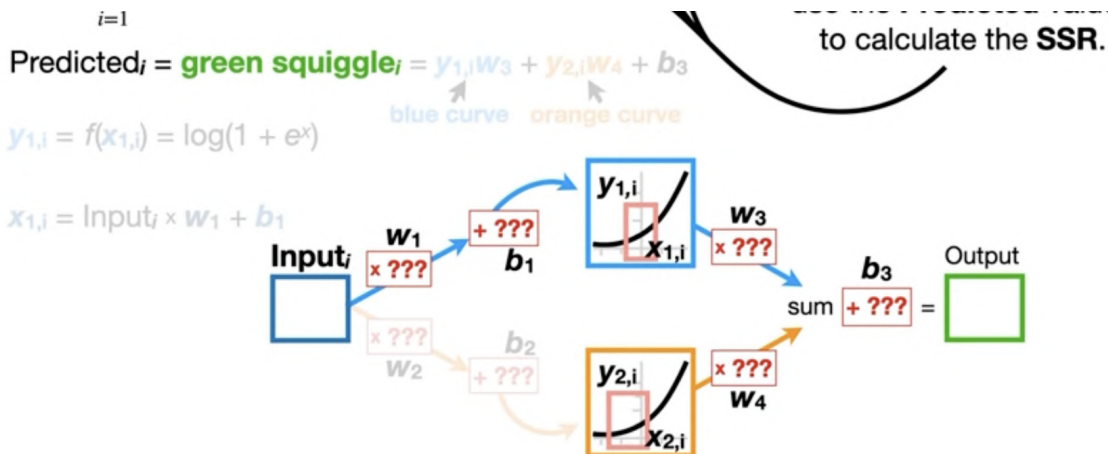


## Backpropagation Details Pt. 2: Going bonkers with The Chain Rule



layer ~~xxx~~

$y(1)$

$A(x_1, 0)$   
 $B(x_2, 1)$   
 $C(x_3, 0)$

input  $w_1, b_1$   $w_2, b_2$   $w_3, b_3$   $w_4, b_4$

Act fun.  $w_3$   $w_4$

sum +  $b_3$

output

1. assume:  
 Value -  $w_1, b_1, w_2, b_2$  are already optimized

2. give  $w_3, w_4, b_3$  random values

3. calculate the sum of the squared residuals. SSR

$$SSR = (0 - \text{Predicted } x_1)^2 + (1 - \text{Predicted } x_2)^2 + (0 - \text{Predicted } x_3)^2$$

4. optimize  $\frac{dSSR}{db_3} = \frac{dSSR}{d\text{Predicted}} \times \frac{d\text{Predicted}}{db_3}$

fancy notation part:

$x_{1,i}$  Act  $\rightarrow y_{1,i}$   
 $x_{2,i}$  funct  $\rightarrow y_{2,i}$

5. Predicted = green squiggle<sub>i</sub> = blue + orange +  $b_3$

②  $= y_{1,i}w_3 + y_{2,i}w_4 + b_3$

6. since:  $SSR = \sum_{i=1}^{n=3} (\text{observed}_i - \text{Predicted}_i)^2$

$$\frac{dSSR}{dw_3} = \frac{dSSR}{d\text{Predicted}} \times \frac{d\text{Predicted}}{dw_3}$$

chain rule time!

$$\frac{dSSR}{dw_4} = \frac{dSSR}{d\text{Predicted}} \times \frac{d\text{Predicted}}{dw_4}$$

$$\frac{dSSR}{db_3} = \frac{dSSR}{d\text{predicted}} \times \frac{d\text{Predicted}}{db_3}$$

7. derivative of ①:

$$\begin{aligned} \frac{dSSR}{d\text{predicted}} &= \sum_{i=1}^{n=3} 2 \times (\text{observed}_i - \text{Predicted}_i) \times (-1) \\ &= \sum_{i=1}^{n=3} -2 \times (\text{observed}_i - \text{Predicted}_i) \end{aligned}$$

8. derivative of

$$\begin{aligned} \frac{d\text{Predicted}}{dw_3} &= (y_{1,i}w_3 + y_{2,i}w_4 + b_3)^{-1} \\ &= y_{1,i} \end{aligned}$$



$$\frac{d \text{ Predicted}}{dw_4} = y_{2,i}$$

$$\frac{d \text{ Predicted}}{b_3} = 1$$

9. chain rule time:

$$\frac{dSSR}{dw_3} = \sum_{i=1}^{n=3} -2 (\text{observed}_i - \text{Predicted}_i) \times y_{1,i}$$

$$\frac{dSSR}{dw_2} = \sum_{i=1}^{n=3} -2 (\text{observed}_i - \text{Predicted}_i) \times y_{2,i}$$

$$\frac{dSSR}{db_3} = \sum_{i=1}^{n=3} -2 (\text{observed}_i - \text{Predicted}_i) \times 1$$

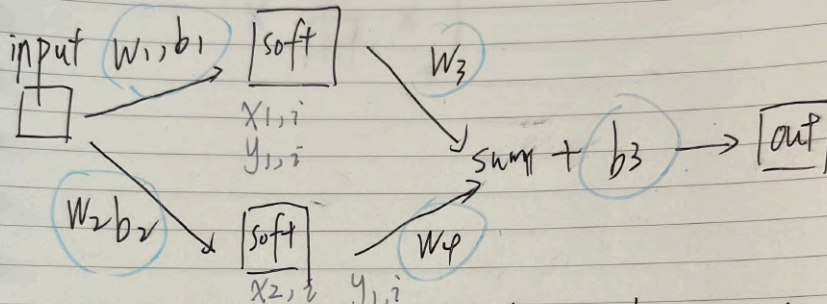
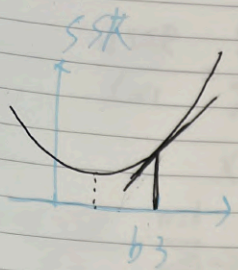
10. Stepsize  $\equiv$

$$SS = \frac{dSSR}{dw_3} \times \text{learning rate}$$

$$\text{New } w_3 = 0.36 - \text{Stepsize}$$

11. repeat the process until the Predictions no longer improve very much, or reach a maximum number of steps





1. Now Let's optimize  $w_1, b_1, w_2, b_2, w_3, w_4, b_3$

$$2. \frac{dSSR}{dw_1} = \frac{dSSR}{d\text{Predicted}} \times \frac{d\text{Predicted}}{dy_1} \times \frac{dy_{1,i}}{dx_{1,i}} \times \frac{dx_{1,i}}{dw_1}$$

$$\textcircled{1} SSR = \sum_{i=1}^{n=3} (\text{observed}_i - \text{Predicted}_i)^2 \Rightarrow \sum_{i=1}^{n=3} -2(\text{observed}_i - \text{Predicted}_i)$$

$$\textcircled{2} \text{Predicted}_i = y_{1,i} w_3 + y_{2,i} w_4 + b_3 \Rightarrow w_3$$

$$\textcircled{3} y_{1,i} = f(x_{1,i}) = \log(1 + e^x) \Rightarrow \frac{e^x}{1 + e^x}$$

$$\textcircled{4} x_{1,i} = \text{input}_i \times w_1 + b_1 \Rightarrow \text{input}_i$$

$$3. \frac{dSSR}{dw_1} = \sum_{i=1}^{n=3} -2(\text{observed}_i - \text{Predicted}_i) \times w_3 \times \frac{e^x}{1 + e^x} \times \text{Input}_i$$

$$\frac{dSSR}{db_1} = \sum_{i=1}^{n=3} -2(\text{observed}_i - \text{Predicted}_i) \times w_3 \times \frac{e^x}{1 + e^x} \times 1$$