Logistics Regression

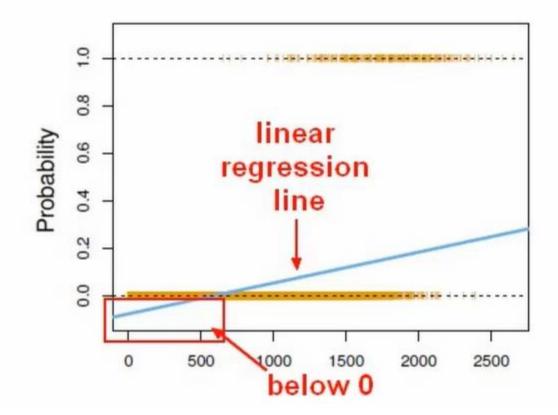
We want to learn about Logistic Regression as a method for **Classification**.

Some examples of classification problems:

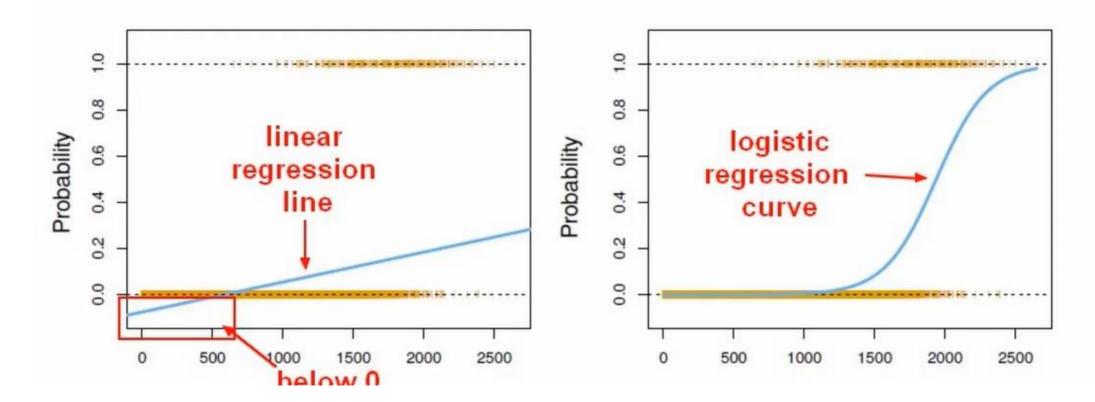
- Spam versus "Ham" emails
- Loan Default (yes/no)
- Disease Diagnosis

Above were all examples of Binary Classification

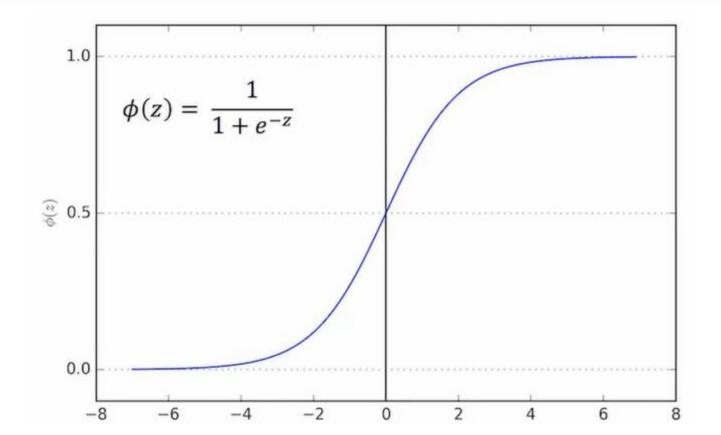
We can't use a normal linear regression model on binary groups. It won't lead to a good fit:



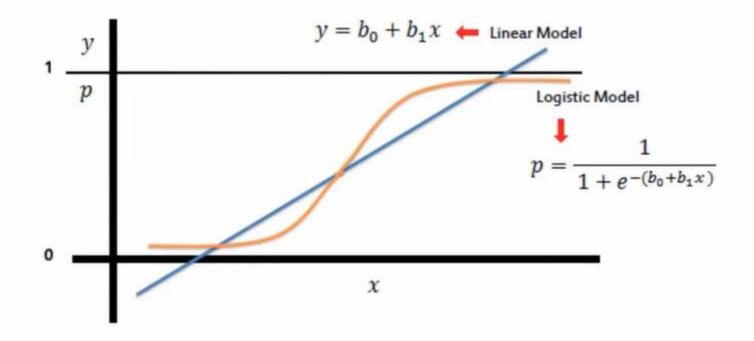
Instead we can transform our linear regression to a logistic regression curve.



The Sigmoid (aka Logistic) Function takes in any value and outputs it to be between 0 and 1.

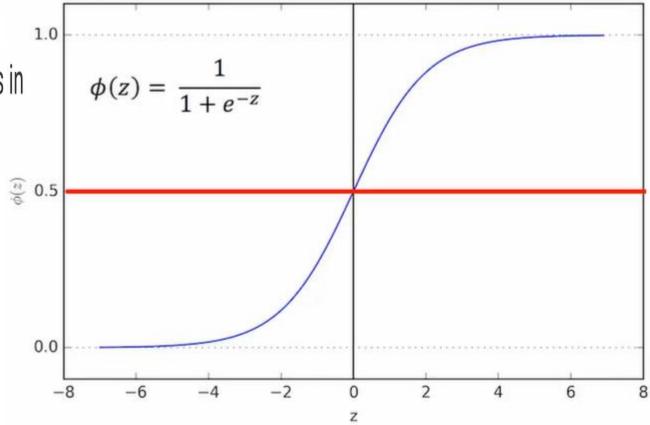


This means we can take our Linear Regression Solution and place it into the Sigmoid Function.



This results in a probability from 0 to 1 of belonging in the 1 class.

We can set a cutoff point at 0.5, anything below it results in class 0, anything above is class 1.



We can use a confusion matrix to evaluate our model. For example, imagine testing for disease.

	Predicted:	Predicted:
n=165	NO	YES
Actual:		
NO	50	10
Actual:		
YES	5	100

Example: Test for presence of disease

NO = negative test = False = 0

YES = positive test = True = 1

n=165	Predicted: NO	Predicted: YES	
Actual: NO	TN = 50	FP = 10	60
Actual: YES	FN = 5	TP = 100	105
	55	110	

Basic Terminology:

- True Positives (TP)
- True Negatives (TN)
- False Positives (FP)
- False Negatives (FN)

n=165	Predicted: NO	Predicted: YES	
Actual: NO	TN = 50	FP = 10	60
Actual: YES	FN = 5	TP = 100	105
	55	110	

Accuracy:

- Overall, how often is it correct?
- (TP + TN) / total = 150/165 = 0.91

n=165	Predicted: NO	Predicted: YES	
Actual: NO	TN = 50	FP = 10	60
Actual: YES	FN = 5	TP = 100	105
	55	110	

Misclassification Rate (Error Rate):

- Overall, how often is it wrong?
- (FP + FN) / total = 15/165 = 0.09

