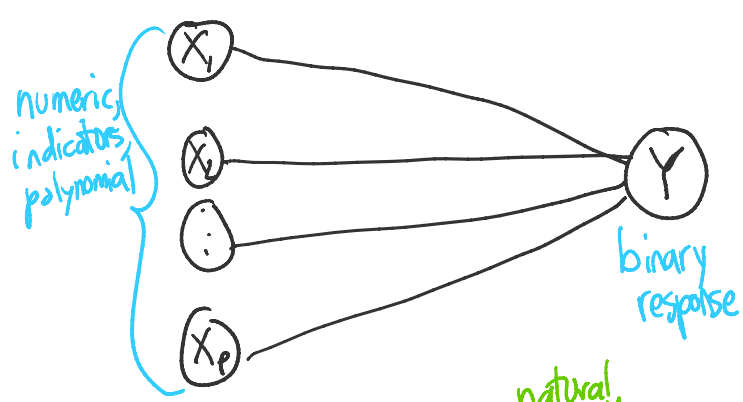


Machine Learning for binary responses

- The choice of activation function depends on the range of possible values of the predicted response



$$\hat{Y} = f^{-1}(\hat{\beta}_0 + \hat{\beta}_1 X_1 + \hat{\beta}_2 X_2 + \dots + \hat{\beta}_p X_p)$$

What is our activation function?

$$\hat{Y} = P(Y=1) \quad 0 \leq \hat{Y} \leq 1$$

We need an activation f^{-1} (f^{-1}) that can take a real number and give us a number between 0 & 1.

$$f(\hat{Y}) = \text{logit}(\hat{Y}) = \log\left(\frac{\hat{Y}}{1-\hat{Y}}\right)$$

↑ natural logarithm

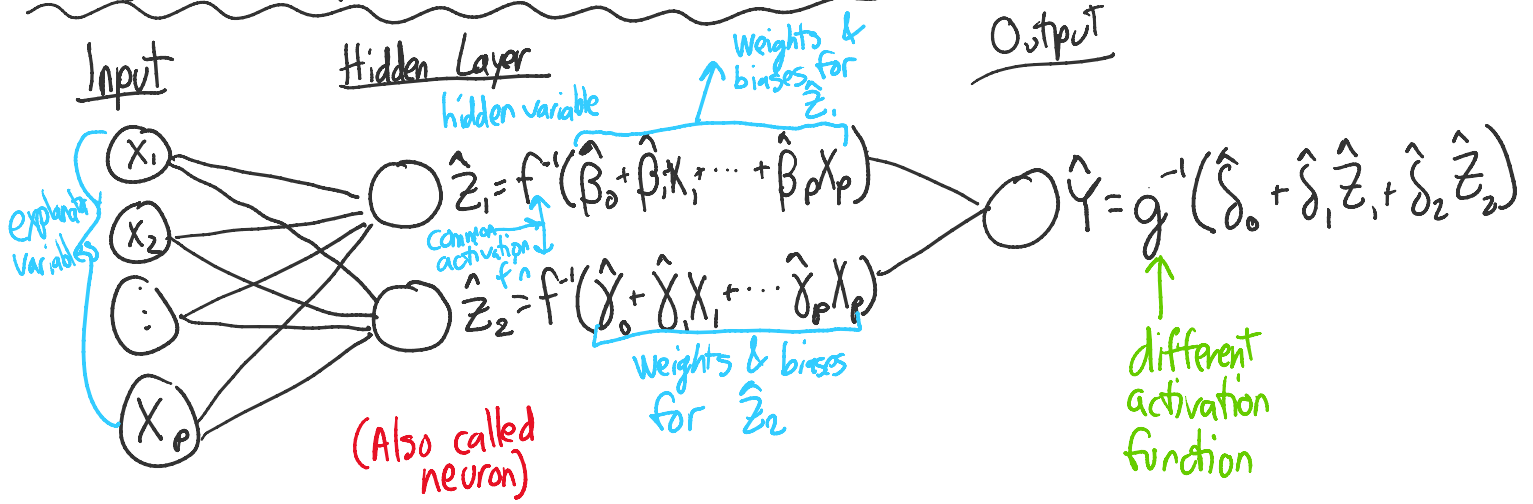
log odds ratio

$$f^{-1}(x) = \frac{\exp(x)}{1 + \exp(x)}$$

inverse limit function

$$\Rightarrow \hat{Y} = \frac{\exp\{\hat{\beta}_0 + \hat{\beta}_1 X_1 + \dots + \hat{\beta}_p X_p\}}{1 + \exp\{\hat{\beta}_0 + \hat{\beta}_1 X_1 + \dots + \hat{\beta}_p X_p\}}$$

$$f^{-1}(x) = \frac{\exp(x)}{1 + \exp(x)} \quad \left. \begin{array}{l} \text{log odds ratio} \\ \text{inverse logit function} \end{array} \right\} \Rightarrow \hat{Y} = \frac{\exp\{\hat{\beta}_0 + \hat{\beta}_1 X_1 + \dots + \hat{\beta}_p X_p\}}{1 + \exp\{\hat{\beta}_0 + \hat{\beta}_1 X_1 + \dots + \hat{\beta}_p X_p\}}$$



This type of algorithm is called a neural network

↳ This algorithm tries to mimic the functionality of the human brain!

Notes about neural networks

- You will NOT be tested on complex neural networks
- You can have as many hidden layers & as many neurons in each layer as you want (but don't want too many)
- Logistic Regression and Linear regression are special types of neural networks with no hidden layers

Major Drawback of Neural Networks

- You lose interpretability of model parameters for perhaps only moderately improved predictions