

# Deep Neural Networks

## Machine Learning and Pattern Recognition

(Largely based on slides from Fei-Fei Li & Justin Johnson & Serena Yeung)

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# CNNs Architectures

INPUT 32x32

C1: feature maps 6@28x28

C3: f. maps 16@10x10

S2: f. maps 6@14x14

S4: f. maps 16@5x5

C5: layer 120

F6: layer 84

OUTPUT 10

Convolutions

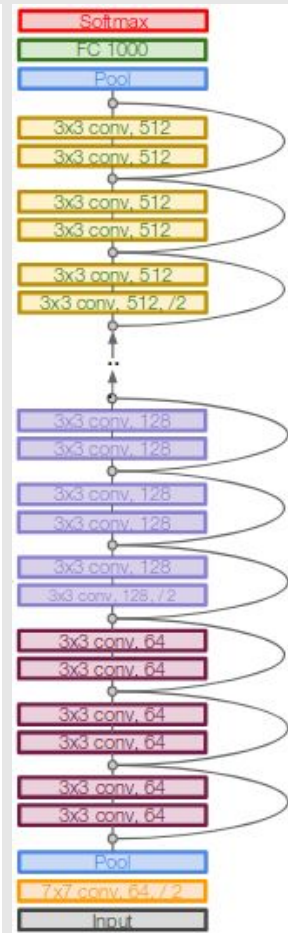
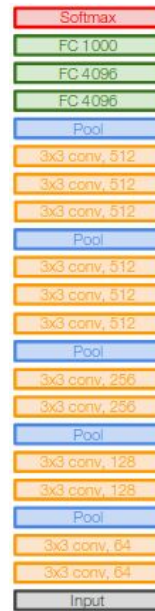
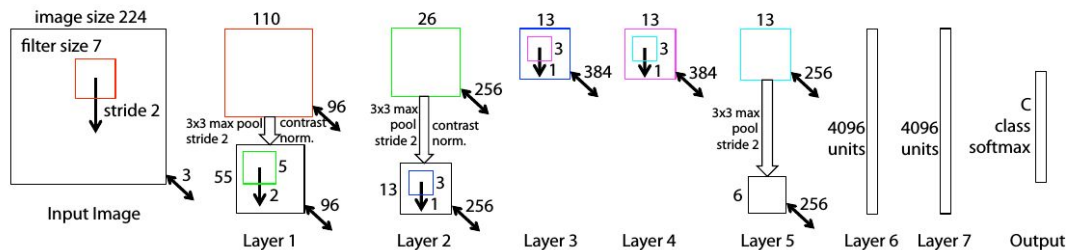
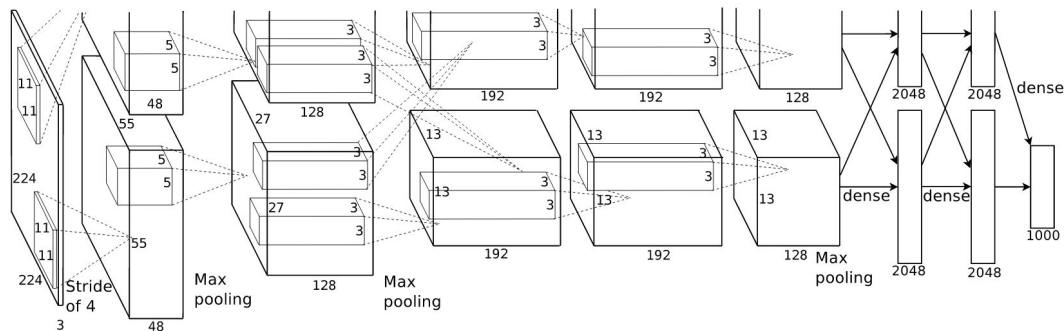
Subsampling

Convolutions

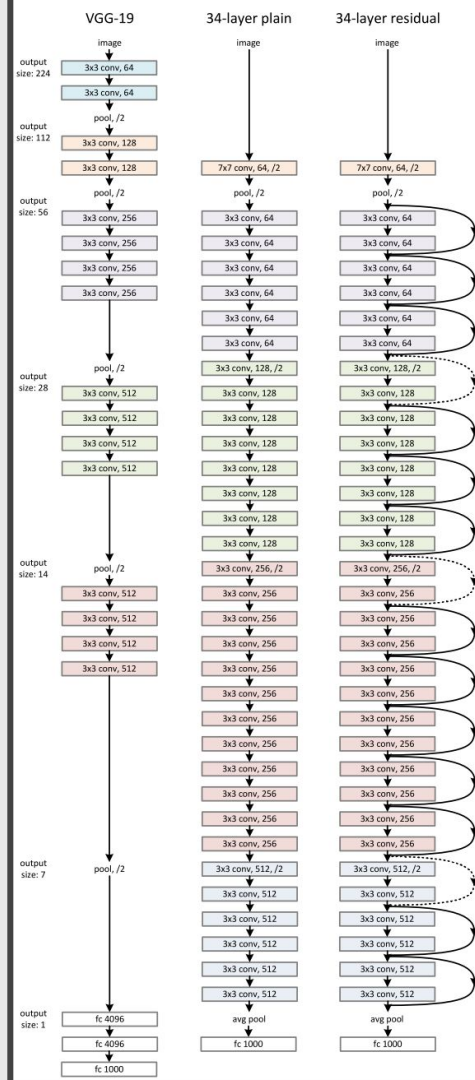
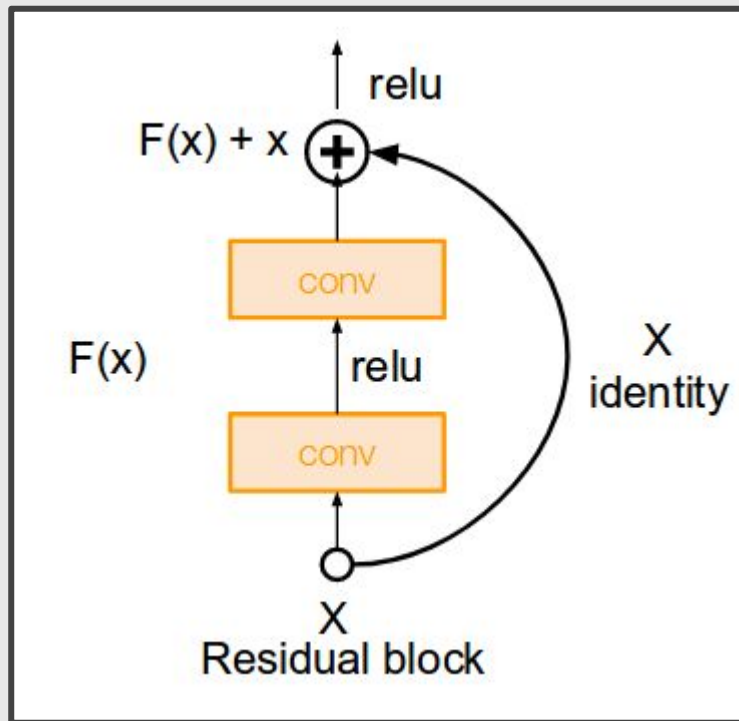
Subsampling

Full connection

Full connection



# ResNet [He et al., 2015]



# ResNet [He et al., 2015]

For deeper networks  
(**ResNet-50+**), use  
“bottleneck” layer to  
improve efficiency  
(similar to GoogLeNet)

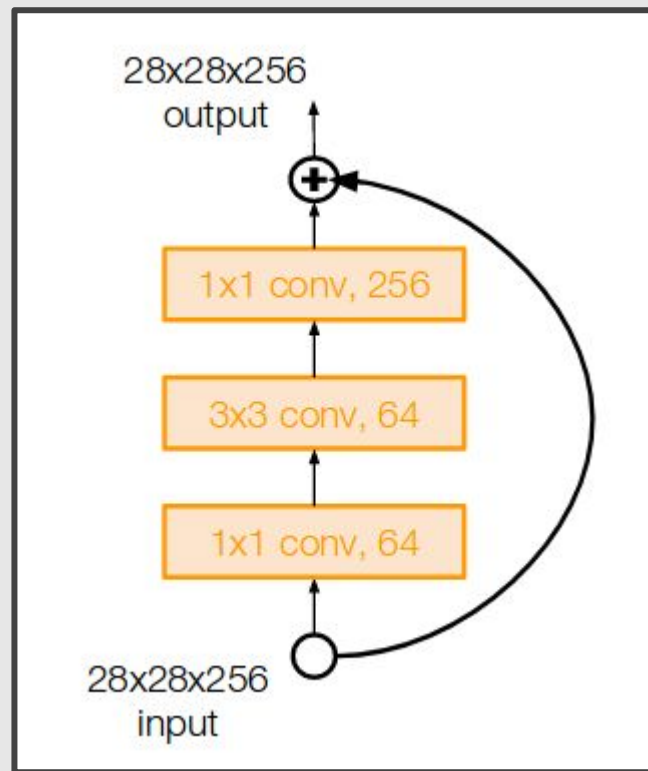
1x1 conv, 256 filters projects  
back to 256 feature maps  
(28x28x256)



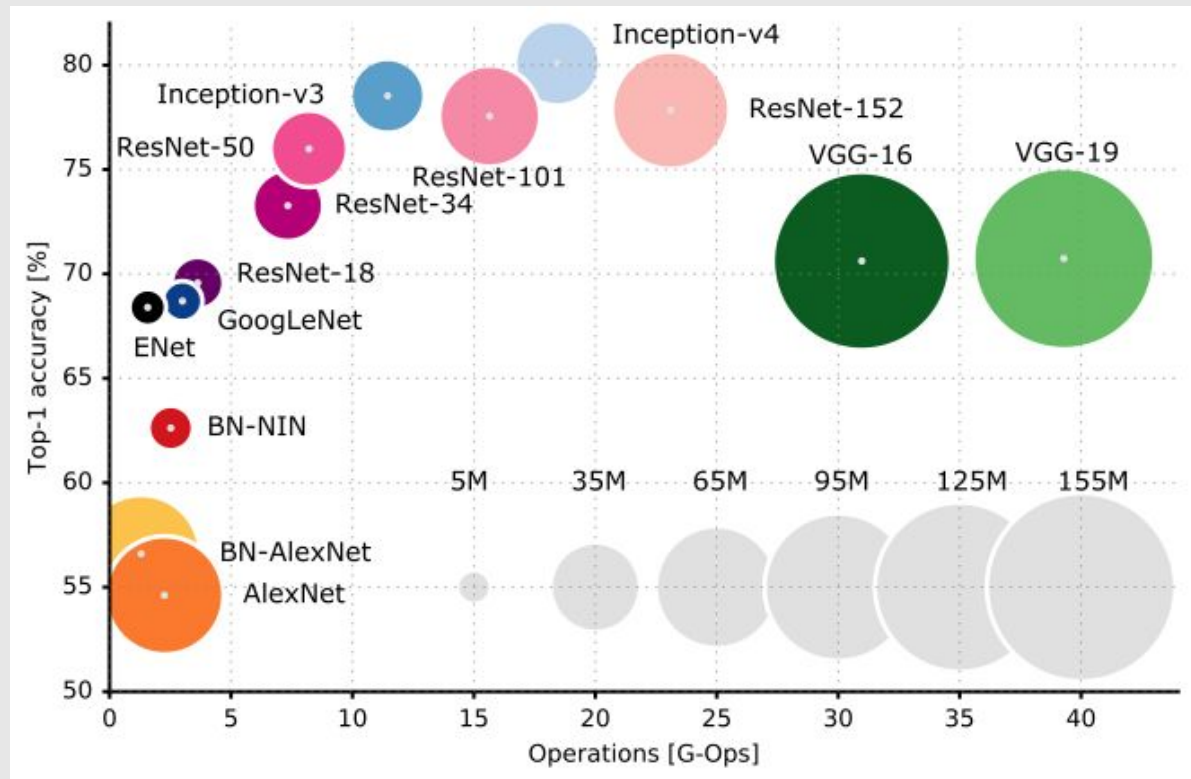
3x3 conv operates over  
only 64 feature maps



1x1 conv, 64 filters  
to project to  
28x28x64



The size of the blobs is proportional to the number of network parameters.



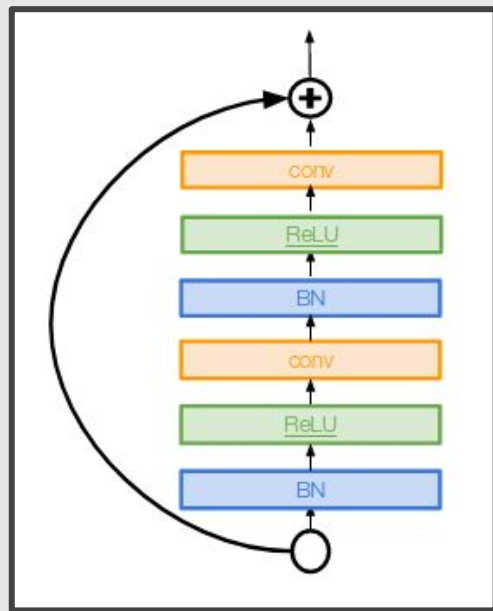
<https://medium.com/towards-data-science/neural-network-architectures-156e5bad51ba>

# Other CNNs Architectures

# Improving ResNet ...

## Identity Mappings in Deep Residual Networks [He et al., 2016]

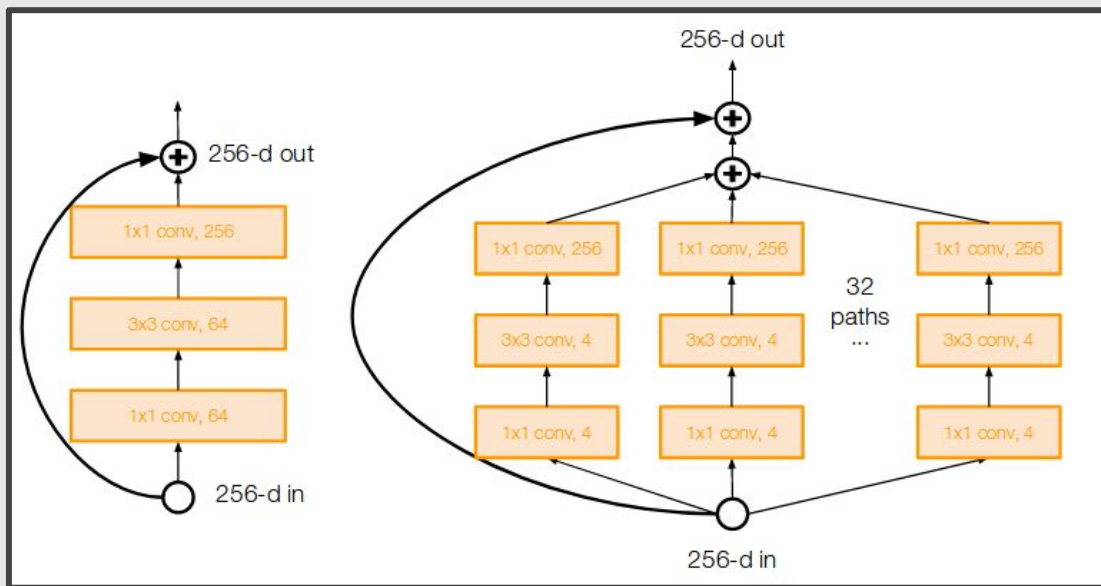
- Improved ResNet block design from creators of ResNet
- Creates a more direct path for propagating information throughout network (moves activation to residual mapping pathway)
- Gives better performance





# Improving ResNet ...

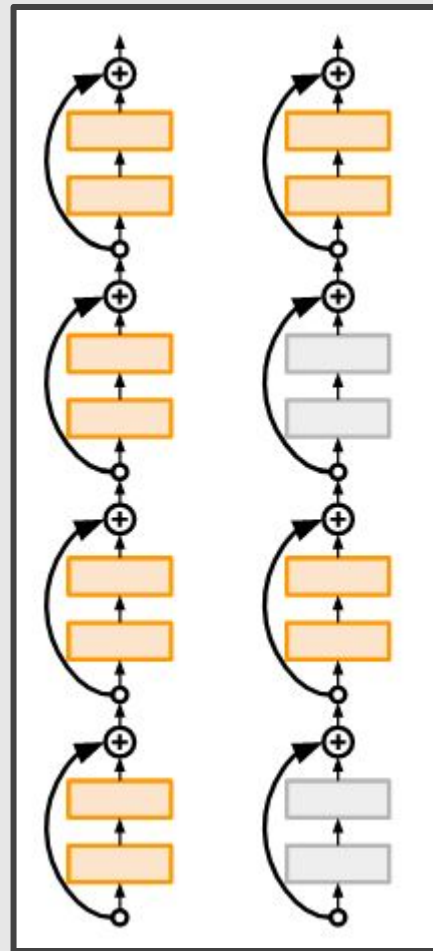
Aggregated Residual Transformations for Deep Neural Networks (**ResNeXt**) [Xie et al., 2016]



# Improving ResNet ...

Deep Networks with Stochastic Depth  
[Huang et al., 2016]

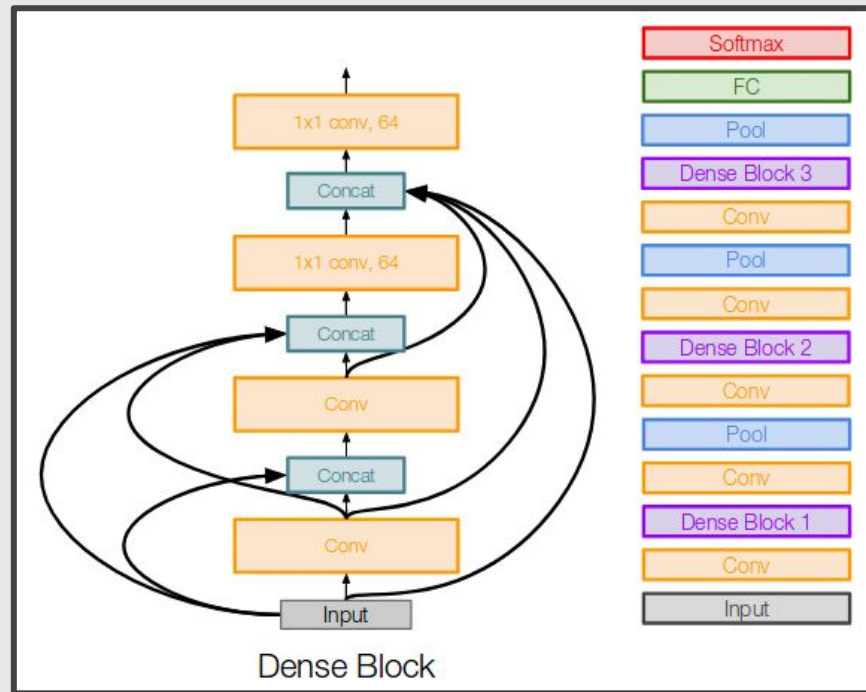
- Motivation: reduce vanishing gradients
- Randomly drop a subset of layers during each training pass
- Bypass with identity function



# Beyond ResNet ...

## Densely Connected Convolutional Networks (**DenseNet**) [Huang et al., 2017]

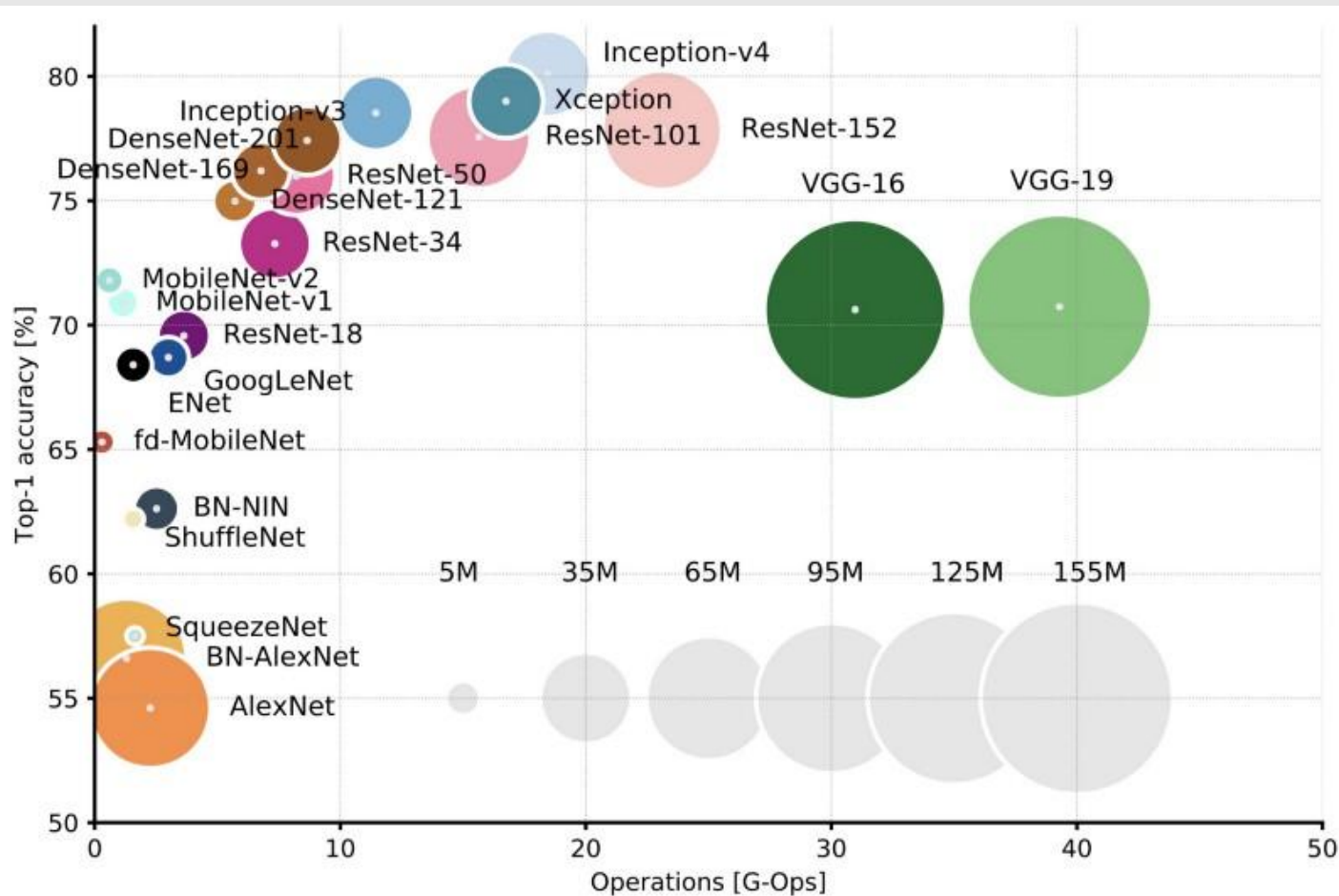
- Dense blocks where each layer is connected to every other layer in feedforward fashion
- Alleviates vanishing gradient, strengthens feature propagation, encourages feature reuse



# Summary

- VGG, GoogLeNet, ResNet all in wide use, available in model zoos
- ResNet current best default
- Trend towards extremely deep networks
- Significant research centers around design of layer / skip connections and improving gradient flow
- Even more recent trend towards examining necessity of depth vs. width and residual connections

The size of the blobs is proportional to the **number of network parameters**.



# Transfer Learning

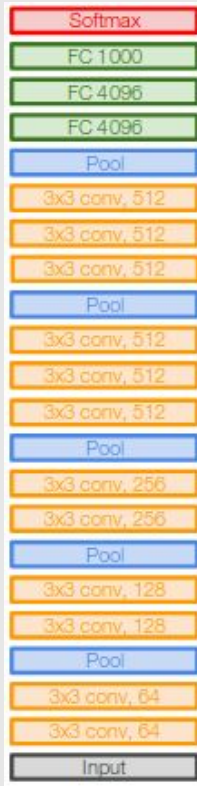
“You need a lot of a data if you want to train/use CNNs”

# Transfer Learning

“You need a lot of data if you  
want to train these CNNs”



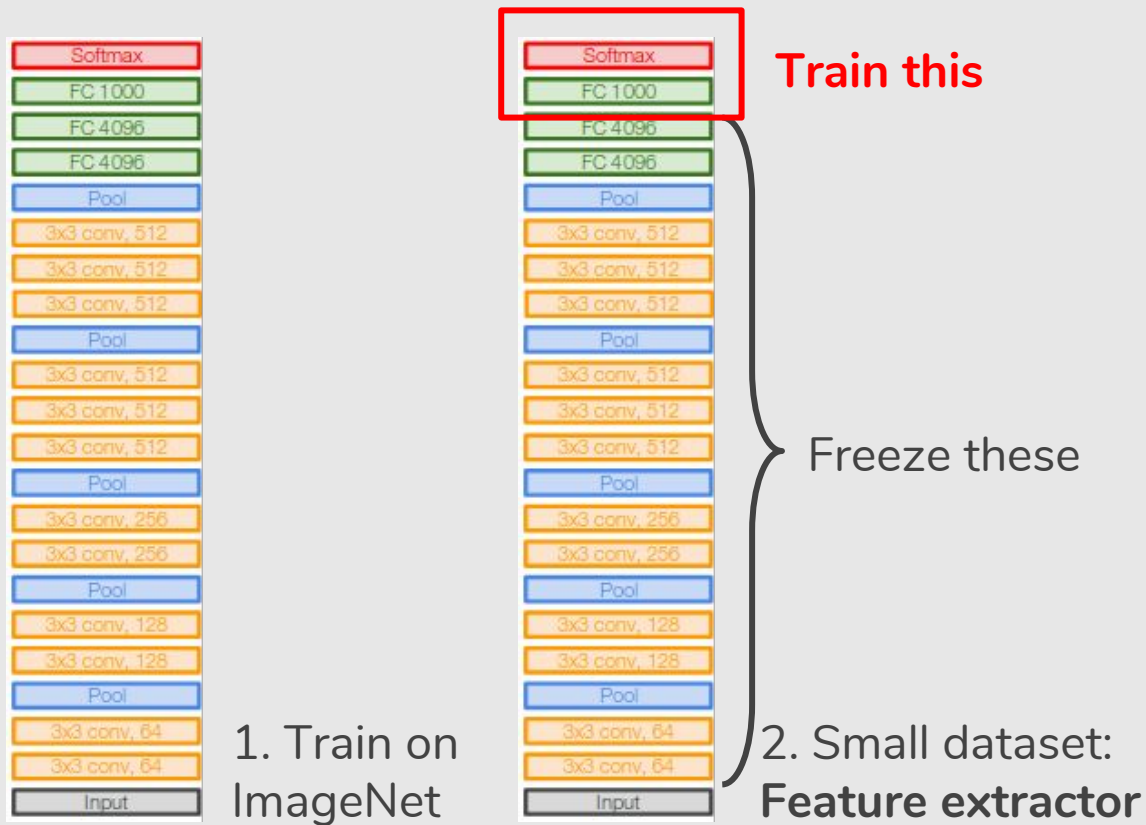
# Transfer Learning with CNNs



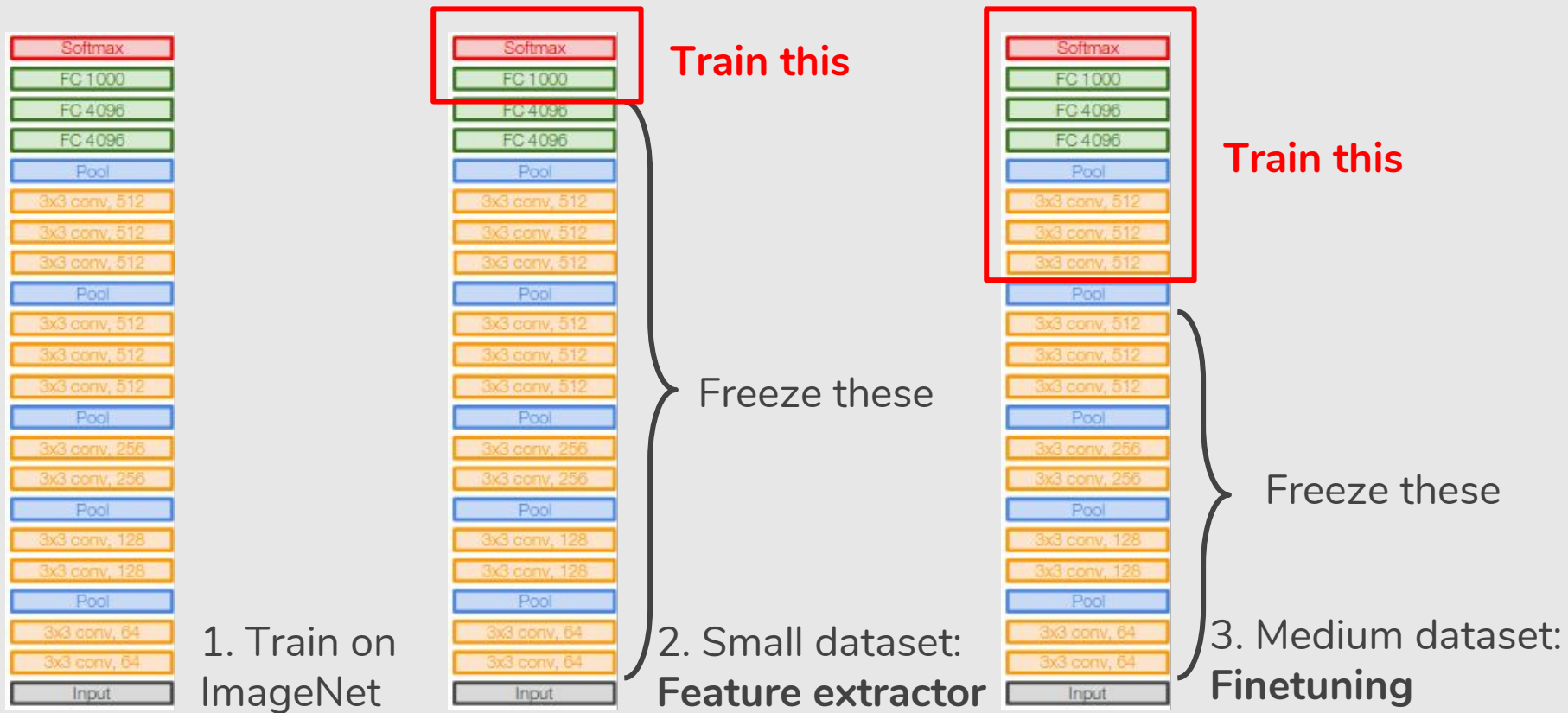
1. Train on  
ImageNet



# Transfer Learning with CNNs



# Transfer Learning with CNNs



# References

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## Machine Learning Books

- Hands-On Machine Learning with Scikit-Learn and TensorFlow, Chap. 11 & 13

## Machine Learning Courses

- <https://www.coursera.org/learn/neural-networks>
- “The 3 popular courses on Deep Learning”:  
<https://medium.com/towards-data-science/the-3-popular-courses-for-deeplearning-ai-ac37d4433bd>