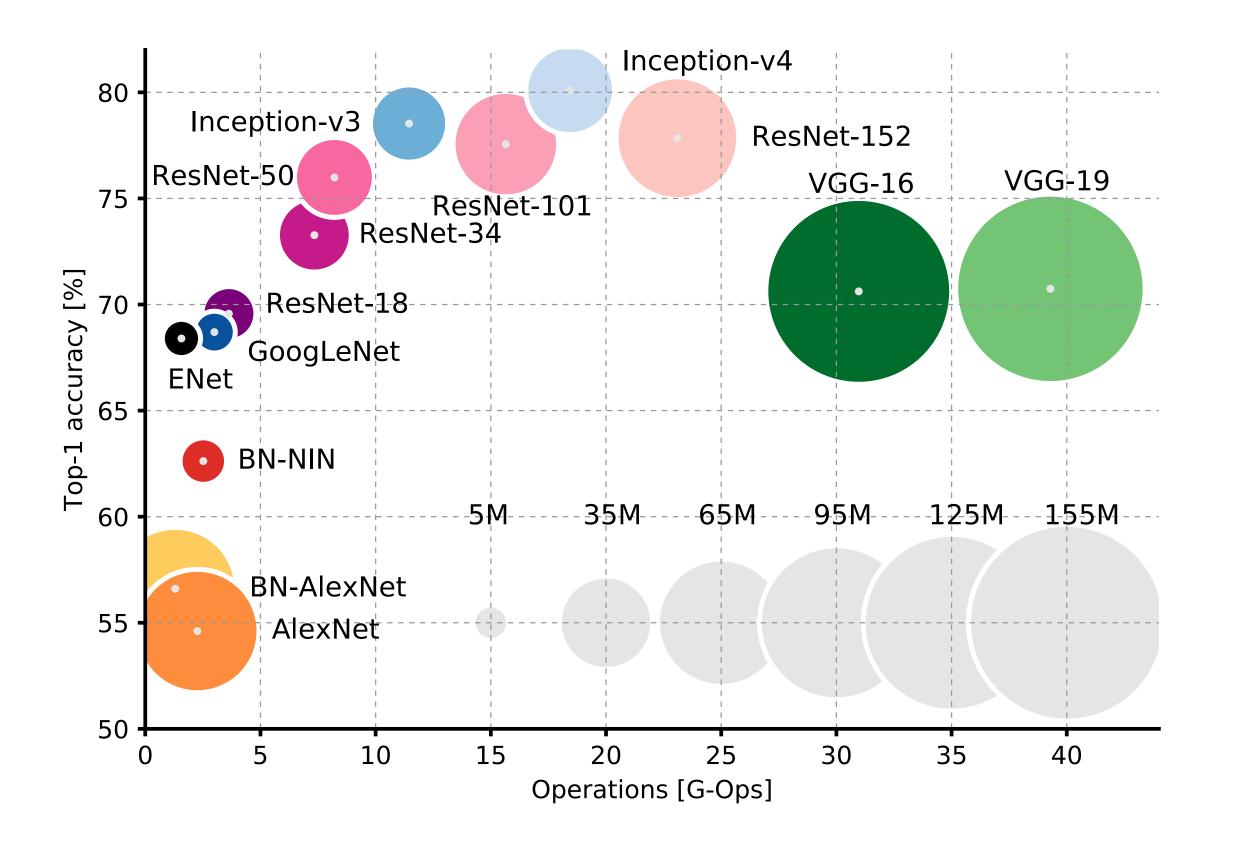
# 7.3

Convolutional Neural Networks Architectures

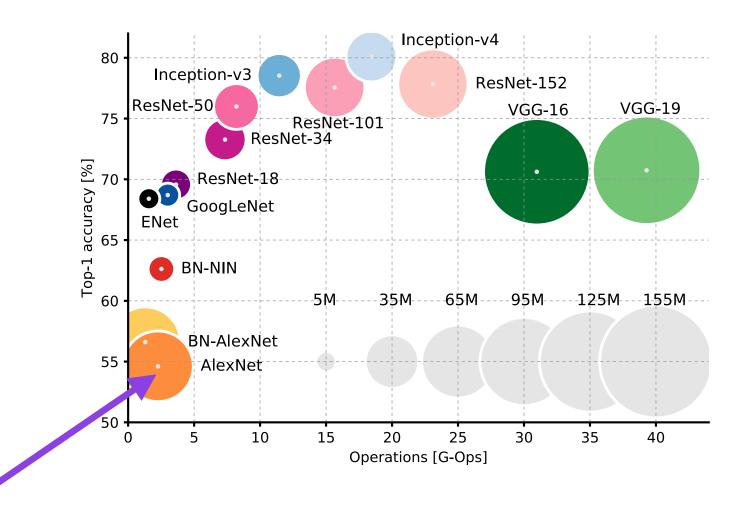
Part 3: A Whirlwind Tour Of Key Architecture Ideas

Sebastian Raschka and the Lightning Al Team

While this graphic is from 2016, many influential ideas or on this paper

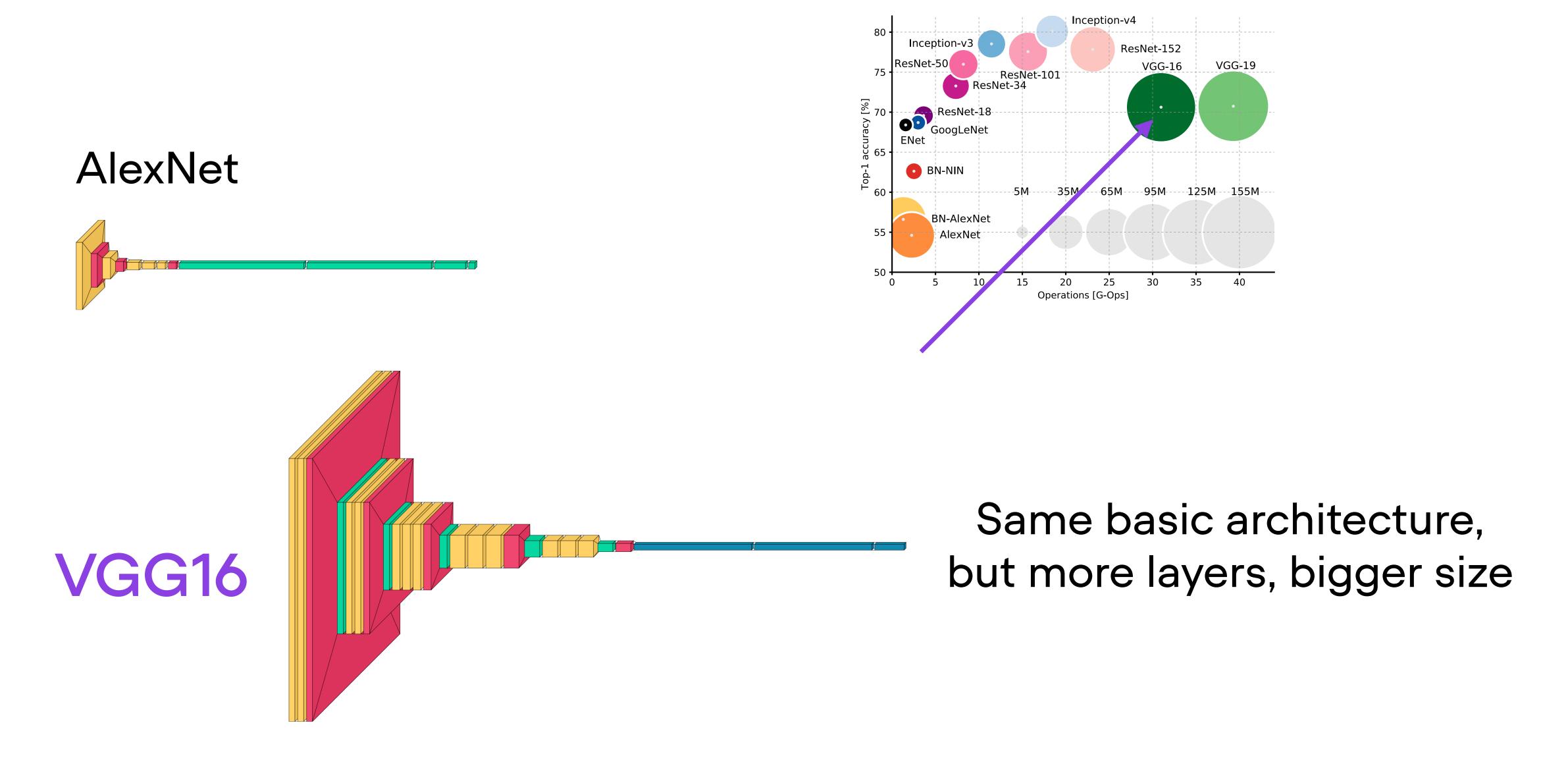


Canziani, Paszke, Culurciello (2016). An analysis of deep neural network models for practical applications. https://arxiv.org/abs/1605.07678



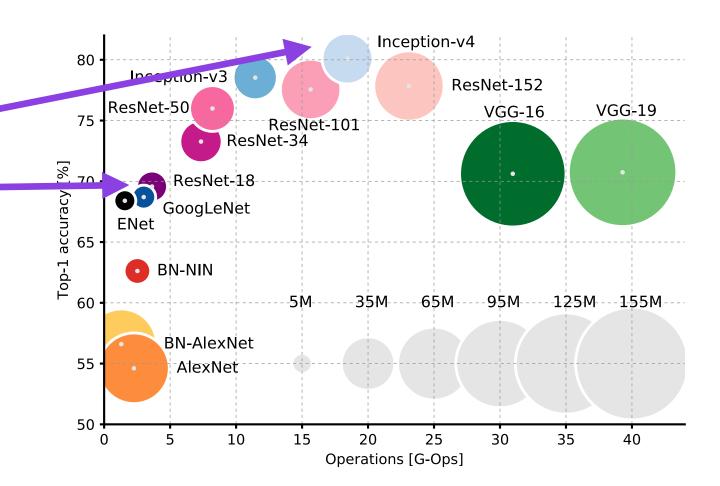
#### AlexNet

- "Birthday" of deep learning
- Won ImageNet 2012 competition
- One of the first CNNs utilizing GPUs for efficient training



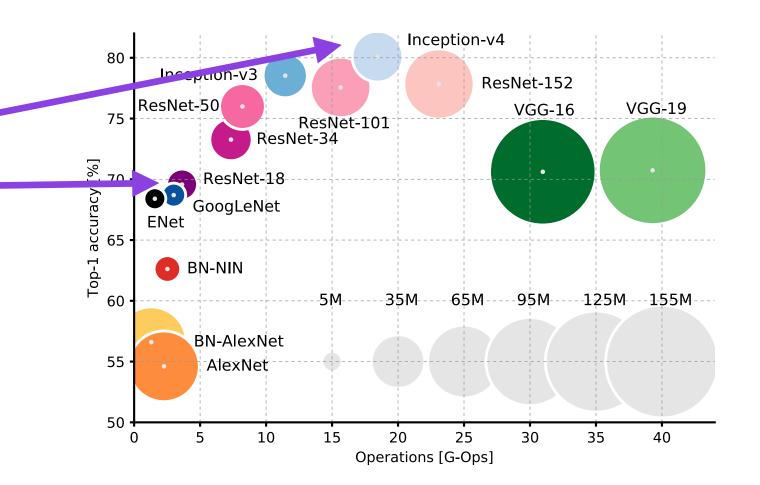
Se Simonyan and Zisserman (2014). Very deep convolutional networks for large-scale image recognition. https://arxiv.org/abs/1409.1556 in G.A.

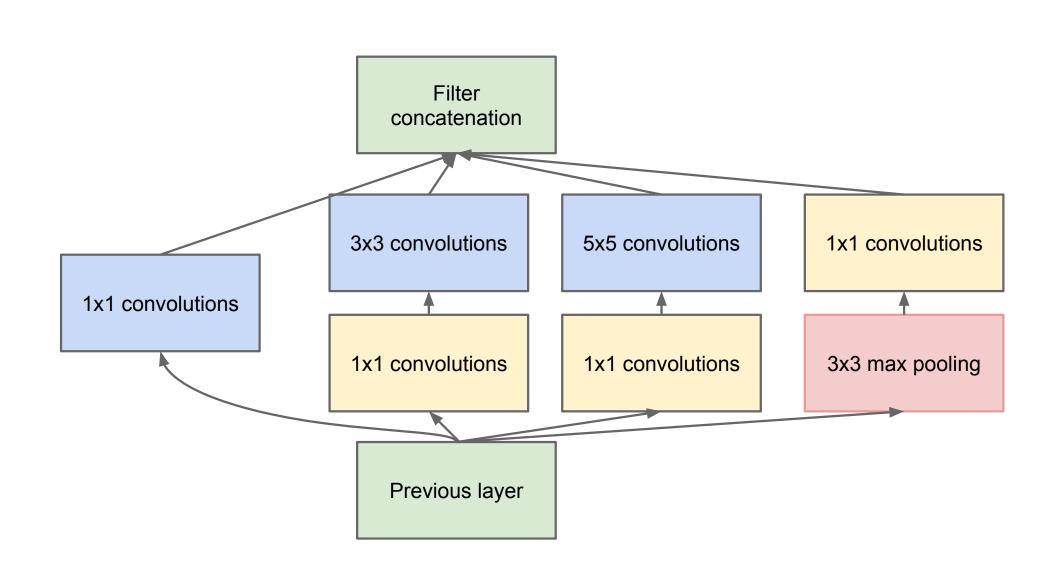
### GoogLeNet/Inception



#### GoogLeNet/Inception

Inception modules: use multiple convolutional layers with smaller kernels in parallel

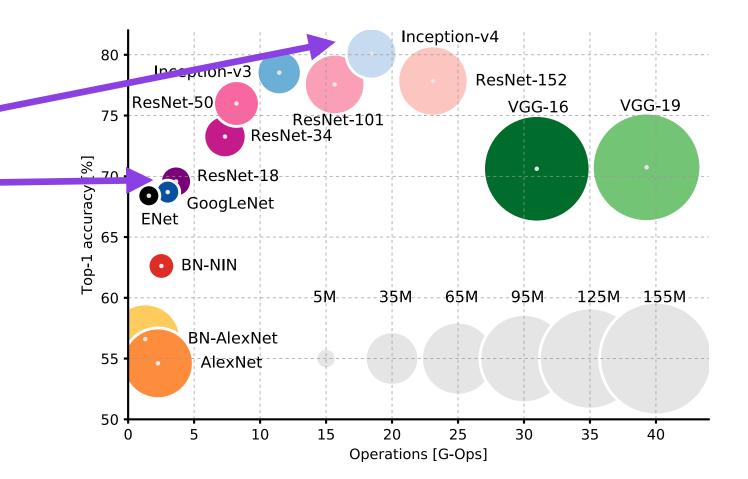


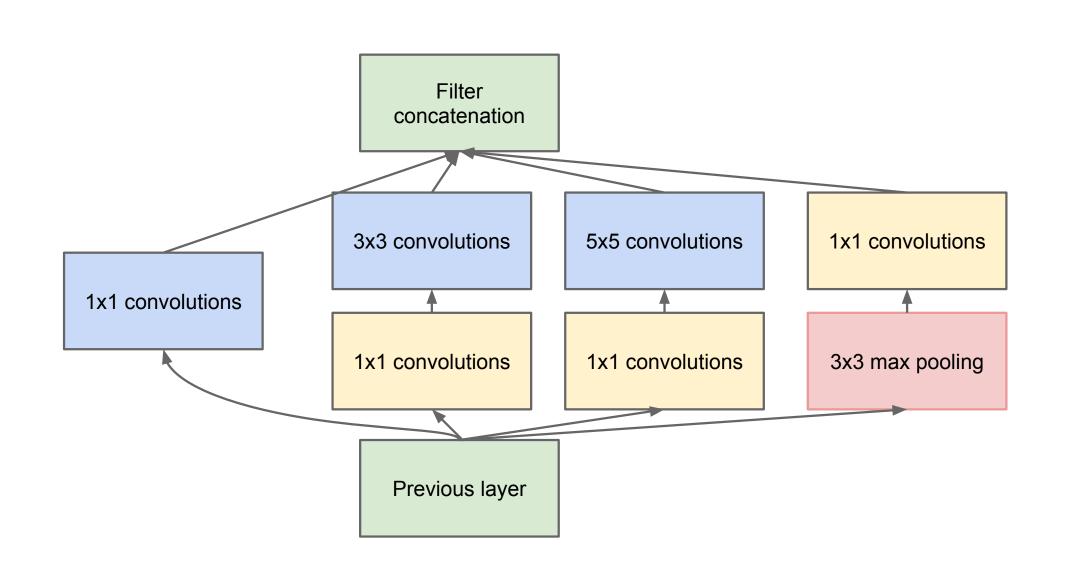


#### GoogLeNet/Inception

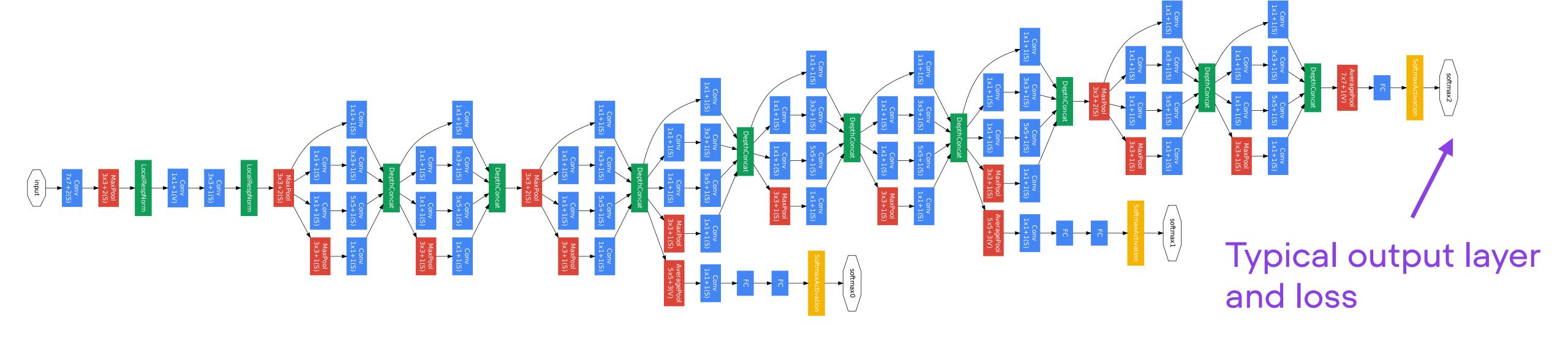
Inception modules: use multiple convolutional layers with smaller kernels in parallel

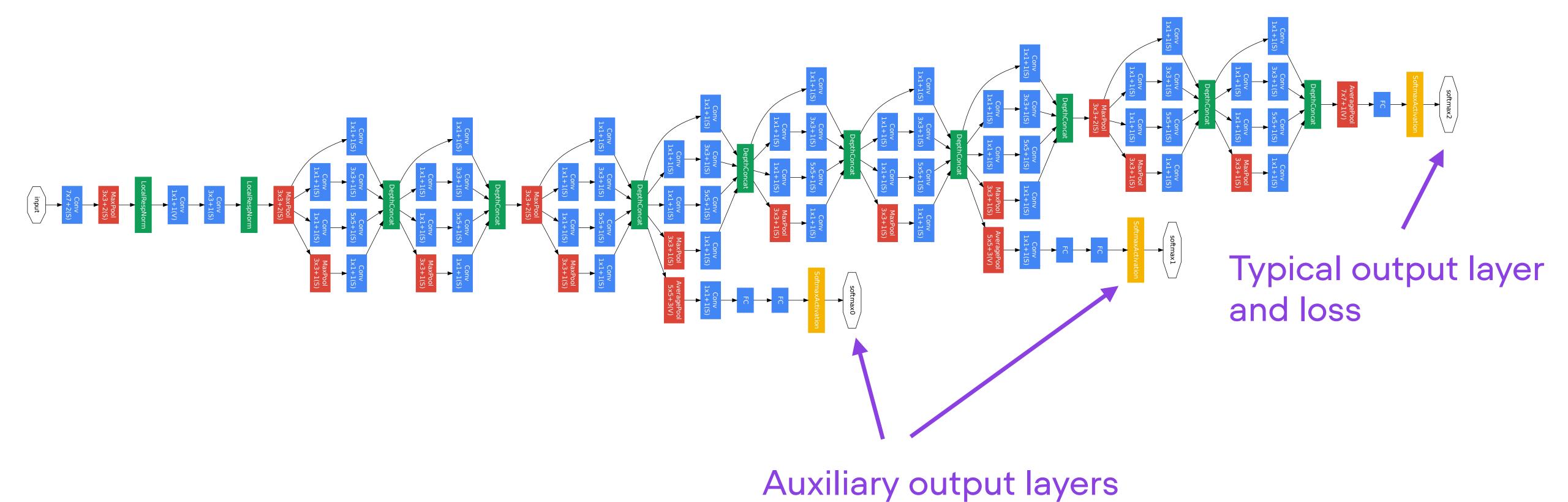
- keeps model smaller
- extracts features at various scales





## ResNet-152 VGG-16 VGG-19 GoogLeNet/Inception GoogLeNet BN-NIN ---125M ---155M BN-AlexNet Operations [G-Ops]



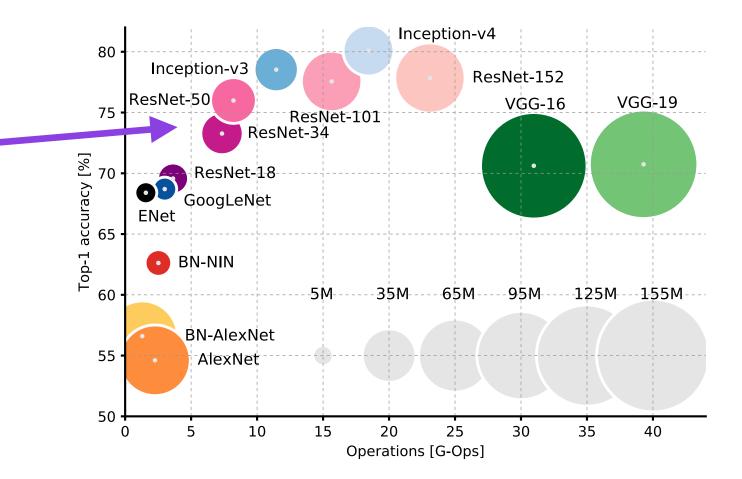


and losses

#### ResNet-34, 50, 101

Even more layers (but smaller)

Key idea: skip connections!

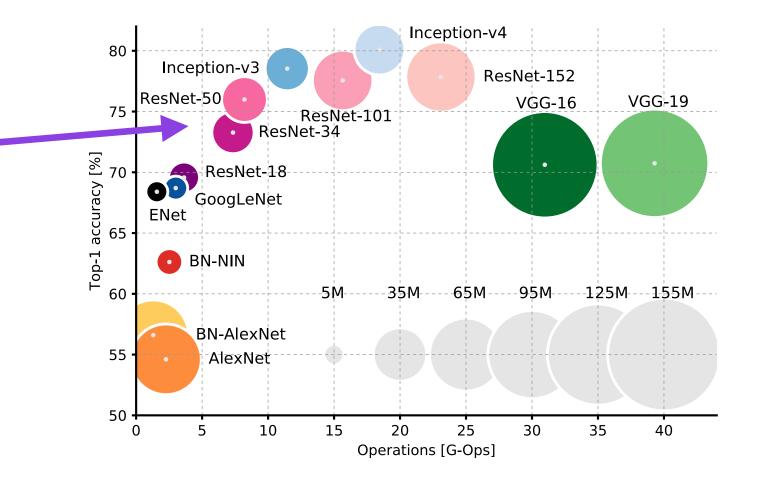


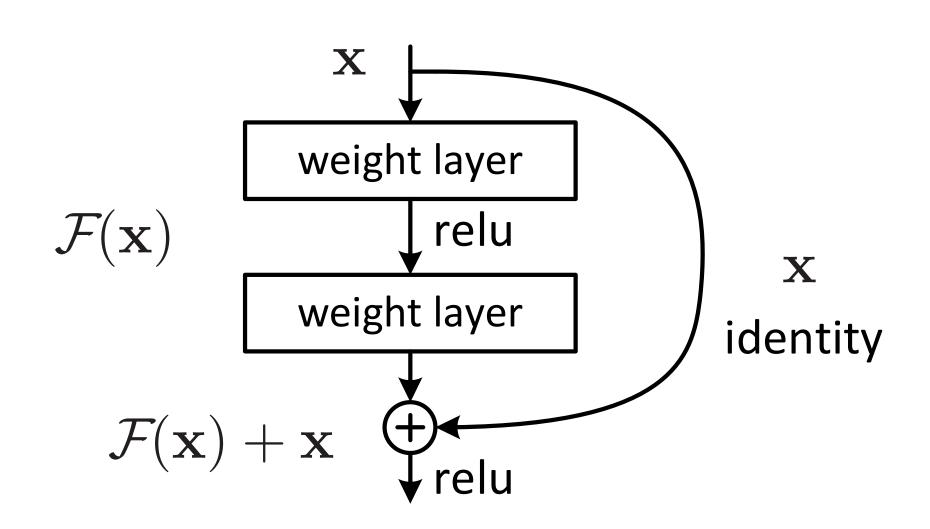
### ResNet-34, 50, 101

Even more layers (but smaller)

Key idea: skip connections!

Can ignore "bad" layers if layers Stronger signal during backprop



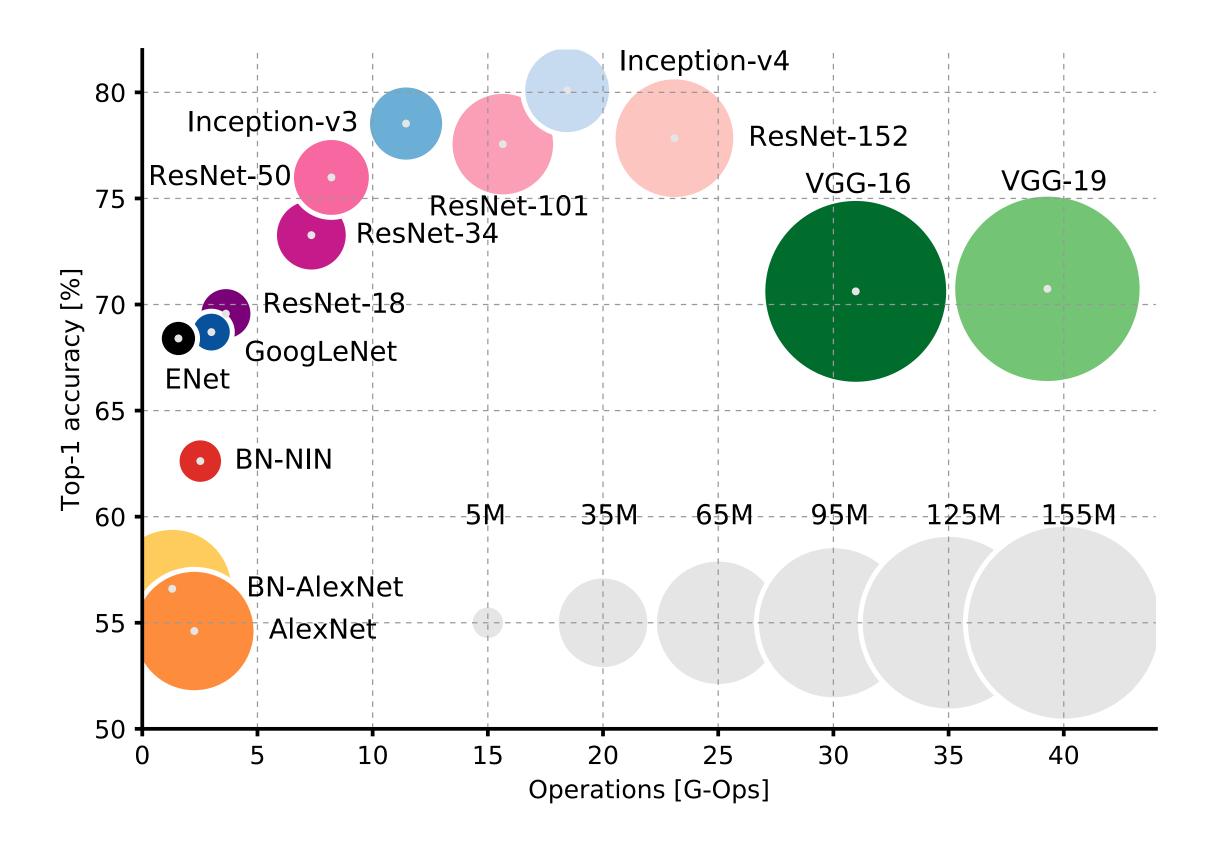


Sebastian Raschka

Deep Learning Fundamentals, Unit 7

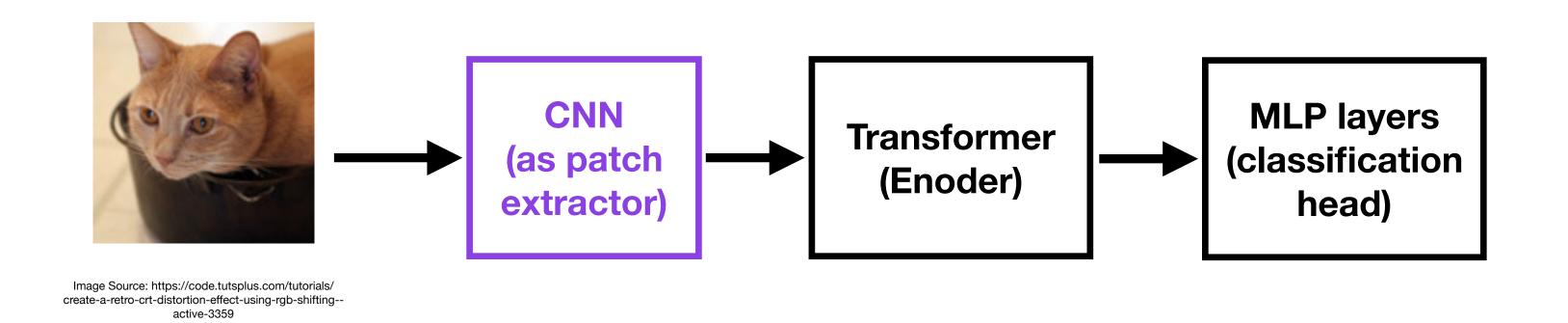
Lightning Al

Many additional architectures were created in recent years



#### Convolutional Vision Transformer Hybrids

Combining convolutional networks and transformers based on self-attention (next unit)



Next: Let's finally train CNNs ourselves!