Chapter 9 - Ex1: Iris - Simple Linear Regression

Cho dữ liệu iris.xls

Yêu cầu: Thực hiện linenear regression để từ pentalwidth => dự đoán pentallength

- 1. Đọc dữ liệu, trực quan hóa dữ liệu.
- Tạo X_train, X_test, y_train, y_test từ dữ liệu đọc được là 2 cột pentalwidth (inputs) và pentallength (outputs) với tỷ lệ dữ liệu test là 0.2
- 3. Áp dụng linrear regression
- 4. Vẽ hình. Nhận xét kết quả
- 5. Nếu pentalwidth là 1.5 => pentallength là bao nhiêu?
- url = 'https://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.data'
 (https://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.data')

```
In [1]: # from google.colab import drive
# drive.mount("/content/gdrive", force_remount=True)
# %cd '/content/gdrive/My Drive/MDS5_2022/Practice_2022/Chapter9/'
```

```
In [2]: import pandas as pd
iris = pd.read_excel("Iris.xls", encoding='utf-8')
```

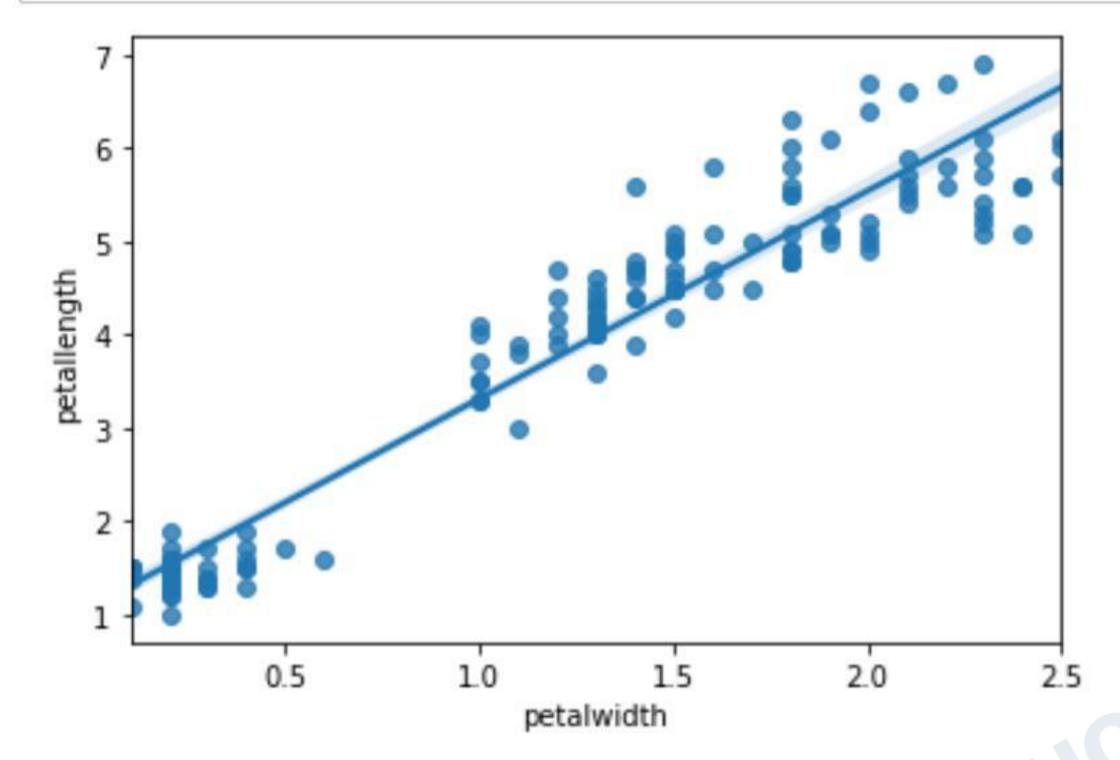
In [3]: iris.head()

Out[3]:

iris	petalwidth	petallength	sepalwidth	sepallength	
Iris-setosa	0.2	1.4	3.5	5.1	0
Iris-setosa	0.2	1.4	3.0	4.9	1
Iris-setosa	0.2	1.3	3.2	4.7	2
Iris-setosa	0.2	1.5	3.1	4.6	3
Iris-setosa	0.2	1.4	3.6	5.0	4

```
In [4]: import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [5]: sns.regplot(data=iris, x='petalwidth', y='petallength')
plt.show()
```



```
In [6]: inputs = iris[['petalwidth']]
  inputs.head()
```

Out[6]:

petalwidth				
0	0.2			
1	0.2			
2	0.2			
3	0.2			
4	0.2			

```
In [7]: outputs = iris[['petallength']]
  outputs.head()
```

Out[7]:

	petallength
0	1.4
1	1.4
2	1.3
3	1.5
4	1.4

```
In [8]: import numpy as np
          from sklearn import datasets, linear_model
          from sklearn.metrics import mean_squared_error, r2_score
 In [9]: from sklearn.model_selection import train_test_split
          X_train, X_test, y_train, y_test = train_test_split(inputs, outputs,
                                                                test_size=0.20) # train data.

    https://scikit-learn.org/stable/modules/generated/sklearn.linear_model.LinearRegression.html

             (https://scikit-learn.org/stable/modules/generated/sklearn.linear_model.LinearRegression.html)
In [10]: # huan Luyen bang du lieu train data
          regr1 = linear_model.LinearRegression()
          regr1 = regr1.fit(X_train, y_train) # train model
In [11]: # kiem tra va du doan voi test data
          y_pred = regr1.predict(X_test)
In [12]: df = pd.DataFrame({'Actual': pd.DataFrame(y test.values)[0].values,
                              'Prediction': pd.DataFrame(y pred)[0].values})
          df.head()
Out[12]:
             Actual Prediction
                6.0
                     5.107060
                1.7
                     1.754246
                     4.436497
                4.7
                5.6
                     6.448185
                3.9
                     4.212976
In [13]: # The mean squared error
          print("Mean squared error: %.2f"
                % mean_squared_error(outputs, regr1.predict(inputs)))
          # Explained variance score: 1 is perfect prediction
          print('Variance score: %.2f' % regr1.score(inputs, outputs))
          Mean squared error: 0.23
          Variance score: 0.93
In [14]: # Score = 93% => model fits with ~ 93% data => This is suitable model.
In [15]: # Check the score of train and test
          regr1.score(X test, y test) # test
```

Out[15]: 0.9458695247784931

```
In [16]: regr1.score(X_train, y_train) # train
Out[16]: 0.9211926788648582
In [17]: # Both training data and testing data have high score. => Choose this model.
In [18]: # The coefficients
          m=regr1.coef_[0] # chi co 1 m
          b=regr1.intercept_
          print('Coefficients: \n', m)
          print('Interceft: \n', b)
         Coefficients:
           [2.23520927]
          Interceft:
           [1.08368322]
In [19]: # Cung cap x \Rightarrow y_{hat} = mx + b \Rightarrow ket qua du doan
          # hoac dung ten_model.predict(x) => y_hat
In [20]: \# reg_line = [(m* float(x)) + b for x in np.array(inputs)]
          reg_line = regr1.predict(inputs)
In [21]: x_{now} = [[1.5]]
          y_now = regr1.predict(x_now)
          print(y_now)
          [[4.43649712]]
```

```
In [22]: # Plot outputs
    plt.scatter(X_train, y_train, color='green', label="Training Data")
    plt.scatter(X_test, y_test, color='red', label= "Test Data")
    plt.scatter(x_now, y_now, color='black', label= "New Data")
    plt.plot(inputs,reg_line, color="blue", linewidth=1)

    plt.xlabel("petalwidth")
    plt.ylabel("petalheight")
    plt.legend()
    plt.show()
```

