

# Logistic Regression

# Binary Classification

- $X$  is one or more attributes
- $y$  is in  $\{0,1\}$
- Example
  - $x$  - size of tumor
  - $y$  - malignant? (yes or no)
  - $\hat{y}$  - predicted value of  $y$  (sometimes  $y'$ )

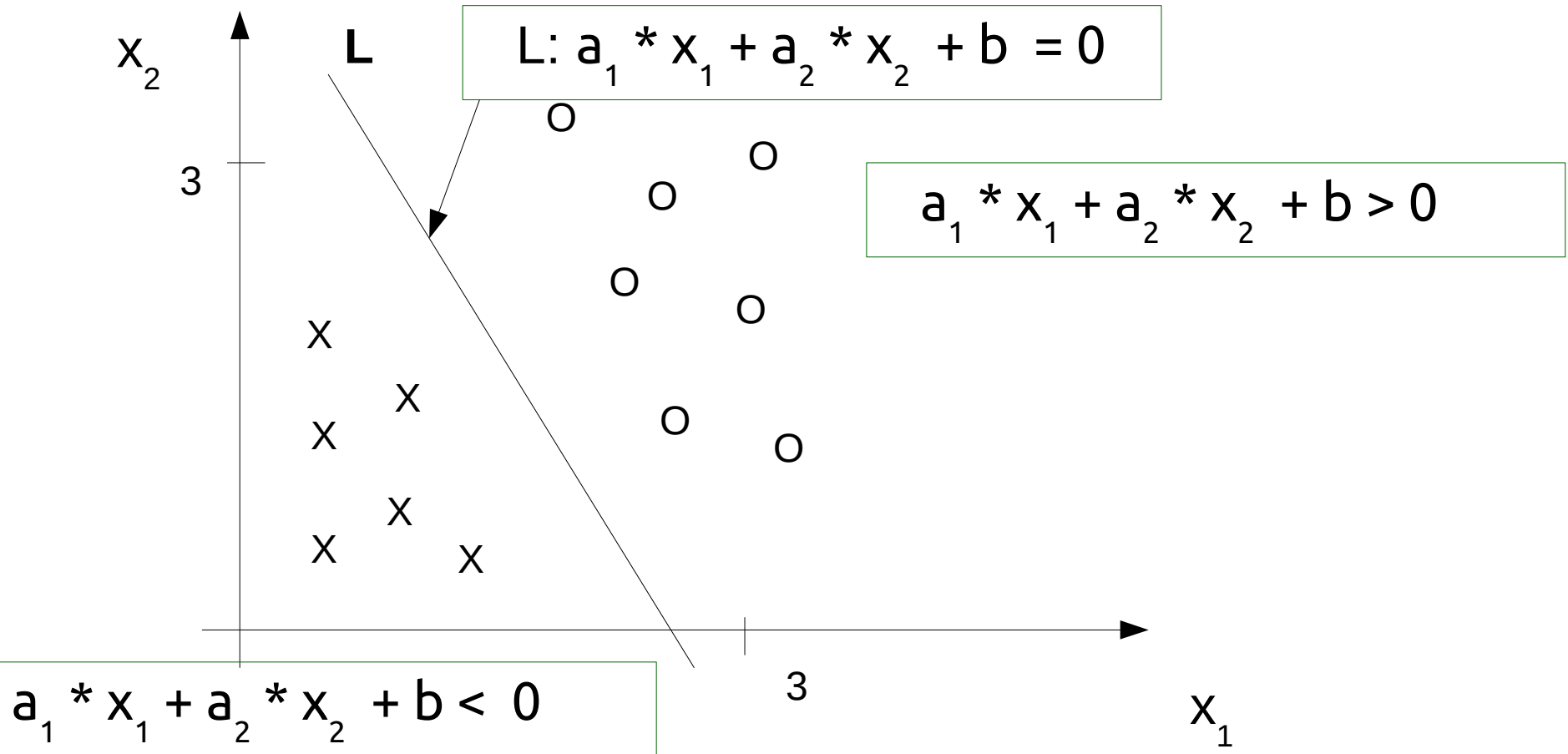
# Classifiers

- Probabilistic Classifier – outputs probabilities of an instance belonging to a particular class.
- For binary classification output the probability  $p$  of a positive.
- Discrete classifier – outputs a predicted class rather than a probability.

# Probabilistic Classifier

- Logistic Regression builds a Probabilistic Classifier.
- So  $\hat{y}$  (or  $p$ ) is between 0 and 1.
- $\hat{y}$  is the probability of a positive.
- (A positive is an event of interest, for example having a medical condition.)

# Example - Two Features



# Separation Line

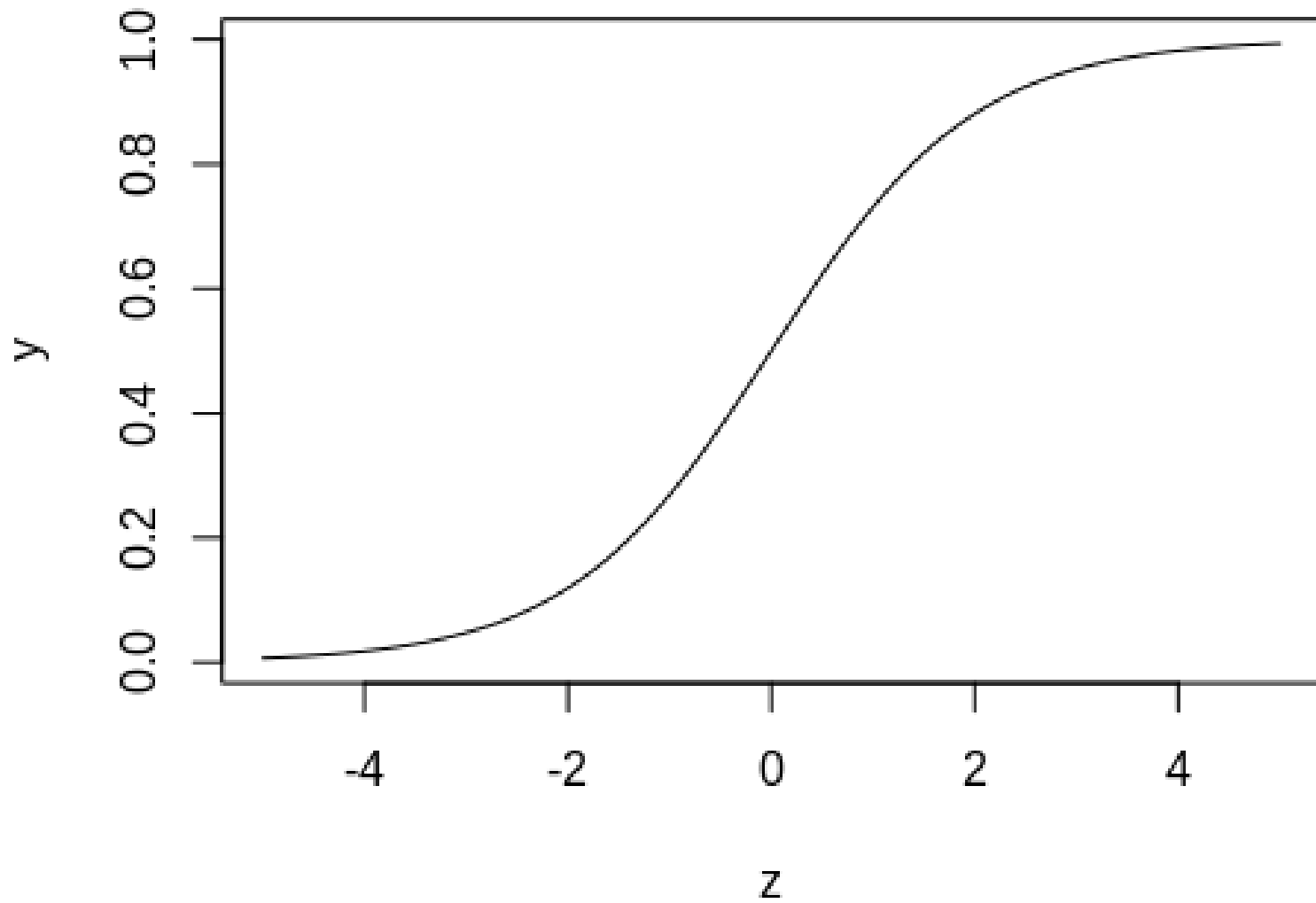
- Suppose for a moment we have the value of the parameters  $a$ ,  $b_1$  and  $b_2$ .
- A value of  $a_1 * x_1 + a_2 * x_2 + b < 0$  indicates an x.
- A value of  $a_1 * x_1 + a_2 * x_2 + b > 0$  indicates an o.
- However the output now is between  $-\infty$  and  $\infty$ .
- What we need is a value between 0 and 1.

# Sigmoid Function

$$y = \frac{1}{(1 + e^{-z})}$$

- Also called the logistic function
- Maps all  $z$  values to values between 0 and 1
- Its a way to maps values between  $-\infty$  and  $\infty$  to values between 0 and 1.

## Sigmoid Function





# Logistic Regression

- Want  $0 \leq \hat{y} \leq 1$
- $z = a_1 * x_1 + .. a_n * x_n + b$
- $z$  can take values between  $-\infty$  and  $\infty$
- Apply a function  $g(z)$  to  $z$  to give values between 0 and 1.

$$z = \sum_k a_k x_k + b$$

$$\hat{y} = g(z) = \frac{1}{(1 + e^{-z})}$$

# Interpretation

- Can now interpret  $\hat{y}$  as a probability of a positive value (1).
- For example,  $\hat{y} = 0.7$ 
  - 70% chance of tumor being malignant.
  - 30% chance of it not being malignant.

# Prediction/ Threshold

- In order to make a prediction
  - choose threshold (say 0.5)
  - if  $\hat{y} > \text{threshold}$ , predict  $y = 1$ .
- [Notice that  $\hat{y} = g(z) > 0.5$  when  $z > 0$ ]
- Thresholds will be examined in detail in the next section.
- A threshold converts a probabilistic classifier to discrete classifier.