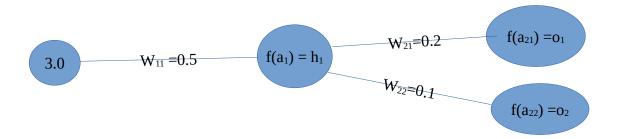
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Activation function:

$$f(x) = 2.0 * x$$

derivative
$$\frac{df(x)}{dx} = 2.0$$

Forward pass:

$$a_1 = w_{11} * x_{11}$$

 $h_1 = f(a_1) = 2 * a_1$

$$a_{21} = w_{21} * h_1$$

$$a_{22} = w_{22} * h_1$$

$$o_1 = f(a_{21}) = 2.0 * a_{21}$$

 $o_2 = f(a_{22}) = 2.0 * a_{22}$;

with real values:

$$a_1 = w_{11} * x_{11} = 0.5 * 3.0 = 1.5$$

 $h_1 = f(a_1) = 2.0 * a_1 = 2.0 * 1.5 = 3.0$

$$a_{21} = w_{21} * h_1 = 0.2 * 3 = 0.6$$

 $a_{22} = w_{22} * h_1 = 0.1 * 3.0 = 0.3$

$$o_1 = f(a_{21}) = 2.0 * a_{21} = 1.2$$

 $o_2 = f(a_{22}) = 2.0 * a_{22} = 0.6$

Target values: [2.0, 1.0]

Errors = $[e_1, e_2]$

Loss function: mean squared error (mse)

$$e_1 = (2.0-1.2)^2$$

 $e_2 = (1.0-0.6)^2$

Backward pass:

1. Calculate derivative of $\frac{\partial e_1}{\partial w_{21}}$

$$\frac{\partial e_1}{\partial w_{21}} = \frac{\partial e_1}{\partial o_1} * \frac{\partial o_1}{\partial a_{21}} * \frac{\partial a_{21}}{\partial w_{21}}$$

Calculation:

$$\frac{\partial e_1}{\partial o_1} = 2.0 * 0.8 * (-1)$$

$$\frac{\partial o_1}{\partial a_{21}} = 2.0$$

$$\frac{\partial a_{21}}{\partial w_{21}} = h_1 = 3.0$$

Together:

$$\frac{\partial e_1}{\partial w_{21}} = \frac{\partial e_1}{\partial o_1} * \frac{\partial o_1}{\partial a_{21}} * \frac{\partial a_{21}}{\partial w_{21}} = 2.0 * 0.8 * (-1) * 2.0 * 3 = -9.6$$

Manual derivative check:

$$a_1 = w_{11} * x_{11} = 0.5 * 3.0 = 1.5$$

 $h_1 = f(a_1) = 2.0 * a_1 = 2.0 * 1.5 = 3.0$

$$a_{21} = (w_{21} + 0.000001) * h_1 = (0.2 + 0.000001) * 3 = 0.6$$

$$o_1 = f(a_{21}) = 2.0 * a_{21} = 1.200006$$

$$\frac{(2.0-1.200006)^2-(2.0-1.2)^2}{0.000001}=-9,599964\approx-9.6$$

2. Calculate derivative of $\frac{\partial e_2}{\partial w_{22}}$

$$\frac{\partial e_2}{\partial w_{22}} = \frac{\partial e_2}{\partial o_2} * \frac{\partial o_2}{\partial a_{22}} * \frac{\partial a_{22}}{\partial w_{22}}$$

$$\frac{\partial e_2}{\partial o_2} = 2.0 * 0.4 * (-1)$$

$$\frac{\partial o_2}{\partial a_{22}} = 2.0$$

$$\frac{\partial a_{22}}{\partial w_{22}} = h_1 = 3.0$$

Together:

$$\frac{\partial e_2}{\partial w_{22}} = \frac{\partial e_2}{\partial o_2} * \frac{\partial o_2}{\partial a_{22}} * \frac{\partial a_{22}}{\partial w_{22}} = 2.0 * 0.4 * (-1) * 2.0 * 3 = -4.8$$

Manual derivative check:

$$a_1 = w_{11} * x_{11} = 0.5 * 3.0 = 1.5$$

 $h_1 = f(a_1) = 2.0 * a_1 = 2.0 * 1.5 = 3.0$

$$a_{22} = (w_{22} + 0.00001) * h_1 = (0.1 + 0.00001) * 3.0 = 0.300003$$

 $o_2 = f(a_{22}) = 2.0 * a_{22} = 0.600006$

$$\frac{(1.0-0.600006)^2-(1.0-0.6)^2}{0.000001}=-4,799964\approx-4.8$$

Recollection:

Forward pass:

$$a_1 = w_{11} * x_{11}$$

 $h_1 = f(a_1) = 2 * a_1$

$$a_{21} = w_{21} * h_1$$

 $a_{22} = w_{22} * h_1$

$$o_1 = f(a_{21}) = 2.0 * a_{21}$$

 $o_2 = f(a_{22}) = 2.0 * a_{22}$

$$e_1 = (2.0-1.2)^2$$

 $e_2 = (1.0-0.6)^2$

2. Calculate derivative of $\frac{\partial e_{total}}{\partial w_{11}} = \frac{\partial e_1}{\partial w_{11}} + \frac{\partial e_2}{\partial w_{11}}$

Calculate first $\frac{\partial e_1}{\partial w_{11}}$

$$\frac{\partial e_1}{\partial w_{11}} = \frac{\partial e_1}{\partial o_1} * \frac{\partial o_1}{\partial a_{21}} * \frac{\partial a_{21}}{\partial h_1} * \frac{\partial h_1}{\partial a_1} * \frac{\partial a_1}{\partial w_{11}}$$

Derivative is already calculated in the output layer:

$$\frac{\partial e_1}{\partial o_1} * \frac{\partial o_1}{\partial a_{21}}$$

$$\frac{\partial e_1}{\partial o_1} = 2.0 * 0.8 * (-1)$$

$$\frac{\partial o_1}{\partial a_{21}} = 2.0$$

So need to further calculate the other values:

$$\frac{\partial e_1}{\partial w_{11}} = \frac{\partial e_1}{\partial o_1} * \frac{\partial o_1}{\partial a_{21}} * \frac{\partial a_{21}}{\partial h_1} * \frac{\partial h_1}{\partial a_1} * \frac{\partial a_1}{\partial w_{11}}$$

$$\frac{\partial a_{21}}{\partial h_1} = w_{21} = 0.2$$

$$\frac{\partial h_1}{\partial a_1} = 2.0$$

$$\frac{\partial a_1}{\partial w_{11}} = x \, 11 = 3.0$$

Altogethera

$$\frac{\partial e_1}{\partial w_{11}} = \frac{\partial e_1}{\partial o_1} * \frac{\partial o_1}{\partial a_{21}} * \frac{\partial a_{21}}{\partial h_1} * \frac{\partial h_1}{\partial a_1} * \frac{\partial a_1}{\partial w_{11}} = 2.0 * 0.8 * (-1) * 2.0 * 0.2 * 2.0 * 3.0 = -3.84$$

Manual derivative check:

$$a_1 = (w_{11} + 0.000001) * x_{11} = 1.500003$$

 $h_1 = f(a_1) = 2.0 * a_1 = 2.0 * 1.500003 = 3.000006$
 $a_{21} = w_{21} * h_1 = 0.2 * 3.000006 = 0.6000012$
 $o_1 = f(a_{21}) = 2.0 * a_{21} = 2.0 * a_{21} = w_{21} * h_1 = 2.0 * 0.6000012 = 1,2000024$

$$\frac{(2.0-1.2000024)^2-(2.0-1.2)^2}{0.000001} = -3,83999424 \approx -3.84$$

Calculate then
$$\frac{\partial e_2}{\partial w_{11}} = \frac{\partial e_2}{\partial o_2} * \frac{\partial o_2}{\partial a_{22}} * \frac{\partial a_{22}}{\partial h_1} * \frac{\partial h_1}{\partial a_1} * \frac{\partial a_1}{\partial w_{11}}$$

$$\frac{\partial e_2}{\partial o_2} = 2.0 * 0.4 * (-1)$$

$$\frac{\partial o_2}{\partial a_{22}} = 2.0$$

$$\frac{\partial a_{22}}{\partial h_1} = w_{22} = 0.1$$

$$\frac{\partial h_1}{\partial a_1} = 2.0$$

$$\frac{\partial a_1}{\partial w_{11}} = 3.0$$

Altogether:

$$\frac{\partial e_2}{\partial w_{11}} = \frac{\partial e_2}{\partial o_2} * \frac{\partial o_2}{\partial a_{22}} * \frac{\partial a_{22}}{\partial h_1} * \frac{\partial h_1}{\partial a_1} * \frac{\partial a_1}{\partial w_{11}} = 2.0 * 0.4 * (-1) * 2.0 * 0.1 * 2.0 * 3.0 = -0.96$$

Manual derivative check:

$$a_1 = (w_{11} + 0.000001) * x_{11} = 1.500003$$

 $h_1 = f(a_1) = 2.0 * a_1 = 2.0 * 1.500003 = 3.000006$

$$\begin{aligned} &a_{22} = w_{22} * h_1 = 0.1 * 3.000006 = 0.3000006 \\ &o_2 = f(a_{22}) = 2.0 * a_{22} = 2.0 * 0.3000006 = 0.6000012 \end{aligned}$$

$$\frac{(1.0-0.6000012)^2-(1.0-0.6)^2}{0.000001} = -0.95999856 \approx -0.96$$

$$\frac{\partial e_{total}}{\partial w_{11}} = \frac{\partial e_1}{\partial w_{11}} + \frac{\partial e_2}{\partial w_{11}} = -3.84 + (-0.96) = -4.8$$

Conclusion:

As one can see the partial derivative of the output layer is propagated in direction of the input layer. Thus a general backpropagation rule can be formulated as:

Gradient o_{ji} is a partial derivative that is used to update the weight with index ji, in other words the weight from hidden neuron j to the output neuron i.

gradient
$$o_{ji} = \frac{\partial e_j}{\partial o_j} * \frac{\partial o_j}{\partial a_{ji}}$$