Decision Boundary

In order to get our discrete 0 or 1 classification, we can translate the output of the hypothesis function as follows:

$$\mathbf{h}_{\theta}(\mathbf{x}) \ge 0.5 \rightarrow y = 1$$

$$\mathbf{h}_{\theta}(\mathbf{x}) < 0.5 \rightarrow y = 0$$

The way our logistic function g behaves is that when its input is greater than or equal to zero, its output is greater than or equal to 0.5:

$$g(z) \ge 0.5$$

when $z \ge 0$

Remember.

$$z=0$$
, $\mathbf{e^0}=1\Rightarrow g(z)=1/2$
 $z\to\infty$, $\mathbf{e^{-\omega}}\to 0\Rightarrow g(z)=1$
 $z\to-\infty$, $\mathbf{e^{\omega}}\to\infty\Rightarrow g(z)=0$

So if our input to g is $\theta^T X$, then that means:

$$\mathbf{h}_{\theta}(x) = g(\theta^T x) \ge 0.5$$

When $\theta^T x \ge 0$

From these statements we can now say:

$$\theta^T \mathbf{x} \ge 0 \Rightarrow y = 1
\theta^T \mathbf{x} < 0 \Rightarrow y = 0$$

The **decision boundary** is the line that separates the area where y = 0 and where y = 1. It is created by our hypothesis function.

Example:

5
$$\theta = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$
 $y=1 \text{ If } 5+(-1) x_1+0 x_2 \ge 0$

$$5-x_1 \ge 0$$

$$-x_1 \ge -5$$

$$x_1 \le 5$$

In this case, our decision boundary is a straight vertical line placed on the graph where $\mathbf{x_1} = \mathbf{5}$, and everything to the left of that denotes y = 1, while everything to the right denotes y = 0.

Again, the input to the sigmoid function g(z) (e.g. $\theta^T X$) doesn't need to be linear, and could be a function that describes a circle (e.g. $z = \theta_0 + \theta_1 x_1^2 + \theta_2 x_1^2$) or any shape to fit our data.