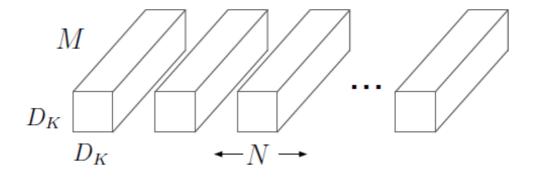
MobileNets: Efficient Convolutional Neural Networks for Mobile Vision Applications

Howard et al. MobileNets: Efficient Convolutional Neural Networks for Mobile Vision Applications. arXiv 2017.

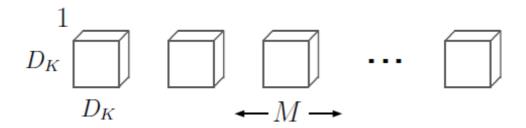
Introduction

This paper proposed a compressing and accelerating method with trading of accuracy.

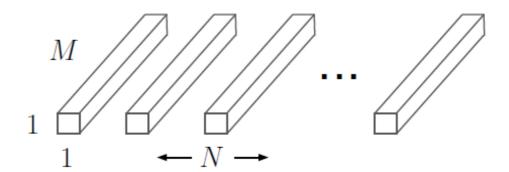
Method



(a) Standard Convolution Filters



(b) Depthwise Convolutional Filters



(c) 1×1 Convolutional Filters called Pointwise Convolution in the context of Depthwise Separable Convolution

Figure 2. The standard convolutional filters in (a) are replaced by two layers: depthwise convolution in (b) and pointwise convolution in (c) to build a depthwise separable filter.

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Basic

Replace **standard convolution filters** with combination of **depthwise convolution filters** and **pointwise convolution filters** and the time complexity can be reduce:

- $\bullet \quad \hbox{Original Time Complexity: } D^2_K NMD^2_F$
- ullet Modified Time Complexity: $(D_K^2+N)MD_F^2$

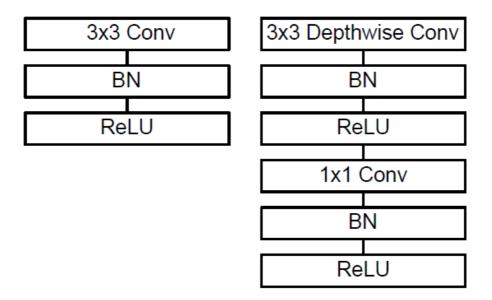


Figure 3. Left: Standard convolutional layer with batchnorm and ReLU. Right: Depthwise Separable convolutions with Depthwise and Pointwise layers followed by batchnorm and ReLU.

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Result

| Model | ImageNet Accuracy | Mult-Add(Million) | Parameters(Million) |
|-----------|-------------------|-------------------|---------------------|
| MobileNet | 70% | 569 | 4.2 |
| GoogleNet | 69% | 1550 | 6.8 |
| VGG16 | 71% | 15300 | 138 |