

Simple Linear Regression

⇒ When we have a single input attribute (x) and we want to use linear regression, this is called Simple Linear Regression.

⇒ If we had multiple input attributes (eg. x_1, x_2, x_3 etc.) This would be called multiple linear regression.

⇒ With Simple Linear Regression ~~Model~~ we want to model our data as follows:-

$$Y = B_0 + B_1 x$$

↓ ↓ ↓
O/P Variable Coefficient Input Variable

Technically B_0 is called the intercept and B_1 is called the slope.

⑦
⇒ The goal is to find the best estimates for the coefficients to minimize the errors in predicting y from x .

⇒ First we estimate the value of B_1 as:

$$B_1 = \frac{\text{sum}((x_i - \text{mean}(x)) * (y_i - \text{mean}(y)))}{\text{sum}((x_i - \text{mean}(x))^2)}.$$

$\text{mean}()$ → average value for the variable in our dataset.

⇒ we can calculate B_0 using B_1 and some statistics from our dataset as follows:

$$B_0 = \text{mean}(y) - B_1 * \text{mean}(x).$$

Estimating the Slope (B_1)

First we need to calculate the mean value of x and y .

$$\text{mean} = \frac{1}{n} \times \text{sum}(x)$$

$n \rightarrow$ in the no. of values (5 in this case).
We can use the AVERAGE() function in your spreadsheet.

$$\begin{aligned}\text{mean}(x) &= 3 \\ \text{mean}(y) &= 2.8\end{aligned}$$

$$\frac{1}{5} \times 15 = 3$$

$$\frac{1}{5} \times 14 = 2.8$$

	X	Y
1	1	1
2	2	3
3	4	3
4	3	2
5	5	5

Now we calculate the error of each variable from the mean

1	X	mean(x)	$X - \text{mean}(X)$
2	1	3	-2
3	2	3	-1
4	3	3	0
5	4	3	1
6	5	3	2

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	y	$\text{mean}(y)$	$y - \text{mean}(y)$
1	1	2.8	-1.8
2	1	2.8	-1.8
3	3	2.8	0.2
4	3	2.8	0.2
5	2	2.8	-0.8
6	5	2.8	2.2

\Rightarrow Now we have all parts for calculating the numerator. All we need to do is multiple the error for each x with error for each y . and calculate the sum of multiplications

	$x - \text{mean}(x)$	$y - \text{mean}(y)$	multiplier
1	-2	-1.8	3.6
2	-2	-1.8	3.6
3	-1	0.2	-0.2
4	-1	0.2	-0.2
5	1	-0.8	-0.8
6	3	2.2	6.6
			<u>4.4</u>
			7

$$\begin{array}{r}
 3.6 \\
 + 6.6 \\
 \hline
 10.2 \\
 - 3.2 \\
 \hline
 7.0
 \end{array}$$

⇒ summing the final column we have the sum as 0.7

⇒ now we calculate the bottom part of the eqⁿ for calculating B_1

	$x - \text{mean}(x)$	Squared
1	-2	4
2	-1	1
3	1	1
5	0	0
8	2	4

$$\frac{7}{10} = 0.7$$

$$B_1 = 8/10 = 0.8$$

now estimating the intercept (B_0)

$$B_0 = \text{mean}(y) - B_1 \times \text{mean}(x)$$

$$B_0 = 2.8 - 0.8 \times 3$$

$$B_0 = 0.4$$

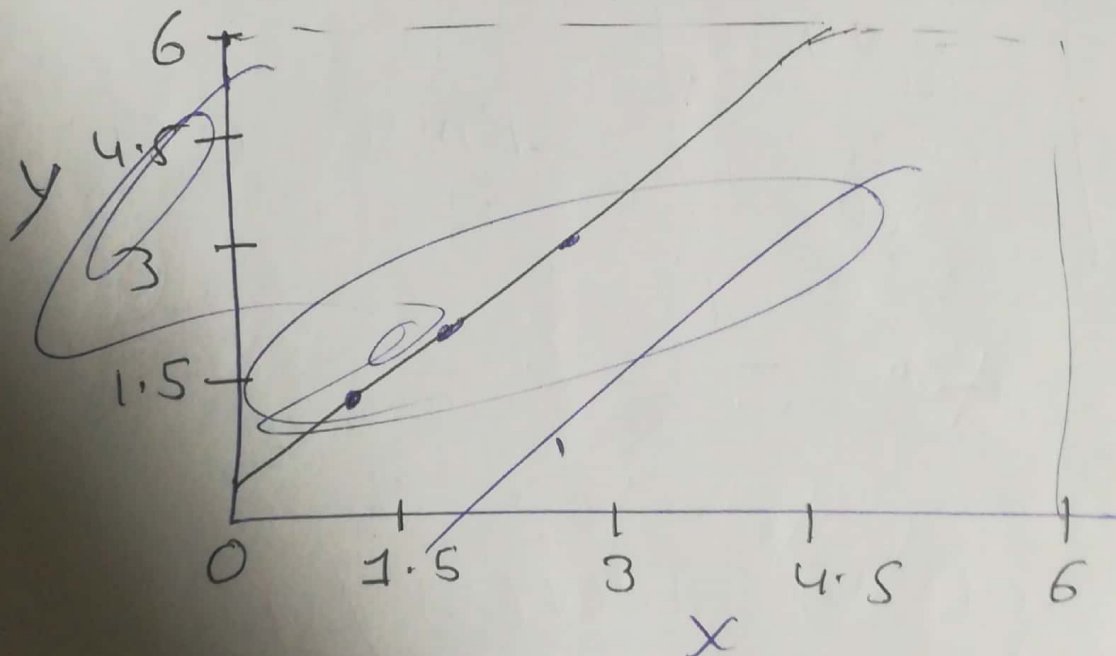
making Predictions.

$$y = B_0 + B_1 \times x$$

$$y = 0.7 + 0.8 \times x$$

Let's try out the model by making Predictions for our training data.

1	x	y	Predicted y
2	1	1	1.2
3	2	3	2
4	3	3	3.6
5	4	2	2.8
6	5	5	4.4



Estimating the Error.

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Root mean Squared Error or RMSE

$$RMSE = \sqrt{\frac{\sum (p_i - y_i)^2}{n}}$$

\downarrow Predicted value \downarrow actual value.

i	Predicted	y	error
1	1.2	1	0.2
2	2	3	-1
3	3.6	3	0.6
4	2.8	2	0.8
5	4.4	5	0.6

error	Squared error
0.2	0.04
-1	1
0.6	0.36
0.8	0.64
-0.6	0.36

$$RMSE = \underline{\underline{0.692}}$$

Shortcut:-

$$B_1 = \frac{\text{Corr}(X, Y) * \text{Stdev}(Y)}{\text{Stdev}(X)}$$

\downarrow
 Pearson's correlation coefficient.