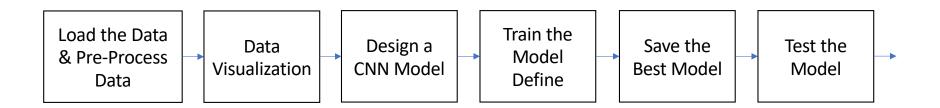
Probabilistic Machine Learning and Al

Outline of the lecture

This lecture introduces you to the fascinating subject of classification and regression with convolutional neural networks.

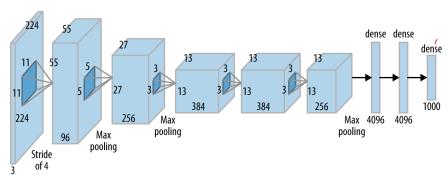
- Different CNN Models and Datasets < http://www.image-net.org/challenges/LSVRC/ >
 - AlexNet
 - VGGNet
 - ResNet
- Transfer Learning
- Visualizing Convolution Neural Network

Process



AlexNet

AlexNet is the first deep architecture and was introduced by Geoffrey Hinton and his colleagues. It is a very simple layout but a substantial network architecture that consist of convolution and pooling layers placed one by one which is completely connected at the top. One of the most remarkable features of this architecture is the pace at which it performs various tasks; it has the ability to speed up the training by 10 times through GPU. The network which was given by Geoffrey could be used for classification with 1000 possible categories. It also uses ReLU for nonlinearity functions and data augmentation techniques which comprise of various reflections, patch extractions and image translations. It also applies drop out layers to overcome the overfitting training data problem. Though presently we have more updated architecture but AlexNet is still applied for the deep neural network for different tasks.

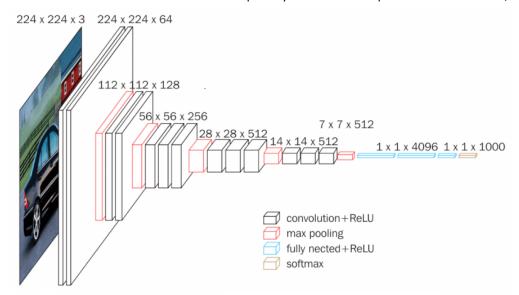


| | AlexNet Network - Structural Details | | | | | | | | | | | | |
|-------|--------------------------------------|-----|----|-----|-----|----------|--------|-----|-------|---------|------------|------|------------|
| | Input | | | utp | out | Layer | Stride | Pad | Kerne | el size | in | out | # of 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 |
| | fc6 1 1 9216 4096 | | | | | | | | | | 37752832 | | |
| | | | | | | fc7 | | | 1 | 1 | 4096 | 4096 | 16781312 |
| fc8 | | | | | | | | | 1 | 1 | 4096 | 1000 | 4097000 |
| Total | | | | | | | | | | | 62,378,344 | | |

https://papers.nips.cc/paper/4824-imagenet-classification-with-deep-convolutional-neural-networks.pdf

VGG Net

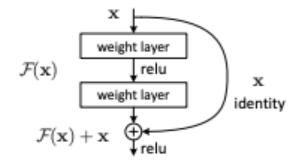
The Researchers at Visual Graphics Group at Oxford gave the VGC Net architecture. The network is basically of pyramidal shape with deep top layers and the bottom layers are placed closer to the image are wide. It is one of the advisable architecture for benchmarking on a particular task. It consists of convolutional layer preceded by pooling layer due to which the layers are narrower. Pre-trained networks of this particular architecture are easily available on the internet due to which it is highly used. Though it is slow to train, if training needs to be followed from a very basic level. It works considerably well on both image classification and localization tasks. The number of filters doubles after each maxpool layer due to which spatial dimension shrinks, but grows in terms of depth.



| VGG16 - Structural Details | | | | | | | | | | | | | |
|----------------------------|-------------|-----|-------|--------|-----|------|-----------|--------|--------|-------------|-------|------|-----------|
| # | Input Image | | | output | | | Layer | Stride | Kernel | | in | out | Param |
| 1 | 224 | 224 | 3 | 224 | 224 | 64 | conv3-64 | 1 | 3 | 3 | 3 | 64 | 1792 |
| 2 | 224 | 224 | 64 | 224 | 224 | 64 | conv3064 | 1 | 3 | 3 | 64 | 64 | 36928 |
| | 224 | 224 | 64 | 112 | 112 | 64 | maxpool | 2 | 2 | 2 | 64 | 64 | 0 |
| 3 | 112 | 112 | 64 | 112 | 112 | 128 | conv3-128 | 1 | 3 | 3 | 64 | 128 | 73856 |
| 4 | 112 | 112 | 128 | 112 | 112 | 128 | conv3-128 | 1 | 3 | 3 | 128 | 128 | 147584 |
| | 112 | 112 | 128 | 56 | 56 | 128 | maxpool | 2 | 2 | 2 | 128 | 128 | 65664 |
| 5 | 56 | 56 | 128 | 56 | 56 | 256 | conv3-256 | 1 | 3 | 3 | 128 | 256 | 295168 |
| 6 | 56 | 56 | 256 | 56 | 56 | 256 | conv3-256 | 1 | 3 | 3 | 256 | 256 | 590080 |
| 7 | 56 | 56 | 256 | 56 | 56 | 256 | conv3-256 | 1 | 3 | 3 | 256 | 256 | 590080 |
| | 56 | 56 | 256 | 28 | 28 | 256 | maxpool | 2 | 2 | 2 | 256 | 256 | 0 |
| 8 | 28 | 28 | 256 | 28 | 28 | 512 | conv3-512 | 1 | 3 | 3 | 256 | 512 | 1180160 |
| 9 | 28 | 28 | 512 | 28 | 28 | 512 | conv3-512 | 1 | 3 | 3 | 512 | 512 | 2359808 |
| 10 | 28 | 28 | 512 | 28 | 28 | 512 | conv3-512 | 1 | 3 | 3 | 512 | 512 | 2359808 |
| | 28 | 28 | 512 | 14 | 14 | 512 | maxpool | 2 | 2 | 2 | 512 | 512 | 0 |
| 11 | 14 | 14 | 512 | 14 | 14 | 512 | conv3-512 | 1 | 3 | 3 | 512 | 512 | 2359808 |
| 12 | 14 | 14 | 512 | 14 | 14 | 512 | conv3-512 | 1 | 3 | 3 | 512 | 512 | 2359808 |
| 13 | 14 | 14 | 512 | 14 | 14 | 512 | conv3-512 | 1 | 3 | 3 | 512 | 512 | 2359808 |
| | 14 | 14 | 512 | 7 | 7 | 512 | maxpool | 2 | 2 | 2 | 512 | 512 | 0 |
| 14 | 1 | 1 | 25088 | 1 | 1 | | | | 1 | 1 | 25088 | 4096 | 102764544 |
| 15 | 1 | 1 | 4096 | 1 | 1 | 4096 | | | 1 | 1 | 4096 | 4096 | 16781312 |
| 16 | 1 | 1 | 4096 | 1 | 1 | 1000 | fc | | 1 | 1 | 4096 | 1000 | 4097000 |
| Total | | | | | | | | | | 138,423,208 | | | |

Simonyan, Karen, and Andrew Zisserman. "Very deep convolutional networks for large-scale image recognition." *arXiv preprint arXiv:1409.1556* (2014).

Residual Networks



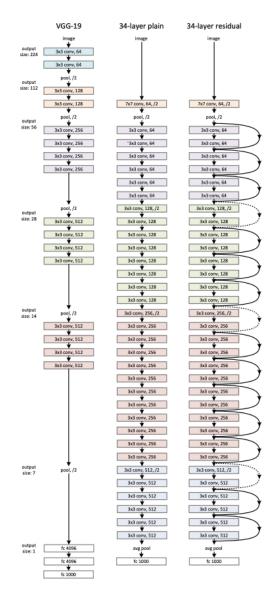
Practice assignment:

Please read the following link:

http://neuralnetworksanddeeplearning.com/chap5.html

and write 1 page on vanishing gradient

https://arxiv.org/pdf/1512.03385v1.pdf



CNN Benchmarks

https://github.com/jcjohnson/cnn-benchmarks

Transfer Learning

Transfer learning involves taking a **pre-trained neural network** and adapting the neural network to a new, different data set. Depending on both:

- The size of the new data set, and
- The similarity of the new data set to the original data set

Transfer Learning

The approach for using transfer learning will be different. There are four main cases:

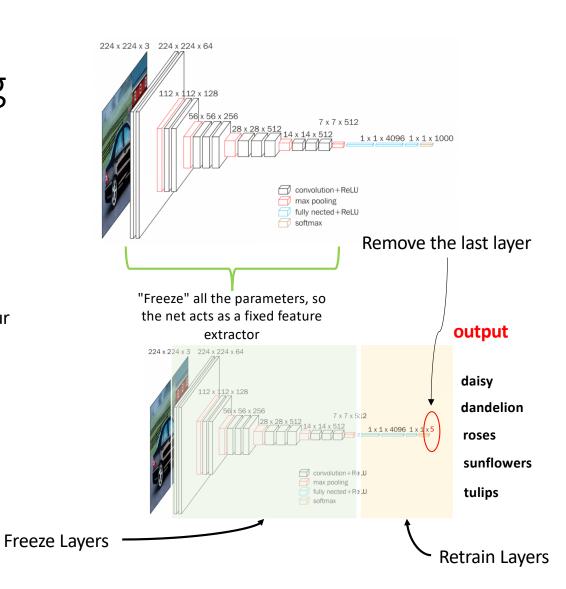
- New data set is small, new data is similar to original training data.
- New data set is small, new data is different from original training data.
- New data set is large, new data is similar to original training data.
- New data set is large, new data is different from original training data.

More to read: https://cs231n.github.io/transfer-learning/#tf

Ex. Transfer Learning

Sample Transfer Learning Steps:

- Load in a pre-trained VGG16 model
- "Freeze" all the parameters, so the net acts as a fixed feature extractor
- Remove the last layer
- Replace the last layer with a linear classifier of our own
- Train FC layers using the data set



Assignment: Visualizing and Understanding Convolutional Networks

Please read the following Paper by Zeiler and Fergus (NYU)

https://cs.nyu.edu/~fergus/papers/zeilerECCV2014.pdf

Here is Matt Zeiler's presentation

https://www.youtube.com/watch?v=ghEmQSxT6tw

Code: https://github.com/FHainzl/Visualizing Understanding CNN Implementation

Assignment 3: Implement this paper in Pytorch