

ANN Project (Assignment 1)

Q1.

- Forward Propagation: - In this process training data is passed through (in the forward direction) through the hidden layers and the respective activation functions with the aim of training the neural network.
- Backward Propagation: - Given an ANN and an error function Backpropagation is the calculation of gradient proceeds backwards through the network update the model weights based on feedback of the previous runs.
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Q2. Let $W[4][5]$ be the Weight matrix and $B[1][5]$ be the Bias. $X[1][4]$ are the inputs and $A[1][5]$ are the Hidden Layers

Forward Propagation

Then $A[1][5] = X_i[1][j] * W[j][5] + B[1][5]$ which is matrix multiplication via looping where i (inputs) varies from 0 to 3, j (weights) varies from 0 to 4.

Output = $\sum \text{Activation}(A[i][j])$ (Final Step)

Backward Propagation

Calculating derivative by chain rule by using error/loss/cost function at the output stage and moving backwards through the layer using appropriate weights.

Q3.

- Sigmoid: - $f(x) = 1/(1+\exp(-x))$, $f'(x) = \exp(-x)/(1+\exp(-x))^2$
- Relu: - $f(x) = \max(0, x)$, $f'(x) = \{0, \text{ if } x < 0 \text{ (undefined at } x = 0,)$
 $1, \text{ if } x > 0\}$
- Leaky Relu: - $f(x) = 0.01x$ for $x < 0$ and x for $x \geq 0$. $f'(x) = 0.01$ for $x < 0$ and 1 for $x \geq 0$
- Tanh: - $f(x) = e^x - e^{-x} / e^x + e^{-x}$ and $f'(x) = 4 / (e^x + e^{-x})^2$
- Softmax: - $f(x)(i) = e^{x_i} / \sum_k e^{x_k}$ for all $k = 1$ to N and $f'(x) = f(x)(i) * [\delta_{ij} - f(x)(j)]$

Where δ represents the Kronecker delta function