

index

January 23, 2022

1 Complete Regression - Lab

1.1 Introduction

By now, you have created all the necessary functions to calculate the slope, intercept, best-fit line, prediction, and visualizations. In this lab you will put them all together to run a regression experiment and calculate the model loss.

1.2 Objectives

You will be able to:

- Perform a linear regression using self-constructed functions
- Calculate the coefficient of determination using self-constructed functions
- Use the coefficient of determination to determine model performance

1.3 The formulas

Slope:

$$\hat{m} = \frac{\bar{x} * \bar{y} - \overline{xy}}{(\bar{x})^2 - \bar{x}^2} \quad (1)$$

Intercept:

$$\hat{c} = \bar{y} - \hat{m}\bar{x} \quad (2)$$

Prediction:

$$\hat{y} = \hat{m}x + \hat{c} \quad (3)$$

R-Squared:

$$R^2 = 1 - \frac{SS_{RES}}{SS_{TOT}} = 1 - \frac{\sum_i (y_i - \hat{y}_i)^2}{\sum_i (y_i - \bar{y}_i)^2} \quad (4)$$

Use the Python functions created earlier to implement these formulas to run a regression analysis using x and y as input variables.

```
[2]: import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from matplotlib import style
style.use('ggplot')
```

```
%matplotlib inline
```

```
[47]: # Combine all the functions created so far to run a complete regression
      ↪ experiment.
      # Produce an output similar to the one shown below.

X = np.array([1, 2, 3, 4, 5, 6, 7, 8, 9, 10], dtype=np.float64)
Y = np.array([7, 7, 8, 9, 9, 10, 10, 11, 11, 12], dtype=np.float64)

def m_and_c(X,Y):

    m = np.round((np.mean(X) * np.mean(Y) - np.mean(X*Y))
                  / (np.mean(X)**2 - np.mean(X*X)), 2)
    c = np.round(np.mean(Y) - m*np.mean(X), 2)
    return m, c

def reg(X,Y):
    m, c = m_and_c(X,Y)
    reg = np.array([m*i + c for i in X])
    return reg

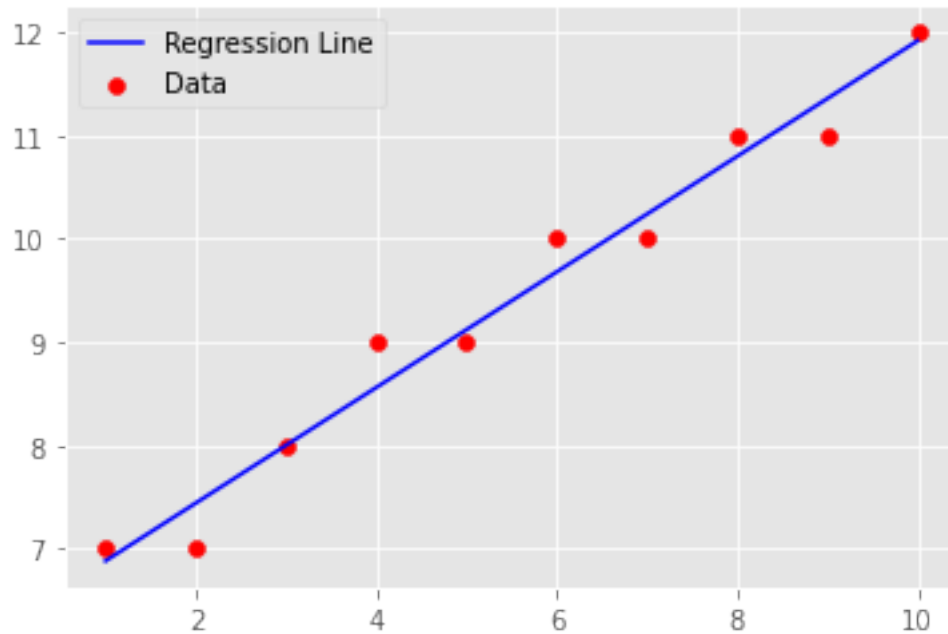
def R2_f(X,Y):

    num = np.sum((Y - reg(X,Y))**2)
    dnum = np.sum((Y - np.mean(Y))**2)

    R2 = np.round(1 - num / dnum, 2)
    return R2
```

```
[48]: # Basic Regression Diagnostics
      # -----
      # Slope: 0.56
      # Y-Intercept: 6.33
      # R-Squared: 0.97
      # -----
      # Model:  $Y = 0.56 * X + 6.33$ 

plt.scatter(X,Y, color = "Red", label = "Data")
plt.plot(X, reg(X,Y), color = "Blue", label = "Regression Line")
plt.legend();
```

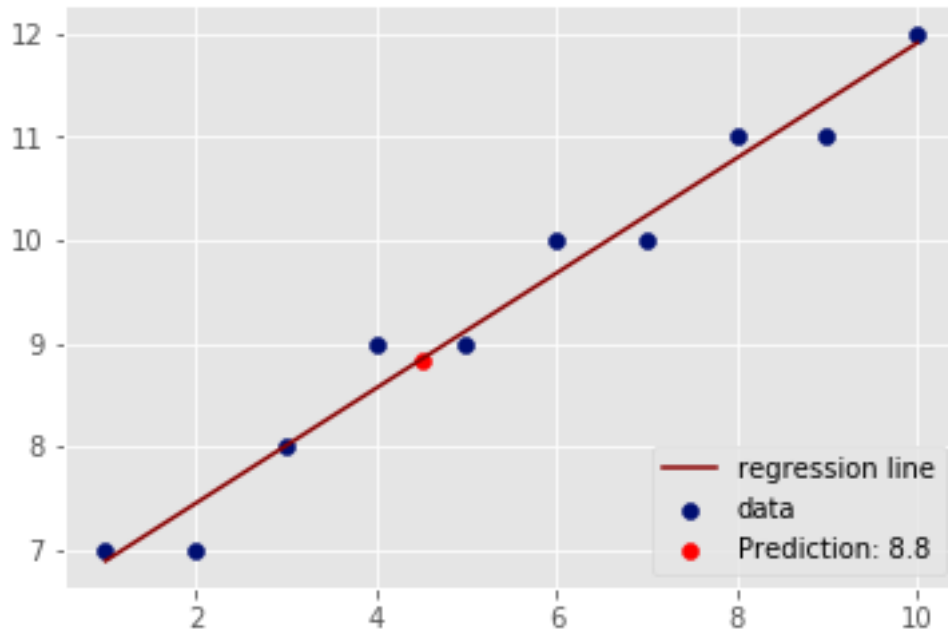


1.4 Make Predictions

Predict and plot the value of y using regression line above for a new value of $x = 4.5$.

```
[49]: # Make prediction for x = 4.5 and visualize on the scatter plot
x = 4.5
m, c = m_and_c(X,Y)
y_predict = m_and_c(X,Y)[0] * x + m_and_c(X,Y)[1]
y_predict

plt.scatter(X,Y, color = "Red", label = "Data")
plt.scatter(x,y_predict, color = "Black", label = "Prediction", marker = "X")
plt.plot(X, reg(X,Y), color = "Blue", label = "Regression Line")
plt.legend();
```



1.5 Level up - Optional

Load the “heightweight.csv” dataset. Use the height as an independent and weight as a dependent variable and draw a regression line to data using your code above. Calculate your R-Squared value for the model and try to predict new values of y.

```
[50]: import pandas as pd

df = pd.read_csv("heightweight.csv")
df.head()
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20 entries, 0 to 19
Data columns (total 2 columns):
#   Column  Non-Null Count  Dtype
---  ------  -
0   height  20 non-null         int64
1   weight  20 non-null         int64
dtypes: int64(2)
memory usage: 448.0 bytes
```

```
[53]: h = np.array(df.height)

w = np.array(df.weight)
```

```

num = np.sum((w - reg(h,w))**2)
dnum = np.sum((w - np.mean(w))**2)

R2 = np.round(1 - num / dnum, 2)
# R2

R2_f(h,w)

```

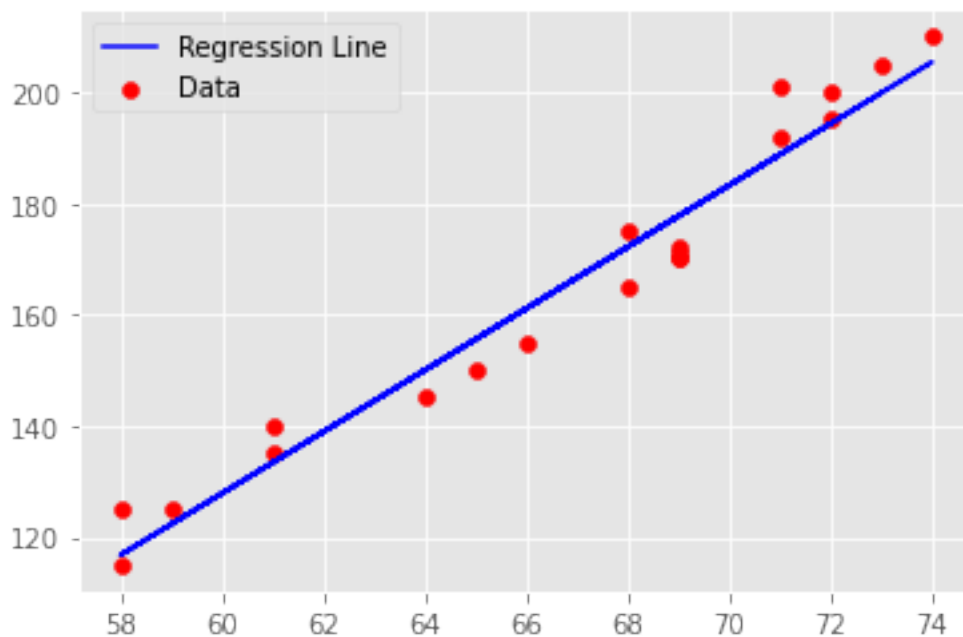
[53]: 0.96

[35]: m,c = m_and_c(h,w)

```

plt.scatter(h,w, color = "Red", label = "Data")
plt.plot(h, reg(h,w), color = "Blue", label = "Regression Line")
plt.legend();

```



1.6 Summary

In this lab, we ran a complete simple regression analysis experiment using functions created so far. Next up, you'll learn how you can use Python's built-in modules to perform similar analyses with a much higher level of sophistication.

[]: