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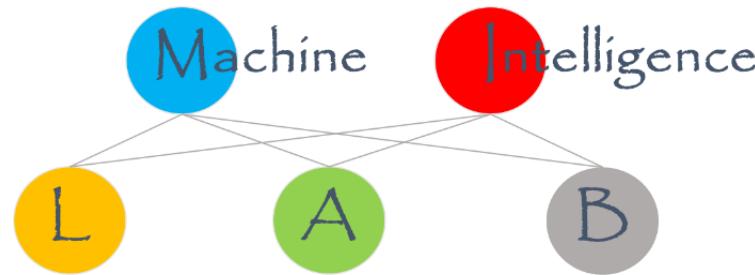
# Visual Recognition – Part 2

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Several slides are adapted from related courses or tutorials.  
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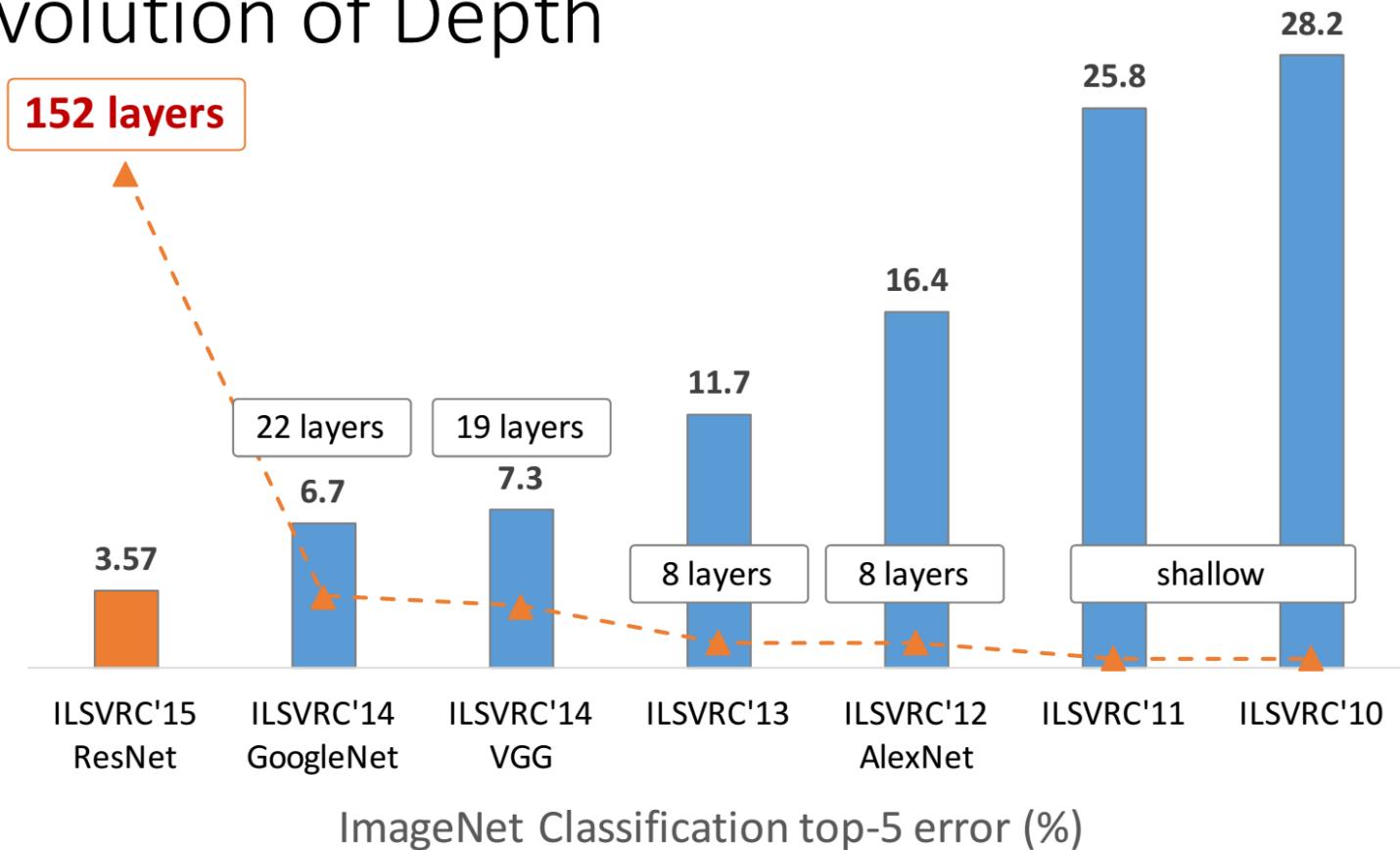
# Outline

- Why DL suddenly works? (AlexNet, 2012)
- Can it go deeper? (ResNet, 2015)
- Further extensions (DenseNet etc.)

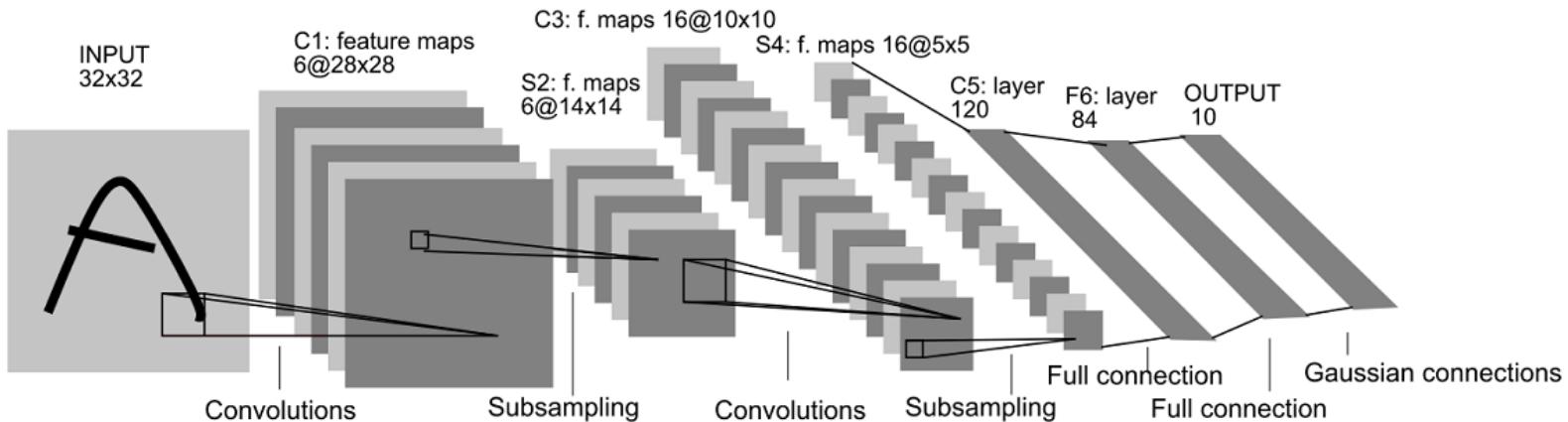
# AlexNet

- Named after Alex Krizhevsky, proposed in 2012

## Revolution of Depth



# LeNet-5

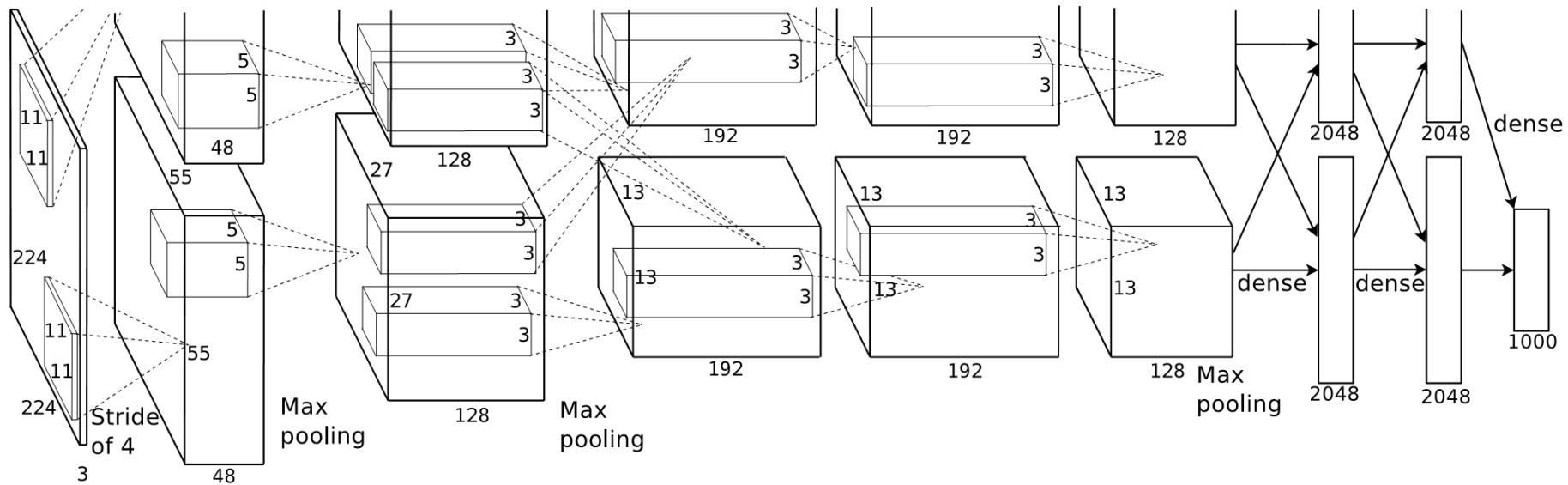


- Input: 32x32 pixel image. Largest character is 20x20  
(All important info should be in the center of the receptive field of the highest level feature detectors)
- Cx: Convolutional layer
- Sx: Subsample layer
- Fx: Fully connected layer
- Black and White pixel values are normalized:  
E.g. White = -0.1, Black = 1.175 (Mean of pixels = 0, Std of pixels = 1)

**Y. LeCun, L. Bottou, Y. Bengio, and P. Haffner. Gradient-based learning applied to document recognition.  
Proceedings of the IEEE, november 1998.**

# AlexNet

- Much larger than LeNet-5
- Trained on two GTX 580 GPU
- Largest networks at its time
- Utilize multiple engineering tricks (dropout, ReLU)



Alex Krizhevsky et al., ImageNet Classification with Deep Convolutional Neural Networks, NIPS 2012

## Why DL Suddenly works?

*...It may be that the primary barriers to the success of neural networks were psychological (practitioners did not expect neural networks to work, so they did not make a serious effort to use neural networks)...*

-- Goodfellow et al. “deep Learning”

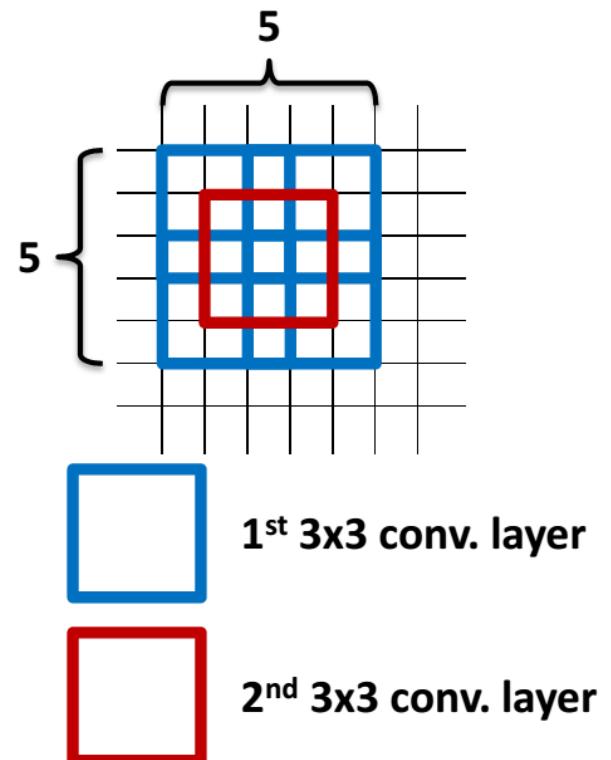
# Why DL Suddenly works? – My Two Cents

- Emerging of big visual data
- GPU -> large network
- New engineering tricks (dropout, ReLU etc.)

# VGG Net

Why 3x3 layers?

- Stacked conv. layers have a large receptive field
  - two 3x3 layers – 5x5 receptive field
  - three 3x3 layers – 7x7 receptive field
- More non-linearity
- Less parameters to learn
  - ~140M per net



image

# Network Design

conv-64

conv-64

maxpool

conv-128

conv-128

maxpool

conv-256

conv-256

maxpool

conv-512

conv-512

maxpool

conv-512

conv-512

maxpool

FC-4096

FC-4096

FC-1000

softmax

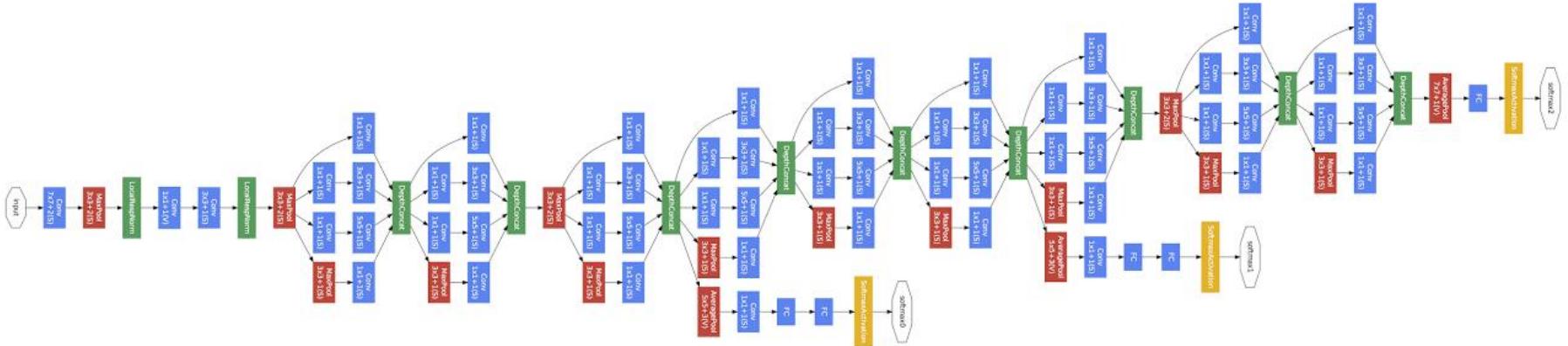
## Key design choices:

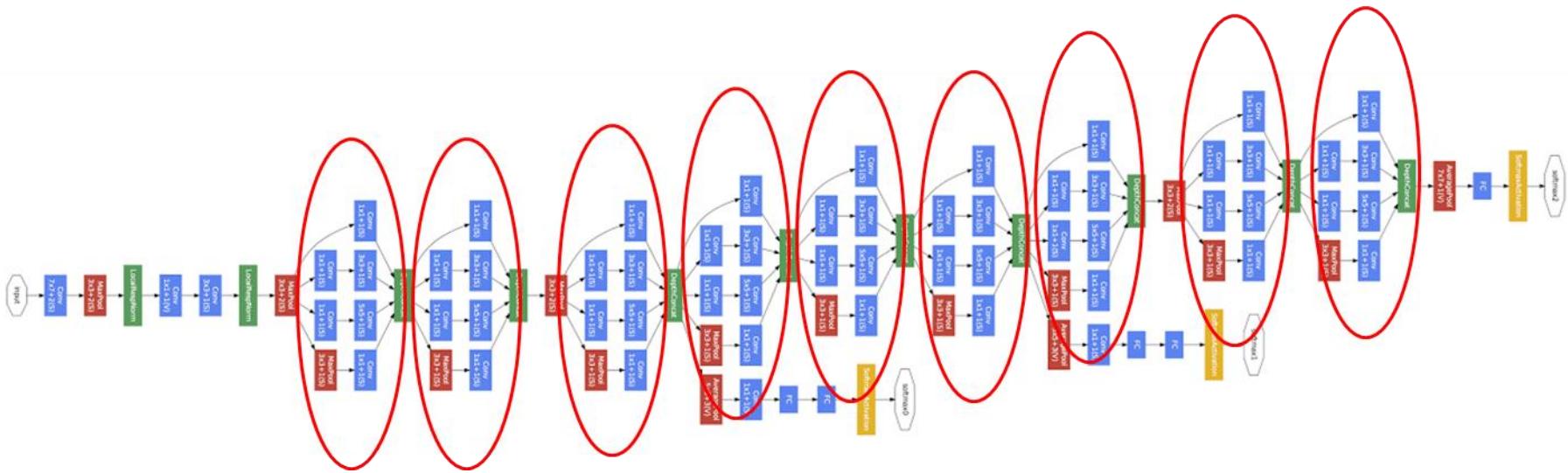
- 3x3 conv. kernels – very small
- conv. stride 1 – no loss of information

## Other details:

- Rectification (ReLU) non-linearity
- 5 max-pool layers (x2 reduction)
- no normalisation
- 3 fully-connected (FC) layers

# GoogLeNet



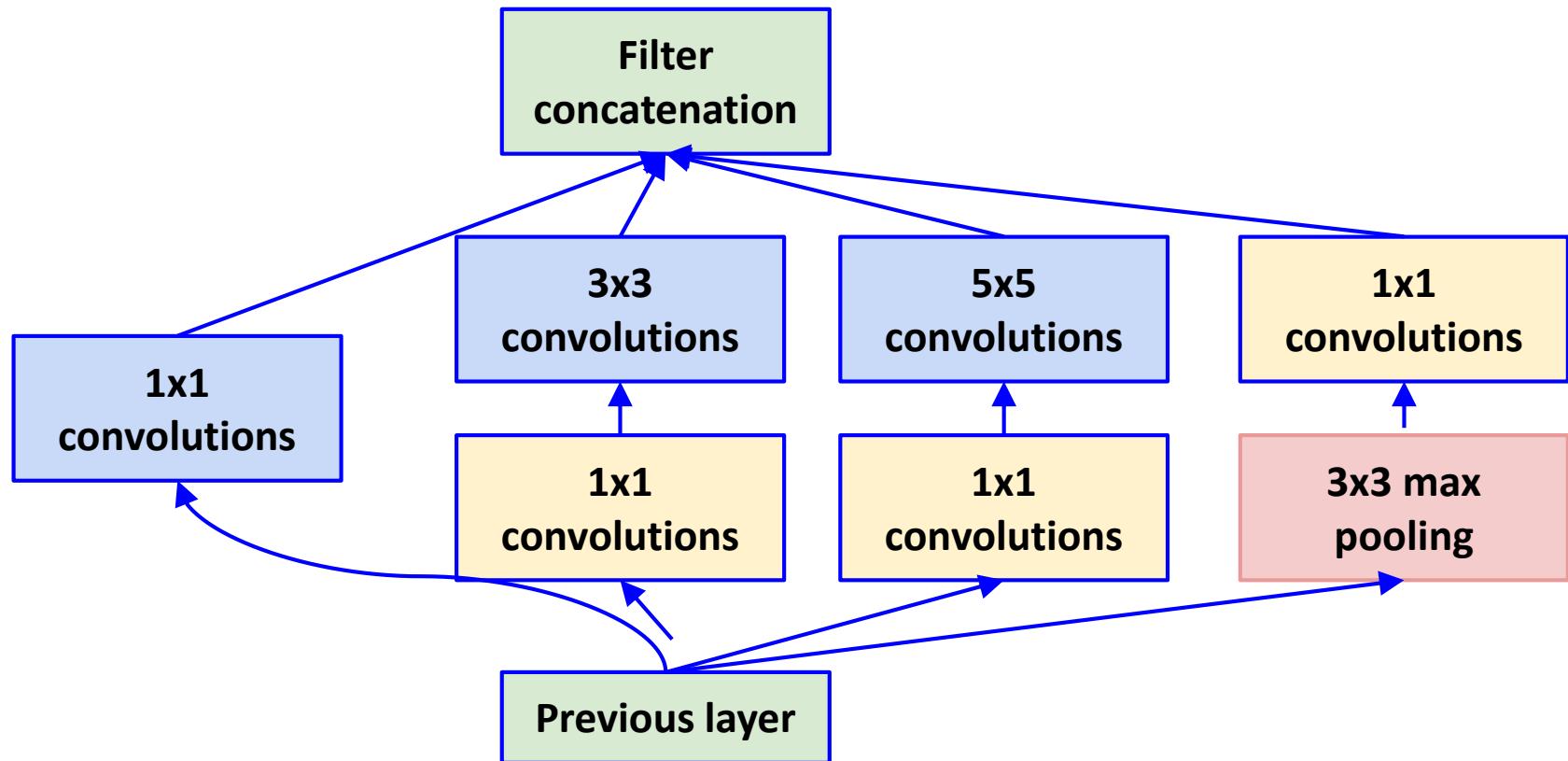


# Inception

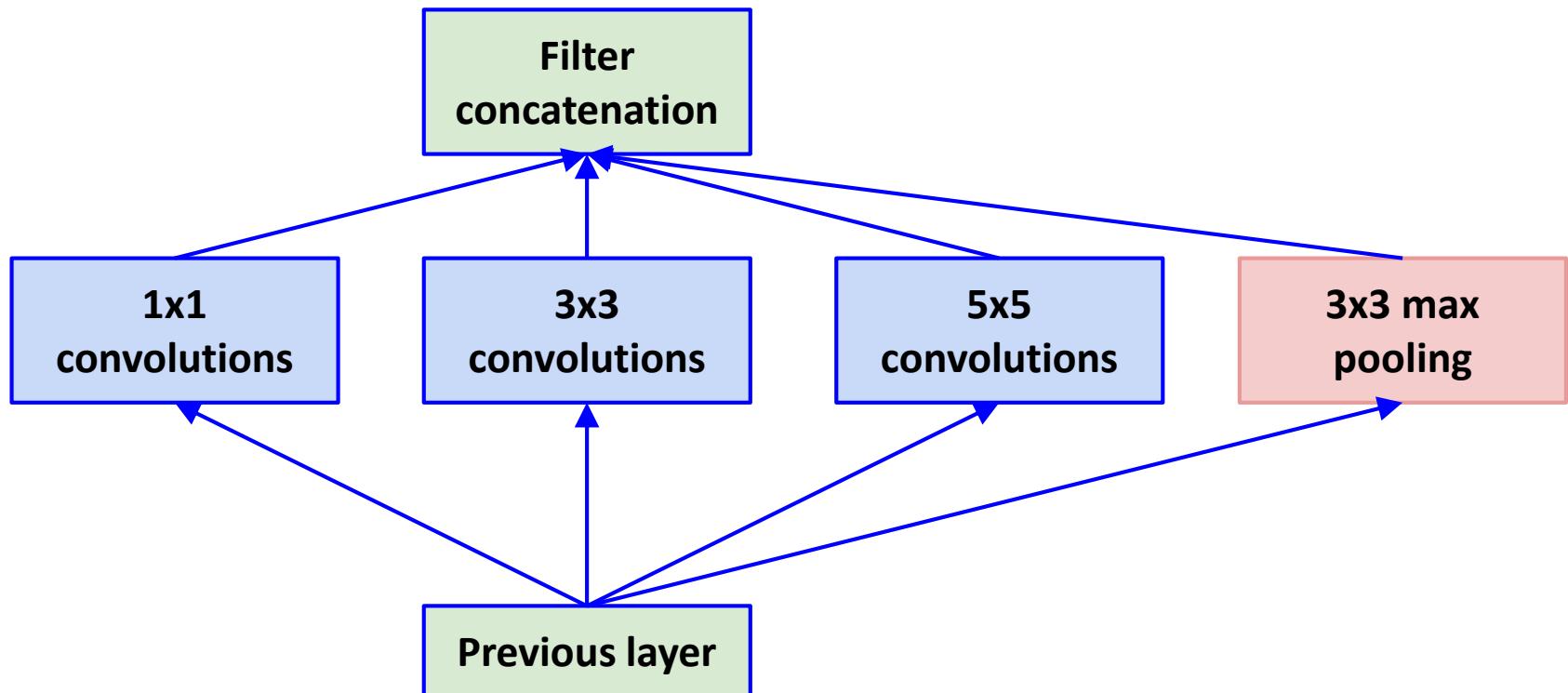
## Network in a network in a network...

**Convolution**  
**Pooling**  
**Softmax**  
**Other**

# Inception module



# Naive idea (does not work!)



# ResNet

- See He Kaiming's ICML tutorial

# DenseNet

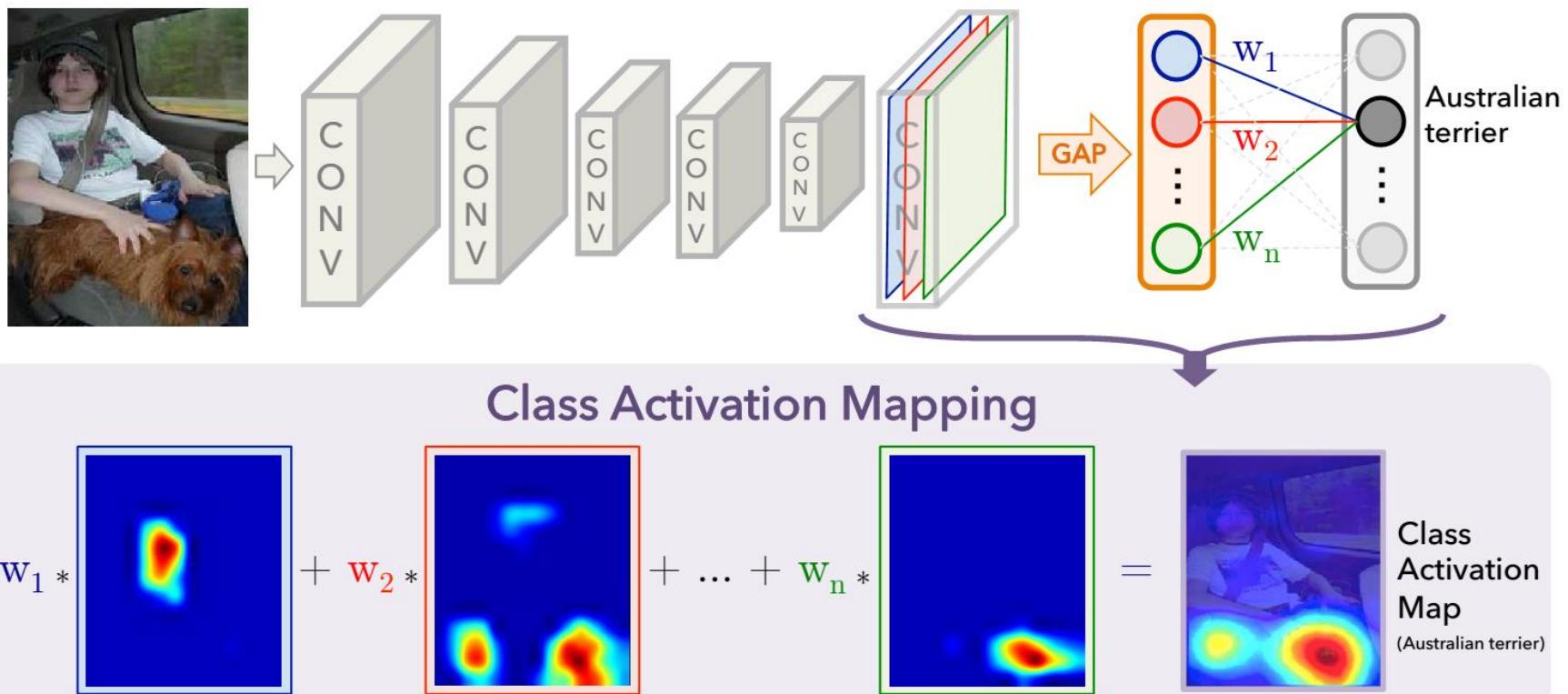
- See DenseNet's CVPR slides

## Dual Path Network

- See DualPathNet's CVPR slides

# Class Activation Map (CAM)

- #### ▪ Global Average Pooling



# Class Activation Map (CAM)

- Examples

