

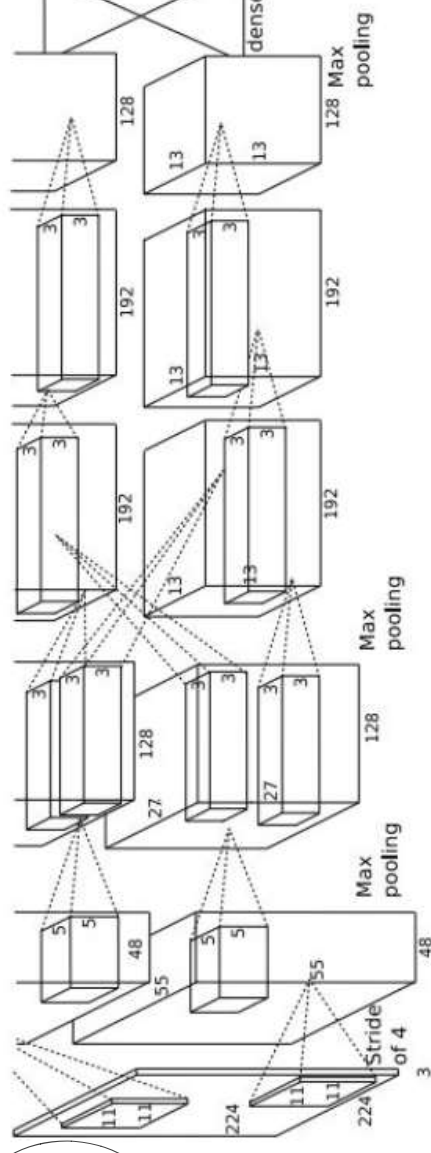
Architectures

Case Study: AlexNet

[Krizhevsky et al. 2012]

Architecture:

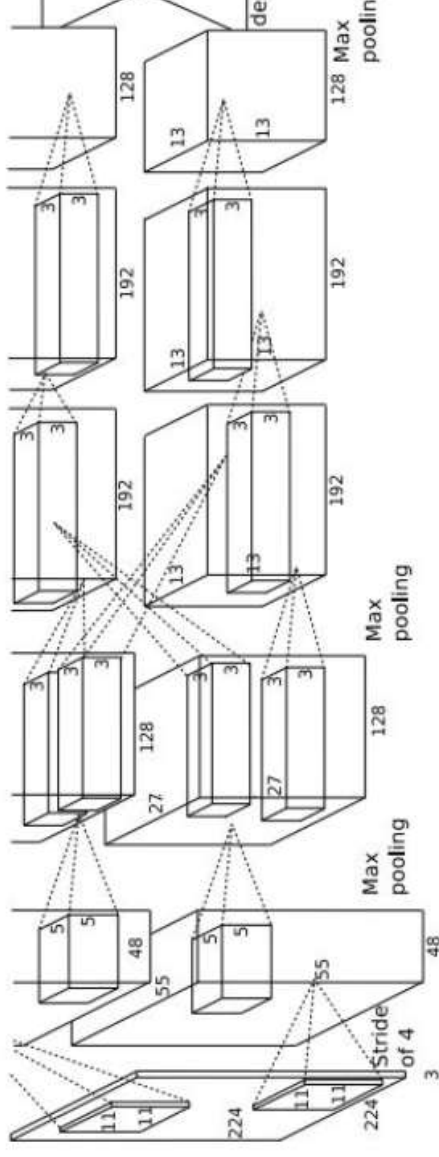
CONV1 \hookrightarrow relu.
 MAX POOL1 \hookrightarrow
~~NORM1~~
 CONV2 \hookrightarrow
 MAX POOL2 \hookrightarrow
 NORM2
 CONV3 \hookrightarrow
 CONV4 \hookrightarrow
 CONV5 \hookrightarrow
 Max POOL3 \hookrightarrow
 FC6 }
 FC7 }
 FC8 }



Architectures

Case Study: AlexNet

[Krizhevsky et al. 2012]



Input: 227x227x3 images

First layer (CONV1): 96 11x11 filters applied at stride 4

=>

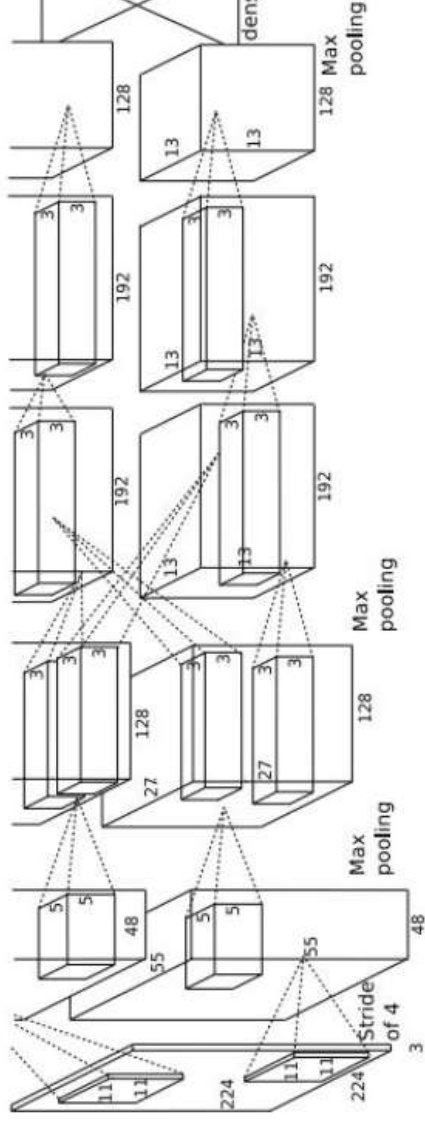
Q: what is the output volume size? Hint: $(227-11)/4+1 = 55$

$$W' = (W - F + 2P)$$

Architectures

Case Study: AlexNet

[Krizhevsky et al. 2012]



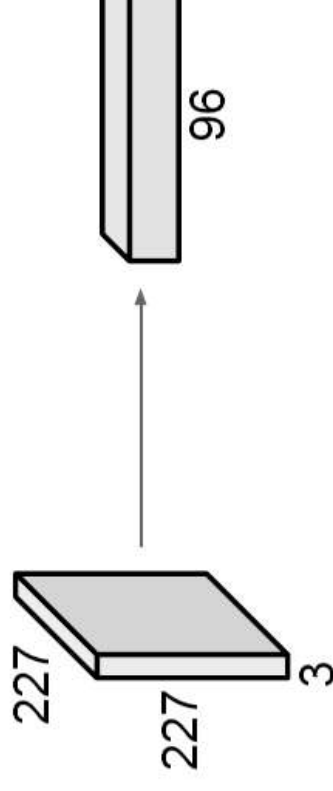
Input: 227x227x3 images

First layer (CONV1): 96 11x11 filters applied at stride 4

=>

Output volume **[55x55x96]**

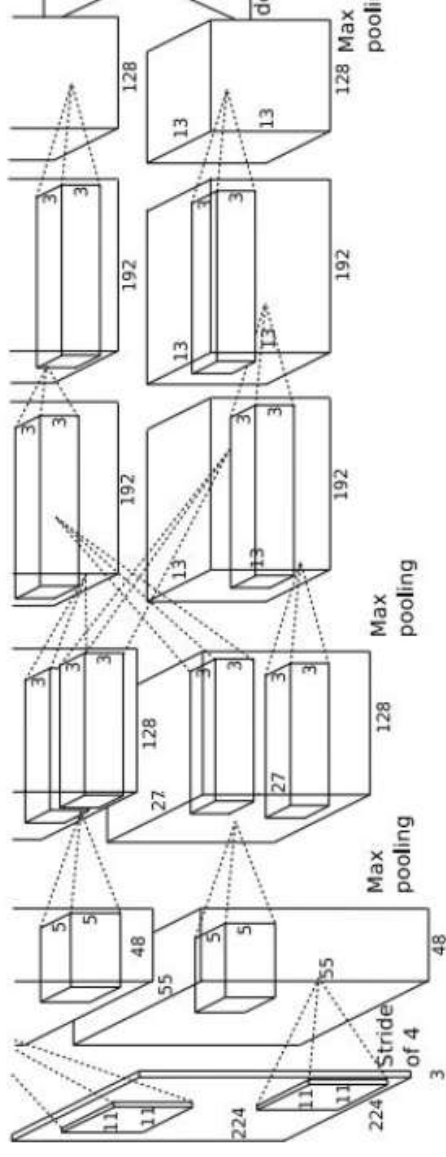
$$W' = (W - F + 2P)$$



Architectures

Case Study: AlexNet

[Krizhevsky et al. 2012]



Input: 227x227x3 images

First layer (CONV1): 96 11x11 filters applied at stride 4

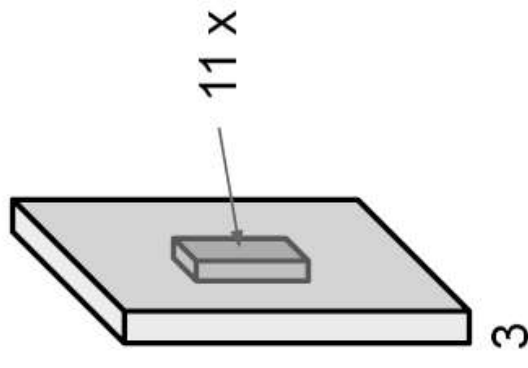
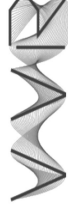
=>

Output volume **[55x55x96]**

Q: What is the total number of parameters in this layer?

$$11 \times 11 \times 3 \times 96$$

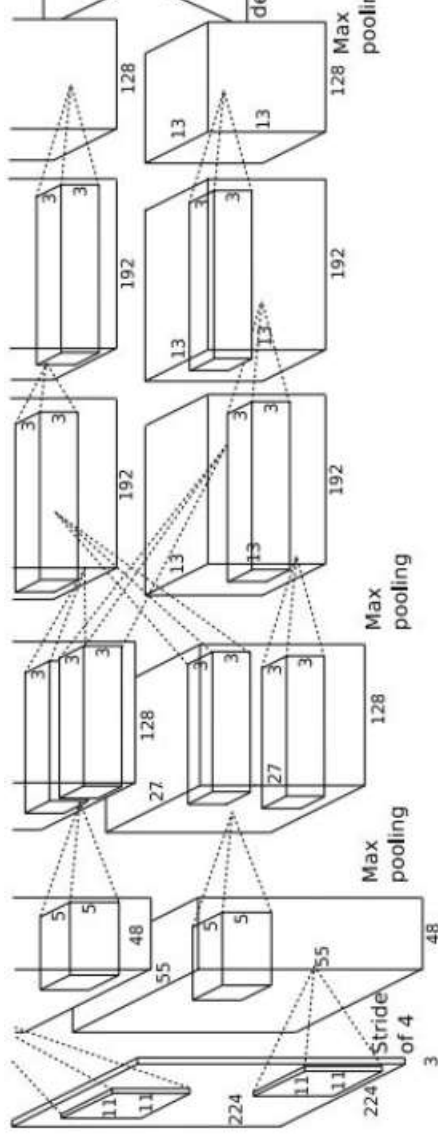
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Architectures

Case Study: AlexNet

[Krizhevsky et al. 2012]



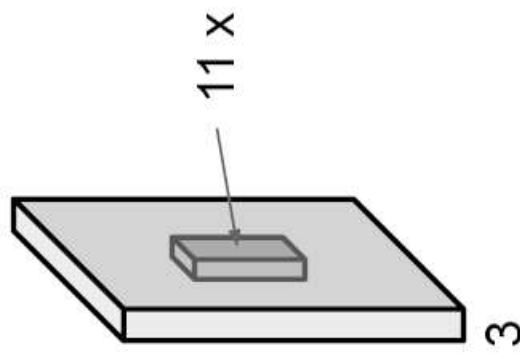
Input: 227x227x3 images

First layer (CONV1): 96 11x11 filters applied at stride 4

=>

Output volume **[55x55x96]**

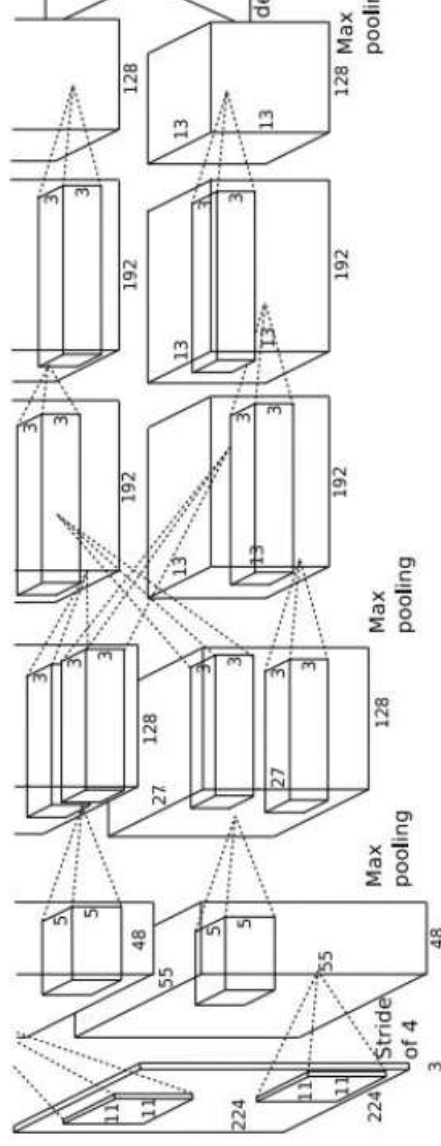
Parameters: $(11 \cdot 11 \cdot 3 + 1) \cdot 96 = 35K$



Architectures

Case Study: AlexNet

[Krizhevsky et al. 2012]



Input: 227x227x3 images

After CONV1: 55x55x96

Second layer (POOL1): 3x3 filters applied at stride 2

Q: what is the output volume size? Hint: $(55-3)/2+1 \neq 27$

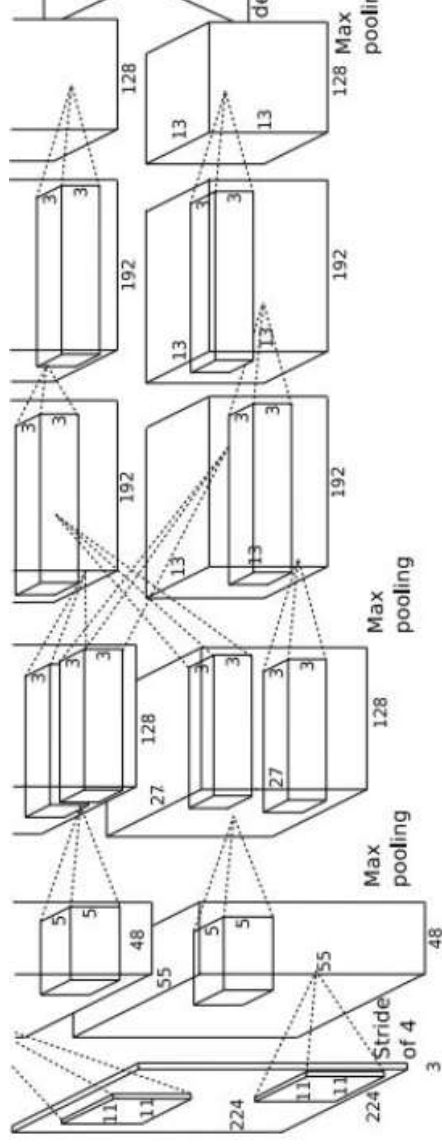
$$W' = (W - F + 2P)$$

27x27x128

Architectures

Case Study: AlexNet

[Krizhevsky et al. 2012]



Input: 227x227x3 images

After CONV1: 55x55x96

Second layer (POOL1): 3x3 filters applied at stride 2

Output volume: 27x27x96

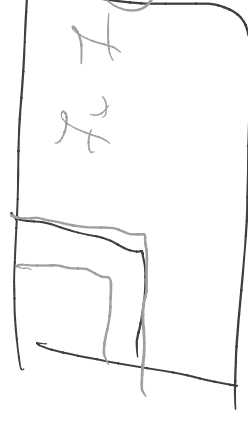
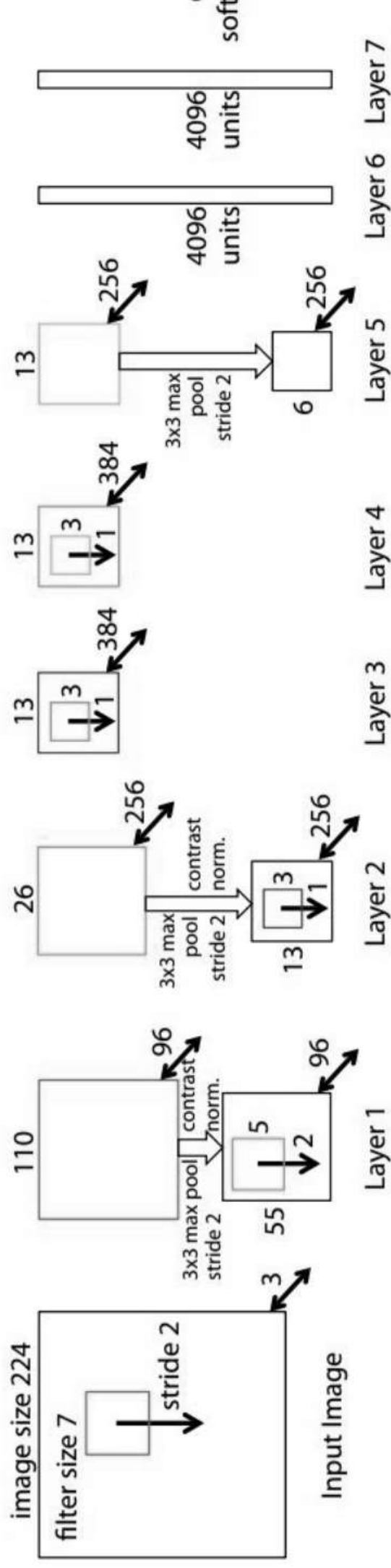
Q: what is the number of parameters in this layer?

$$W' = (W - F + 2P)$$

Architectures

ZFNet

[Zeiler and Fergus, 2013]



AlexNet but:

CONV1: change from (11x11 stride 4) to (7x7 stride 2)

CONV3,4,5: instead of 384, 384, 256 filters use 512, 1024, 512

ImageNet top 5 error: 16.4% ->

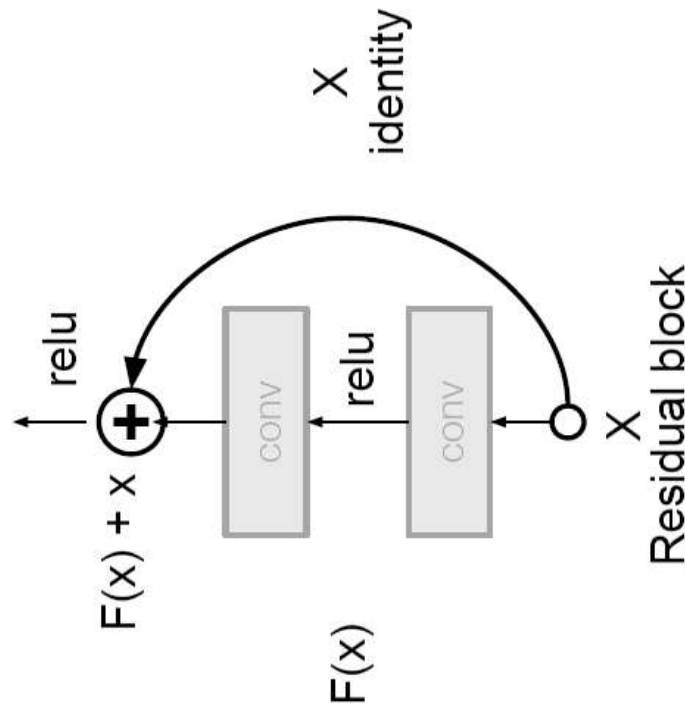
Architectures

Case Study: ResNet

[He et al., 2015]

Very deep networks using residual connections

- 152-layer model for ImageNet
- ILSVRC'15 classification winner (3.57% top 5 error)
- Swept all classification and detection competitions in ILSVRC'15 and COCO'15!

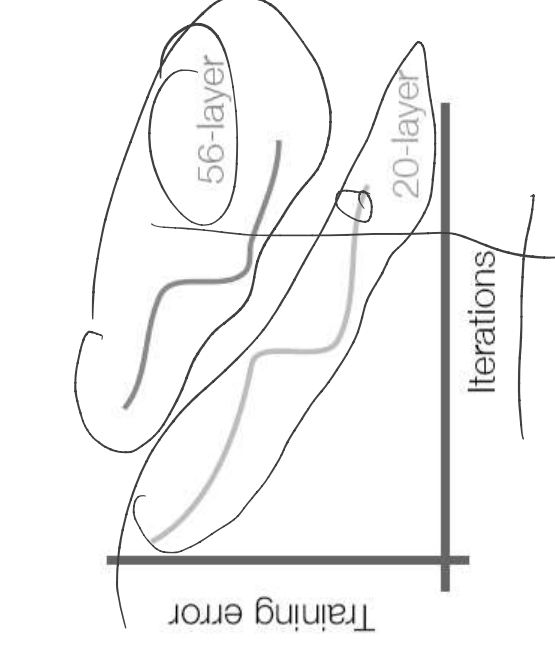
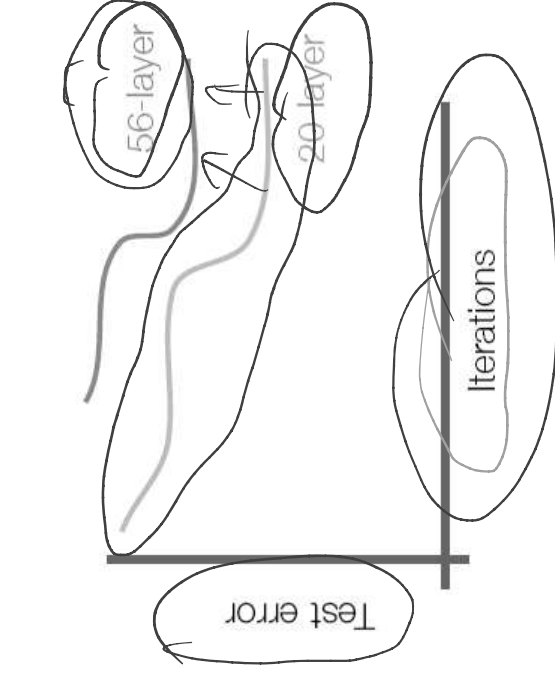


Architectures

Case Study: ResNet

[He et al., 2015]

What happens when we continue stacking deeper layers on a “plain” convolutional neural network?



56-layer model performs worse on both test and training error

-> The deeper model performs worse, but it's not caused by overfitting!

Architectures

Case Study: ResNet

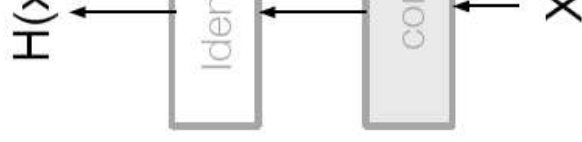
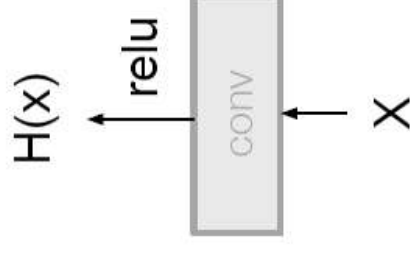
[He et al., 2015]

Fact: Deep models have more representation power (more parameters) than shallower models.

Hypothesis: the problem is an *optimization* problem, deeper models are harder to optimize

What should the deeper model learn to be at least as good as the shallower model?

A solution by construction is copying the learned layers from the shallower model and setting additional layers to identity mapping.

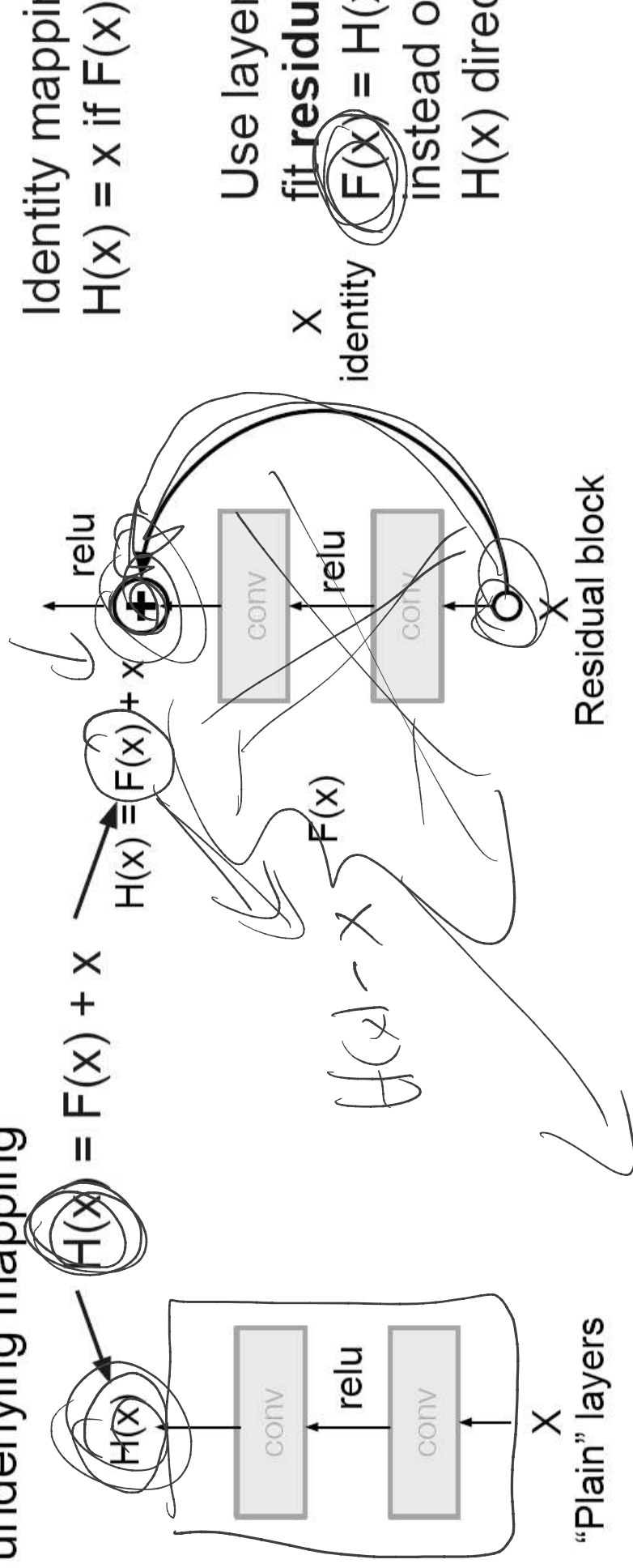


Architectures

Case Study: ResNet

[He et al., 2015]

Solution: Use network layers to fit a residual mapping instead of directly trying to fit the desired underlying mapping



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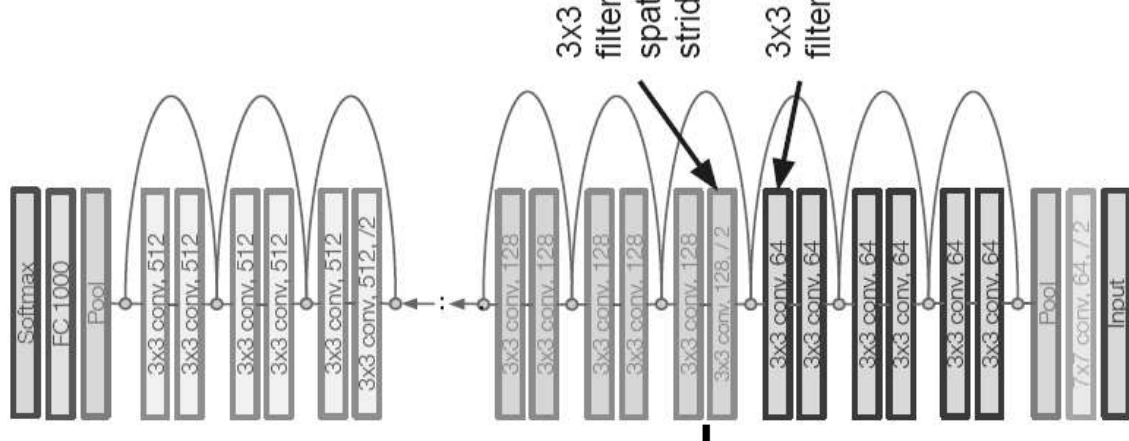
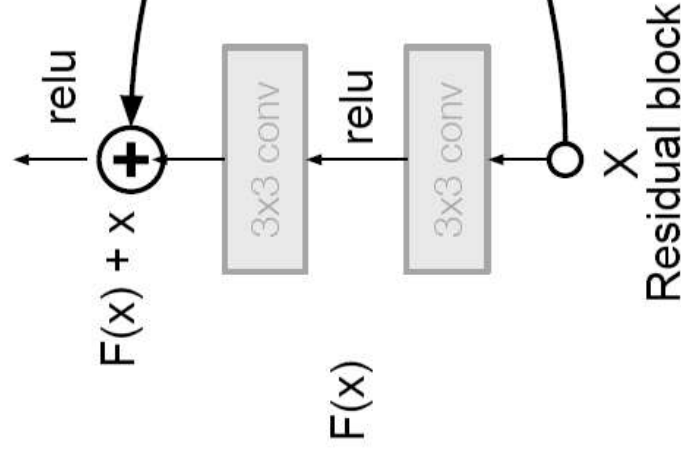
<http://www.vvr.ece.upat.edu>

Architectures

Case Study: ResNet

[He et al., 2015]

- Full ResNet architecture:
- Stack residual blocks
 - Every residual block has two 3x3 conv layers
 - Periodically, double # of filters and downsample spatially using stride 2 (/2 in each dimension)
- Reduce the activation volume by half.

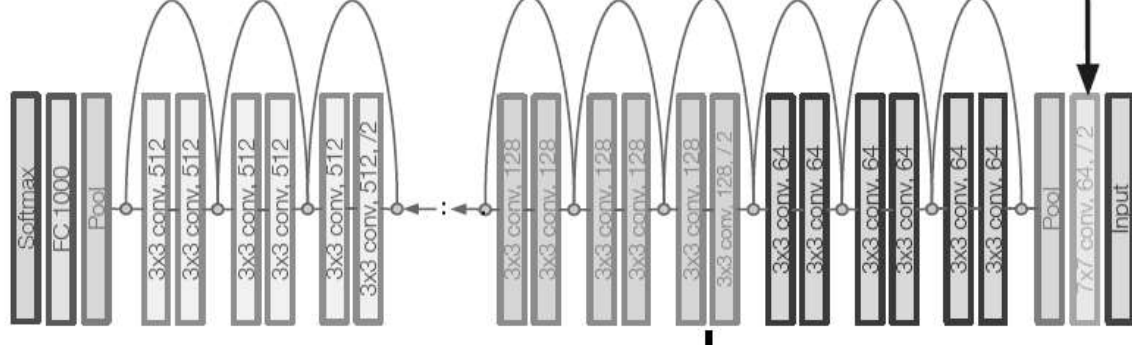
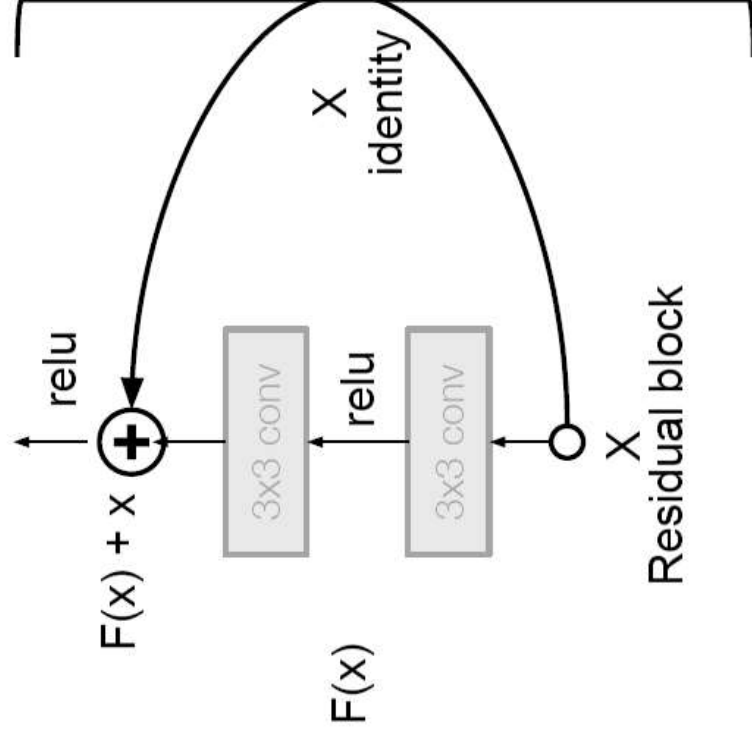


Architectures

Case Study: ResNet

[He et al., 2015]

- Full ResNet architecture:
- Stack residual blocks
 - Every residual block has two 3x3 conv layers
 - Periodically, double # of filters and downsample spatially using stride 2 (/2 in each dimension)
 - Additional conv layer at the beginning (stem)



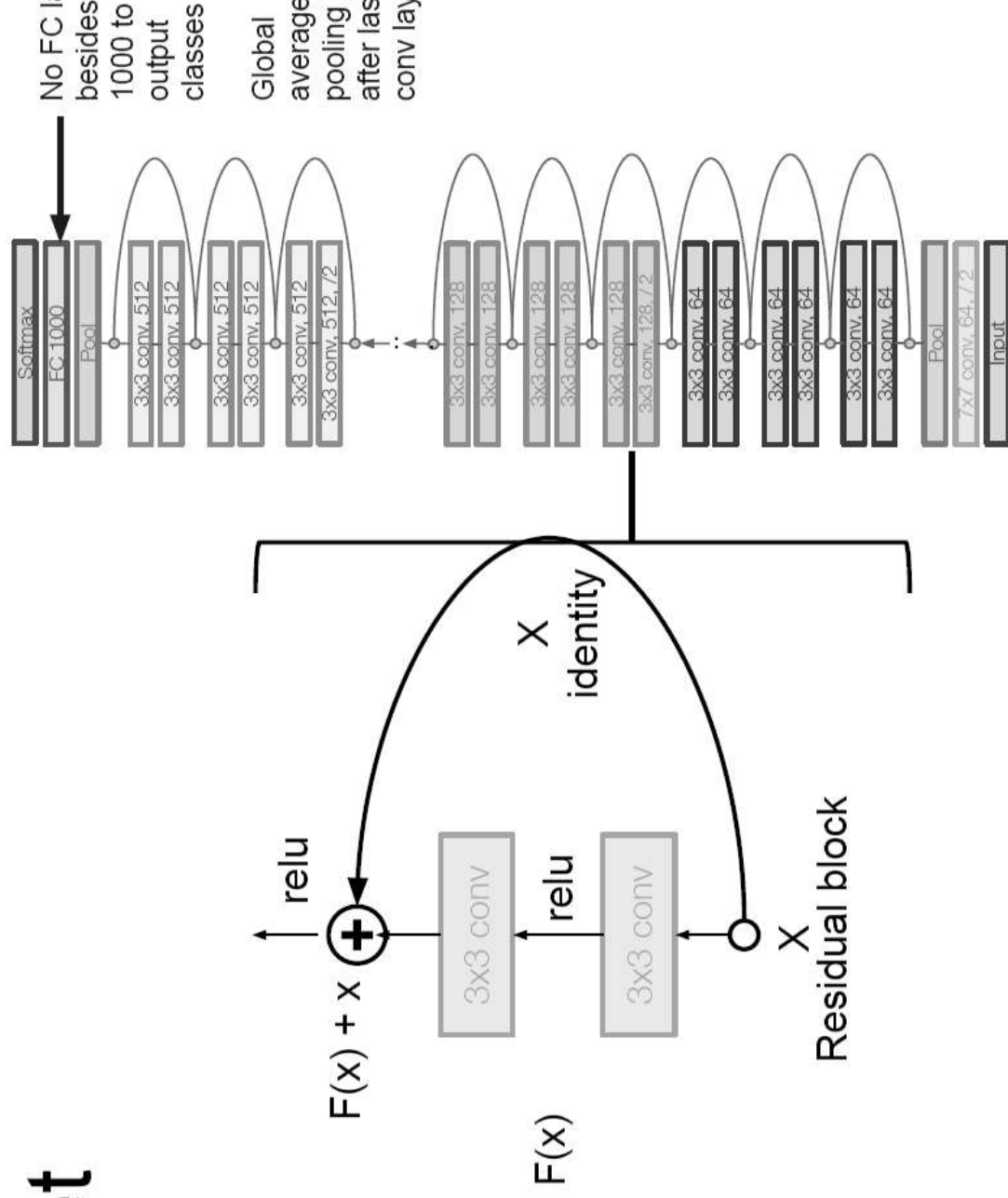
Architectures

Case Study: ResNet

[He et al., 2015]

Full ResNet architecture:

- Stack residual blocks
- Every residual block has two 3x3 conv layers
- Periodically, double # of filters and downsample spatially using stride 2 (/2 in each dimension)
- Additional conv layer at the beginning (stem)
- No FC layers at the end (only FC 1000 to output classes)
- (In theory, you can train a ResNet with input image of variable sizes)



Architectures

Case Study: ResNet

[He et al., 2015]

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