



Proposed in 2012

#### ImageNet Classification with Deep Convolutional Neural Networks

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#### Abstract

We trained a large, deep convolutional neural network to classify the 1.2 million high-resolution images in the ImageNet LSVRC-2010 contest into the 1000 different classes. On the test data, we achieved top-1 and top-5 error rates of 37.5% and 17.0% which is considerably better than the previous state-of-the-art. The neural network, which has 60 million parameters and 650,000 neurons, consists of five convolutional layers, some of which are followed by max-pooling layers, and three fully-connected layers with a final 1000-way softmax. To make training faster, we used non-saturating neurons and a very efficient GPU implementation of the convolution operation. To reduce overfitting in the fully-connected layers we employed a recently-developed regularization method called "dropout" that proved to be very effective. We also entered a variant of this model in the ILSVRC-2012 competition and achieved a winning top-5 test error rate of 15.3%, compared to 26.2% achieved by the second-best entry.



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- It has 8 layers with learnable parameters





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- Architecture details: Analytics
  - o 5 convolutional layers with combination of max pooling



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  - 3 Fully connected layers



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- Also used dropouts
- Trained on imagenet dataset



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
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Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
Input	-	-	-	-	227 x 227 x 3	-





Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
Input	-	-	-	-	227 x 227 x 3	-
Conv 1	96	11 x 11	4	-	55 x 55 x 96	ReLU





Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
Input	-	-	-	-	227 x 227 x 3	-
Conv 1	96	11 x 11	4	-	55 x 55 x 96	ReLU
Max Pool 1	-	3 x 3	2		27 x 27 x 96	-





Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
Input	-	-	-	-	227 x 227 x 3	-
Conv 1	96	11 x 11	4	-	55 x 55 x 96	ReLU
Max Pool 1	-	3 x 3	2		27 x 27 x 96	-
Conv 2	256	5 x 5	naly	[[C25]	27 x 27 x 256	ReLU
		VV	/idhy	a		



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
Input	-	-	-	-	227 x 227 x 3	-
Conv 1	96	11 x 11	4	-	55 x 55 x 96	ReLU
Max Pool 1	-	3 x 3	2		27 x 27 x 96	-
Conv 2	256	5 x 5	naly		27 x 27 x 256	ReLU
Max Pool 2	- /	3 x 3	/ 2 /	a	13 x 13 x 256	-



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
Input	-	-	-	-	227 x 227 x 3	-
Conv 1	96	11 x 11	4	-	55 x 55 x 96	ReLU
Max Pool 1	-	3 x 3	2		27 x 27 x 96	-
Conv 2	256	5 x 5	analy	LL C25	27 x 27 x 256	ReLU
Max Pool 2	- /	3 x 3	/ic²hy	a	13 x 13 x 256	-
Conv 3	384	3 x 3	10119	1	13 x 13 x 384	ReLU



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
Input	-	-	-	-	227 x 227 x 3	-
Conv 1	96	11 x 11	4	-	55 x 55 x 96	ReLU
Max Pool 1	-	3 x 3	2		27 x 27 x 96	-
Conv 2	256	5 x 5	analy		27 x 27 x 256	ReLU
Max Pool 2	- /	3 x 3	/ic²hy	a	13 x 13 x 256	-
Conv 3	384	3 x 3	10119	1	13 x 13 x 384	ReLU
Conv 4	384	3 x 3	1	1	13 x 13 x 384	ReLU



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Input	-	-	-	-	227 x 227 x 3	-
Conv 1	96	11 x 11	4	-	55 x 55 x 96	ReLU
Max Pool 1	-	3 x 3	2		27 x 27 x 96	-
Conv 2	256	5 x 5	lnaly		27 x 27 x 256	ReLU
Max Pool 2	- /	3 x 3	/ic2hy	a	13 x 13 x 256	-
Conv 3	384	3 x 3	10119	1	13 x 13 x 384	ReLU
Conv 4	384	3 x 3	1	1	13 x 13 x 384	ReLU
Conv 5	256	3 x 3	1	1	13 x 13 x 256	ReLU



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
Input	-	-	-	-	227 x 227 x 3	-
Conv 1	96	11 x 11	4	-	55 x 55 x 96	ReLU
Max Pool 1	-	3 x 3	2		27 x 27 x 96	-
Conv 2	256	5 x 5	<b>Analy</b>	[[C25]	27 x 27 x 256	ReLU
Max Pool 2	- /	3 x 3	/id2hy	2	13 x 13 x 256	-
Conv 3	384	3 x 3	1	1	13 x 13 x 384	ReLU
Conv 4	384	3 x 3	1	1	13 x 13 x 384	ReLU
Conv 5	256	3 x 3	1	1	13 x 13 x 256	ReLU
Max Pool 3	-	3 x 3	2	-	6 x 6 x 256	-



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
Input	-	-	-	-	227 x 227 x 3	-
Conv 1	96	11 x 11	4	-	55 x 55 x 96	ReLU
Max Pool 1	-	3 x 3	2		27 x 27 x 96	-
Conv 2	256	5 x 5	analy		27 x 27 x 256	ReLU
Max Pool 2	- /	3 x 3	/idhy	a	13 x 13 x 256	-
Conv 3	384	3 x 3	10119	1	13 x 13 x 384	ReLU
Conv 4	384	3 x 3	1	1	13 x 13 x 384	ReLU
Conv 5	256	3 x 3	1	1	13 x 13 x 256	ReLU
Max Pool 3	-	3 x 3	2	-	6 x 6 x 256	-
Dropout 1	rate = 0.5	-	-	-	6 x 6 x 256	-



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
-	-	-	-	-	-	-
-	-	-	-	-	-	-
-	-				-	-
Dropout 1	rate = 0.5	7 /-	lnaly	tics	6 x 6 x 256	-
		$\vee$ $\vee$	/idhy	а		



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
-	-	-	-	-	-	-
-	-	-	-	-	-	-
-	-		-		-	-
Dropout 1	rate = 0.5	7 /-	lnaly	tics	6 x 6 x 256	-
Fully Connected 1	- /	V - V	/idhv	a -	4096	ReLU



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
-	-	-	-	-	-	-
-	-	-	-	-	-	-
-	-		-		-	-
Dropout 1	rate = 0.5	// /-	Inaly	tics	6 x 6 x 256	-
Fully Connected 1	- /	V - 1	/idhv	a	4096	ReLU
Dropout 2	rate = 0.5	-	icity		4096	-



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
-	-	-	-	-	-	-
-	-	-	-	-	-	-
-	-				-	-
Dropout 1	rate = 0.5	/\ /-	analy	tics	6 x 6 x 256	-
Fully Connected 1	_ /	V · 1	/idhv	a	4096	ReLU
Dropout 2	rate = 0.5	-	icity		4096	-
Fully Connected 2	-	-	-	-	4096	ReLU



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
-	-	-	-	-	-	-
-	-	-	-	-	-	-
-	-				-	-
Dropout 1	rate = 0.5	7 /-	lnaly	tics	6 x 6 x 256	-
Fully Connected 1	- /	V- V	/idhv	a	4096	ReLU
Dropout 2	rate = 0.5	-	idity	-	4096	-
Fully Connected 2	-	-	-	-	4096	ReLU
Fully Connected 3	-	-	-	-	1000	Softmax



Layer	# filters / neurons	Filter size	Stride	Padding	Size of feature map	Activation function
-	-	-	-	-	-	-
-	-	-	-	-	-	-
-	-		- -		-	-
Dropout 1	rate = 0.5	// /-	Inaly	tics	6 x 6 x 256	-
Fully Connected 1	- /	V · \	/idhv	a	4096	ReLU
Dropout 2	rate = 0.5	-	icity	-	4096	-
Fully Connected 2	-	-	-	-	4096	ReLU
Fully Connected 3	-	-	-	-	1000	Softmax

Number of Parameters = 62.3 million







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- Takes RGB image as input





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  - 3 Fully connected layers
  - **ReLU** activation function
  - Dropouts
  - Softmax classifier



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- Takes RGB image as input
- Architecture details:
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  - **ReLU** activation function
  - Dropouts
  - Softmax classifier
- Total number of parameters: 62.3 million





