

ML: Experiment 1

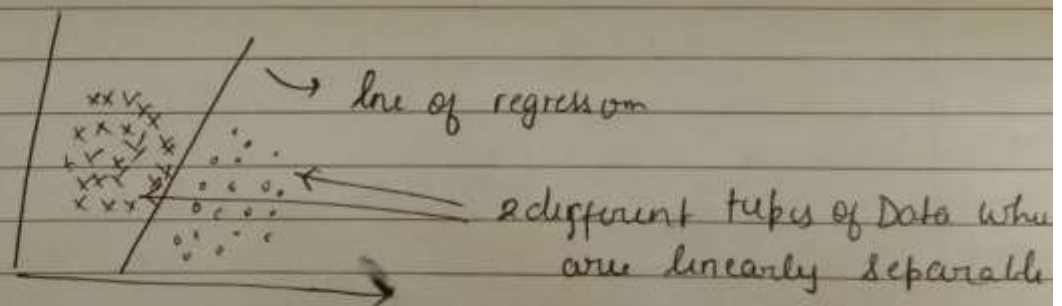
Aim: To implement Linear Regression

Theory:

LR is one of easiest & most popular ML Algos. It is a Statistical Method used for predictive analysis. It makes prediction for continuous / real / numeric Variables such as sales, salary, age etc. It shows linear relationship between dependent (Y) & one or more independent Variable (X).

It finds how values of X causes a change in the output variable Y.

This model provides sloped line representing relationship between variables.



Mathematically it is represented as

$$y = w_0 + w_1 x$$

y = independent variable

w_0 = Y intercept

w_1 = Slope of line

x = Independent variable.

Linear Regression:

- Simple - Single Independent Variable is used
- Multiple - More than one independent Variable used

• Conclusion

Hence we have implemented linear regression using both statistical method & machine learning.

ML- Lab 1 Linear regression

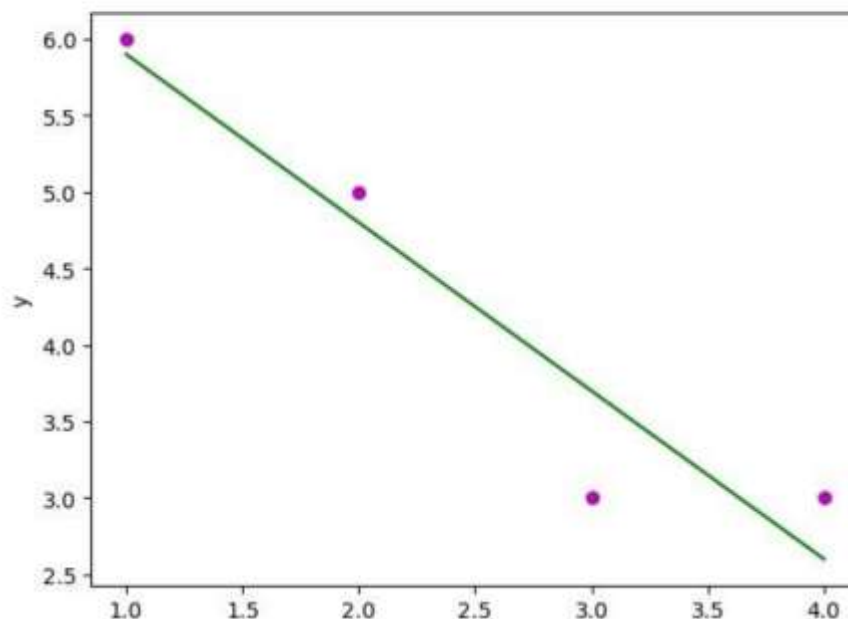
```
[2]: import numpy as np
import matplotlib.pyplot as plt
```

```
[3]: def estimate_coeff(x, y):
    n = np.size(x)
    mean_x = np.mean(x)
    mean_y = np.mean(y)
    SS_xy = np.sum(y * x) - n * mean_y * mean_x
    SS_xx = np.sum(x * x) - n * mean_x * mean_x
    SS_yx2 = mean_y * np.sum(x * x) - mean_x * np.sum(x * y)
    SS_x = np.sum(x * x) - n * mean_x * mean_x
    w_1 = SS_xy / SS_xx
    w_0 = SS_yx2 / SS_x
    return (w_0, w_1)
```

```
[4]: def plot_regression_line(x, y, w):
    plt.scatter(x, y, color = "m", marker = "o", s = 30)
    y_pred = w[0] + w[1] * x
    plt.plot(x, y_pred, color = "g")
    plt.xlabel('x')
    plt.ylabel('y')
    plt.show()
```

```
[6]: x = np.array([1, 2, 3, 4])
y = np.array([6, 5, 3, 3])
w = estimate_coeff(x, y)
print("Estimated coefficients - \nw_0 = {}\nw_1 = {}".format(w[0], w[1]))
print("The equation is : y = {} + {}x\n".format(w[0], w[1]))
```

```
Estimated coefficients w_0 =
7.0 w_1 = -1.1 The equation
is : y = 7.0 + -1.1x
```



```
[8]: import pandas as pd

[73]: X = np.array([2,3,4,5,6,7,8,9,10])
      y = np.array([1,3,6,9,11,13,15,17,20])

[78]: w_0 = 0.1
      w_1 = 0.2
      alpha = 0.01
      epochs = 100

[79]: for epoch in range(epochs):
      y_pred = w_0 + w_1 * X # predicted values
      error = y_pred - y # difference between predicted and actual values
```

```
# Update weights using gradient descent
w_0 -= alpha * np.mean(error) # update intercept
w_1 -= alpha * np.mean(error * X) # update slope
```

```
[81]: print("Intercept (w_0):", w_0)
      print("Slope (w_1):", w_1)

      # Plot the original data points
      plt.scatter(X, y, color='blue', label='Original data')

      # Plot the linear regression line
      plt.plot(X, w_0 + w_1 * X, color='red', label='Manual linear regression')

      plt.xlabel('X')
      plt.ylabel('Y')
      plt.title('Manual Linear Regression')
      plt.legend()
      plt.show()
```

```
Intercept (w_0): -0.2099873738061369
Slope (w_1):
1.8768486428539883
```

