

反向传播

单个训练样本的损失函数:

$$C(\mathbf{w}, \mathbf{b}) = \frac{1}{2} \|\mathbf{y} - \mathbf{o}\|^2 = \frac{1}{2} \sum_{i=0}^1 (\mathbf{y}_i - \mathbf{o}_i)^2 \quad (33)$$

损失函数 MSE:

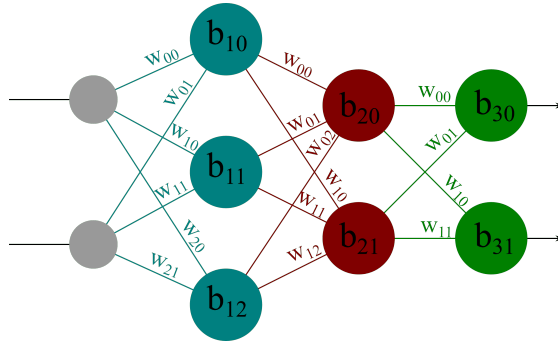
$$L(\mathbf{w}, \mathbf{b}) = \sum_{n=1}^N C^n(\mathbf{w}, \mathbf{b}) \quad (34)$$

对每一个训练样本得到的梯度分量求和，即可得到总的梯度分量:

$$\begin{aligned} \frac{\partial L(\mathbf{w}, \mathbf{b})}{\partial w} &= \sum_{n=1}^N \frac{\partial C^n(\mathbf{w}, \mathbf{b})}{\partial w} \\ \frac{\partial L(\mathbf{w}, \mathbf{b})}{\partial b} &= \sum_{n=1}^N \frac{\partial C^n(\mathbf{w}, \mathbf{b})}{\partial b} \end{aligned} \quad (35)$$

损失函数对第 $i = 1$ 层的其中一个权重 w_{10} 的偏导数为:

$$\frac{\partial C}{\partial w_{00}} = \frac{\partial C}{\partial z_{10}} \frac{\partial z_{10}}{\partial w_{00}} \quad (36)$$



前向传播计算 $\frac{\partial z}{\partial w}$

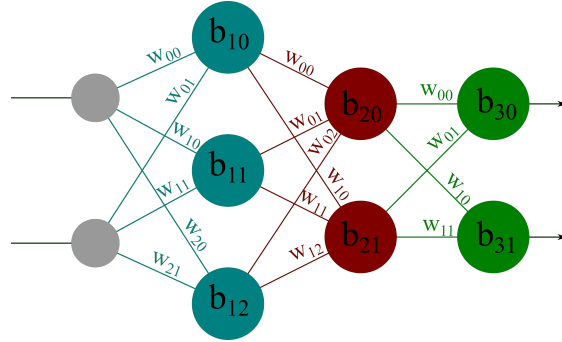
由于激活 $a_{10} = \sigma(z_{10})$ 的中间变量 $z_{10} = w_{00}a_{00} + w_{01}a_{01} + b_{10}$ ，因此有 $\frac{\partial z}{\partial w_{00}} = a_{00}$ ；于是可以得到这一层中连接到该神经元的权重 w 对应的 $\frac{\partial z}{\partial w}$ 分别为 a_{00}, a_{01} ；那么整个第 i 层所有权重

$$\begin{bmatrix} w_{00} & w_{01} \\ w_{10} & w_{11} \\ w_{20} & w_{21} \end{bmatrix} \quad (37)$$

对应的 $\frac{\partial z}{\partial w}$ 为，每一组(行)对应的值是上一层的激活，每一行都是一样的，这些值通过前向传播直接就算出来了:

$$\begin{bmatrix} a_{00} & a_{01} \\ a_{00} & a_{01} \\ a_{00} & a_{01} \end{bmatrix} \quad (38)$$

反向传播计算 $\frac{\partial C}{\partial z}$



$$\frac{\partial C}{\partial z_{10}} = \frac{\partial C}{\partial a_{10}} \frac{\partial a_{10}}{\partial z_{10}} \quad (39)$$

因为 $a_{10} = \sigma(z_{10})$, 因此

$$\frac{\partial a_{10}}{\partial z_{10}} = \sigma'(z_{10}) \quad (40)$$

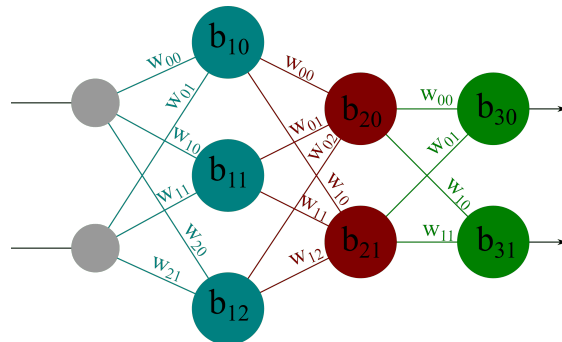
那么这一层所有节点对应的 $\frac{\partial a}{\partial z}$ 构成向量为

$$[\sigma'(z_{10}), \sigma'(z_{11}), \sigma'(z_{12})]^\top \quad (41)$$

z 在前向传播的时候可以记录下来, 而 $\sigma'(z)$ 也是知道的。

还剩下一项:

$$\frac{\partial C}{\partial a_{10}} = \frac{\partial C}{\partial z_{20}} \frac{\partial z_{20}}{\partial a_{10}} + \frac{\partial C}{\partial z_{21}} \frac{\partial z_{21}}{\partial a_{10}} \quad (42)$$



由于 $z_{20} = w_{00}a_{10} + w_{01}a_{11} + w_{02}a_{12}$, 因此 $\frac{\partial z_{20}}{\partial a_{10}} = w_{00}$, 同理 $\frac{\partial z_{21}}{\partial a_{10}} = w_{10}$, 于是上式变成:

$$\frac{\partial C}{\partial a_{10}} = \frac{\partial C}{\partial z_{20}} w_{00} + \frac{\partial C}{\partial z_{21}} w_{10} \quad (43)$$

同理, 这一层其他神经元对应

$$\begin{aligned}\frac{\partial C}{\partial a_{11}} &= \frac{\partial C}{\partial z_{20}} w_{01} + \frac{\partial C}{\partial z_{21}} w_{11} \\ \frac{\partial C}{\partial a_{12}} &= \frac{\partial C}{\partial z_{20}} w_{02} + \frac{\partial C}{\partial z_{21}} w_{12}\end{aligned}\tag{44}$$

写成矩阵的形式是这样的

$$\begin{bmatrix} \frac{\partial C}{\partial a_{10}} \\ \frac{\partial C}{\partial a_{11}} \\ \frac{\partial C}{\partial a_{12}} \end{bmatrix} = \begin{bmatrix} w_{00} & w_{10} \\ w_{01} & w_{11} \\ w_{02} & w_{12} \end{bmatrix} \begin{bmatrix} \frac{\partial C}{\partial z_{20}} \\ \frac{\partial C}{\partial z_{21}} \end{bmatrix}\tag{45}$$

注意到

$$\begin{bmatrix} w_{00} & w_{10} \\ w_{01} & w_{11} \\ w_{02} & w_{12} \end{bmatrix} = \begin{bmatrix} w_{00} & w_{01} & w_{02} \\ w_{10} & w_{11} & w_{12} \end{bmatrix}^\top\tag{46}$$

这个关系会给我们带来很大的方便。

回到

$$\begin{aligned}\frac{\partial C}{\partial z_{10}} &= \frac{\partial C}{\partial a_{10}} \frac{\partial a_{10}}{\partial z_{10}} \\ &= \sigma'(z_{10}) \frac{\partial C}{\partial a_{10}} \\ &= \sigma'(z_{10}) \left(w_{00} \frac{\partial C}{\partial z_{20}} + w_{10} \frac{\partial C}{\partial z_{21}} \right)\end{aligned}\tag{47}$$

再回到

$$\begin{aligned}\frac{\partial C}{\partial w_{00}} &= \frac{\partial C}{\partial z_{10}} \frac{\partial z_{10}}{\partial w_{00}} \\ &= a_{00} \frac{\partial C}{\partial z_{10}} \\ &= a_{00} \sigma'(z_{10}) \left(w_{00} \frac{\partial C}{\partial z_{20}} + w_{10} \frac{\partial C}{\partial z_{21}} \right)\end{aligned}\tag{48}$$

$$\frac{\partial C}{\partial w_{00}^i} = a_{00} \sigma'(z_{10}) \left(w_{00}^j \frac{\partial C}{\partial z_{20}} + w_{10}^j \frac{\partial C}{\partial z_{21}} \right)\tag{49}$$

汇总一下:

$$\nabla \mathbf{w}^i = \begin{bmatrix} a_{00} & a_{01} \\ a_{00} & a_{01} \\ a_{00} & a_{01} \end{bmatrix} \odot \begin{bmatrix} \frac{\partial C}{\partial z_{10}} \\ \frac{\partial C}{\partial z_{11}} \\ \frac{\partial C}{\partial z_{12}} \end{bmatrix}\tag{50}$$

其中,

$$\begin{aligned}
\begin{bmatrix} \frac{\partial C}{\partial z_{10}} \\ \frac{\partial C}{\partial z_{11}} \\ \frac{\partial C}{\partial z_{12}} \end{bmatrix} &= \begin{bmatrix} \sigma'(z_{10}) \\ \sigma'(z_{11}) \\ \sigma'(z_{12}) \end{bmatrix} \odot \begin{bmatrix} \frac{\partial C}{\partial a_{10}} \\ \frac{\partial C}{\partial a_{11}} \\ \frac{\partial C}{\partial a_{12}} \end{bmatrix} \\
&= \begin{bmatrix} \sigma'(z_{10}) \\ \sigma'(z_{11}) \\ \sigma'(z_{12}) \end{bmatrix} \odot \left(\begin{bmatrix} w_{00} & w_{10} \\ w_{01} & w_{11} \\ w_{02} & w_{12} \end{bmatrix} \begin{bmatrix} \frac{\partial C}{\partial z_{20}} \\ \frac{\partial C}{\partial z_{21}} \end{bmatrix} \right) \\
&= \begin{bmatrix} \sigma'(z_{10}) \\ \sigma'(z_{11}) \\ \sigma'(z_{12}) \end{bmatrix} \odot \left(\begin{bmatrix} w_{00} & w_{01} & w_{02} \\ w_{10} & w_{11} & w_{12} \end{bmatrix}^T \begin{bmatrix} \frac{\partial C}{\partial z_{20}} \\ \frac{\partial C}{\partial z_{21}} \end{bmatrix} \right) \\
&= \begin{bmatrix} \sigma'(z_{10}) \\ \sigma'(z_{11}) \\ \sigma'(z_{12}) \end{bmatrix} \odot \left(\mathbf{w}^j{}^T \begin{bmatrix} \frac{\partial C}{\partial z_{20}} \\ \frac{\partial C}{\partial z_{21}} \end{bmatrix} \right)
\end{aligned} \tag{51}$$

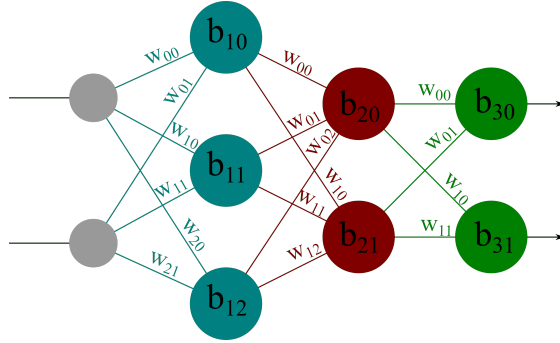
即

$$\frac{\partial C}{\partial \mathbf{z}_1} = \sigma'(\mathbf{z}_1) \odot \left(\mathbf{w}^2{}^T \frac{\partial C}{\partial \mathbf{z}_2} \right) \tag{52}$$

于是:

$$\nabla \mathbf{w}^1 = \begin{bmatrix} a_{00} & a_{01} \\ a_{00} & a_{01} \\ a_{00} & a_{01} \end{bmatrix} \odot \begin{bmatrix} \frac{\partial C}{\partial z_{10}} \\ \frac{\partial C}{\partial z_{11}} \\ \frac{\partial C}{\partial z_{12}} \end{bmatrix} = \begin{bmatrix} a_{00} & a_{01} \\ a_{00} & a_{01} \\ a_{00} & a_{01} \end{bmatrix} \odot \left\{ \begin{bmatrix} \sigma'(z_{10}) \\ \sigma'(z_{11}) \\ \sigma'(z_{12}) \end{bmatrix} \odot \left(\mathbf{w}^2{}^T \begin{bmatrix} \frac{\partial C}{\partial z_{20}} \\ \frac{\partial C}{\partial z_{21}} \end{bmatrix} \right) \right\} \tag{53}$$

即只要算出每个神经元的 $\frac{\partial c}{\partial z}$ 即可算出梯度的改变量，先把最后一层的 $\frac{\partial c}{\partial z}$ 算出，一层层就可往前算出了。



输出层,

$$C(\mathbf{w}, \mathbf{b}) = \frac{1}{2} \|\mathbf{y} - \mathbf{o}\|^2 = \frac{1}{2} \sum_{i=0}^1 (\mathbf{y}_i - \mathbf{o}_i)^2 = \frac{1}{2} (y_0^2 - 2y_0o_0 + o_0^2 + y_1^2 - 2y_1o_1 + o_1^2) \tag{54}$$

$$o_0 = \sigma(z_{30}) \tag{55}$$

$$\frac{\partial C}{\partial z_{30}} = \frac{\partial C}{\partial o_0} \frac{\partial o_0}{\partial z_{30}} = (o_0 - y_0) \sigma'(z_{30}) \tag{56}$$

同理

$$\frac{\partial C}{\partial z_{31}} = \frac{\partial C}{\partial o_1} \frac{\partial o_1}{\partial z_{31}} = (o_1 - y_1) \sigma'(z_{31}) \tag{57}$$

$$\frac{\partial C}{\partial \mathbf{z}_3} = (\mathbf{o} - \mathbf{y})\sigma'(\mathbf{z}_3) \quad (58)$$

总结

相邻两层*i, j*关键量的计算关系:

$$\frac{\partial C}{\partial \mathbf{z}_i} = \sigma'(\mathbf{z}_i) \odot \left(\mathbf{w}^{j\top} \frac{\partial C}{\partial \mathbf{z}_j} \right) \quad (59)$$

最后一层若*j=0*, 即第*j*层是输出层:

$$\frac{\partial C}{\partial \mathbf{z}_0} = (\mathbf{o} - \mathbf{y})\sigma'(\mathbf{z}_0) \quad (60)$$

第*i*层*w*的改变量:

$$\nabla \mathbf{w}^i = [\mathbf{a}_h, \mathbf{a}_h, \mathbf{a}_h]^\top \odot \frac{\partial C}{\partial \mathbf{z}_i} \quad (61)$$

对于偏置

仿照类似的方法可以得到:

$$\frac{\partial C}{\partial b_{10}} = \frac{\partial C}{\partial z_{10}} \frac{\partial z_{10}}{\partial b_{10}} \quad (62)$$

前向传播计算 $\frac{\partial z}{\partial b}$

$a_{10} = \sigma(z_{10})$; $z_{10} = w_{00}a_{00} + w_{01}a_{01} + b_{10}$; 因此, $\frac{\partial z_{10}}{\partial b_{10}} = 1$

那么,

$$\frac{\partial \mathbf{z}_1}{\partial \mathbf{b}_1} = [1, 1, 1]^\top \quad (63)$$

反向传播计算 $\frac{\partial C}{\partial z}$

$$\nabla \mathbf{b}^i = \frac{\partial C}{\partial \mathbf{z}_i} \quad (64)$$