Chapter 2 - Ex1: Iris - Simple Linear Regression

Cho dữ liệu iris.xls

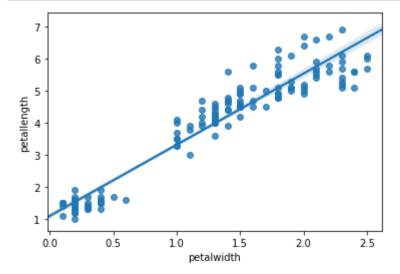
Yêu cầu: Thực hiện linenear regression để từ pentalwidth => dự đoán pentallength

- 1. Đọc dữ liệu, trực quan hóa dữ liệu.
- 2. 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')

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

```
In [3]: import matplotlib.pyplot as plt
import seaborn as sns
```

In [4]: sns.regplot(data=iris, x='petalwidth', y='petallength')
plt.show()



```
In [5]: inputs = iris[['petalwidth']]
inputs.head()
```

Out[5]:

	petalwidth
0	0.2
1	0.2
2	0.2
3	0.2
4	0.2

```
In [6]: outputs = iris[['petallength']]
  outputs.head()
```

Out[6]:

	petallength		
0	1.4		
1	1.4		
2	1.3		
3	1.5		
4	1.4		

```
In [7]: import numpy as np
    from sklearn import datasets, linear_model
    from sklearn.metrics import mean_squared_error, r2_score
```

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

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```
In [9]: # huan Luyen bang du lieu train data
regr1 = linear_model.LinearRegression()
regr1 = regr1.fit(X_train, y_train) # train model
```

```
In [10]: # kiem tra va du doan voi test data
y_pred = regr1.predict(X_test)
```

Out[11]:

	Actual	Prediction
0	1.5	1.282524
1	1.4	1.282524
2	5.1	4.685630
3	1.4	1.509398
4	5.6	6.500620

```
In [12]: # 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 [13]: # Score = 93% => model fits with ~ 93% data => This is suitable model.
In [14]: # Check the score of train and test
         regr1.score(X_test, y_test) # test
Out[14]: 0.9218903415517908
In [15]: regr1.score(X_train, y_train) # train
Out[15]: 0.927627469315894
In [16]: # Both training data and testing data have high score. => Choose this model.
In [17]: # The coefficients
         m=regr1.coef_[0] # chi co 1 m
         b=regr1.intercept_
          print('Coefficients: \n', m)
         print('Interceft: \n', b)
         Coefficients:
          [2.26873709]
         Interceft:
          [1.05565075]
In [18]: \# Cung cap x \Rightarrow y hat = mx + b \Rightarrow ket qua du doan
         # hoac dung ten model.predict(x) \Rightarrow y hat
In [19]: \# reg line = [(m^* float(x)) + b for x in np.array(inputs)]
         reg line = regr1.predict(inputs)
In [20]:
         x now = [[1.5]]
         y_now = regr1.predict(x_now)
         print(y_now)
         [[4.45875639]]
```

In [21]: # 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()

