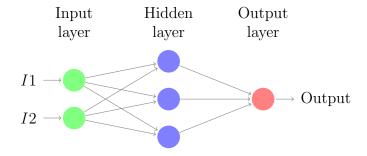
1 Feedforward: Building a ReLU neural network

1.1 Draw a netwrok



1.2 Write out the mathematical equation for the output of this network

$$\hat{y} = \sigma(h_1v_1 + h_2v_2 + h_3v_3 + c_1)$$

$$h_1 = ReLU(x_1w_{11} + x_2w_{21} + b_1)$$

$$h_2 = ReLU(x_1w_{12} + x_2w_{22} + b_2)$$

$$h_3 = ReLU(x_1w_{13} + x_2w_{23} + b_3)$$

2 Gradient Descent

2.1 What are the partial derivatives of f with respect to x and to y?

$$\frac{\partial f(x,y)}{\partial x} = -3x^2 - 200(y^2 - x)$$
$$\frac{\partial f(x,y)}{\partial y} = 400 * y * (y^2 - x)$$

3 Backprop

3.1 For the same network in Question 1, derive expressions of the gradient of the Loss function with respect to each of the model parameters

Suppose the Loss function is Binary Cross Entropy since it's a binary classification problem.

$$\frac{\partial L}{\partial V} = \frac{\partial L}{\partial \hat{y}} \cdot \frac{\partial \hat{y}}{\partial out} \cdot \frac{\partial out}{\partial V} = \sum_{i=1}^{n} \left(\frac{y}{\hat{y}} - \frac{1-y}{1-\hat{y}}\right) \cdot \sigma'(out) \cdot h_i$$

$$\frac{\partial L}{\partial c} = \frac{\partial L}{\partial \hat{y}} \cdot \frac{\partial \hat{y}}{\partial out} \cdot \frac{\partial out}{\partial c} = (\frac{y}{\hat{y}} - \frac{1-y}{1-\hat{y}}) \cdot \sigma'(out)$$

$$\frac{\partial L}{\partial W} = \frac{\partial L}{\partial \hat{y}} \cdot \frac{\partial \hat{y}}{\partial out} \cdot \frac{\partial out}{\partial H} \cdot \frac{\partial h_1}{\partial out'} \cdot \frac{\partial out'}{\partial W} = \sum_{i=1}^n \sum_{j=1}^m (\frac{y}{\hat{y}} - \frac{1-y}{1-\hat{y}}) \cdot \sigma'(out) \cdot v_i \cdot ReLU'(out'_i) \cdot x_j$$

$$\frac{\partial L}{\partial B} = \frac{\partial L}{\partial \hat{y}} \cdot \frac{\partial \hat{y}}{\partial out} \cdot \frac{\partial out}{\partial H} \cdot \frac{\partial h_1}{\partial out'} \cdot \frac{\partial out'}{\partial B} = \sum_{i=1}^n (\frac{y}{\hat{y}} - \frac{1-y}{1-\hat{y}}) \cdot \sigma'(out) \cdot v_i \cdot ReLU'(out'_i)$$