

1) Linear Regression.

Used to find relationship between two variables
 one independent (input) and one dependent (output)
 if fit a straight line ($y = mx + c$) to the data

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
x = np.array([1, 2, 3, 4, 5, 6, 7, 8, 9, 10])
```

```
y = np.array([50, 55, 65, 70, 75, 78, 80, 85, 87, 90])
```

```
x_mean = np.mean(x)
```

```
y_mean = np.mean(y)
```

```
m = (sum((x - x_mean) * (y - y_mean)) /
      sum((x - x_mean) ** 2))
```

```
c = y_mean - m * x_mean
```

```
def Predict(x):
```

```
    return m * x + c
```

```
plt.scatter(x, y, color='blue', label="Actual Data")
```

```
y_pred = Predict(x)
```

```
plt.plot(x, y_pred, color='red', label="Regression Line")
```

```
plt.xlabel("Hours Studied")
```

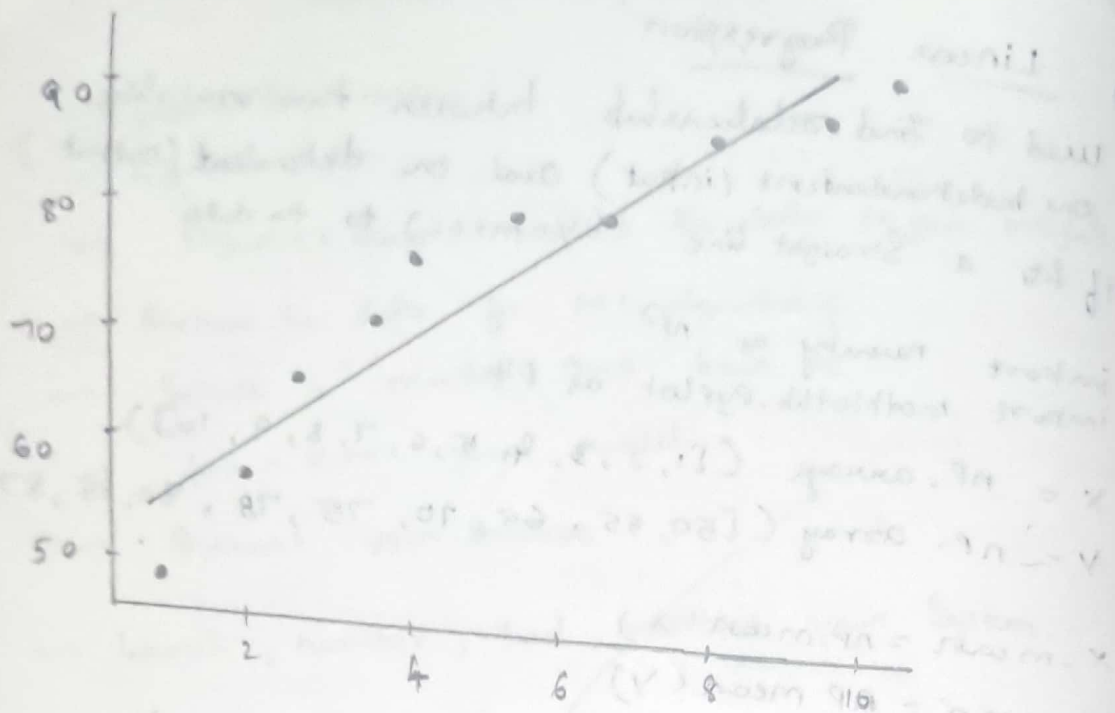
```
plt.ylabel("Test Score")
```

```
plt.title("Linear Regression: Hours vs Score")
```

```
plt.legend()
```

```
plt.show()
```

```
Print ("Prediction for 'Hours of Study':", Predict(11))
```



Linear regression: Hours vs Score
 Prediction for 11 hrs of Study: 97.4

• Actual data
 — Regression line

2) Multiple Regression

import numpy as np

import matplotlib.pyplot as plt

from mpl_toolkits.mplot3d import Axes3D

```
X = np.array([
    [100, 2], [1000, 3], [1200, 3],
    [1500, 4], [1800, 4], [2000, 5], [2200, 5],
    [2800, 6], [3000, 6]
])
```

```
Y = np.array([150, 180, 200, 240, 280, 310, 340,
              370, 400, 420]) # Prices in $1000
```

```
X = np.column_stack((np.ones((len(X))), X))
```

```
B = np.linalg.pinv(X.T.dot(X)).dot(X.T).dot(Y)
```

```
def Predict (left, beds):
```

```
fig = plt.figure(left, beds):
```

return B[0] + B[1] * Sqft + B[2] * beds

fig = plt.figure()

ax = fig.add_subplot(111, projection='3d')

Sqft_vals = x[:, 1]

bed_vals = x[:, 2]

Price_vals = y

ax.scatter(Sqft_vals, bed_vals, Price_vals, color='blue',
label="Actual Data")

Sqft_grid, bed_grid = np.meshgrid(np.linspace(800,
3000, 10), np.linspace(2, 5, 5))

Price_grid = predict(Sqft_grid, bed_grid)

ax.plot_surface(Sqft_grid, bed_grid, Price_grid,
color='red', alpha=0.5)

ax.set_xlabel("Square Footage")

ax.set_ylabel("Bedrooms")

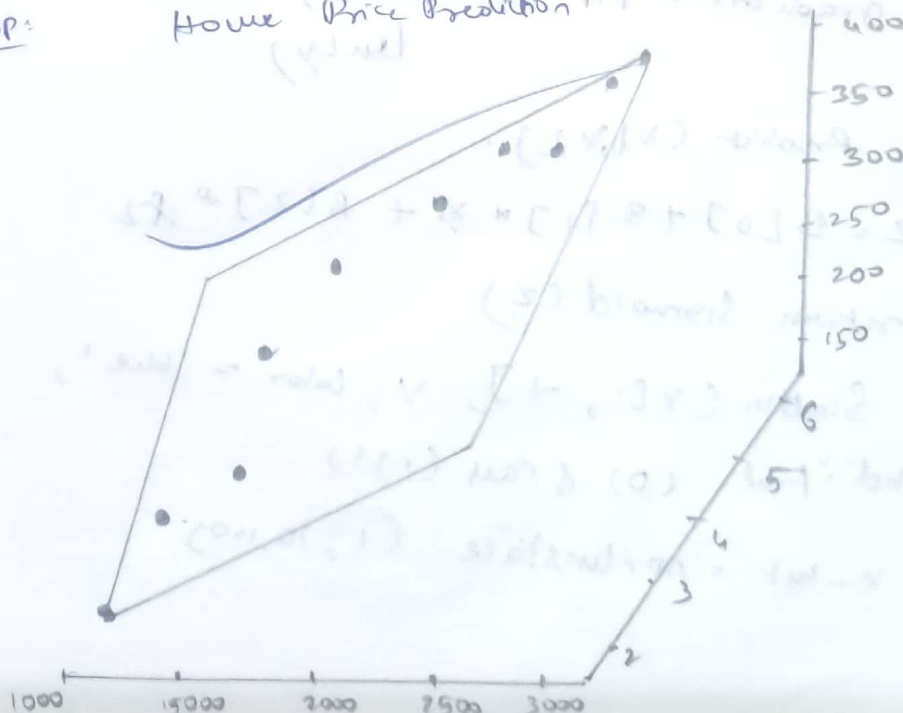
ax.set_zlabel("Price (\$1000)")

ax.set_title("Multiple Linear Regression:
House Price Prediction")

plt.show()

Print ("Prediction for 2600 Sqft, 5 bedrooms:",
predict(2600, 5))

op: House Price Prediction



3) logistic regression

import numpy as np
import matplotlib.pyplot as plt

def Sigmoid(z):
 return 1 / (1 + np.exp(-z))

x = np.array([
 [1, 50], [2, 55], [3, 60], [4, 62],
 [5, 65], [6, 70], [7, 75], [8, 78],
 [9, 85], [10, 90]])

y = np.array([0, 0, 0, 0, 1, 1, 1, 1, 1, 1])
0 = fail, 1 = Pass

X = np.c_[np.ones((len(x), 1)), x]
B = np.zeros(X.shape[1])

lr = 0.01

epochs = 2000

for i in range(epochs):

prediction = sigmoid(np.dot(X, B))

gradient = np.dot(X.T, (prediction - y)) /
 len(y)

def predict(x1, x2):

z = B[0] + B[1] * x1 + B[2] * x2

return Sigmoid(z)

plt.scatter(X[:, 1:], y, color = 'blue',
 label = 'Fail (0) & Pass (1)')

x_test = np.linspace(1, 10, 100)

Y_test = predict(x, 6)

plt.plot(x_test, Y_test)

"Logistic Regression"

plt.xlabel("Hours Studied")

plt.ylabel("Probability")

plt.legend()

plt.show()

print("Probability of
studied 8-10 hours")

predict(8, 10)

Logistic Regression

This Studied 8

$Y_{-test} = [predict(x, 65)] \text{ for } x \text{ in } X_{-test}]$

plt.plot(X_test, Y_test, color='red', label =

"Logistic Regression Curve")

plt.xlabel("Hours Studied")

plt.ylabel("Probability of Passing")

plt.legend()

plt.show()

print("Probability of Passing for this
Student's 75 Per Score : ",
predict(7, 75))

Logistic Regression : Pass/Fail Prediction for
this Student's 75 (X, Y)

(X_train, Y_train) = data_train

(X_test, Y_test) = data_test

(X_train_scaled, Y_train_scaled) = scale(X_train, Y_train)

(X_test_scaled, Y_test_scaled) = scale(X_test, Y_test)

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