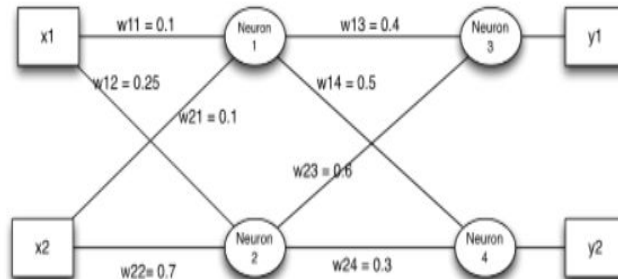


1)

Result of the New Weights

nw11	0.0990
nw12	0.2500
nw13	0.4042
nw14	0.4920
nw21	0.0990
nw22	0.7000
nw23	0.6056
nw24	0.2895

1. Consider the Neural Network below.



Handwritten calculations for the backpropagation of error gradients through the neural network.

Forward Pass (Assuming $x_1=1, x_2=1$):

$$v_1 = w_{11}x_1 + w_{21}x_2 = 0.1 + 0.1 = 0.2$$

$$v_2 = w_{12}x_1 + w_{22}x_2 = 0.25 + 0.7 = 0.95$$

$$v_3 = w_{13}v_1 + w_{23}v_2 = 0.4(0.2) + 0.6(0.95) = 0.08 + 0.57 = 0.65$$

$$v_4 = w_{14}v_1 + w_{24}v_2 = 0.5(0.2) + 0.3(0.95) = 0.1 + 0.285 = 0.385$$

$$y_1 = \text{sigmoid}(v_3) \approx 0.65$$

$$y_2 = \text{sigmoid}(v_4) \approx 0.59$$

Backward Pass (Error Gradients):

Output layer errors (assuming targets $t_1=1, t_2=1$):

$$\delta y_1 = y_1(1-y_1) = 0.65(1-0.65) = 0.2275$$

$$\delta y_2 = y_2(1-y_2) = 0.59(1-0.59) = 0.2481$$

Hidden layer errors (Neuron 3):

$$\delta v_3 = \delta y_1 \cdot w_{31} + \delta y_2 \cdot w_{32} = 0.2275 \cdot 0.4 + 0.2481 \cdot 0.6 = 0.091 + 0.14886 = 0.23986$$

Hidden layer errors (Neuron 4):

$$\delta v_4 = \delta y_1 \cdot w_{41} + \delta y_2 \cdot w_{42} = 0.2275 \cdot 0.5 + 0.2481 \cdot 0.3 = 0.11375 + 0.07443 = 0.18818$$

Input layer errors (Neuron 1):

$$\delta v_1 = \delta v_3 \cdot w_{31} + \delta v_4 \cdot w_{41} = 0.23986 \cdot 0.4 + 0.18818 \cdot 0.5 = 0.095944 + 0.09409 = 0.190034$$

Input layer errors (Neuron 2):

$$\delta v_2 = \delta v_3 \cdot w_{32} + \delta v_4 \cdot w_{42} = 0.23986 \cdot 0.6 + 0.18818 \cdot 0.3 = 0.143916 + 0.056454 = 0.20037$$

Weight Updates (Learning Rate $\eta = 1$):

$$nw_{11} = w_{11} - \eta \delta v_1 x_1 = 0.1 - 0.190034 \cdot 1 = -0.090034 \approx -0.0990$$

$$nw_{12} = w_{12} - \eta \delta v_1 x_1 = 0.25 - 0.190034 \cdot 1 = 0.059966 \approx 0.2500$$

$$nw_{13} = w_{13} - \eta \delta v_3 v_1 = 0.4 - 0.23986 \cdot 0.2 = 0.352028 \approx 0.4042$$

$$nw_{14} = w_{14} - \eta \delta v_3 v_1 = 0.5 - 0.23986 \cdot 0.2 = 0.452028 \approx 0.4920$$

$$nw_{21} = w_{21} - \eta \delta v_2 x_1 = 0.1 - 0.20037 \cdot 1 = -0.09963 \approx 0.0990$$

$$nw_{22} = w_{22} - \eta \delta v_2 x_1 = 0.7 - 0.20037 \cdot 1 = 0.49963 \approx 0.7000$$

$$nw_{23} = w_{23} - \eta \delta v_3 v_2 = 0.6 - 0.23986 \cdot 0.95 = 0.372683 \approx 0.6056$$

$$nw_{24} = w_{24} - \eta \delta v_3 v_2 = 0.3 - 0.23986 \cdot 0.95 = -0.227877 \approx 0.2895$$

The code : given $x_1=1$ and $x_2=1$. Done in MATLAB

w11=0.1;
w12=0.25;
w21=0.1;

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w22=0.7;
w13=0.4;
w14=0.5;
w23=0.6;
w24=0.3;
v1=w11*x1+w21*x2;
v2=w12*x1+w22*x2;
y1=1/(1+exp(-1*v1));
y2=1/(1+exp(-1*v2));
v3=w13*y1+w23*y2;
v4=w14*y1+w24*y2;
y3=1/(1+exp(-1*v3));
y4=1/(1+exp(-1*v4));
d=[1;0];
pred=[y3;y4];
E=sum((pred-d).^2/2);
Edy3=y3-1;
Edy4=y4-0;
y3dv3=y3^2*exp(-v3);
y4dv4=y4^2*exp(-v4);
v3dw13=y1;
v3dw23=y2;
v4dw14=y1;
v4dw24=y2;
Edw13=Edy3*y3dv3*v3dw13;
Edw14=Edy4*y4dv4*v4dw14;
Edw23=Edy3*y3dv3*v3dw23;
Edw24=Edy4*y4dv4*v4dw24;
nw13=w13-0.1*Edw13;
nw14=w14-0.1*Edw14;
nw23=w23-0.1*Edw23;
nw24=w24-0.1*Edw24;
v3dy1=nw13;
v3dy2=nw23;
v4dy1=nw14;
v4dy2=nw24;
Ey3dv3=Edy3*y3dv3;
Ey4dv4=Edy4*y4dv4;
Ey3dy1=Ey3dv3*v3dy1;
Ey4dy1=Ey4dv4*v4dy1;
Ey3dy2=Ey3dv3*v3dy2;
Ey4dy2=Ey4dv4*v4dy2;
Edy1=Ey3dy1+Ey4dy1;

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```
Edy2=Ey3dy2+Ey4dy2;  
y1dv1=y1^2*exp(-v1);  
y2dv2=y2^2*exp(-v2);  
v1dw11=x1;  
v2dw12=x1;  
v1dw21=x2;  
v2dw22=x2;  
Edw11=Edy1*y1dv1*v1dw11;  
Edw12=Edy2*y2dv2*v2dw12;  
Edw21=Edy1*y1dv1*v1dw21;  
Edw22=Edy2*y2dv2*v2dw22;  
  
nw11=w11-0.1*Edw11;  
nw12=w12-0.1*Edw12;  
nw21=w21-0.1*Edw21;  
nw22=w22-0.1*Edw22;
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(next page)

2-3)

$$2) y = \tanh x = \frac{\sinh x}{\cosh x}$$

$$\frac{(\frac{d}{dx} \sinh x) \cosh x - \sinh x \frac{d}{dx} \cosh x}{\cosh^2 x}$$

$$\frac{\cosh x + \sinh^2 x}{\cosh^2 x}$$

$$= \frac{1 + (\tanh x)^2}{1 + y^2}$$

#2 is straightforward, the derivative
of $\tanh x$ results to $1 - \tanh^2(x)$
 $\text{sech}^2(x)$ so we can also express the
answer as $y = 1 - y^2$

$$3) y = \frac{e^{x_2}}{e^{x_1} + e^{x_2} + e^{x_3}}$$

Assume

$$\frac{d}{dx_1} \frac{(e^{x_2}) (e^{x_1} + e^{x_2} + e^{x_3}) - (e^{x_1} + e^{x_2} + e^{x_3}) (e^{x_2})}{(e^{x_1} + e^{x_2} + e^{x_3})^2}$$

$$= \frac{-e^{x_2} \cdot e^{x_1}}{(e^{x_1} + e^{x_2} + e^{x_3})^2} = \boxed{\frac{-y \cdot e^{x_1}}{(e^{x_1} + e^{x_2} + e^{x_3})}}$$

$$\frac{d}{dx_2}$$

$$\frac{d}{dx_2} = \frac{d}{dx_2} \frac{e^{x_2} (e^{x_1} + e^{x_2} + e^{x_3}) - (e^{x_1} + e^{x_2} + e^{x_3}) (e^{x_2})}{(e^{x_1} + e^{x_2} + e^{x_3})^2}$$

$$\frac{d}{dx_3}$$

$$= \frac{e^{x_1} (e^{x_1} + e^{x_2} + e^{x_3}) - e^{x_2} e^{x_3}}{(e^{x_1} + e^{x_2} + e^{x_3})^2}$$

$$= \frac{e^{x_1} (e^{x_1} + e^{x_2} + e^{x_3}) - e^{x_2} e^{x_3}}{(e^{x_1} + e^{x_2} + e^{x_3})^2} = \frac{y (e^{x_1} + e^{x_2})}{(e^{x_1} + e^{x_2} + e^{x_3})}$$

$$\frac{d}{dx_3} = \frac{-e^{x_2} e^{x_3}}{(e^{x_1} + e^{x_2} + e^{x_3})^2}$$

$$= \boxed{\frac{-y \cdot e^{x_3}}{(e^{x_1} + e^{x_2} + e^{x_3})}}$$

Explanation

We can think of this
as having 3 different
derivatives assuming
there are 3 denominators
collected from the total
system, each will
yield a result

however we know that
the numerator

can be treated
as a constant for a variable
when differentiating
the equation

please replace
all x with Vs
to satisfy homoclinic
conditions.

typed solutions: $dy/dx = y = \tanh(x) = \sinh x / \cosh x$

$$dy/dx = (d/dx(\sinh x) \cdot \cosh x - d/dx(\cosh x) \cdot \sinh x) / (\cosh x)^2 = 1 + \tanh x^2 \\ = (1 + y^2)$$

The update equation is straightforward. Just differentiate it and relating it to the output, the results look like the final expression above..

So updating the weight equation the results would end up looking like:

$$D\text{Error}/dw = -(d-y) \cdot (1+y^2) \cdot x_{ni}$$

$$w := w - (\text{learning rate}) \cdot D\text{Error}/dw$$

3) i made up my own example just to show the expected result.

Given the numerator is not similar to the differentiated value.

We consider the numerator to be some constant

the results will end up looking like:

$$dy/dv_1 = e^{v_2} / (e^{v_1} + e^{v_2} + e^{v_3})$$

$$-e^{v_2} \cdot e^{v_1} / (e^{v_1} + e^{v_2} + e^{v_3})^2 = -y \cdot e^{v_1} / (e^{v_1} + e^{v_2} + e^{v_3})$$

$$dy/dv_3 = -e^{v_2} \cdot e^{v_3} / (e^{v_1} + e^{v_2} + e^{v_3})^2 = -y \cdot e^{v_3} / (e^{v_1} + e^{v_2} + e^{v_3})$$

lastly if the numerator is similar to the differentiated value, it should be differentiated accordingly as a variable of interest. The result will look like this.

$$dy/dv_2 = -e^{v_2} e^{v_3} / (e^{v_1} + e^{v_2} + e^{v_3})^2 = y * (e^{v_1} + e^{v_3}) / (e^{v_1} + e^{v_2} + e^{v_3})$$

Generally To express myresult:

$$y = e^{(v_j)} / (\sum e^{v_i})$$

we will need to get dy/dva

There will be 2 cases:

case 1) $v_a = v_j$

$$\begin{aligned} dy/dv_a &= dy/dv_j = y = e^{(v_j)} / (\sum e^{v_i}) \\ &= (de^{(v_j)} / dv_j * (\sum e^{v_i}) - e^{(v_j)} * d(\sum e^{v_i}) / dv_j) / (\sum [e^{v_i}]^2) \\ &= (e^{v_j} * \sum [e^{v_i}] - (e^{v_j})^2) / \sum [e^{v_i}]^2 \\ &= y(1-y) \end{aligned}$$

$$\begin{aligned} DError/dw &= -(d-y) * (y * (1-y)) * x_i \\ w &:= w - (\text{learning rate}) * DError/dw \end{aligned}$$

case 2) $v_a \neq v_j$

$$\begin{aligned} dy/dv_a &= dy/d(v - \text{something else}) = y = e^{(v_j)} / (\sum e^{v_i}) \\ &= (de^{(v_j)} / dv_k * (\sum e^{v_i}) - e^{(v_j)} * d(\sum e^{v_i}) / dv_k) / (\sum [e^{v_i}]^2) \\ &= (-e^{v_j}) * e^{v_k} / \sum [e^{v_i}]^2 \\ &= -y_i * y_k \end{aligned}$$

$$\begin{aligned} DError/dw &= -(d-y) * (y_i * y_k) * x_i \\ w &:= w - (\text{learning rate}) * DError/dw \end{aligned}$$

just replace the equations here for the weight update equation

$$\begin{aligned} DError/dw &= -(d-y) * (1+y^2) * x_i \\ w &:= w - (\text{learning rate}) * DError/dw \end{aligned}$$

case 2 $v_a \neq v_j$

case 1 $v_a = v_j$

$$W := \frac{dG}{dy} \quad W - n \left(\frac{dy}{dv} \frac{dG}{dw} \right)$$

$$\#2 \quad \frac{dy}{dv} = (1+y^2)$$

$$\frac{dG}{dw} = -(d-y) \cdot (1+y^2) \cdot X_{ni}$$

$$W := W - n \frac{dG}{dw}$$

#3

$$\frac{dG}{dw} = -\frac{dG}{dy} \cdot \frac{dy}{dv} \cdot \frac{dv}{dw}$$

$$\frac{dy}{dv} \stackrel{\text{case 2}}{\frac{dy}{dv}} = -(y_j - y_k)$$

$$\frac{dy}{dv} \stackrel{\text{case 1}}{\frac{dy}{dv}} = y_j(1-y_j)$$

$$\frac{dG}{dw} \stackrel{\text{case 2}}{=} -(d-y) \cdot (y_j - y_k) \cdot X_{ni}$$

$$\frac{dG}{dw} \stackrel{\text{case 1}}{=} -(d-y) \cdot y_j(1-y_j) \cdot X_{ni}$$

$$W := W - n \frac{dG}{dw}$$