

# VGG (Visual Geometry Group)

Very Deep Convolutional Networks For Large-Scale Image Recognition

- 亮点: 通过堆叠多个  $3 \times 3$  的卷积核来替代大尺度卷积核 (减少所需参数)

② 论文中提到, 可以通过堆叠两个  $3 \times 3$  的卷积核替代  $5 \times 5$  的卷积核, 堆叠三个  $3 \times 3$  的卷积核替代  $7 \times 7$  的卷积核

Feature map:  $F=1$   
 Conv  $3 \times 3$  (3):  $F = (1-1) \times 1 + 3 = 3$   
 Conv  $3 \times 3$  (2):  $F = (3-1) \times 1 + 3 = 5$   
 Conv  $3 \times 3$  (1):  $F = (5-1) \times 1 + 3 = 7$   
 # 所需参数 < Input channel = C >  
 $7 \times 7 \times C \times C = 49C^2$   
 $3 \times 3 \times C \times C + 3 \times 3 \times C \times C + 3 \times 3 \times C \times C = 27C^2$

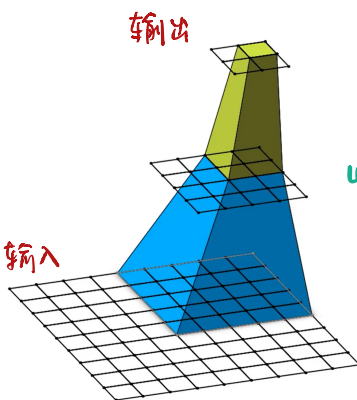
⇒ 拥有相同的感受野

- 般使用 D

receptive field

决定某一层输出结果中一个元素所对应的输入层的区域大小。  
 < 输出 feature map 上的一个单元对应输入层上区域的大小 >

ConvNet Configuration					
A	A-LRN	B	C	D	E
11 weight layers	11 weight layers	13 weight layers	16 weight layers	16 weight layers	19 weight layers
input (224 × 224 RGB image)					
conv3-64	conv3-64 LRN	conv3-64	conv3-64	conv3-64	conv3-64
maxpool					
conv3-128	conv3-128	conv3-128	conv3-128	conv3-128	conv3-128
maxpool					
conv3-256	conv3-256	conv3-256	conv3-256	conv3-256	conv3-256
maxpool					
conv3-512	conv3-512	conv3-512	conv3-512	conv3-512	conv3-512
maxpool					
conv3-512	conv3-512	conv3-512	conv3-512	conv3-512	conv3-512
maxpool					
FC-4096					
FC-1000					
soft-max					



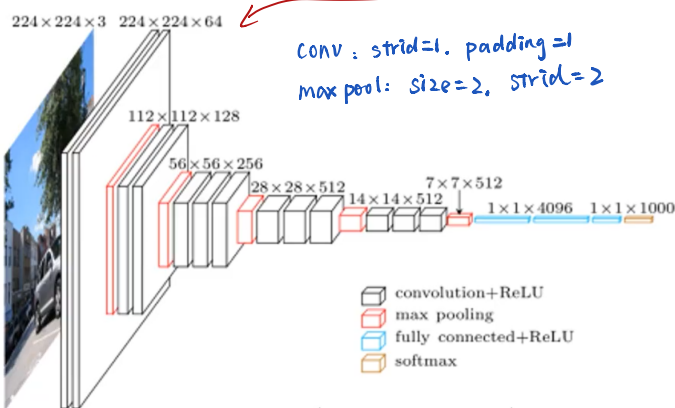
Max Pool  
 Size:  $2 \times 2$   
 Stride: 2  
 Conv1  
 Size:  $3 \times 3$   
 Stride: 2

$$out_{size} = (in_{size} - F_{size} + 2P) / S + 1$$

$$F(i) = (F(i+1) - 1) \times stride + ksize$$

↑  
第 i 层感受野  
↑  
kernel size

eg. Feature map:  $F=1$   
 Pool:  $F = (1-1) \times 2 + 2 = 2$   
 Conv:  $F = (2-1) \times 2 + 3 = 5$



conv:  $stride=1$ ,  $padding=1$   
 max pool:  $size=2$ ,  $stride=2$

$$out_{size} = (in_{size} - F_{size} + 2P) / S + 1$$