

Weight Updates with ReLU Hidden, sigmoid output and binary cross-entropy

- Training Input: $(x_1, x_2) = (1, 1)$
- Target Output: $d = 0.0$

$$z_1^1 = 0.25 * 1 + 0.75 * 1 = 1.0$$

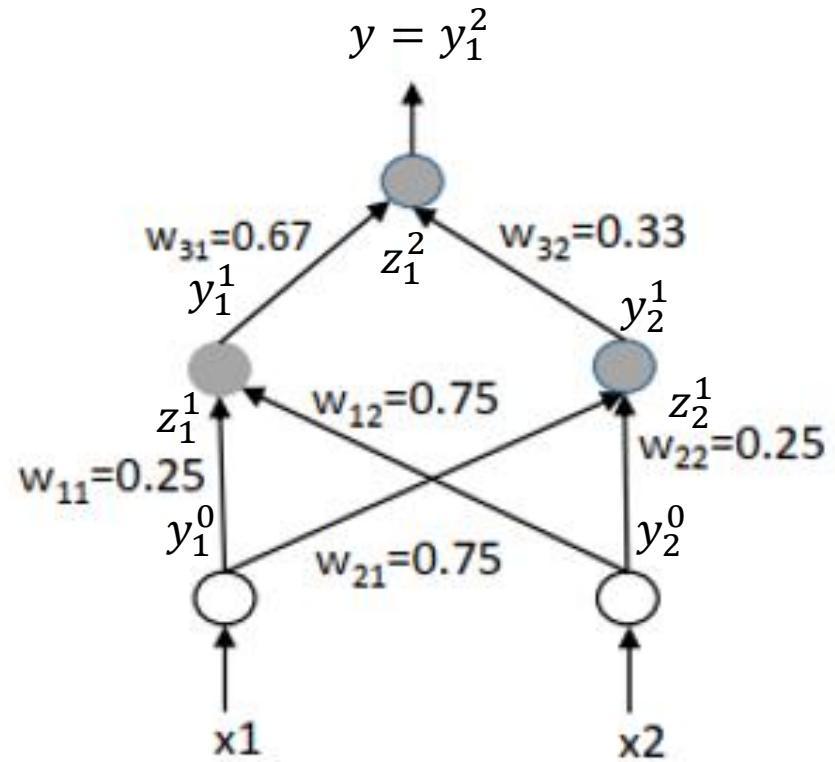
$$z_2^1 = 0.75 * 1 + 0.25 * 1 = 1.0$$

$$y_1^1 = \text{ReLU}(z_1^1) = 1 \quad y_2^1 = \text{ReLU}(z_2^1) = 1$$

$$z_1^2 = 0.67 * 1 + 0.33 * 1 = 1.0$$

$$y_1^2 = \text{sigmoid}(z_1^2) = e / (1 + e)$$

$$\text{div} = -d * \log(y) - (1-d) * \log(1-y)$$



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$$z_1^1 = 0.25*1 + 0.75*1 = 1.0$$

$$z_2^1 = 0.75*1 + 0.25*1 = 1.0$$

$$y_1^1 = \text{ReLU}(z_1^1) = 1 \quad y_2^1 = \text{ReLU}(z_2^1) = 1$$

$$z_1^2 = 0.67*1 + 0.33*1 = 1.0$$

$$y_1^2 = \text{sigmoid}(z_1^2) = e/(1+e)$$

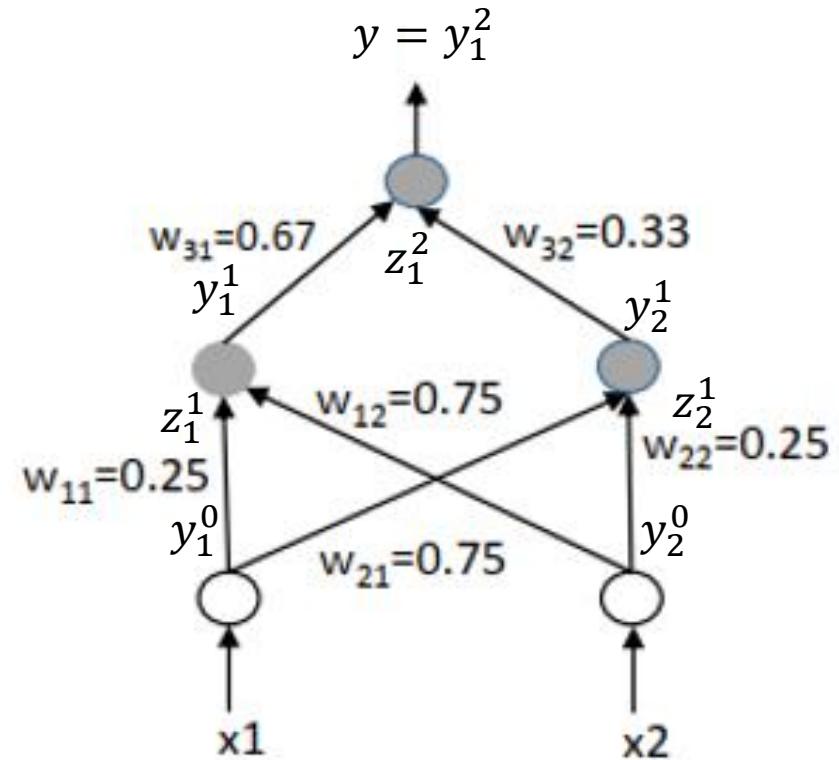
$$\text{div} = -d * \log(y) - (1-d) * \log(1-y)$$

$$\frac{\delta \text{div}}{\delta y} = -d/y + (1-d)/(1-y) = 1/(1-y) = (1+e)$$

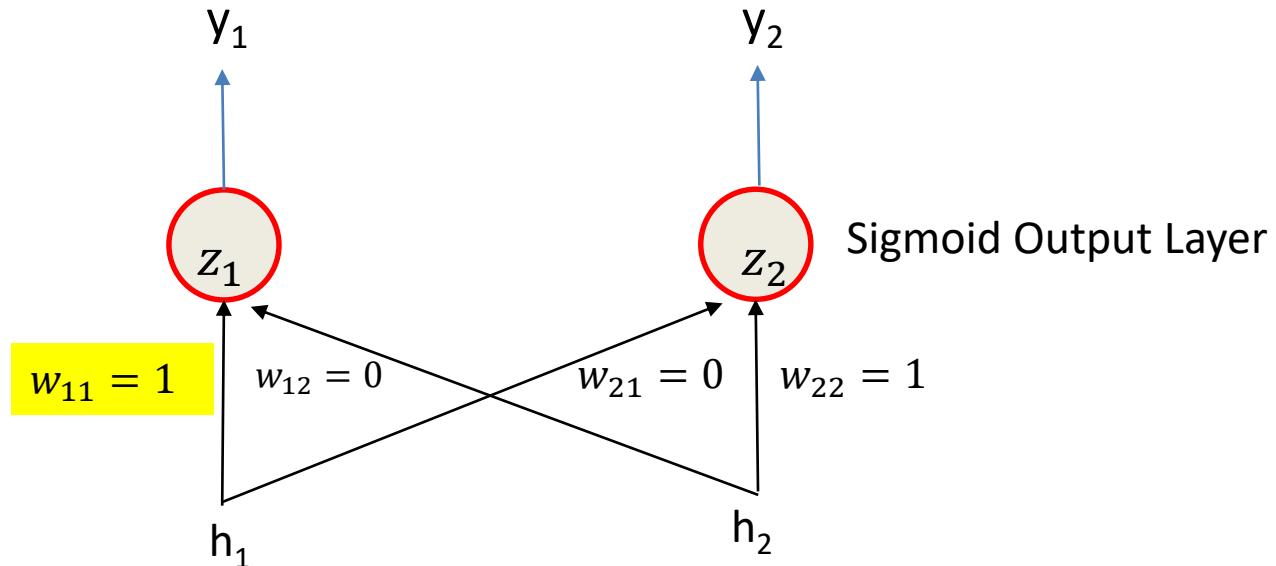
$$\frac{\delta \text{div}}{\delta z_1^2} = \frac{\delta \text{Div}}{\delta y_1^2} * \frac{\delta y_1^2}{\delta z_1^2} = (1+e) * \text{sigmoid}(z_1^2) * (1 - \text{sigmoid}(z_1^2)) = e/(1+e)$$

$$\frac{\delta \text{div}}{\delta w_{31}} = \frac{\delta \text{Div}}{\delta z_1^2} * \frac{\delta z_1^2}{\delta w_{31}} = e/(1+e) * y_1^1 = e/(1+e)$$

$$w_{31} = w_{31} - \eta * \frac{\delta \text{div}}{\delta w_{31}} = 0.67 - 0.1 * e/(1+e) = 0.597$$



Weight Updates with Sigmoid Output



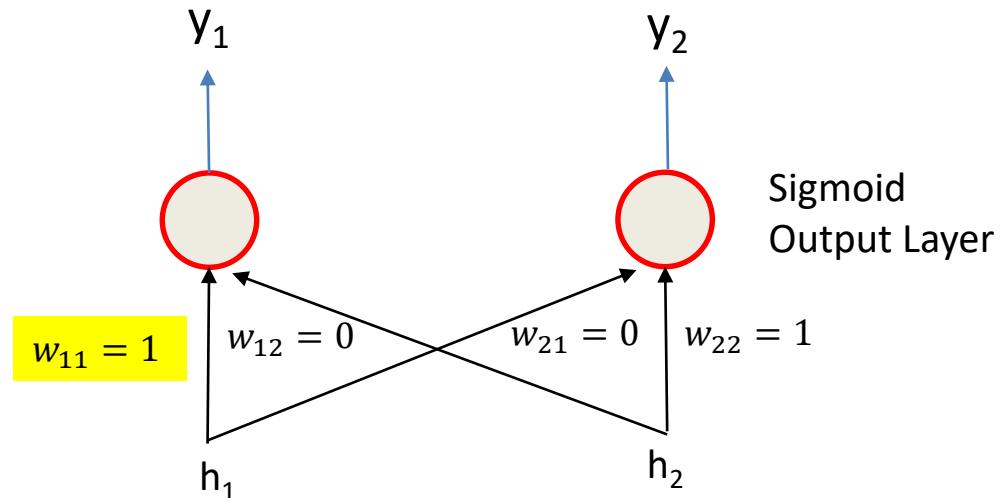
- Training input: $(h_1, h_2) = (1, 0)$ target output: $(d_1, d_2) = (0, 1)$
 - Activation: sigmoid, Bias=0
 - *div* function: cross-entropy
 - Learning rate: 0.1
 - What will be the value of w_{11} in next iteration?
-
- Input to output node $z_1 = w_{11}=1; z_2 = 0$
 - $y_1 = 1 / (1+e^{-1}) = e/(1+e)$ $y_2 = 1 / (1+e^0) = 1/2$
 - $div = -d_1 \log(y_1) - d_2 \log(y_2) = -\log(y_2) = \log 2$

Weight Updates with Sigmoid Outputs ...

$w_{11}(t+1) = ?$

Input $(h_1, h_2) = (1.0, 0.0)$

target output: $(d_1, d_2) = (0, 1)$



Weight Updates with Sigmoid Outputs ...

Input $(h_1, h_2) = (1.0, 0.0)$

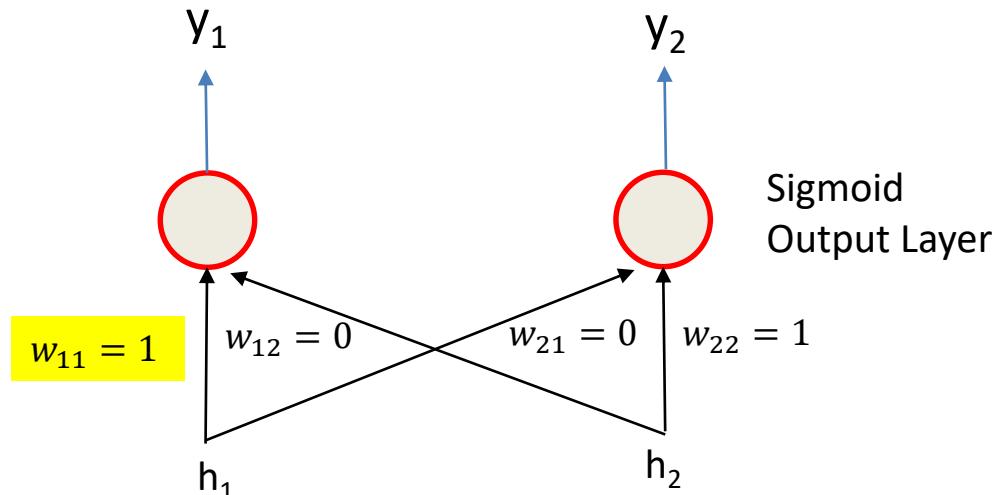
target output: $(d_1, d_2) = (0, 1)$

$w_{11}(t+1) = ?$

$$div = -d_1 \log(y_1) - d_2 \log(y_2)$$

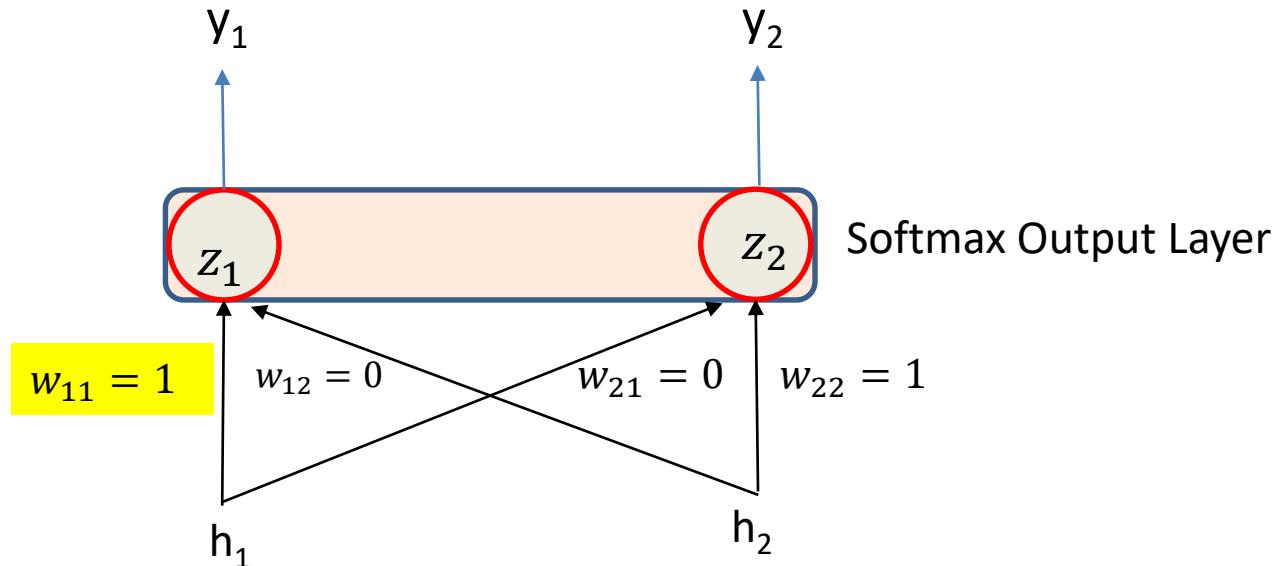
$$\frac{\delta div}{\delta y_1} = -\frac{d_1}{y_1}$$

$$\frac{\delta y_1}{\delta z_1} = \frac{1}{1 + e^{-1}} * \left(1 - \frac{1}{1 + e^{-1}} \right) = \frac{e}{(1 + e)^2}$$



$$\begin{aligned} w_{11}(t+1) &= w_{11}(t) - \eta \frac{\delta div}{\delta w_{11}} \\ &= w_{11}(t) - 0.1 * \frac{\delta div}{\delta y_1} \frac{\delta y_1}{\delta z_1} \frac{\delta z_1}{\delta w_{11}} \\ &= w_{11}(t) - 0.1 * 0 * \frac{e}{(1+e)^2} * h_1 \\ &= 1.0 \end{aligned}$$

Weight Updates with Softmax



- Training input: $(h_1, h_2) = (1, 0)$ target output: $(d_1, d_2) = (0, 1)$
- *div* function: cross-entropy
- Learning rate: 0.1
- Bias = 0
- What will be the value of w_{11} in next iteration?
- Input to softmax node $z_1 = w_{11}=1; z_2 = 0$
- $y_1 = e / (1+e)$
- $y_2 = 1/ (1+e)$
- $div = -d_1 \log(y_1) - d_2 \log(y_2) = -\log(y_2) = \log(1+e) = 1.313$

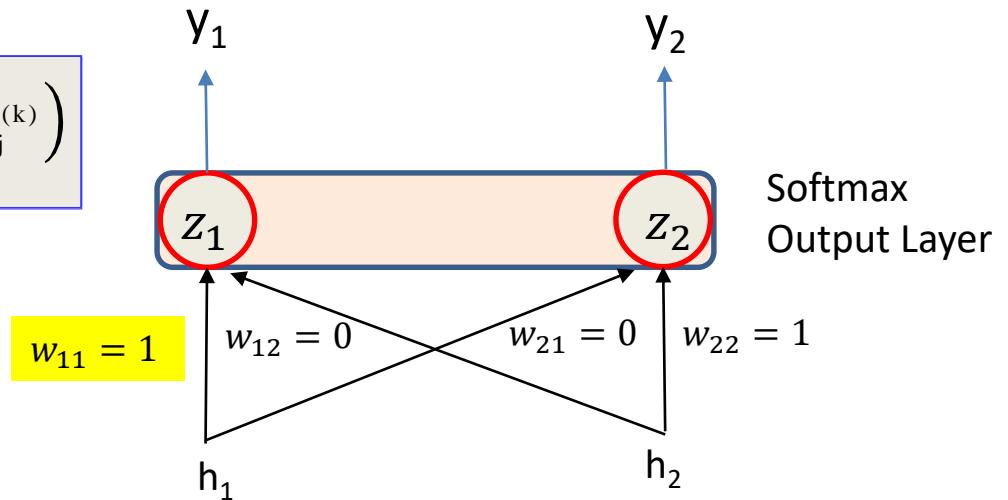
Weight Updates with Softmax ...

$$\frac{\partial Div}{\partial z_i^{(k)}} = \sum_j \frac{\partial Div}{\partial y_j^{(k)}} y_i^{(k)} (\delta_{ij} - y_j^{(k)})$$

$$w_{11}(t+1) = ?$$

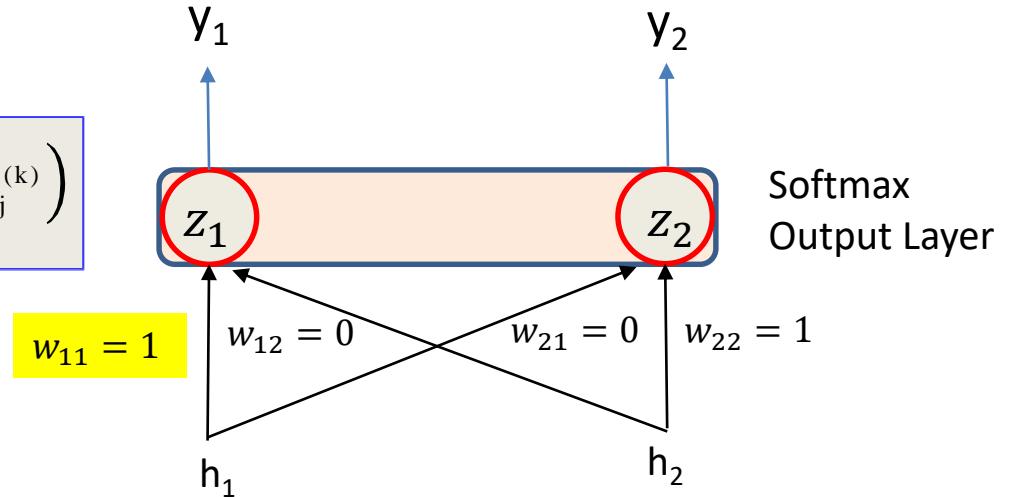
Training input: $(h_1, h_2) = (1, 0)$

Target output: $(d_1, d_2) = (0, 1)$



Weight Updates with Softmax ...

$$\frac{\partial Div}{\partial z_i^{(k)}} = \sum_j \frac{\partial Div}{\partial y_j^{(k)}} y_i^{(k)} (\delta_{ij} - y_j^{(k)})$$



- Change in $w_{11} = \delta w_{11} = -0.1 * ddiv/dw_{11}$
 $= -0.1 * ddiv/dz_1 * dz_1/dw_{11} = -0.1 * ddiv/dz_1 * h_1$
- $ddiv/dy_1 = -d_1/y_1 = 0$ $ddiv/dy_2 = -d_2/y_2 = -1/y_2$
- $ddiv/dz_1 = ddiv/dy_1 * y_1(1-y_1) + ddiv/dy_2 * y_1(-y_2) = -y_1 y_2 / y_2 = -y_1$
- Input to softmax node $z_1 = w_{11}=1; z_2 = 0; y_1 = e / (1+e)$
- $\delta w_{11} = -0.1 * e/(1+e) = -0.0731$