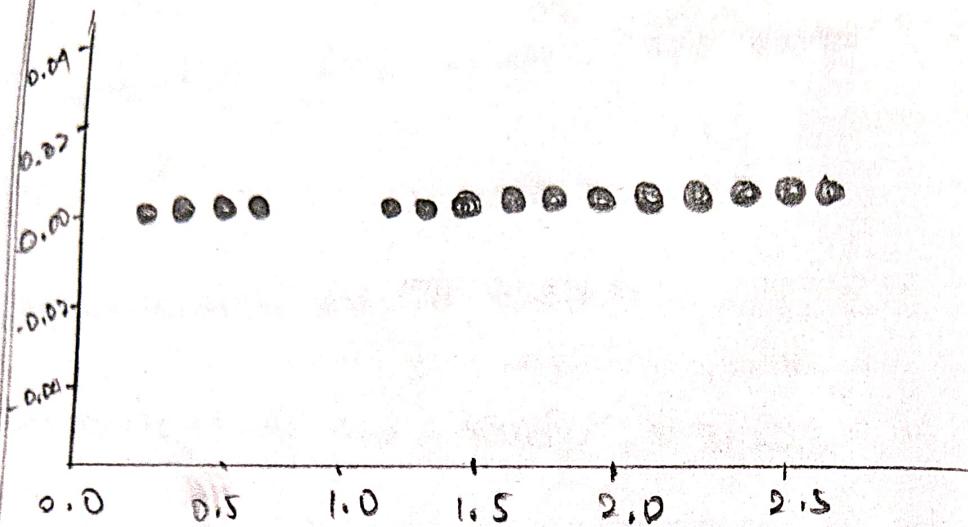
with most 0.004260, min 0.001, max 0.005 4 70

sns.pairplot (df, hue='variety', size=5), background 4

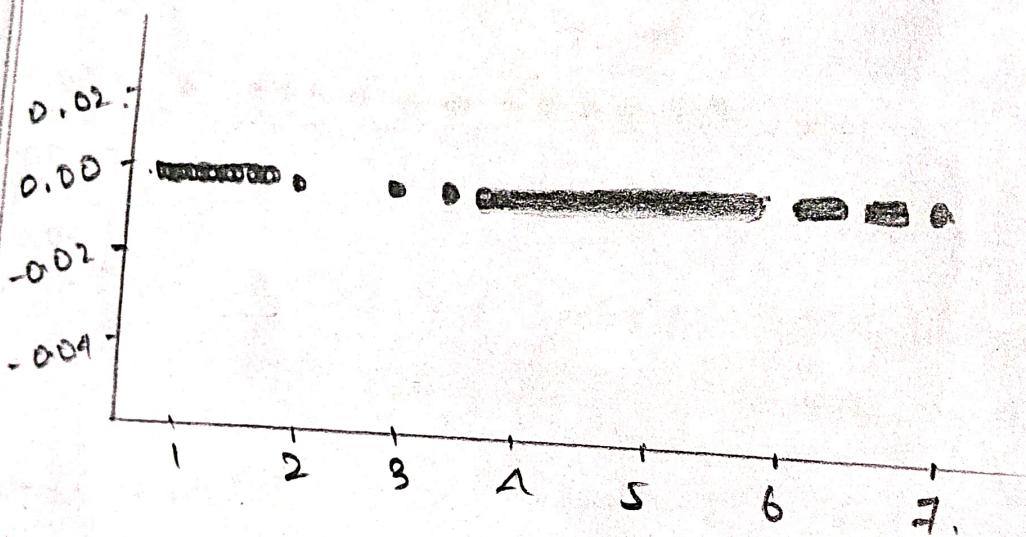
Iteration of scatter because of add background 4

sepal length: Iteration of scatter because of add background 4

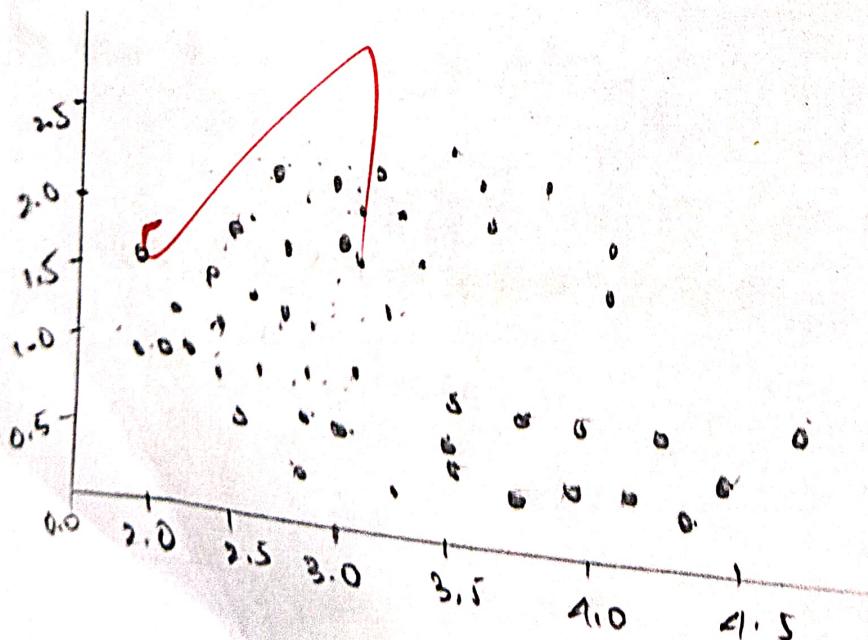
petal width:

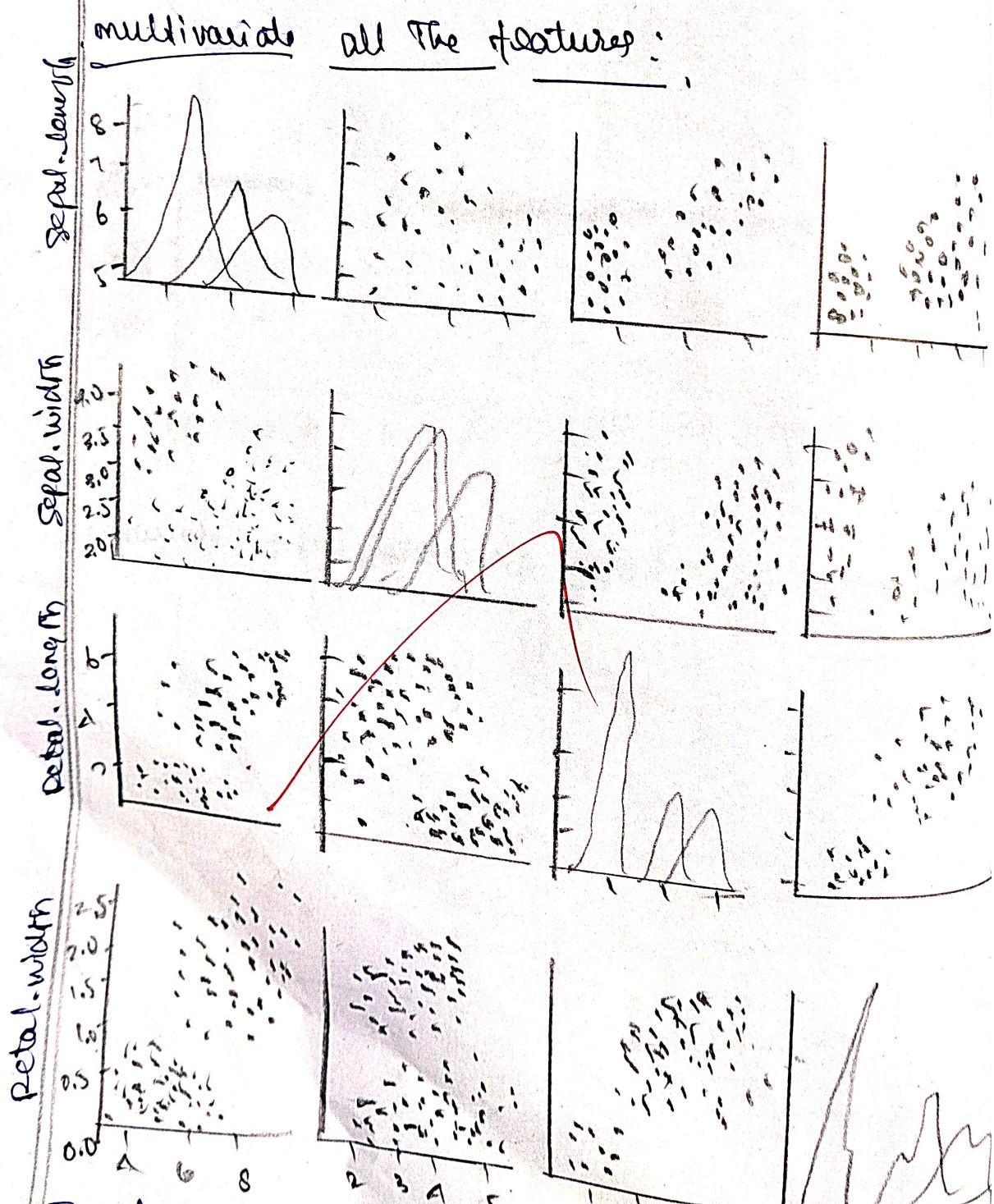
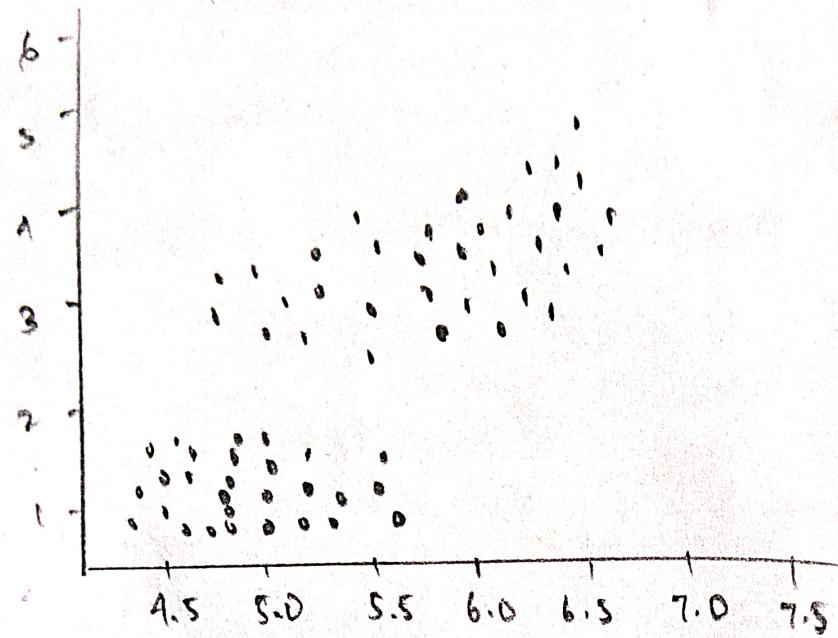


petal length:



bivariate sepal width vs petal width.





Thus the python program to implement univariate, bivariate, and multivariate regression features for the given iris dataset is analyzed and the features are plotted using Scatter plot.