

# *Getting Started with Machine Learning*

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So far now we've covered regression in detail.

Now let's move on to Classification.

In this tutorial we'll cover Logistic Regression.

Well, isn't it Logistic REGRESSION ?

Even though it is Logistic Regression, it is a binary classification problem.

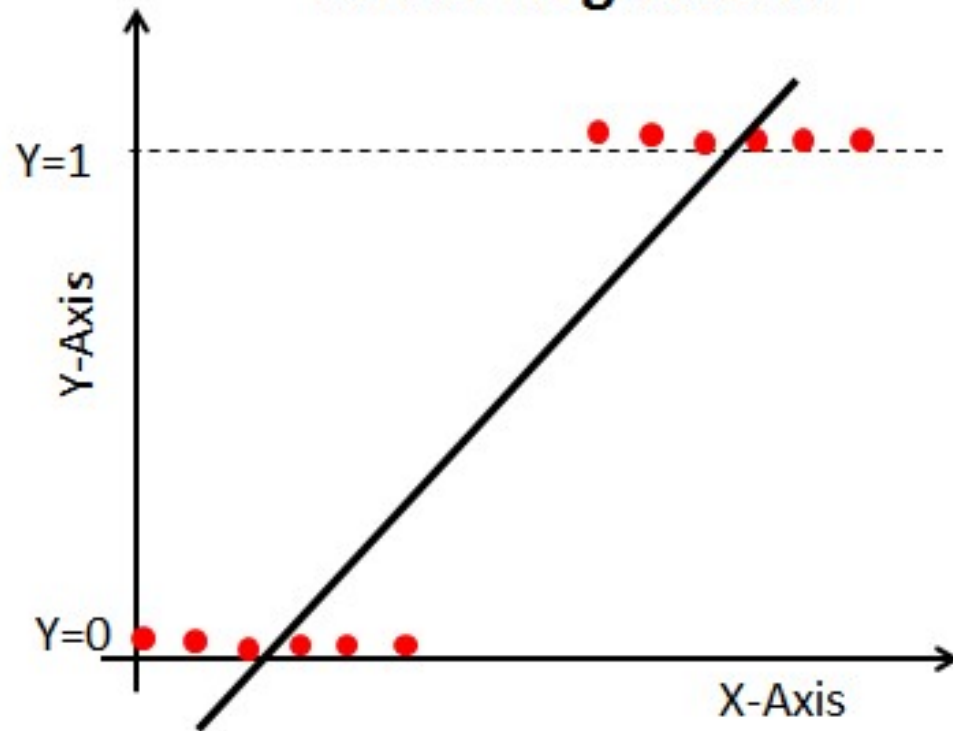
## Formal Definition of Logistic Regression :

In statistics, the logistic model is used to model the probability of a certain class or event existing such as pass/fail, win/lose, alive/dead or healthy/sick. This can be extended to model several classes of events such as determining whether an image contains a cat, dog, lion, etc.

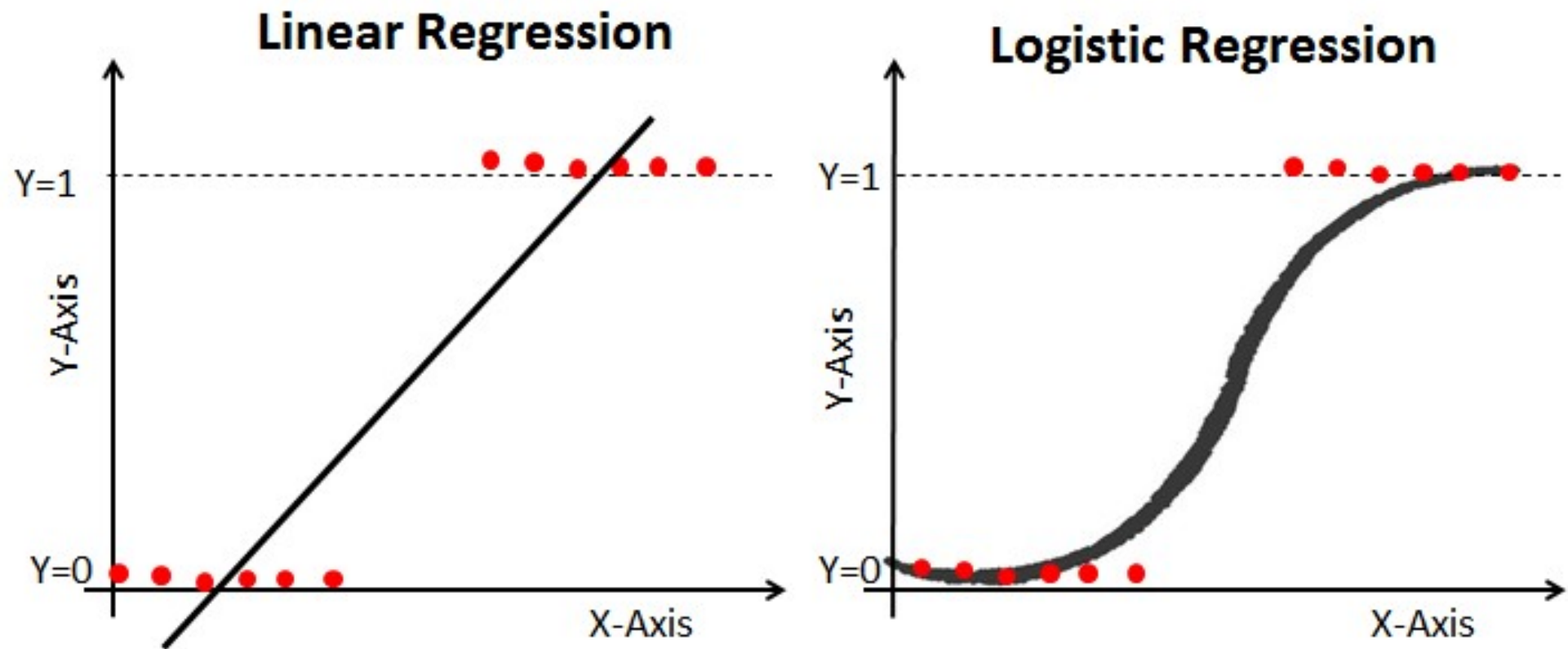
Source :

[https://en.wikipedia.org/wiki/Logistic\\_regression](https://en.wikipedia.org/wiki/Logistic_regression)

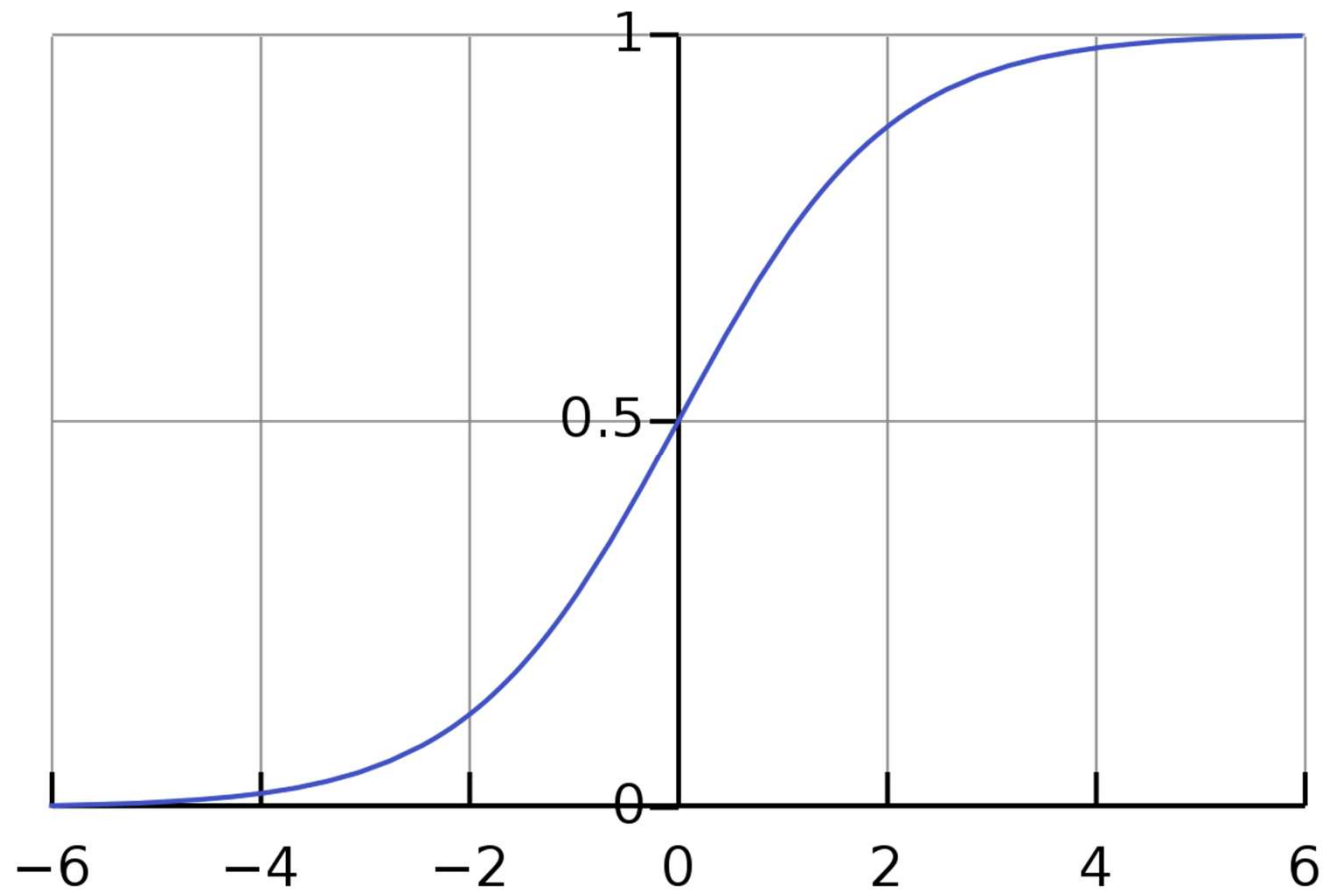
## Linear Regression



Source : <https://www.datacamp.com/community/tutorials/understanding-logistic-regression-python>



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Source : [https://en.wikipedia.org/wiki/Logistic\\_function](https://en.wikipedia.org/wiki/Logistic_function)

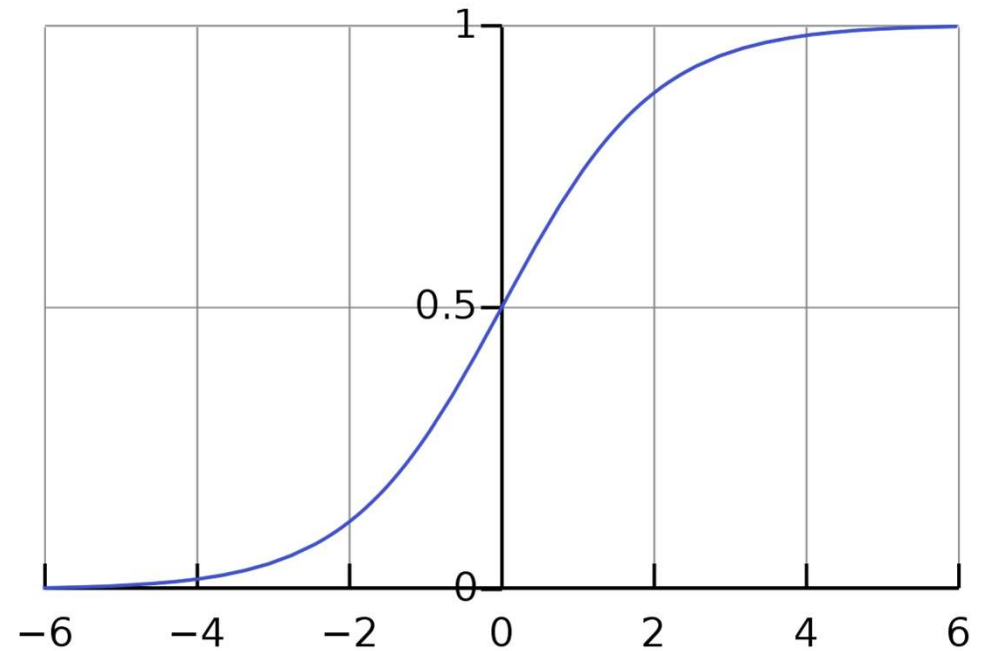
Logistic Function.

$$f(x) = \frac{L}{1 + e^{-k(x-x_0)}}$$

$x_0 = x$  value of midpoint

$L = \text{Max Value}$

$k = \text{growth rate}$



Logistic Function is also known as Sigmoid Function.

$$\sigma(Z) = \frac{1}{1 + e^{-Z}}$$

Where Z is an equation.

In our case we'll use function of Linear Regression and convert into Logistic Regression using Sigmoid.



Logistic Regression is given by :

$$Z = \beta_0 + (\beta_1 * x_i)$$

$$\sigma(Z) = \frac{1}{1 + e^{-Z}}$$

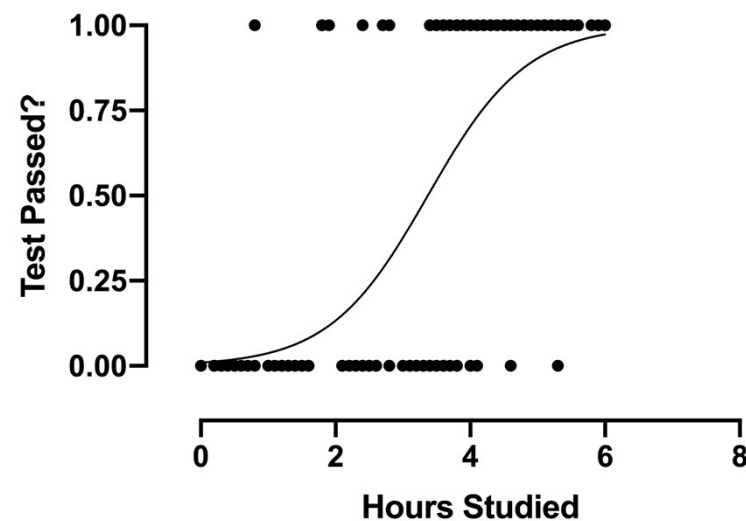
$$\sigma(Z) = \frac{1}{1 + e^{-\{\beta_0 + (\beta_1 * x_i)\}}}$$

$$\hat{y} = \sigma(Z)$$

So  $\hat{y} = \sigma(Z)$  yields Probability.

Probabilities lie between 0-1 i.e 0-100%

So we now set a threshold of .5 i.e 50%



Source : [https://www.graphpad.com/guides/prism/8/curve-fitting/reg\\_simple\\_logistic\\_and\\_linear\\_difference.htm](https://www.graphpad.com/guides/prism/8/curve-fitting/reg_simple_logistic_and_linear_difference.htm)

*So  $\hat{y}$  basically yeilds probability.  
So probability of being 1 & probaility of being 0 should be calculated.*

*For that we set the threshold. Usually threshold is set to 0.5*

$$\hat{y} \geq 0.5 \mid \text{prediction } y = 1$$

$$\hat{y} < 0.5 \mid \text{prediction } y = 0$$

Understanding algorithm in terms of Probability :

$$\sigma(Z) = P(\hat{y} = 1|Z; \{\beta_0, \beta_1\})$$

$$\sigma(Z) = P(\hat{y} = 0|Z; \{\beta_0, \beta_1\})$$

$$P(\hat{y} = 0|Z; \{\beta_0, \beta_1\}) = 1 - P(\hat{y} = 1|Z; \{\beta_0, \beta_1\})$$

## Understanding Logistic Regression with an example :

Considering the prediction of classifying whether the student is going to pass/fail the examination based on number of hours they've devoted to studying.

Hours of Studying (X)	Examination Outcome (y)
0.50	0
1.50	0
2.00	0
4.25	1
3.25	1
5.50	1

Hours of Studying (X)	Examination Outcome (y)	$(x_i - \bar{x})$	$(y_i - \bar{y})$	$(x_i - \bar{x})^2$	$(y_i - \bar{y})^2$	$(x_i - \bar{x}) \cdot (y_i - \bar{y})$
0.50	0	-2.33	-.5	5.4289	.25	1.165
1.50	0	-1.33	-.5	1.7689	.25	.665
2.00	0	-.83	-.5	.6889	.25	.415
4.25	1	1.42	.5	2.0164	.25	.71
3.25	1	.42	.5	.1764	.25	.21
5.50	1	2.67	.5	7.1289	.25	1.335
Sum = 17	Sum = 3	Sum = .02	Sum = 0	Sum = 17.2084	Sum = 1.5	Sum = 4.5

Hours of Studying (X)	Examination Outcome (y)	$(x_i - \bar{x})$	$(y_i - \bar{y})$	$(x_i - \bar{x})^2$	$(y_i - \bar{y})^2$	$(x_i - \bar{x}) \cdot (y_i - \bar{y})$
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$$\text{Mean of } x : \bar{x} = \frac{\sum x_i}{N} = \frac{17}{6} = 2.83$$

$$\text{Mean of } y : \bar{y} = \frac{\sum y_i}{N} = \frac{3}{6} = .5$$

Hours of Studying (X)	Examination Outcome (y)	$(x_i - \bar{x})$	$(y_i - \bar{y})$	$(x_i - \bar{x})^2$	$(y_i - \bar{y})^2$	$(x_i - \bar{x}) \cdot (y_i - \bar{y})$
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$$\beta_1 = \frac{(x_i - \bar{x}) \cdot (y_i - \bar{y})}{(x_i - \bar{x})^2} = \frac{4.5}{17.2084} = 0.2615$$

$$\beta_0 = \bar{y} - (\beta_1 * x_i) = .5 - (.2615 * 2.83) = -.240045$$



Hours of Studying (X)	Examination Outcome (y)	$(x_i - \bar{x})$	$(y_i - \bar{y})$	$(x_i - \bar{x})^2$	$(y_i - \bar{y})^2$	$(x_i - \bar{x}) \cdot (y_i - \bar{y})$
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Sum = 17	Sum = 3	Sum = .02	Sum = 0	Sum = 17.2084	Sum = 1.5	Sum = 4.5

$$Z = \beta_0 + (\beta_1 * x_i) = -.240045 + (.2615 * x_i)$$

$$\sigma(Z) = \frac{1}{1 + e^{-Z}} = \frac{1}{1 + e^{-\beta_0 + (\beta_1 * x_i)}} = \frac{1}{1 + e^{-\{-.240045 + (.2615 * x_i)\}}}$$

$$\hat{y} = \sigma(Z)$$

Hours of Studying (X)	Examination Outcome (y)	$(x_i - \bar{x})$	$(y_i - \bar{y})$	$(x_i - \bar{x})^2$	$(y_i - \bar{y})^2$	$(x_i - \bar{x}) \cdot (y_i - \bar{y})$
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*Predict the status of examination outcome where the student was studying for 2.7 hours on daily basis.*

$$\hat{y} = \sigma(Z) = \frac{1}{1 + e^{-Z}} = \frac{1}{1 + e^{-\beta_0 + (\beta_1 * x_i)}} = \frac{1}{1 + e^{-\{-.24004 \quad (.2615 * 2.7)\}}}$$

$$\hat{y} = .6144 \approx 61.44\%$$

This exceeds the threshold of 50% which indicates the outcome is 1, i.e The student is going to pass based on 2.7 hours of preparation on daily basis.

Hours of Studying (X)	Examination Outcome (y)	$(x_i - \bar{x})$	$(y_i - \bar{y})$	$(x_i - \bar{x})^2$	$(y_i - \bar{y})^2$	$(x_i - \bar{x}) \cdot (y_i - \bar{y})$
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Sum = 17	Sum = 3	Sum = .02	Sum = 0	Sum = 17.2084	Sum = 1.5	Sum = 4.5

*Predict the status of examination outcome where the student was studying for 0.75 hours on daily basis.*

$$\hat{y} = \sigma(Z) = \frac{1}{1 + e^{-Z}} = \frac{1}{1 + e^{-\beta_0 + (\beta_1 * x_i)}} = \frac{1}{1 + e^{-\{-.240045 + (.2615 * .75)\}}}$$

$$\hat{y} = .4890 \approx 48.90\%$$

This does not exceed the threshold of 50% which indicates the outcome is 0, i.e The student is going to fail based on .75 hours of preparation on daily basis.

# Thank You