

CS-F441: SELECTED TOPICS FROM COMPUTER SCIENCE (DEEP LEARNING FOR NLP & CV)

Lecture-KT-12: CNN Architectures, LeNet, AlexNet, ZFNet, VGG



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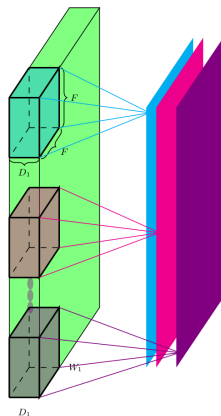
Nov 11, 2019

(Campus @ BITS-Pilani July-Dec 2019)

Recap: Motivation

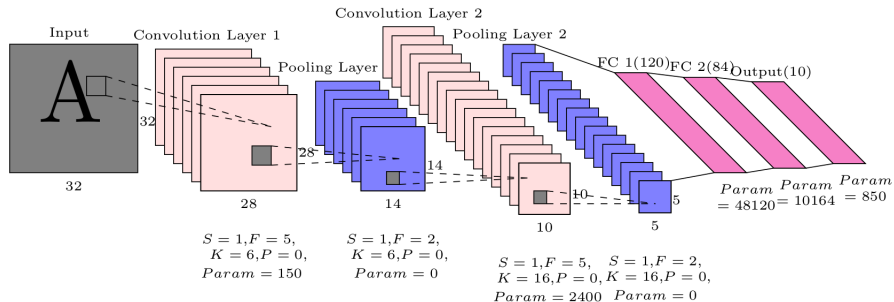
CNN are sparse FC-NN with **weight sharing**

- Multiple filters to get multiple feature maps
- Filter/Kernels are treated as parameters to learning them
- Filters, Padding, Stride and Pooling
- Classification pipeline has components



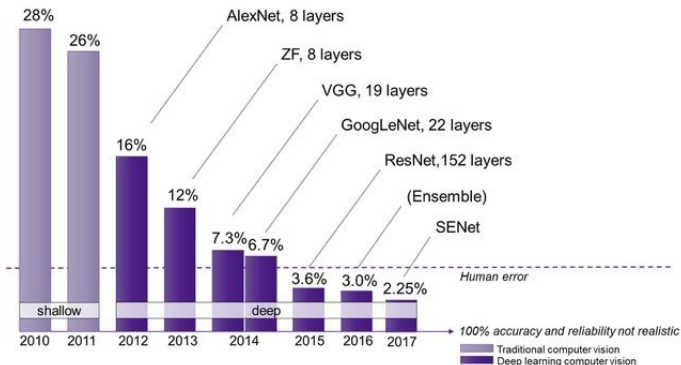
$$H, W = \frac{H - F + 2P}{S} + 1, \frac{S - F + 2P}{S} + 1$$

LeNet-5 for handwritten character recognition ¹



¹Yann Lecun and Leon Bottou and Yoshua Bengio and Patrick Haffner, *Gradient-based learning applied to document recognition*, pp 2278–2324, IEEE-1998

ImageNet ILSVRC²



- (2009) 22K category, 14M images
- Challenge 1000 class, 1431167 images
- HoG, LBP, SVM ...

²Imagenet large scale visual recognition challenge <http://www.image-net.org/challenges/LSVRC/>

AlexNet³

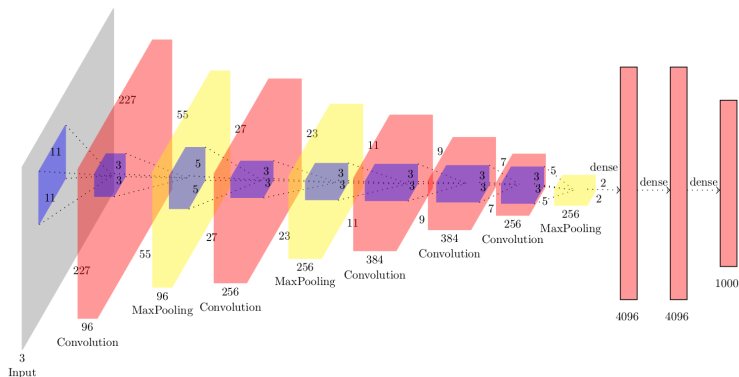


Image size $227 \times 227 \times 3 \rightarrow [69 \text{ F}=11, \text{ S}=4, \text{ P}=0] \rightarrow [\text{MaxPool F}=3, \text{ S}=2] \rightarrow [256 \text{ F}=5, \text{ S}=1, \text{ P}=0] \rightarrow [\text{MaxPool F}=3, \text{ S}=2] \rightarrow [384 \text{ F}=3, \text{ S}=1, \text{ P}=0] \rightarrow [384 \text{ F}=3, \text{ S}=1, \text{ P}=0] \rightarrow [256 \text{ F}=3, \text{ S}=1, \text{ P}=0] \rightarrow [\text{MaxPool F}=3, \text{ S}=2] \rightarrow \text{FC } 4096 \rightarrow \text{FC } 4096 \rightarrow \text{FC } 1000$

27.55 M parameters. Convolution is passes through ReLU activation.

³Cite: 50086, **Imagenet classification with deep convolutional neural networks**, Krizhevsky, Alex and Sutskever, Ilya and Hinton, Geoffrey E, In: Advances in neural information processing systems pages 1097–1105, NIPS-2012

ZFNet⁴

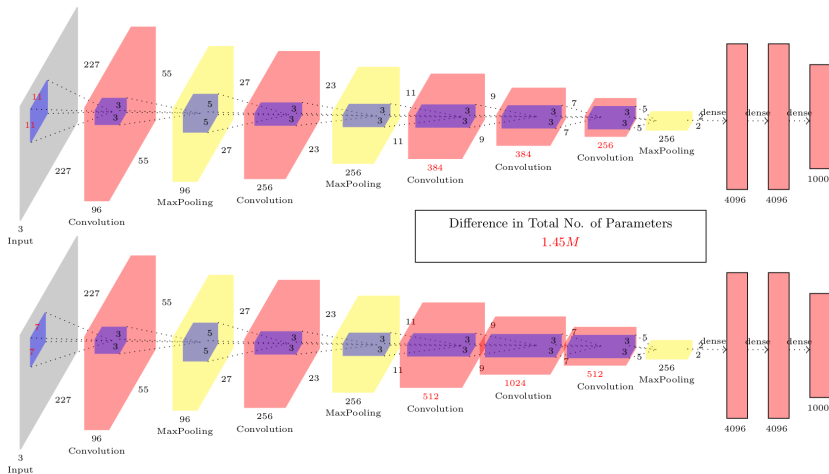
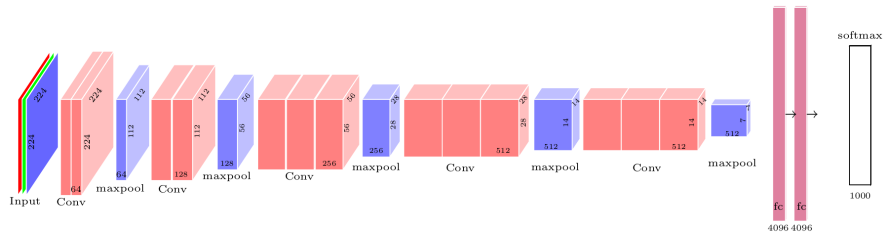


Image size $227 \times 227 \times 3 \rightarrow [69 \text{ F}=7, \text{ S}=4, \text{ P}=0] \rightarrow [\text{MaxPool F}=3, \text{ S}=2] \rightarrow [256 \text{ F}=5, \text{ S}=1, \text{ P}=0] \rightarrow [\text{MaxPool F}=3, \text{ S}=2] \rightarrow [512 \text{ F}=3, \text{ S}=1, \text{ P}=0] \rightarrow [1024 \text{ F}=3, \text{ S}=1, \text{ P}=0] \rightarrow [512 \text{ F}=3, \text{ S}=1, \text{ P}=0] \rightarrow [\text{MaxPool F}=3, \text{ S}=2] \rightarrow \text{FC } 4096 \rightarrow \text{FC } 1000$

⁴ Cite 7560 **Visualizing and understanding convolutional networks**, Zeiler, Matthew D and Fergus, Rob, pages 818–833, European conference on computer vision (ECCV) Springer-2014

VGG16⁵



- 1 Kernel size is always 3×3
- 2 16M parameters in pre-FC and 122 in FC. First FC layer is huge
- 3 Layers represents abstract representation and can be reused (FC or Conv)

⁵ Cite 29548 **Very deep convolutional networks for large-scale image recognition**, Simonyan, Karen and Zisserman, Andrew, pages 818–833, arXiv preprint arXiv(2014):1409.1556 - In: ICLR-2015

Thank You!

Thank you very much for your attention⁶!

Queries ?

⁶Credit: Prof. Mitesh Khapra, Deep Learning(CS7015): Lec 11.4 CNNs for NPTEL 