

Simple

$$y = bx + a$$

$$b = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2} ; a = \bar{y} - b\bar{x}$$

Code:

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$$x = [1, 2, 3, 4, 5]$$

$$y = [12, 15, 13, 14, 16]$$

$$n = \text{len}(x)$$

$$\text{sum}_x = 0$$

$$\text{sum}_y = 0$$

for i in x:

$$\text{sum}_x += i$$

for i in y:

$$\text{sum}_y += i$$

$$\text{mean}_x = \text{sum}_x / n$$

$$\text{mean}_y = \text{sum}_y / n$$

$$\text{numerator} = 0$$

$$\text{denominator} = 0$$

for i in range(n):

$$\text{numerator} += (x[i] - \text{mean}_x) * (y[i] - \text{mean}_y)$$

$$\text{denominator} += (x[i] - \text{mean}_x)^2$$

$$b = \text{numerator} / \text{denominator}$$

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a = mean y - b * mean x
ss = 0
print(b, '-')
for i in range(n):
    predy = b * x[i] + a
    print(y, predy)
    ss += (y[i] - predy) ** 2
print(ss)

```

multiple

$$\beta = (X^T X)^{-1} X^T y$$

$$\hat{y} = X \beta = X (X^T X)^{-1} X^T y$$

Code: import numpy as np  
 from numpy.linalg import inv.

a = [ ]

b = [ ]

y = [ ]

X = [ ]

for i in range(len(a)):

    X.append([ ])

    X[i].append(1)

    X[i].append(a[i])

    X[i].append(b[i])

print(X)

matrix = np.linalg.pinv(X)

#  $\beta_0 + \beta_1 x_1 + \beta_2 x_2 \dots$

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print(x)
matr = np.array(x)
print(matr)
tmp = inv((matr.T).dot(matr))
tmp2 = tmp.dot(matr.T)
print(tmp2)
beta = tmp2.dot(np.array(y))
print(beta)
print(matr.dot(beta))

```

Using Scikit learn: { Same for Simple & multiple }

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from pandas import DataFrame
from sklearn import linear_model

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d2 = {
    } # dictionary data

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df = DataFrame(d2, columns=['y', 'm', 'i', 'u', 'output'])
X = df[['i', 'u']] # like previous example in class
# pass a 2D list only

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y = df['output']

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reg = linear_model.LinearRegression() # regression object

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# fit() # predict()

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reg.fit(X, y) → reg.fit(X.values, y)

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print(reg.intercept_) # a

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print(reg.coef_) # b

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print(reg.predict([[0.75], [0.3]])) # predict for

```

← 2D list

```
print (neg. coe - )  
print (neg. predict([2.75, 1.3])) # predict for  
one random ex:
```

for single & multiple : both  $\rightarrow$  2D list