

$$\mathbf{w} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}, \quad (6.6)$$

and $b = 0$.

We can now walk through the way that the model processes a batch of inputs. Let \mathbf{X} be the design matrix containing all four points in the binary input space, with one example per row:

$$\mathbf{X} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \\ 1 & 1 \end{bmatrix}. \quad (6.7)$$

The first step in the neural network is to multiply the input matrix by the first layer's weight matrix:

$$\mathbf{XW} = \begin{bmatrix} 0 & 0 \\ 1 & 1 \\ 1 & 1 \\ 2 & 2 \end{bmatrix}. \quad (6.8)$$

Next, we add the bias vector \mathbf{c} , to obtain

$$\begin{bmatrix} 0 & -1 \\ 1 & 0 \\ 1 & 0 \\ 2 & 1 \end{bmatrix}. \quad (6.9)$$

In this space, all of the examples lie along a line with slope 1. As we move along this line, the output needs to begin at 0, then rise to 1, then drop back down to 0. A linear model cannot implement such a function. To finish computing the value of \mathbf{h} for each example, we apply the rectified linear transformation:

$$\begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 1 & 0 \\ 2 & 1 \end{bmatrix}. \quad (6.10)$$

This transformation has changed the relationship between the examples. They no longer lie on a single line. As shown in figure 6.1, they now lie in a space where a linear model can solve the problem.

We finish by multiplying by the weight vector \mathbf{w} :

$$\begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \end{bmatrix}. \quad (6.11)$$