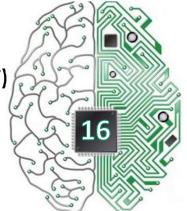
Open Elective Course [OE]

Course Code: CSO507 Winter 2023-24

Lecture#

Deep Learning

Unit-4: Convolutional Neural Networks (Part-IV)



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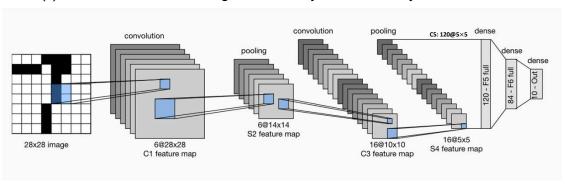
ConvNets: Popular Architectures

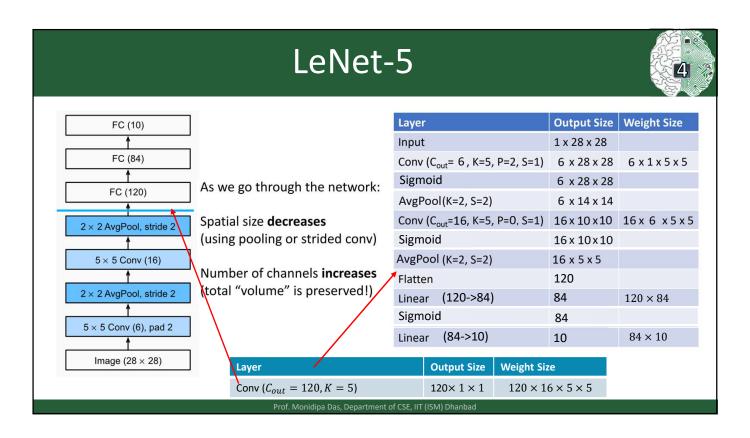
LeNet-5



LeCun et al. in 1998

- At a high level, LeNet (LeNet-5) consists of two parts:
 - (i) a convolutional encoder consisting of two convolutional layers
 - (ii) a dense block consisting of three fully connected layers.

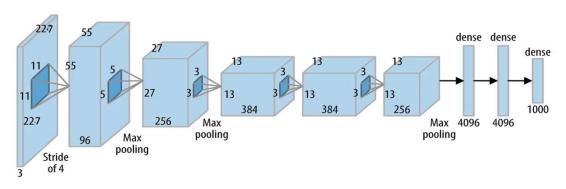




AlexNet



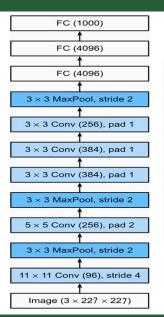
- ImageNet LSVRC-2012 [Krizhevsky et al.]
- The heart of AlexNet is its architecture, comprising of five Convolutional (CONV) layers and three Fully Connected (FC) layers.
- AlexNet opts for the Rectified Linear Unit (ReLU) as its activation function



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<u>AlexNet</u>



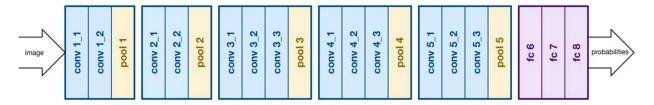


AlexNet Network - Structural Details													
Input		Output			Layer	Stride	Pad	Kerne	el size	in out		# of Param	
227	227	3	55	55	96	conv1	4	0	11	11	3	96	34944
55	55	96	27	27	96	maxpool1	2	0	3	3	96	96	0
27	27	96	27	27	256	conv2	1	2	5	5	96	256	614656
27	27	256	13	13	256	maxpool2	2	0	3	3	256	256	0
13	13	256	13	13	384	conv3	1	1	3	3	256	384	885120
13	13	384	13	13	384	conv4	1	1	3	3	384	384	1327488
13	13	384	13	13	256	conv5	1	1	3	3	384	256	884992
13	13	256	6	6	256	maxpool5	2	0	3	3	256	256	0
	2.					fc6			1	1	9216	4096	37752832
						fc7			1	1	4096	4096	16781312
fc8 1 1 4096 1000 409700										4097000			
	Total 62,378,344												62,378,344

VGGNet



- 16 layers (not including pooling)
- 3x3 convolutions only
- 138 million parameters
- Pre-trained model is available for download and is a standard in the field



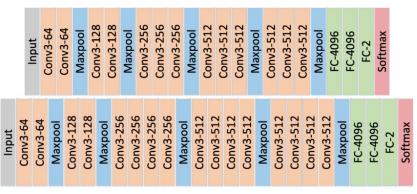
Simonyan & Zisserman (2014)

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VGGNet

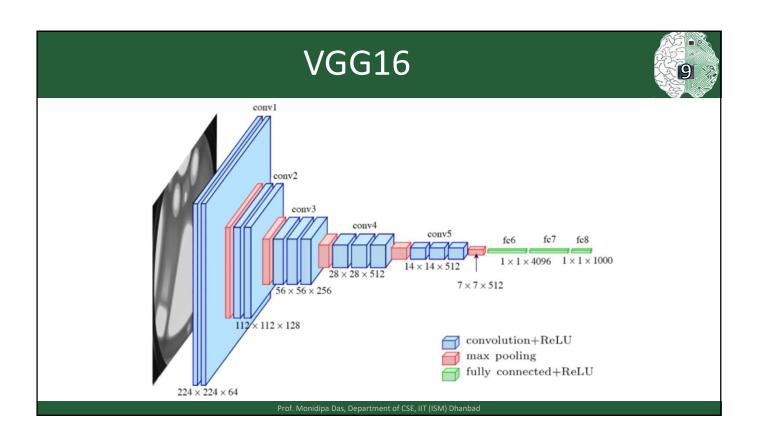


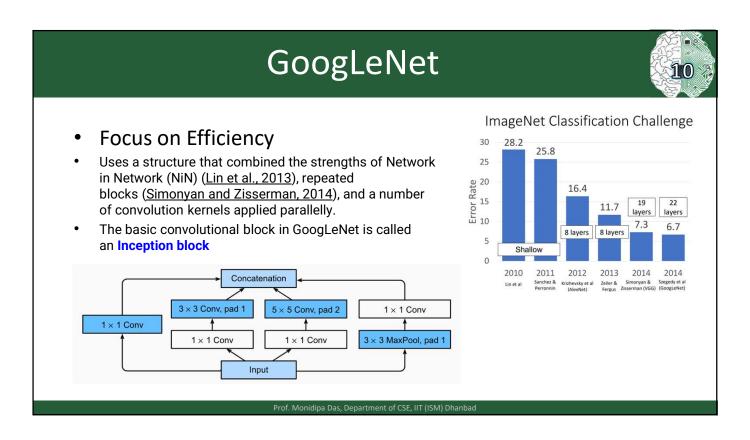
 There exist multiple versions of VGGNet such as VGG16, VGG19, etc., differentiated by the total number of layers in the network.



Network structures of VGG16 (top) and VGG19 (bottom)

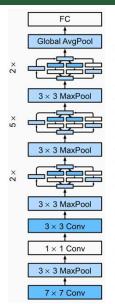






GoogLeNet



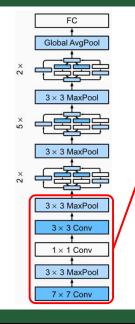


type	patch size/ stride	output size	depth	#1×1	#3×3 reduce	#3×3	#5×5 reduce	#5×5	pool proj	params	ops
convolution	7×7/2	112×112×64	1							2.7K	34M
max pool	3×3/2	56×56×64	0								
convolution	3×3/1	56×56×192	2		64	192				112K	360M
max pool	3×3/2	28×28×192	0								
inception (3a)		28×28×256	2	64	96	128	16	32	32	159K	128M
inception (3b)		28×28×480	2	128	128	192	32	96	64	380K	304M
max pool	3×3/2	14×14×480	0								
inception (4a)		14×14×512	2	192	96	208	16	48	64	364K	73M
inception (4b)		14×14×512	2	160	112	224	24	64	64	437K	88M
inception (4c)		14×14×512	2	128	128	256	24	64	64	463K	100M
inception (4d)		14×14×528	2	112	144	288	32	64	64	580K	119M
inception (4e)		14×14×832	2	256	160	320	32	128	128	840K	170M
max pool	3×3/2	7×7×832	0								
inception (5a)		7×7×832	2	256	160	320	32	128	128	1072K	54M
inception (5b)		7×7×1024	2	384	192	384	48	128	128	1388K	71M
avg pool	7×7/1	1×1×1024	0								
dropout (40%)		1×1×1024	0								
linear		1×1×1000	1							1000K	1M
softmax		1×1×1000	0								

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GoogLeNet





	1	npu	t size		Laye	er	Output size		t size				
Layer	C	ı	1 / W	filters	kernel	stride	pad	С	ŀ	H/W	memory (KB)	params (K)	flop (M)
conv		3	224	64	. 7	2	3	ϵ	54	112	3136	9	11
max-pool		64	112		3	2	1	ϵ	54	56	784	C)
conv		64	56	64	. 1	1	C	(54	56	784	4	1
conv		64	56	192	3	1	1	19	92	56	2352	111	34
max-pool	1	192	56		3	2	1	19	92	28	588	C)

Total from 224 to 28 spatial resolution:

Memory: 7.5 MB

Params: 124K

Compare VGG-16:

Memory: 42.9 MB (5.7x)

Params: 1.1M (8.9x)

