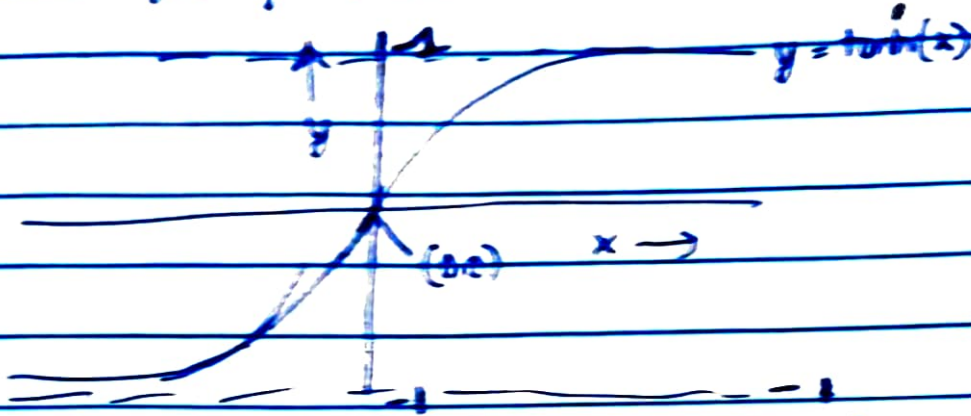
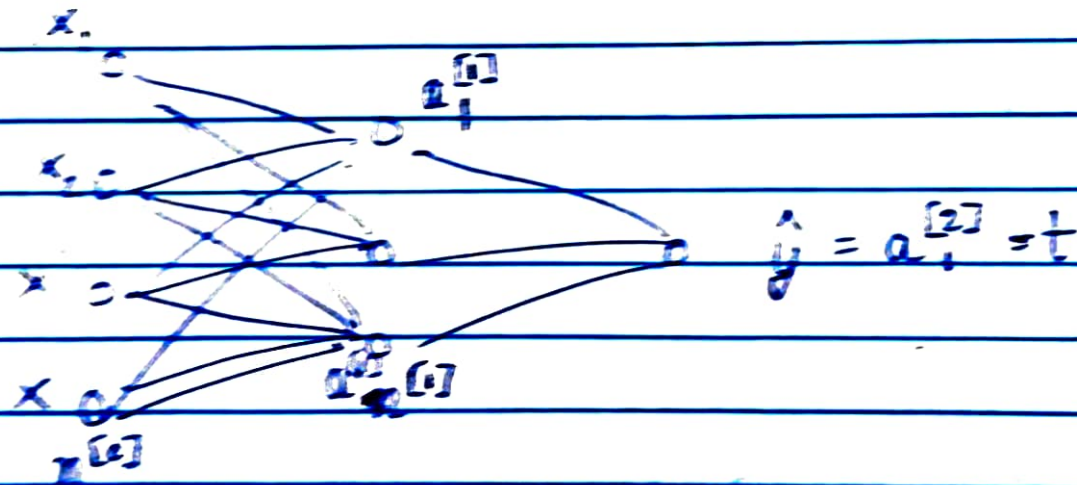


2. This is true, as the range of \tanh is $[-1, 1]$, which is centered around 0, which is the mid point of the range of \tanh .



On the other hand, signed squares values to a range $[0, 1]$ which is centered at 0.5 for the inputs on average, thereby having a better ~~gradient~~ gradient on average.



$$L(t, y) = -(y \log(t) + (1-y) \log(1-t))$$

we have

forward propagation:

$$z^{[1]}(i) = W^{[1]} x^{(i)} + b^{[1]}$$

$$a^{[1]}(i) = \text{ReLU}(z^{[1]}(i)) = \max(z^{[1]}(i), 0)$$

Now, derivative of ReLU =
$$\begin{cases} 1, & x \geq 0 \\ 0, & x < 0 \end{cases} \quad (\text{assumed at } x=0)$$

$$z = \text{ReLU}(x)$$

$$z^{[2]}(i) = W^{[2]} a^{[1]}(i) + b^{[2]} = \text{sigmoid}(x)$$

$$t = a^{[2]} = \sigma(z^{[2]}(i)) \quad (\text{Sigmoid})$$

Vectorizing

$$Z^{[1]} = W^{[1]} X + b^{[1]}$$

$$A^{[1]} = \text{ReLU}(Z^{[1]})$$

$$Z^{[2]} = W^{[2]} X + b^{[2]}$$

$$T = A^{[2]} = \sigma(Z^{[2]})$$

Now, ~~for~~ for update, gradient descent -:

$$W^{[1]} = W^{[1]} - \alpha \frac{\partial L}{\partial W^{[1]}}$$

$$W^{[2]} = W^{[2]} - \alpha \frac{\partial L}{\partial W^{[2]}}$$

$$b^{[1]} = b^{[1]} - \alpha \frac{\partial L}{\partial b^{[1]}}, \quad b^{[2]} = b^{[2]} - \alpha \frac{\partial L}{\partial b^{[2]}}$$

$$\text{Now, } \frac{\partial \mathcal{L}}{\partial W^{[2]}} = \frac{\partial \mathcal{L}}{\partial A^{[2]}} \frac{\partial A^{[2]}}{\partial Z^{[2]}} \frac{\partial Z^{[2]}}{\partial W^{[2]}} \quad (\text{Chain Rule})$$

$$= \frac{\partial \mathcal{L}}{\partial A^{[2]}} = \frac{-Y}{T} + \frac{1-Y}{1-T} = \frac{T-Y}{T(1-T)}$$

$$\frac{\partial A^{[2]}}{\partial Z^{[2]}} = A^{[2]}(1-A^{[2]}) \quad (\because \sigma'(z) = \sigma(z)(1-\sigma(z)))$$

$$\frac{\partial Z^{[2]}}{\partial W^{[2]}} = T(1-T)$$

$$\frac{\partial Z^{[2]}}{\partial W^{[2]}} = \cancel{W} A^{[0]T}$$

$$\Rightarrow \frac{\partial \mathcal{L}}{\partial W^{[2]}} = (T-Y) A^{[0]T}$$



$$\text{Similarly, } \frac{\partial \mathcal{L}}{\partial b^{[2]}} = (T-Y) \quad \text{Vectorized} = \frac{1}{m} \text{np.sum}(dZ, \text{axis}=1, \text{keepdims=True})$$

$$= \frac{\partial \mathcal{L}}{\partial W^{[1]}} = \frac{\partial \mathcal{L}}{\partial Z^{[2]}} \frac{\partial Z^{[2]}}{\partial A^{[1]}} \frac{\partial A^{[1]}}{\partial W^{[1]}} = (T-Y) \text{signum}(A^{[1]}) A^{[0]T}$$