

Aprendizaje Automático Profundo (Deep Learning)





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## Arquitecturas Convolucionales

## Arquitecturas

• Arquitectura = Configuración de capas, #filtros, etc AllConvolutional VGG16

A	A-LRN	B	onfiguration C	D	E
11 weight	11 weight	13 weight	16 weight	16 weight	19 weight
layers	layers	layers	layers	layers	layers
50 5000		nput (224 $\times$ 2)			O 26 VO
conv3-64	conv3-64	conv3-64	conv3-64	conv3-64	conv3-64
	LRN	conv3-64	conv3-64	conv3-64	conv3-64
	N		pool		· · · · · · · · · · · · · · · · · · ·
conv3-128	conv3-128	conv3-128	conv3-128	conv3-128	conv3-128
		conv3-128	conv3-128	conv3-128	conv3-128
	*	max	pool		
conv3-256	conv3-256	conv3-256	conv3-256	conv3-256	conv3-256
conv3-256	conv3-256	conv3-256	conv3-256	conv3-256	conv3-256
			conv1-256	conv3-256	conv3-256
				***************************************	conv3-256
	8 4	max	pool	0 101 25 5 6002	
conv3-512	conv3-512	conv3-512	conv3-512	conv3-512	conv3-512
conv3-512	conv3-512	conv3-512	conv3-512	conv3-512	conv3-512
	1111		conv1-512	conv3-512	conv3-512
					conv3-512
		max	pool		
conv3-512	conv3-512	conv3-512	conv3-512	conv3-512	conv3-512
conv3-512	conv3-512	conv3-512	conv3-512	conv3-512	conv3-512
			conv1-512	conv3-512	conv3-512
		,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,	111111111111111111111111111111111111111	1.1	conv3-512
	•	max	pool		•
		FC-	4096		
		FC-	4096		
		FC-	1000		
		soft-	-max		

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A	ll-CNN-C
3	× 3 conv. 96 ReLU
3	× 3 conv. 96 ReLU
3	× 3 conv. 96 ReLU
	with stride $r=2$
3	× 3 conv. 192 ReLU
3	× 3 conv. 192 ReLU
3	× 3 conv. 192 ReLU
	with stride $r=2$

## MobileNet

Type / Stride	Filter Shape	Input Size	
Conv / s2	$3 \times 3 \times 3 \times 32$	$224 \times 224 \times 3$	
Conv dw / s1	$3 \times 3 \times 32 \text{ dw}$	$112 \times 112 \times 32$	
Conv/s1	$1 \times 1 \times 32 \times 64$	$112 \times 112 \times 32$	
Conv dw / s2	$3 \times 3 \times 64 \text{ dw}$	$112 \times 112 \times 64$	
Conv / s1	$1 \times 1 \times 64 \times 128$	$56 \times 56 \times 64$	
Conv dw / s1	$3 \times 3 \times 128 \text{ dw}$	$56 \times 56 \times 128$	
Conv/s1	$1 \times 1 \times 128 \times 128$	$56 \times 56 \times 128$	
Conv dw / s2	$3 \times 3 \times 128 \text{ dw}$	$56 \times 56 \times 128$	
Conv/s1	$1 \times 1 \times 128 \times 256$	$28 \times 28 \times 128$	
Conv dw / s1	$3 \times 3 \times 256 \text{ dw}$	$28 \times 28 \times 256$	
Conv/s1	$1 \times 1 \times 256 \times 256$	$28 \times 28 \times 256$	
Conv dw / s2	$3 \times 3 \times 256 \text{ dw}$	$28 \times 28 \times 256$	
Conv/s1	$1 \times 1 \times 256 \times 512$	$14 \times 14 \times 256$	
5× Conv dw / s1	$3 \times 3 \times 512 \text{ dw}$	$14 \times 14 \times 512$	
Conv/s1	$1 \times 1 \times 512 \times 512$	$14 \times 14 \times 512$	
Conv dw / s2	$3 \times 3 \times 512 \text{ dw}$	$14 \times 14 \times 512$	
Conv / s1	$1 \times 1 \times 512 \times 1024$	$7 \times 7 \times 512$	
Conv dw / s2	$3 \times 3 \times 1024 \text{ dw}$	$7 \times 7 \times 1024$	
Conv/s1	$1 \times 1 \times 1024 \times 1024$	$7 \times 7 \times 1024$	
Avg Pool / s1	Pool 7 × 7	$7 \times 7 \times 1024$	
FC/s1	$1024 \times 1000$	$1 \times 1 \times 1024$	
Softmax / s1	Classifier	$1 \times 1 \times 1000$	

## ResNet

