



# 3.1 Pre-class Work

## 1. Barber Textbook Exercises

### Exercise 17.1.

1. Give an example of a two-dimensional dataset for which the data are linearly separable, but not linearly independent.
2. Can you find a dataset which is linearly independent but not linearly separable?

### 17.1 Answers

1.  $\{([1,1],0), ([2,2],1)\} \rightarrow$  Can be separated with a single line, however,  $[2,2]$  is a linear combination of  $[1,1]$
2. Not sure.

**Exercise 17.2.** Show that for both Ordinary ~~and Orthogonal~~ Least Squares regression fits to data  $(x^n, y^n), n = 1, \dots, N$  the fitted lines go through the point  $\sum_{n=1}^N (x^n, y^n)/N$ .

$$\sum_{n=1}^N (x^n, y^n)/N = \bar{y}$$

With the regression line, this is estimated by

$$\bar{y} = \frac{1}{N} \sum_{n=1}^N \hat{y}_n + \epsilon_n$$

$$\hat{y}_n = \hat{\beta}_0 + \hat{\beta}_1 x_n$$

Plugging in

$$\bar{y} = \frac{1}{N} \sum_{n=1}^N \hat{\beta}_0 + \hat{\beta}_1 x_n + \hat{\epsilon}_n$$

Error terms add up to zero

We get:

$$\bar{y} = \hat{\beta}_0 + \frac{\hat{\beta}_1 \cdot \sum_{n=1}^N x_n}{N}$$

$$\bar{y} = \hat{\beta}_0 + \hat{\beta}_1 \bar{x}$$

This shows that it passes through the mean

**Exercise 17.6.** The logistic sigmoid function is defined as  $\sigma(x) = e^x/(1+e^x)$ . What is the inverse function,  $\sigma^{-1}(x)$ ?

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

A function  $g$  is the inverse of function  $f$  if for  $y=f(x)$ ,  $x=g(y)$

$$y = e^x (1 + e^x)^{-1}$$

Replace x with y, and y with x

$$x = \frac{e^y}{1+e^y}$$

$$e^y = x + xe^y$$

$$e^y - xe^y = x$$

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

$$y = \ln\left(\frac{x}{(1-x)}\right)$$

## 2. Murphy Textbook Exercises

**Exercise 8.3** Gradient and Hessian of log-likelihood for logistic regression

a. Let  $\sigma(a) = \frac{1}{1+e^{-a}}$  be the sigmoid function. Show that

$$\frac{d\sigma(a)}{da} = \sigma(a)(1 - \sigma(a)) \quad (8.124)$$

b. Using the previous result and the chain rule of calculus, derive an expression for the gradient of the log likelihood (Equation 8.5).

### Answers

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

Apply chain rule:

$$= -(1 + e^{-a})^{-2} \cdot \frac{\partial}{\partial a}(1 + e^{-a})$$

$$= -(1 + e^{-a})^{-2} \cdot -e^{-a}$$

$$= \frac{e^{-a}}{(1+e^{-a})^2} = \frac{1}{1+e^{-a}} \frac{e^{-a}}{1+e^{-a}}$$

$$= \sigma(a)(1 - \sigma(a))$$

b.

$$g(w) = \frac{\partial}{\partial a} LL$$

$$= \sum_{i=1}^N \frac{\partial}{\partial w} [y_i \log \mu_i + (1 - y_i) \log(1 - \mu_i)]$$