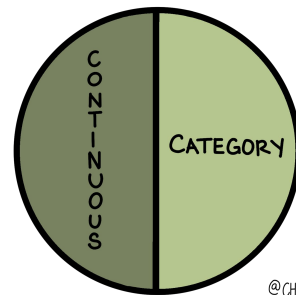


PREDICT

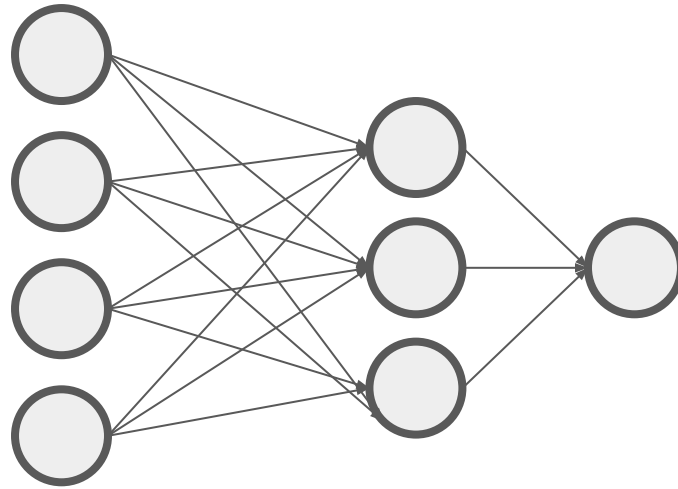


@CHELSEA PARLETT

Intro to Neural Networks

Dr. Chelsea Parlett-Pelleriti

Neural Networks



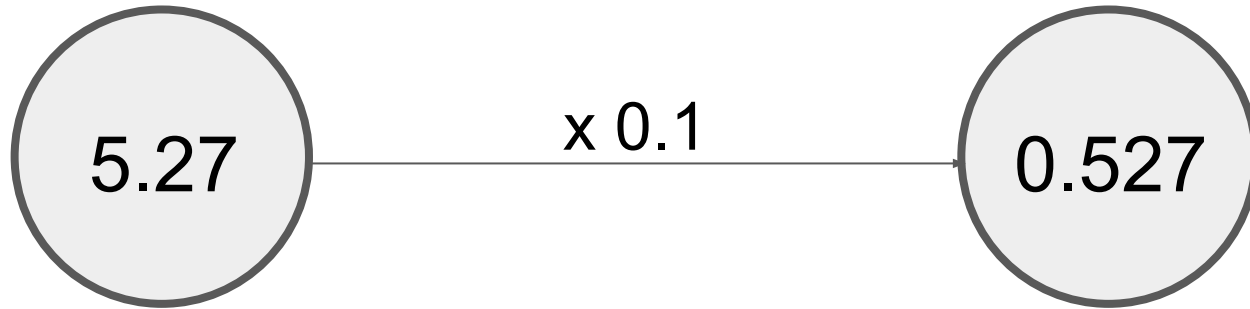
Nodes

Nodes Hold Values



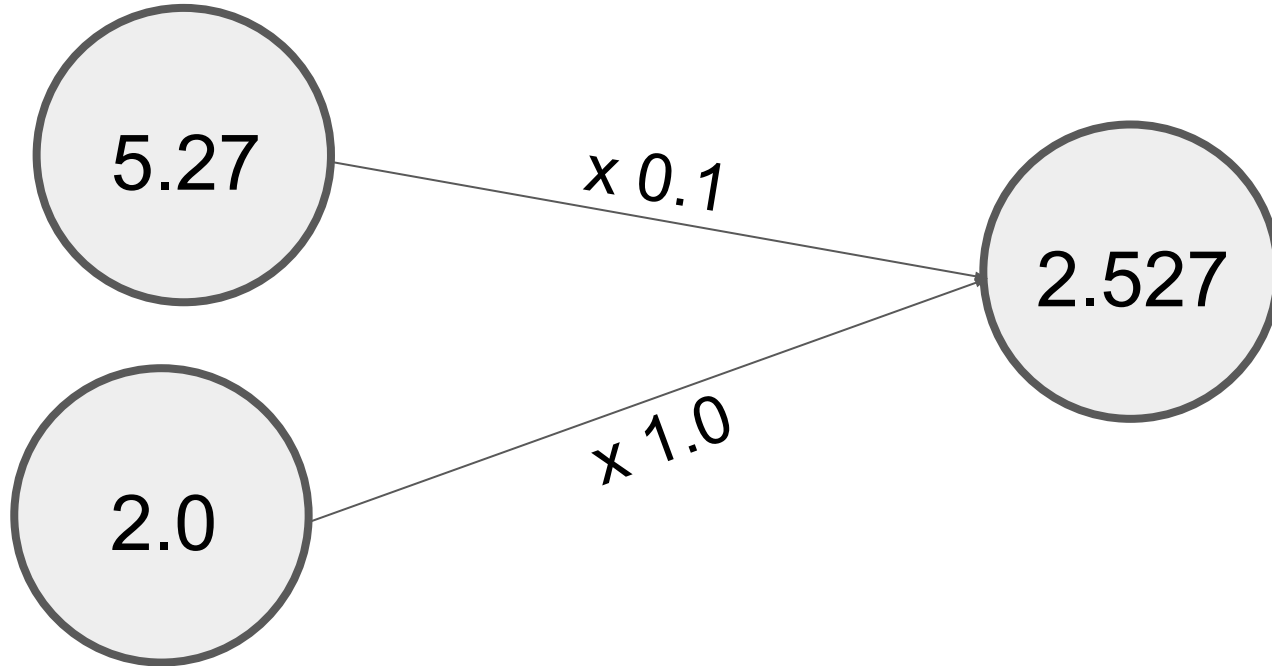
Weights

Weights multiply the number in a previous node and add it to the next node



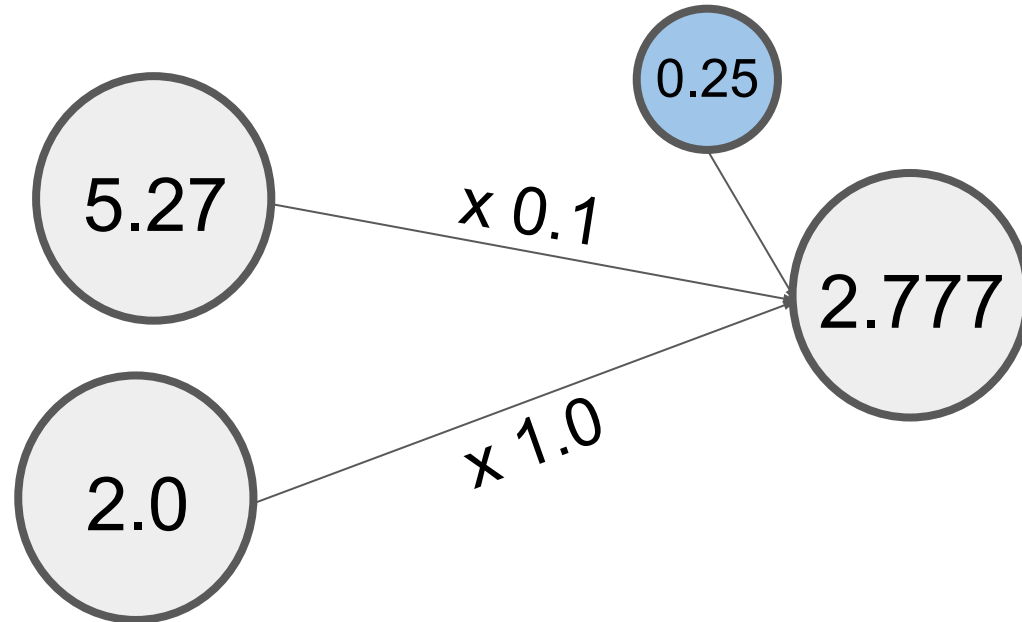
Weights

We can have multiple weights feeding into one node



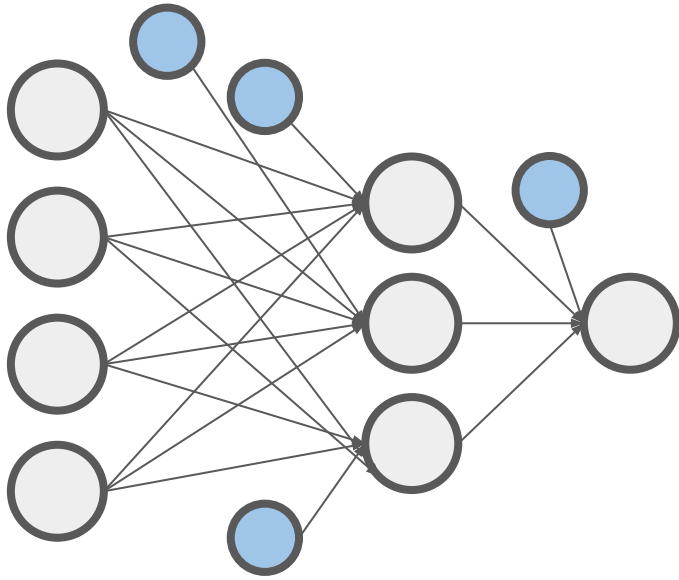
Biases

Biases move the value of a node up (for positive values) or down (for negative values) no matter what the weights and previous nodes' values were

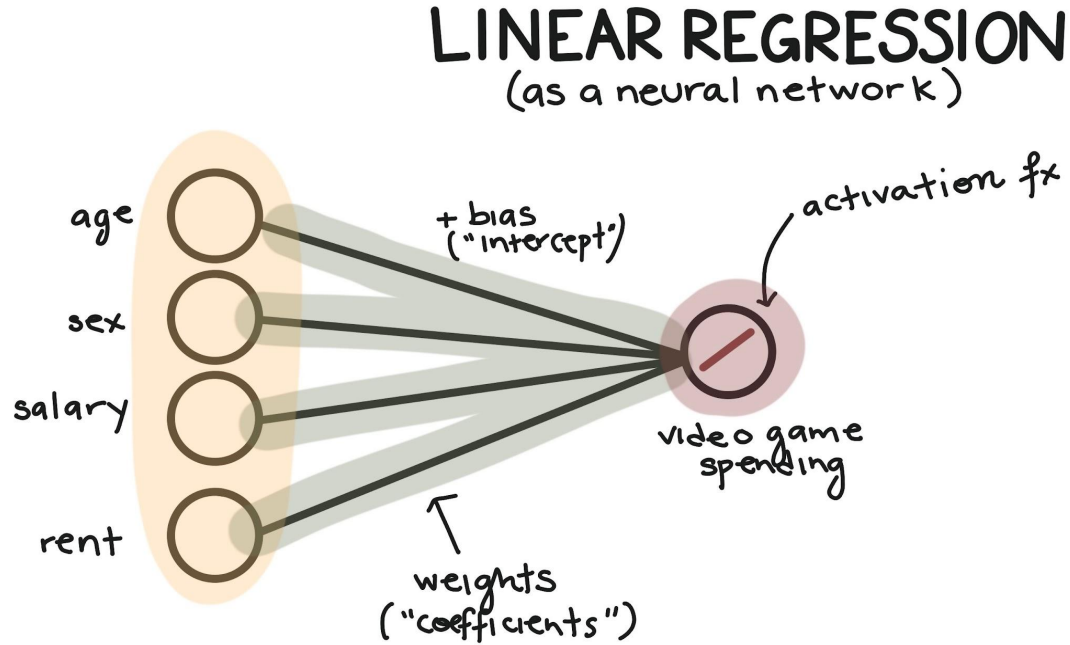


Biases

Together, nodes, weights, and biases make up the core structure of a neural network

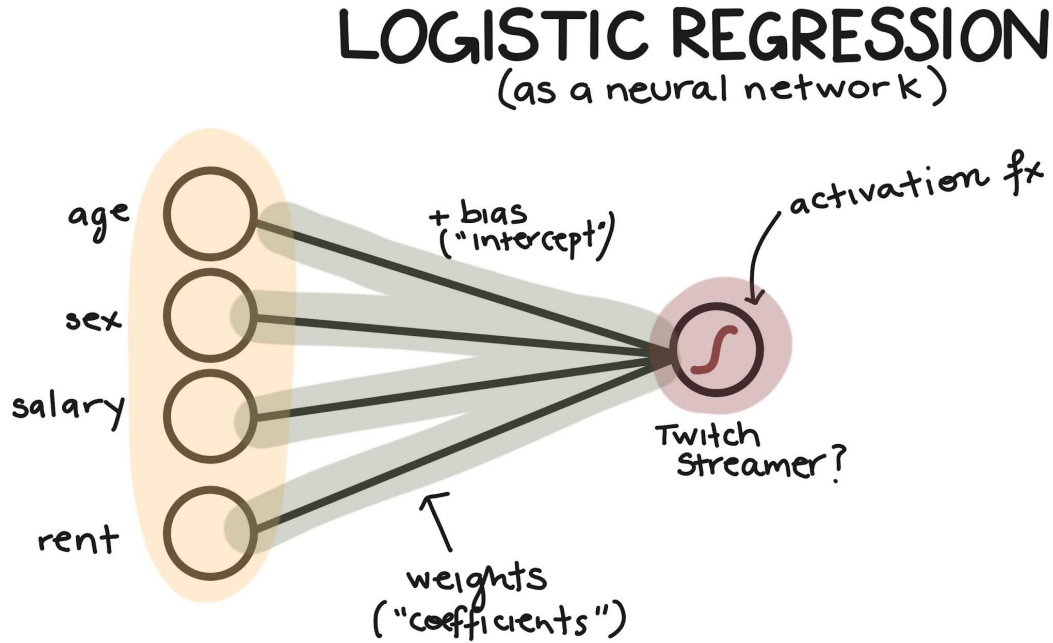


Linear Regression as a NN



$$\text{LOSS: } \sum (x_i - \hat{x})^2$$

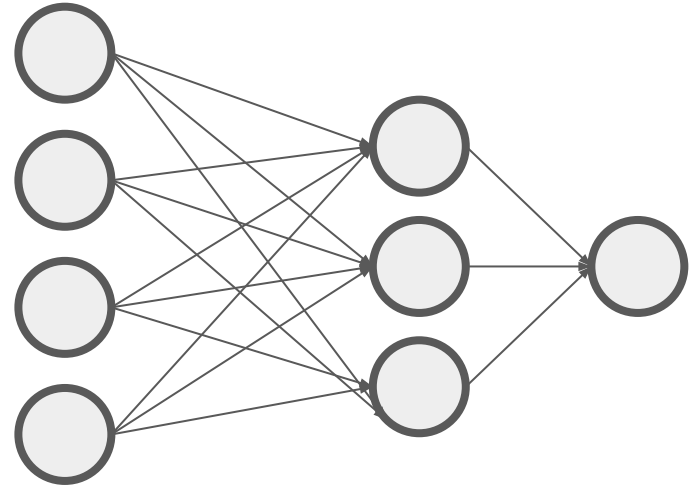
Logistic Regression as a NN



$$\text{Loss: } \sum -y_i \log(\hat{p}_i) - (1 - y_i) \log(1 - \hat{p}_i)$$

Building a FF NN Structure

1. Structure
2. Connections
3. Activations



Common Loss Functions (continuous)

MSE

$$\frac{1}{N} \sum_{i=1}^N (\text{actual} - \text{predicted})^2$$

MAE

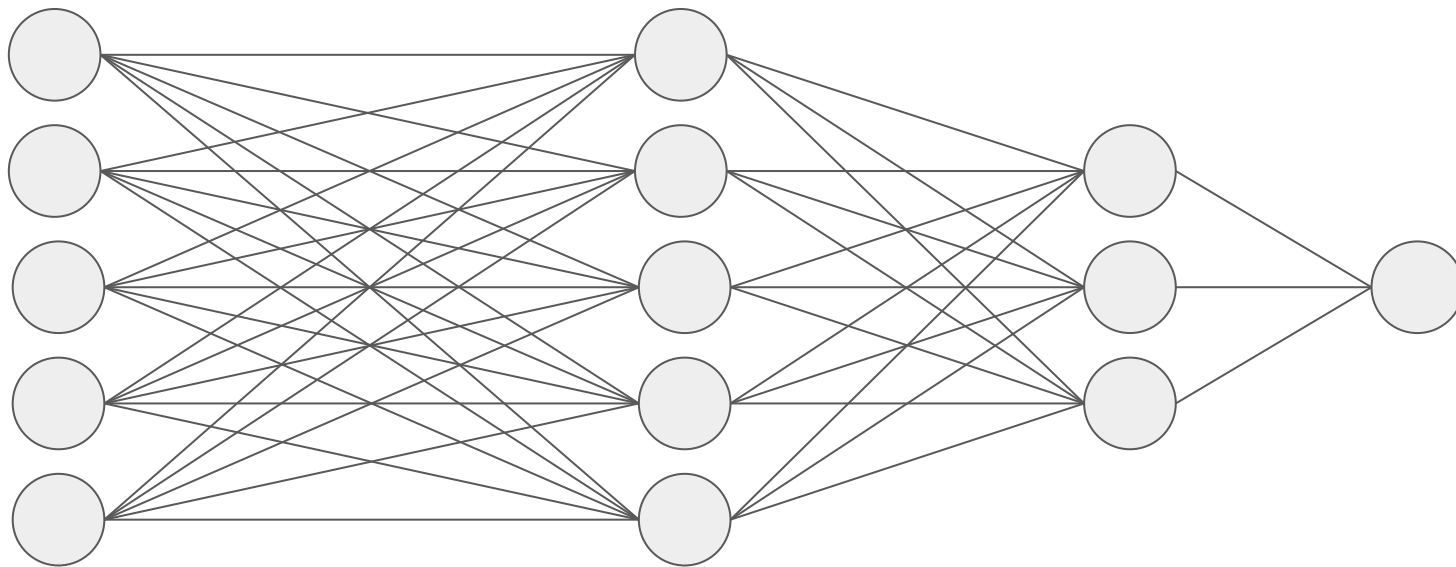
$$\frac{1}{N} \sum_{i=1}^N |\text{actual} - \text{predicted}|$$

Common Loss Functions (categorical)

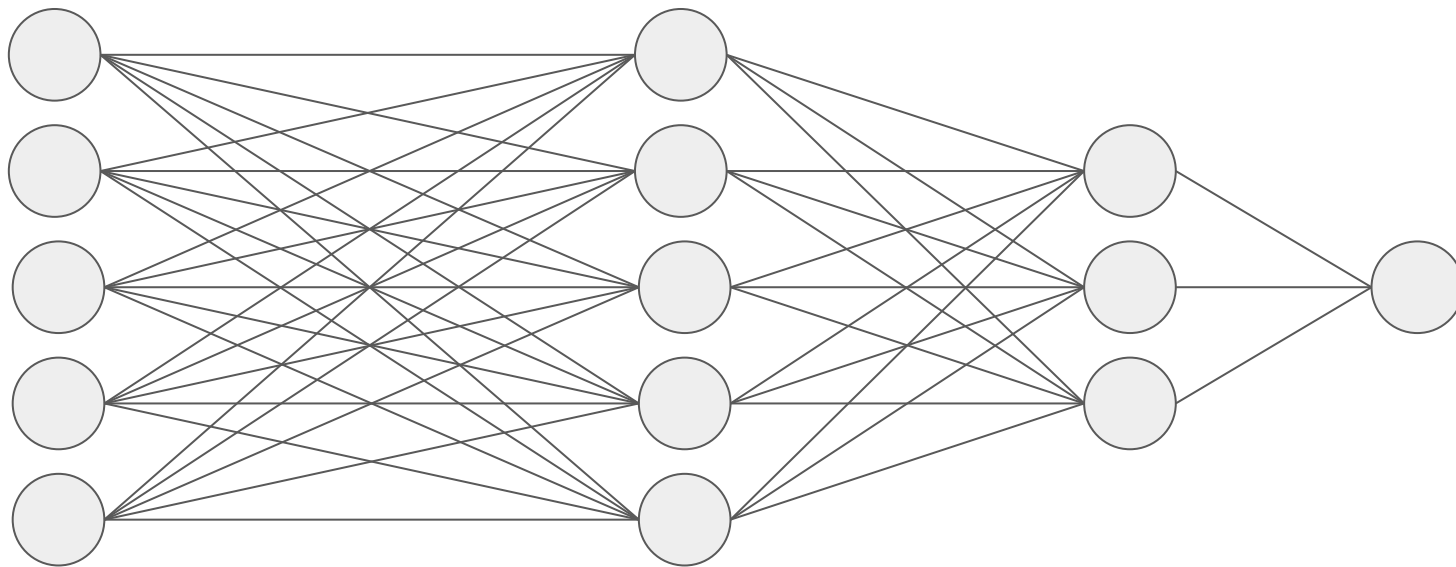
Log Loss/ Binary Cross Entropy

$$-\frac{1}{N} \sum_{i=1}^N y_i \cdot \log(p_i) + (1 - y_i) \cdot \log(1 - p_i)$$

Universal Function Approximation

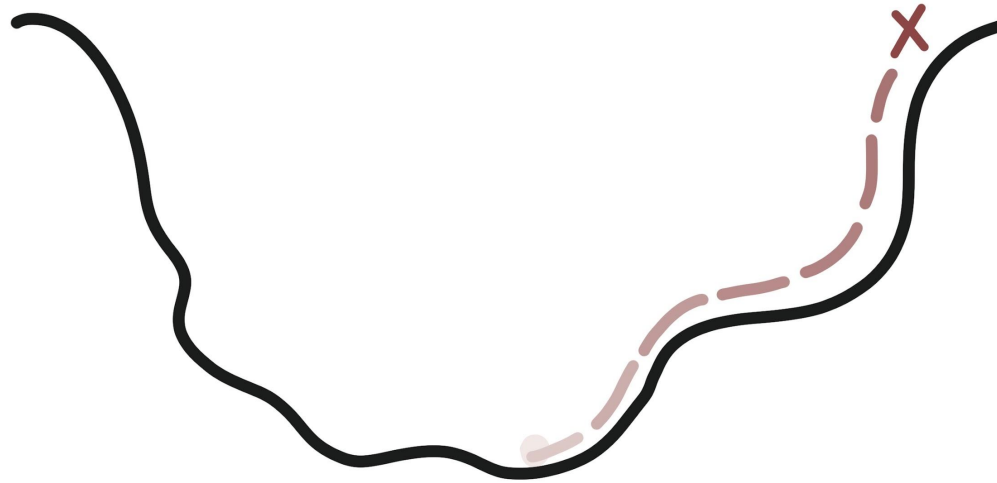


Feature Engineering



Backpropagation/Gradient Descent

1. Which direction goes down the most?
2. Take a step in that direction.
3. Repeat until you get somewhere flat.



@CHELSEAPARLETT

LOSS FUNCTION