

## 3 LDA

Logistic regression involves direction modeling  $P(Y = k|X = x)$  using the logistic function for the case of two response classes. We now consider a less direct approach.

**Idea:**

Why do we need another method when we have logistic regression?

1.

2.

3.

### 3.1 Bayes' Theorem for Classification

Suppose we wish to classify an observation into one of  $K$  classes, where  $K \geq 2$ .

$$\pi_k$$

$$f_k(x)$$

$$P(Y = k|X = x)$$

In general, estimating  $\pi_k$  is easy if we have a random sample of  $Y$ 's from the population.

Estimating  $f_k(x)$  is more difficult unless we assume some particular forms.

## 3.2 $p = 1$

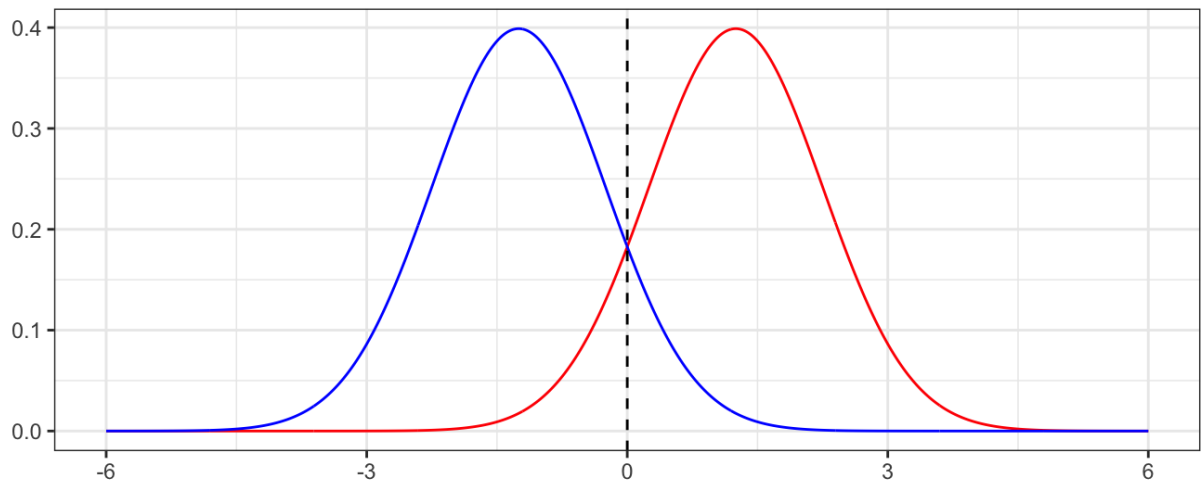
Let's (for now) assume we only have 1 predictor. We would like to obtain an estimate for  $f_k(x)$  that we can plug into our formula to estimate  $p_k(x)$ . We will then classify an observation to the class for which  $\hat{p}_k(x)$  is greatest.

Suppose we assume that  $f_k(x)$  is normal. In the one-dimensional setting, the normal density takes the form

Plugging this into our formula to estimate  $p_k(x)$ ,

We then assign an observation  $X = x$  to the class which makes  $p_k(x)$  the largest. This is equivalent to

**Example 3.1** Let  $K = 2$  and  $\pi_1 = \pi_2$ . When does the Bayes classifier assign an observation to class 1?



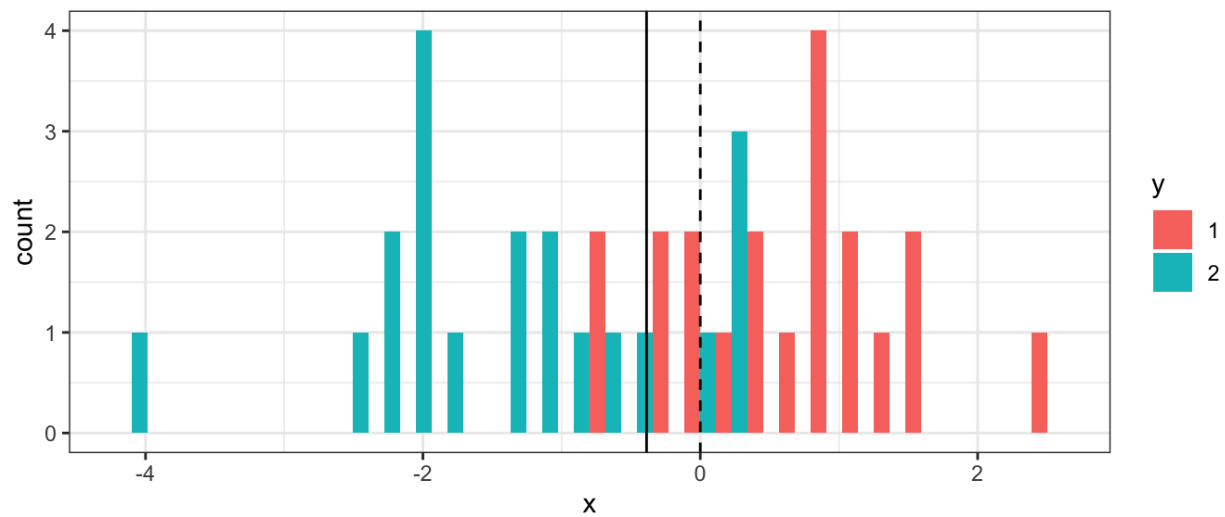
In practice, even if we are certain of our assumption that  $X$  is drawn from a Gaussian distribution within each class, we still have to estimate the parameters

$$\mu_1, \dots, \mu_K, \pi_1, \dots, \pi_K, \sigma^2.$$

The *linear discriminant analysis* (LDA) method approximated the Bayes classifier by plugging estimates in for  $\pi_k, \mu_k, \sigma^2$ .

Sometimes we have knowledge of class membership probabilities  $\pi_1, \dots, \pi_K$  that can be used directly. If we do not, LDA estimates  $\pi_k$  using the proportion of training observations that belong to the  $k$ th class.

The LDA classifier assigns an observation  $X = x$  to the class with the highest value of



```
##      pred
## y      1      2
## 1 18966 1034
## 2  3855 16145
```

The LDA test error rate is approximately 12.22% while the Bayes classifier error rate is approximately 10.52%.

The LDA classifier results from assuming that the observations within each class come from a normal distribution with a class-specific mean vector and a common variance  $\sigma^2$  and plugging estimates for these parameters into the Bayes classifier.