# Computer Version final project theory

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2023年6月25日

首先,设卷积层输入特征图为 X,卷积核为 W,输出特征图为 Y;设池化层的输入特征图为 X,输出特征图为 Y;误差损失函数为 L。

## 1 卷积层

#### 1.1 前向传播

$$Y_{i,j} = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} X_{i+m,j+n} W_{m,n} + b$$

其中,M和N分别为卷积核的高度和宽度。

由于输入特征 3×3, 卷积核与池化核尺寸为 2×2, 所以有:

$$Y_{0,0} = \Sigma(X[0:2,0:2] \cdot W) + b \tag{1}$$

$$Y_{0,1} = \Sigma(X[0:2,1:3] \cdot W) + b \tag{2}$$

$$Y_{1,0} = \Sigma(X[1:3,0:2] \cdot W) + b \tag{3}$$

$$Y_{1,1} = \Sigma(X[1:3,1:3] \cdot W) + b \tag{4}$$

(5)

最后使用激活函数(ReLU):

$$Y = active(Y)$$

### 1.2 反向传播

已知 Y = X \* W + b, 输出梯度为  $\frac{\partial L}{\partial Y}$ 。

先求输入梯度

根据链式求导法则可知:

$$\frac{\partial L}{\partial X} = \frac{\partial L}{\partial Y} \cdot \frac{\partial Y}{\partial X}$$

所以:

$$\frac{\partial L}{\partial X_{i,j}} = \sum_{p=0}^{H-M} \sum_{q=0}^{W-N} \frac{\partial L}{\partial Y_{p,q}} \cdot W_{i-p,j-q}$$

1.2 反向传播 1.2 发向传播

其中, H 和 W 分别为输入特征图的高度和宽度。

所以,针对本次题目要求的 3×3 特征图,有:

$$y_{0,0} = x_{0,0}w_{0,0} + x_{0,1}w_{0,1} + x_{1,0}w_{1,0} + x_{1,1}w_{1,1} + b$$
(6)

$$y_{0,1} = x_{0,1}w_{0,0} + x_{0,2}w_{0,1} + x_{1,1}w_{1,0} + x_{1,2}w_{1,1} + b$$
(7)

$$y_{1,0} = x_{1,0}w_{0,0} + x_{1,1}w_{0,1} + x_{2,0}w_{1,0} + x_{2,1}w_{1,1} + b$$
(8)

$$y_{1,1} = x_{1,1}w_{0,0} + x_{1,2}w_{0,1} + x_{2,1}w_{1,0} + x_{2,2}w_{1,1} + b (9)$$

针对特征图中的  $x_{0,0}, x_{0,2}, x_{2,0}, x_{2,2}$  这类四角的点,以  $x_{2,2}$  为例,有  $\frac{\partial L}{\partial x_{2,2}} = \frac{\partial L}{\partial Y_{1,1}} w_{1,1}$  针对特征图中的  $x_{0,1}, x_{1,0}, x_{1,2}, x_{2,1}$  这类四边的点,以  $x_{1,0}$  为例,有  $\frac{\partial L}{\partial x_{1,0}} = \frac{\partial L}{\partial Y_{0,0}} w_{1,0} + \frac{\partial L}{\partial Y_{1,0}} w_{0,0}$ 

针对中心的  $x_{1,1}$  点,有  $\frac{\partial L}{\partial x_{1,1}} = \frac{\partial L}{\partial Y_{0,0}} w_{1,1} + \frac{\partial L}{\partial Y_{0,1}} w_{1,0} + \frac{\partial L}{\partial Y_{1,0}} w_{0,1} + \frac{\partial L}{\partial Y_{1,1}} w_{1,0}$ 

接下来求权重梯度

根据链式求导法则可知:  $\frac{\partial L}{\partial W} = \frac{\partial L}{\partial Y} \frac{\partial Y}{\partial W}$  也即:

$$\frac{\partial L}{\partial W_{i,j}} = \sum_{p=0}^{H-M} \sum_{q=0}^{W-N} \frac{\partial L}{\partial Y_{p,q}} \cdot X_{i+p,j+q}$$

所以,针对题目要求的 3×3 特征图,有:

$$\frac{\partial L}{\partial w_{0,0}} = \frac{\partial L}{\partial y_{0,0}} \cdot x_{0,0} + \frac{\partial L}{\partial y_{0,1}} \cdot x_{0,1} + \frac{\partial L}{\partial y_{1,0}} \cdot x_{1,0} + \frac{\partial L}{\partial y_{1,1}} \cdot x_{1,1}$$
(10)

$$\frac{\partial L}{\partial w_{0,1}} = \frac{\partial L}{\partial y_{0,0}} \cdot x_{0,1} + \frac{\partial L}{\partial y_{0,1}} \cdot x_{0,2} + \frac{\partial L}{\partial y_{1,0}} \cdot x_{1,1} + \frac{\partial L}{\partial y_{1,1}} \cdot x_{1,2}$$
(11)

$$\frac{\partial L}{\partial w_{1,0}} = \frac{\partial L}{\partial y_{0,0}} \cdot x_{1,0} + \frac{\partial L}{\partial y_{0,1}} \cdot x_{1,1} + \frac{\partial L}{\partial y_{1,0}} \cdot x_{2,0} + \frac{\partial L}{\partial y_{1,1}} \cdot x_{2,1}$$
(12)

$$\frac{\partial L}{\partial w_{1,1}} = \frac{\partial L}{\partial y_{0,0}} \cdot x_{1,1} + \frac{\partial L}{\partial y_{0,1}} \cdot x_{1,2} + \frac{\partial L}{\partial y_{1,0}} \cdot x_{2,1} + \frac{\partial L}{\partial y_{1,1}} \cdot x_{2,2} \tag{13}$$

(14)

可以简化为:

$$\frac{\partial L}{\partial W} = X \cdot Y$$

最后求偏置梯度

根据链式求导法则可知  $\frac{\partial L}{\partial b} = \frac{\partial L}{\partial Y} \frac{\partial Y}{\partial b}$ ,因为 b 只是一个标量,所以得到  $\frac{\partial L}{\partial b} = \Sigma \frac{\partial L}{\partial Y}$  也即:

$$\frac{\partial L}{\partial b} = \sum_{i,j} \frac{\partial L}{\partial Y_{i,j}}$$

由于已知输入特征 3×3, 卷积核与池化核尺寸为 2×2, 所以有:

$$\frac{\partial L}{\partial b} = \frac{\partial L}{\partial Y_{0,0}} + \frac{\partial L}{\partial Y_{0,1}} + \frac{\partial L}{\partial Y_{1,0}} + \frac{\partial L}{\partial Y_{1,1}}$$

## 2 池化层

### 2.1 前向传播

$$Y_{i,j} = \max_{m=0}^{M-1} \max_{n=0}^{N-1} X_{i+m,j+n}$$

其中,M和N分别为池化核的高度和宽度。

由于卷积层输出为 2×2, 池化核大小也为 2×2, 所以:

$$Y_{0,0} = \max_{i=0}^{1} \max_{j=0}^{1} X_{i,j}$$
 (15)

#### 2.2 反向传播

池化层的特点是无参数操作,所以在反向传播过程中,总是将输出梯度传导到最大值所对应的位置(maxpooling)

$$\frac{\partial L}{\partial X_{m,n}} = \begin{cases} \frac{\partial L}{\partial Y_{i,j}}, & \text{if } X_{m,n} = \max_{m=0}^{M-1} \max_{n=0}^{N-1} X_{m,n} \\ 0, & \text{otherwise} \end{cases}$$

由于已知卷积层输出、池化核大小均为 2×2:

$$\frac{\partial L}{\partial X_{m,n}} = \begin{cases} \frac{\partial L}{\partial Y_{i,j}}, & \text{if } X_{m,n} = \max_{m=0}^1 \max_{n=0}^1 X_{m,n} \\ 0, & \text{otherwise} \end{cases}, m \in \{0,1\}, n \in \{0,1\}$$