



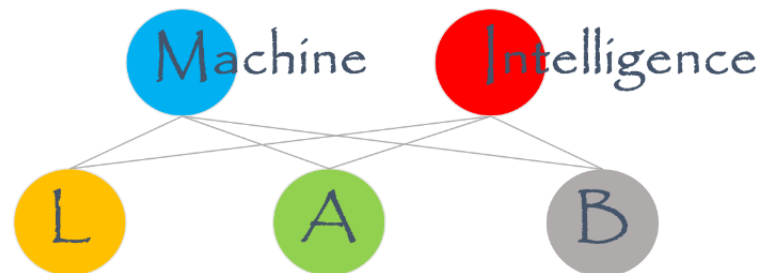
Visual Recognition – Part 2

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Several slides are adapted from related courses or tutorials.
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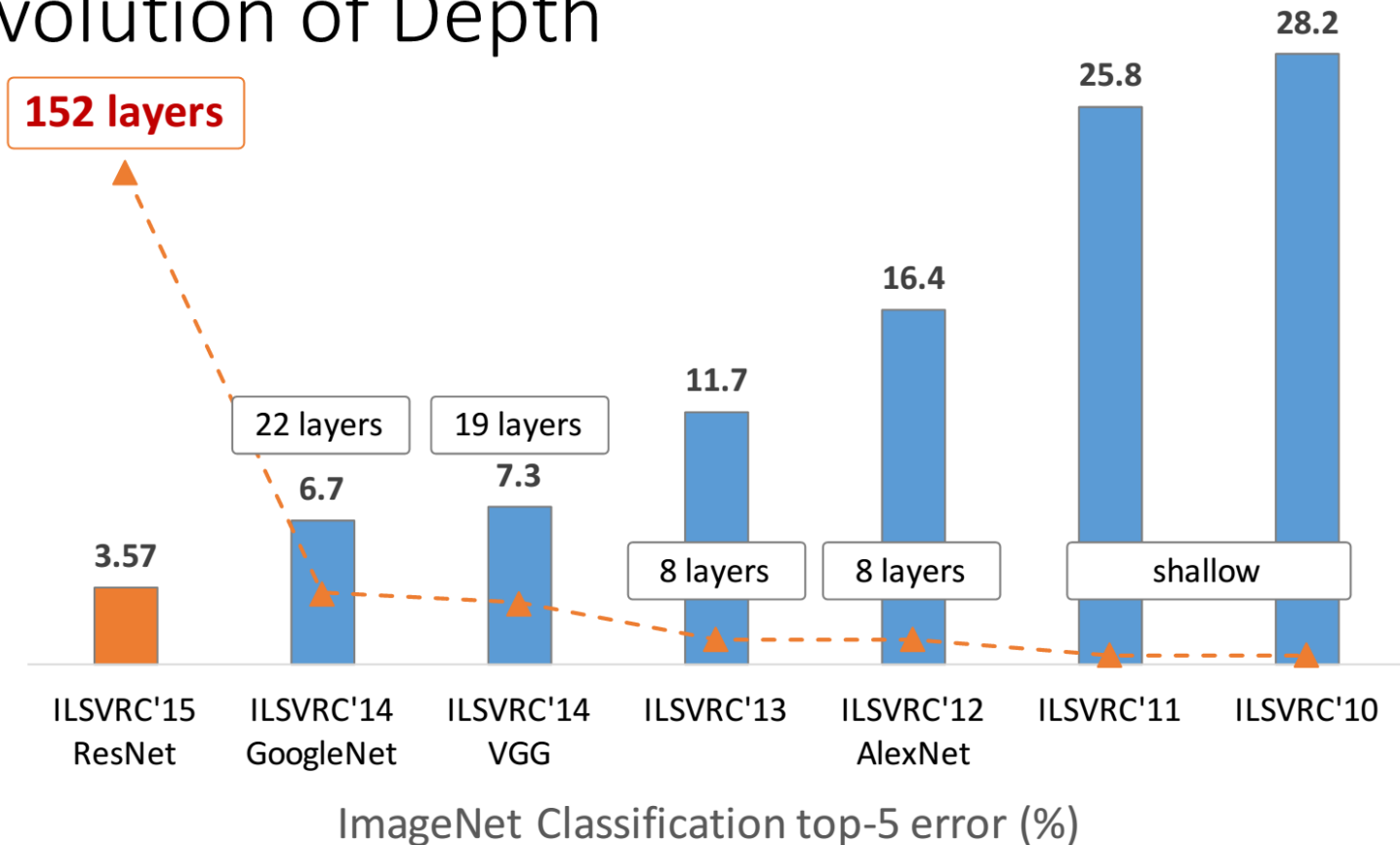
Outline

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

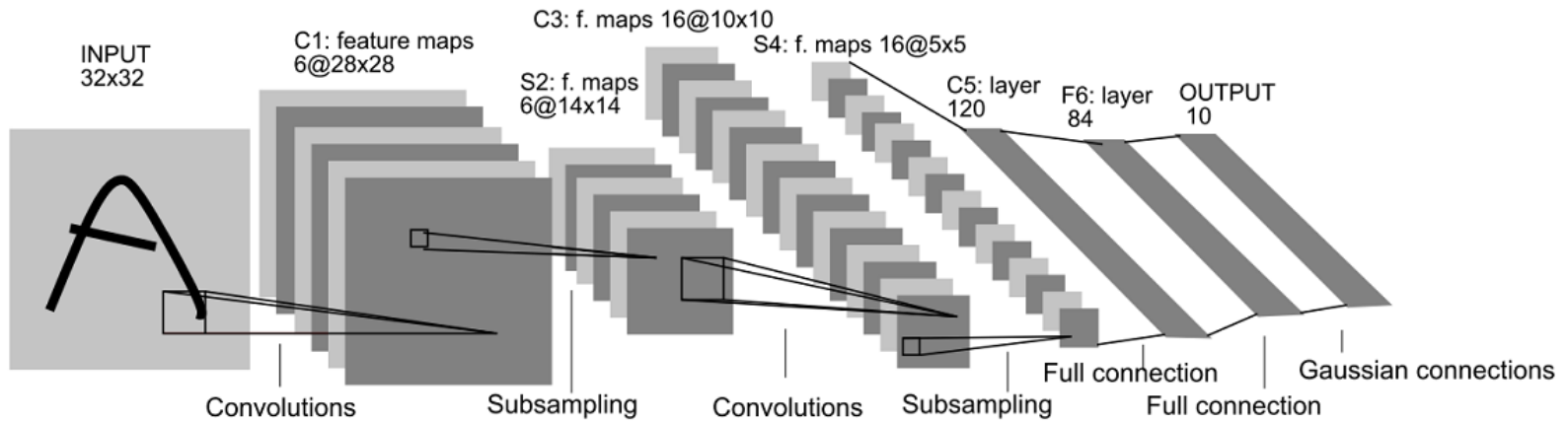
AlexNet

- Named after Alex Krizhevsky, proposed in 2012

Revolution of Depth



LeNet-5

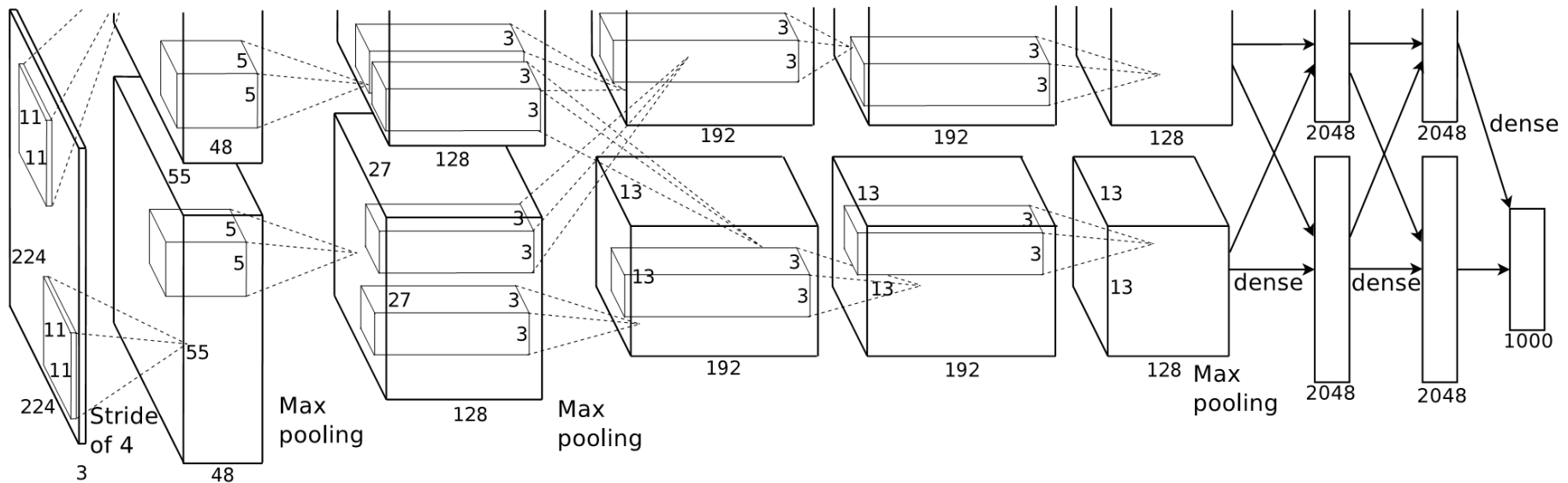


- Input: 32x32 pixel image. Largest character is 20x20
(All important info should be in the center of the receptive field of the highest level feature detectors)
- Cx: Convolutional layer
- Sx: Subsample layer
- Fx: Fully connected layer
- Black and White pixel values are normalized:
E.g. White = -0.1, Black = 1.175 (Mean of pixels = 0, Std of pixels = 1)
14

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"

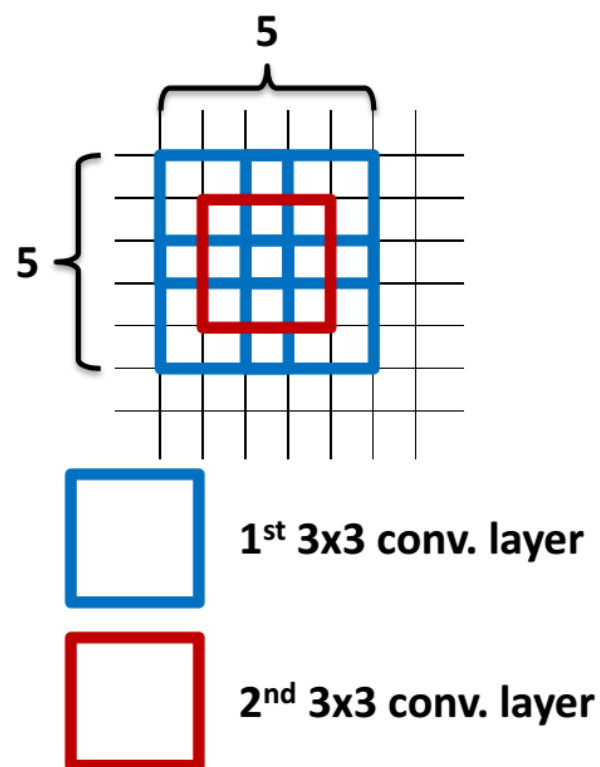
Why DL Suddenly works? – My Two Cents

- 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



Network Design

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

image

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

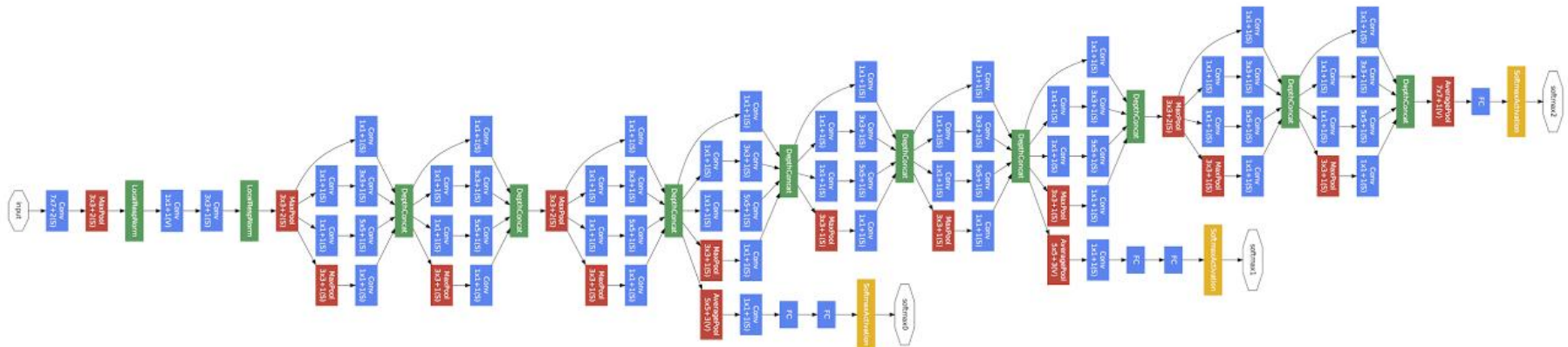
FC-4096

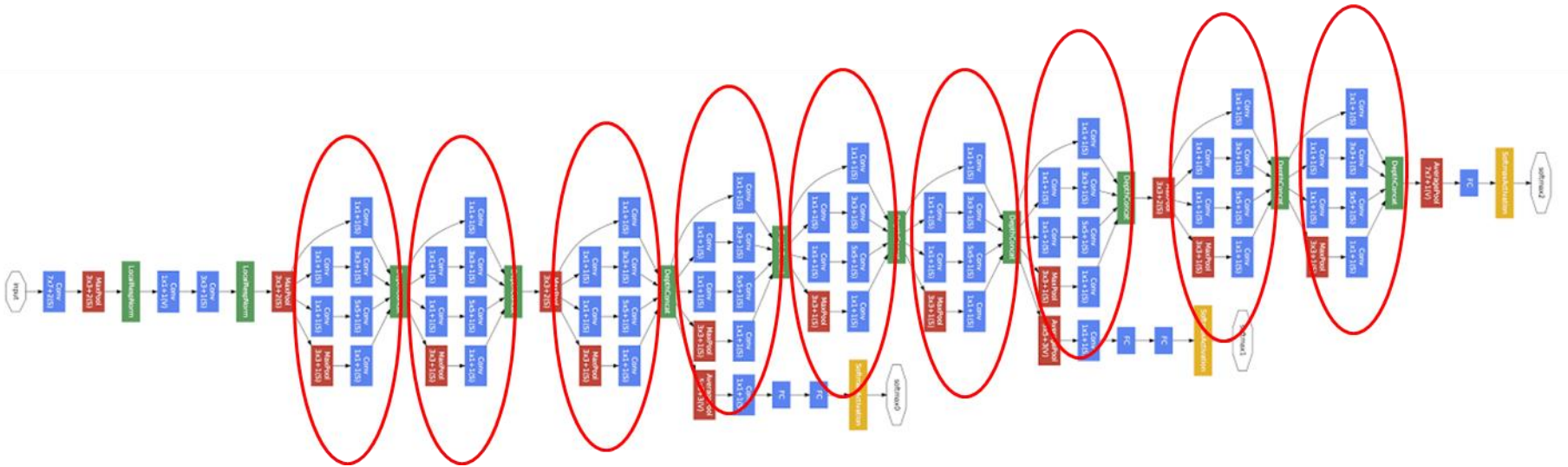
FC-4096

FC-1000

softmax

GoogLeNet



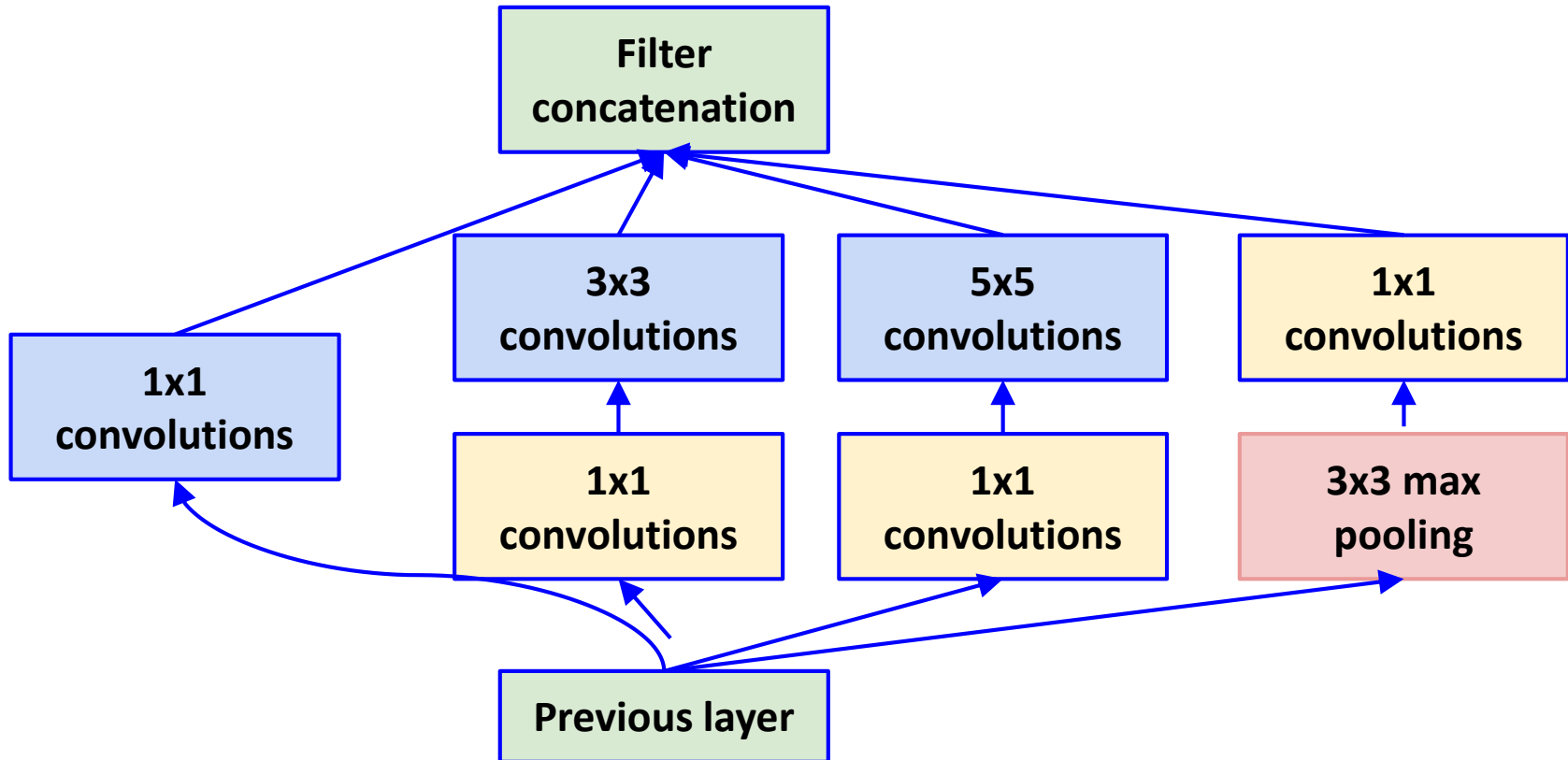


Inception

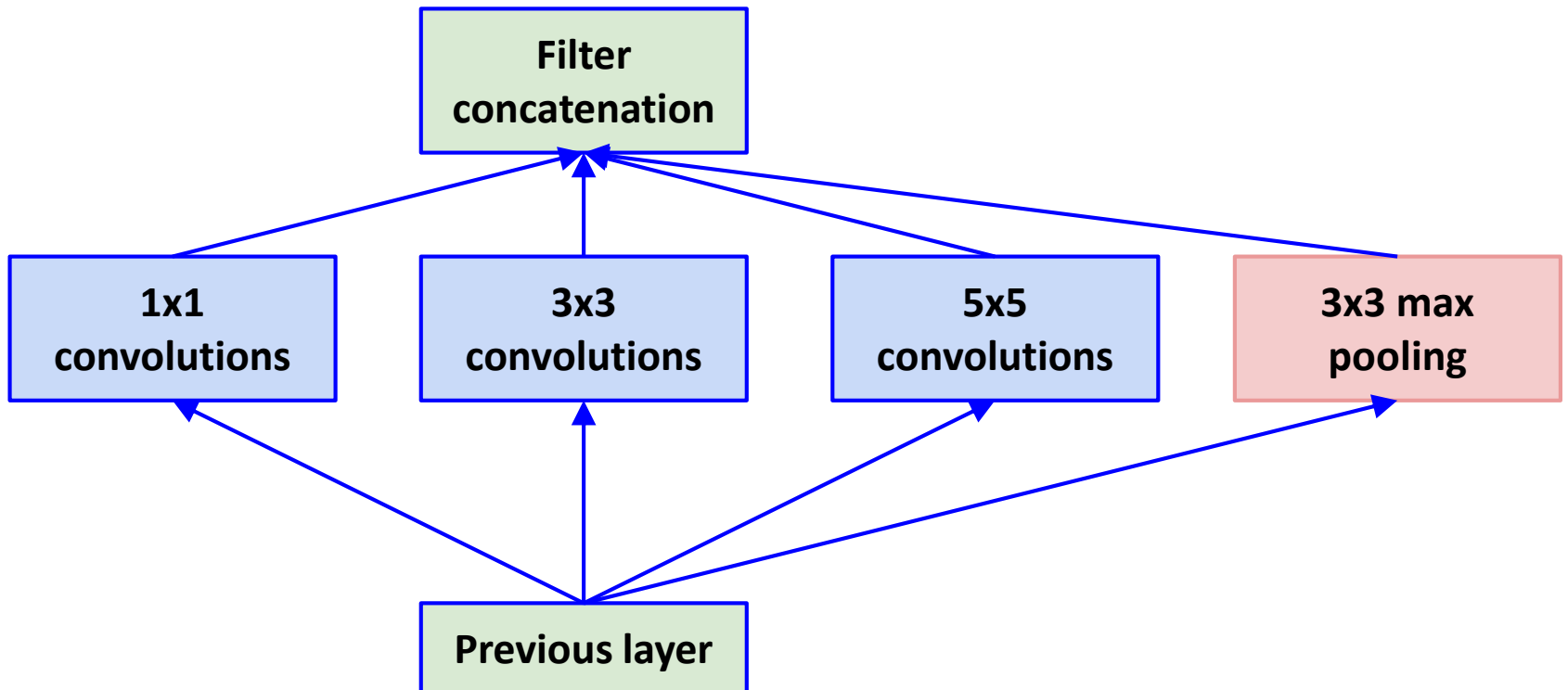
Network in a network in a network...

Convolution
Pooling
Softmax
Other

Inception module



Naive idea (does not work!)



ResNet

- See He Kaiming's ICML tutorial

DenseNet

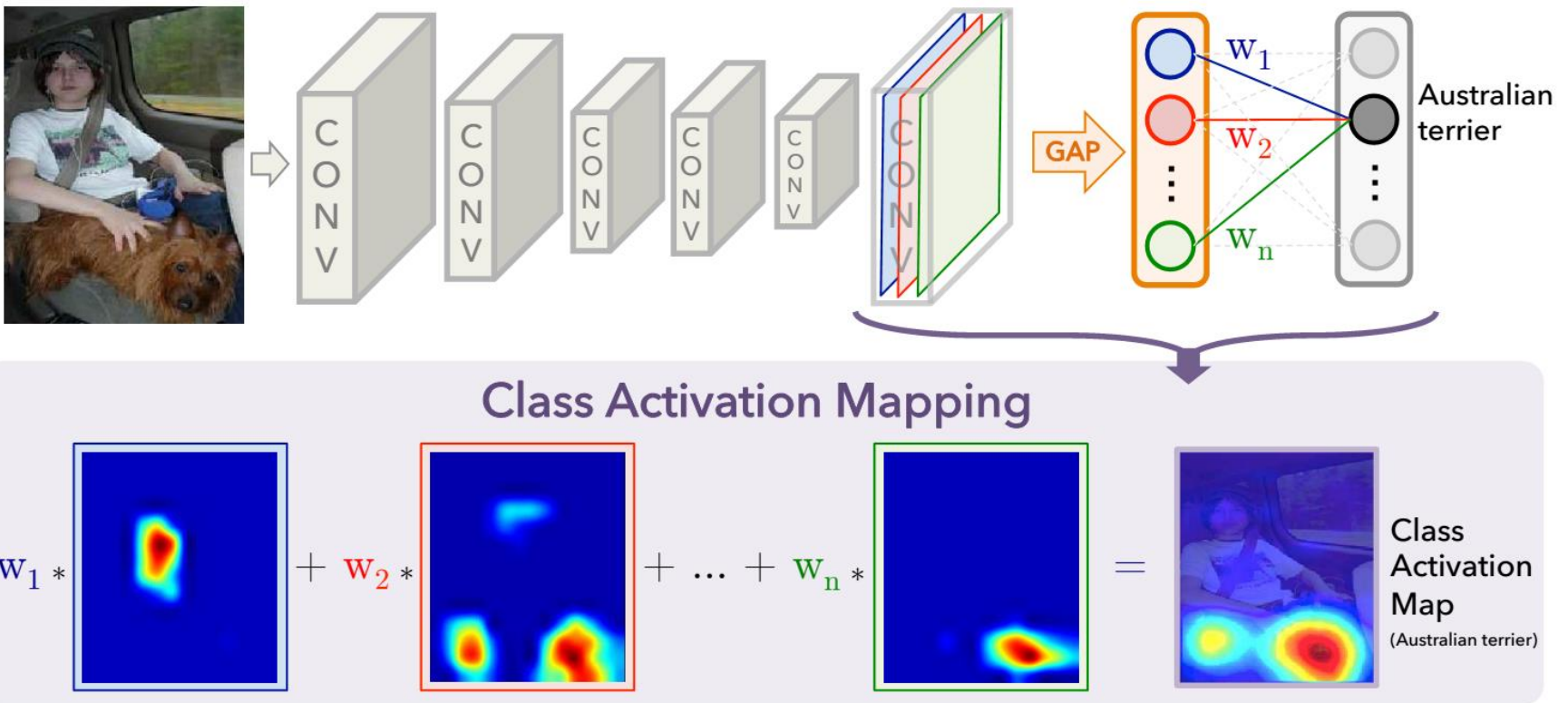
- See DenseNet's CVPR slides

Dual Path Network

- See DualPathNet's CVPR slides

Class Activation Map (CAM)

- Global Average Pooling



Class Activation Map (CAM)

- Examples

