

DNN_05

Convolutional Neural Network-2

Internal

Why Convolution?

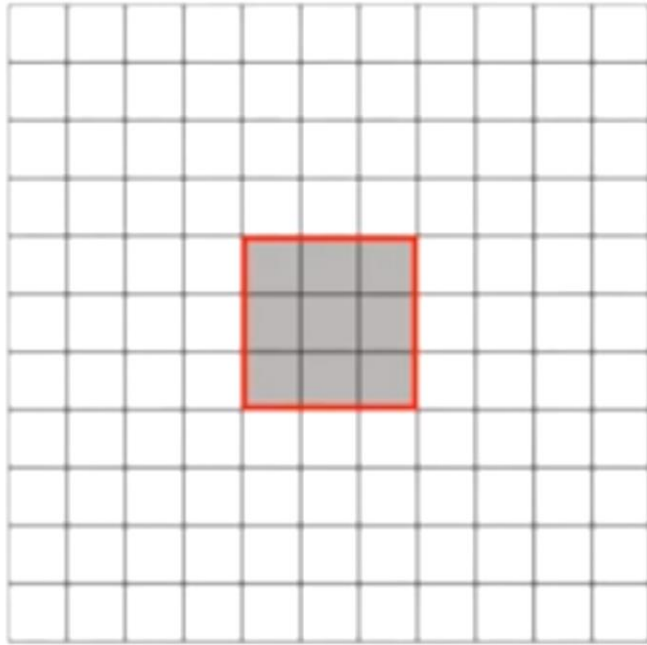
- Parameter sharing .
- Retain Spatial Information.
- Can be scaled for larger Input dimension.
- Can support multiple dimension of Input Image.

Dilated Convolution

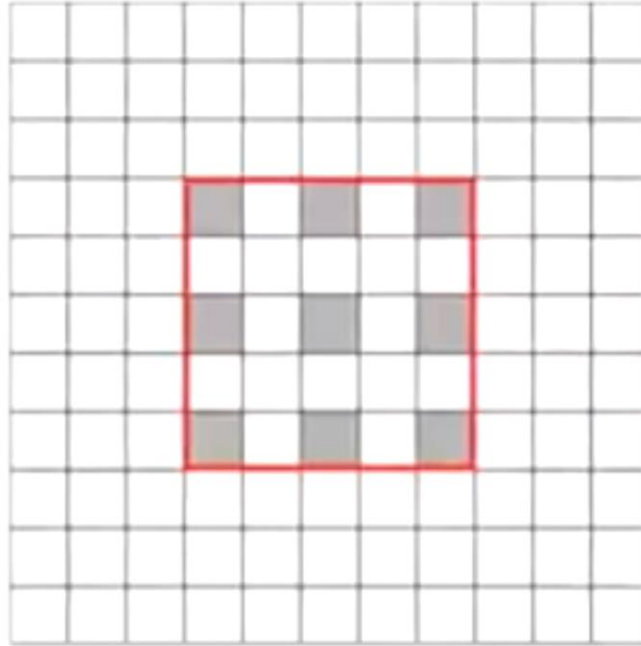
Dilated convolution is just a convolution applied to input with defined gaps. It is a way of increasing receptive field (global view) of a filter.

- Increase Receptive Field of a Filter
- Effective when used in Semantic Segmentation
- Capturing fine Details by processing inputs in higher resolutions.
- Faster Runtime

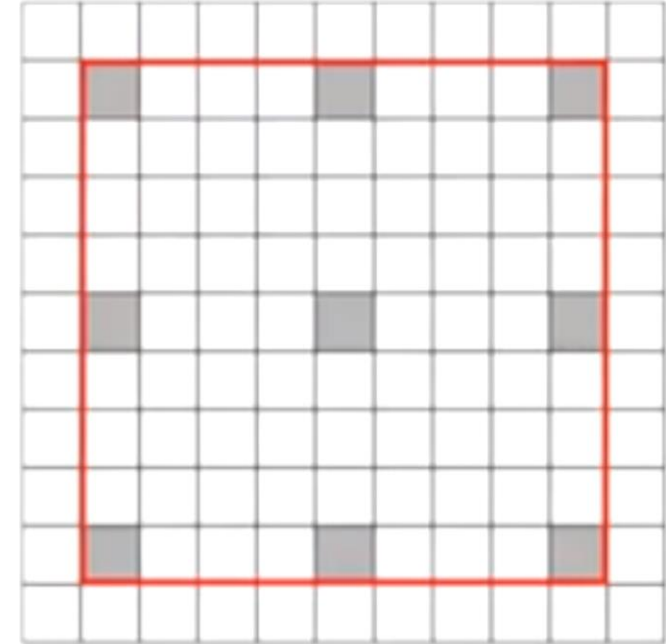
Dilated Convolution



Simple Convolution
Receptive Field : 3×3



Dilation rate $L=2$
Receptive Field 5×5



Dilation rate $L=4$
Receptive Field 9×9

- $L=1$ -> Standard Convolution
- $L=2$ -> Skipping one pixel per input
- $L=4$ -> Skipping three pixels per input

One by One [1 x 1] Convolution

- Also called as Network in Network
- Reduce Number of channel
- Can be used to replace Fully connected Layer
- **Reduce Computational Cost**

One by One [1 x 1] Convolution

1	2	3	6	5	8
3	5	5	1	3	4
2	1	3	4	9	3
4	7	8	5	7	9
1	5	3	7	4	8
5	4	9	8	3	5

6 × 6



6 × 6 × 32

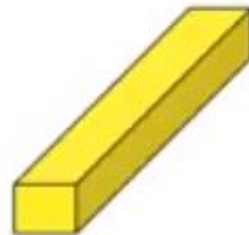
*

2

=

2	4	6	...		

*

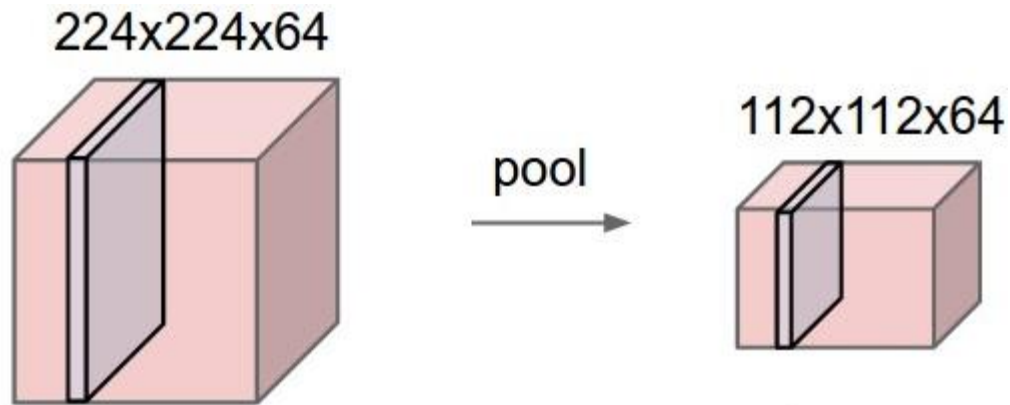


=

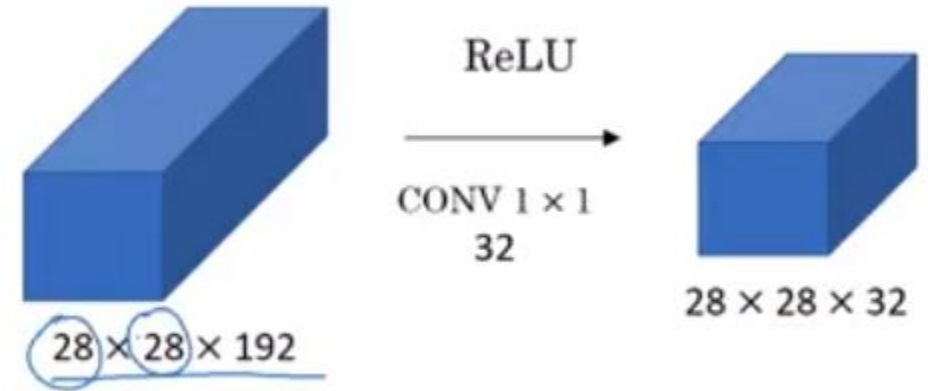
1 × 1 × 32

6 × 6 × # filters

One by One [1 x 1] Convolution



Pooling : Reduce the dimension (W, H)



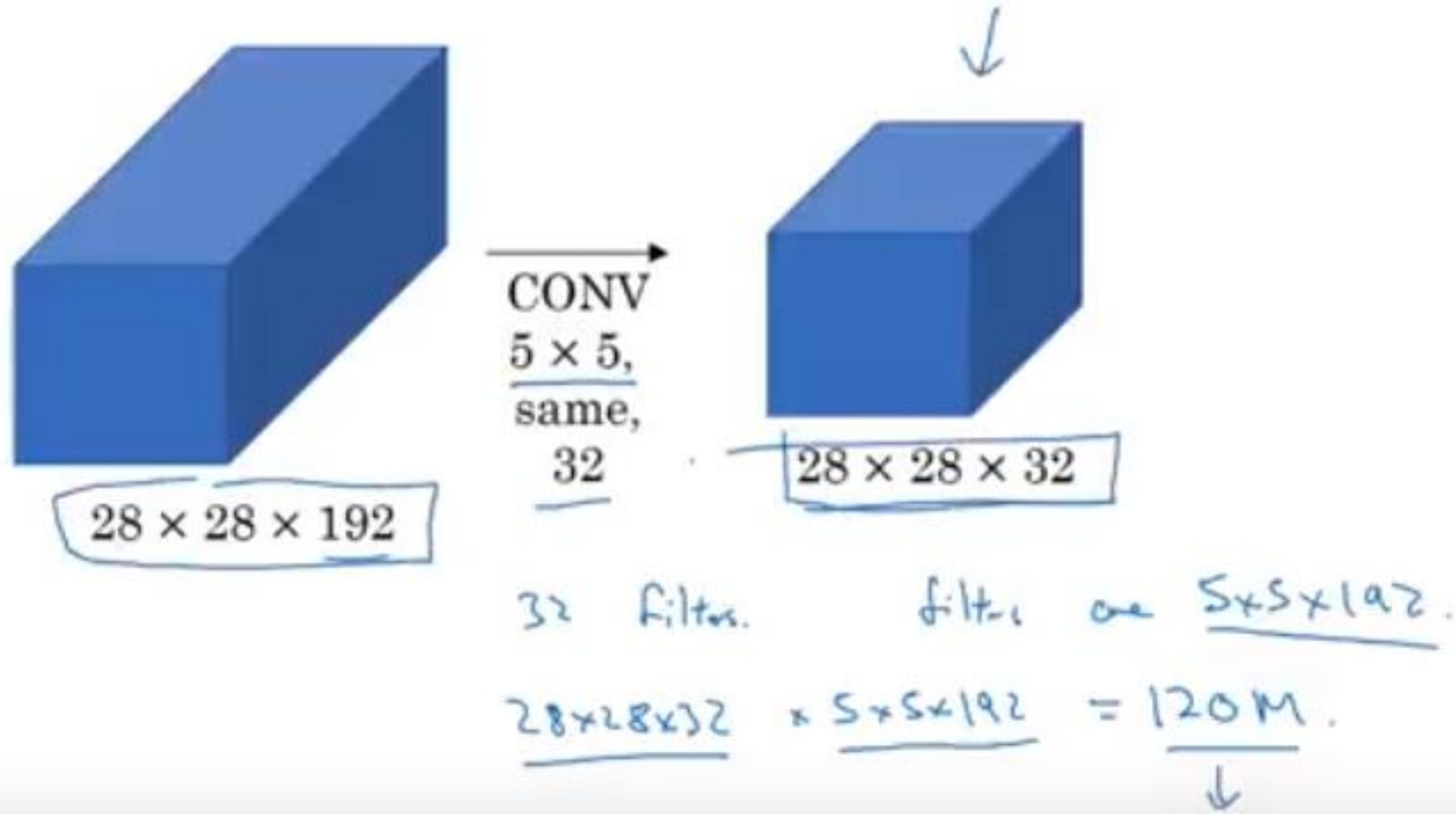
1 x 1 Convolution : Reduce number of channels

Internal

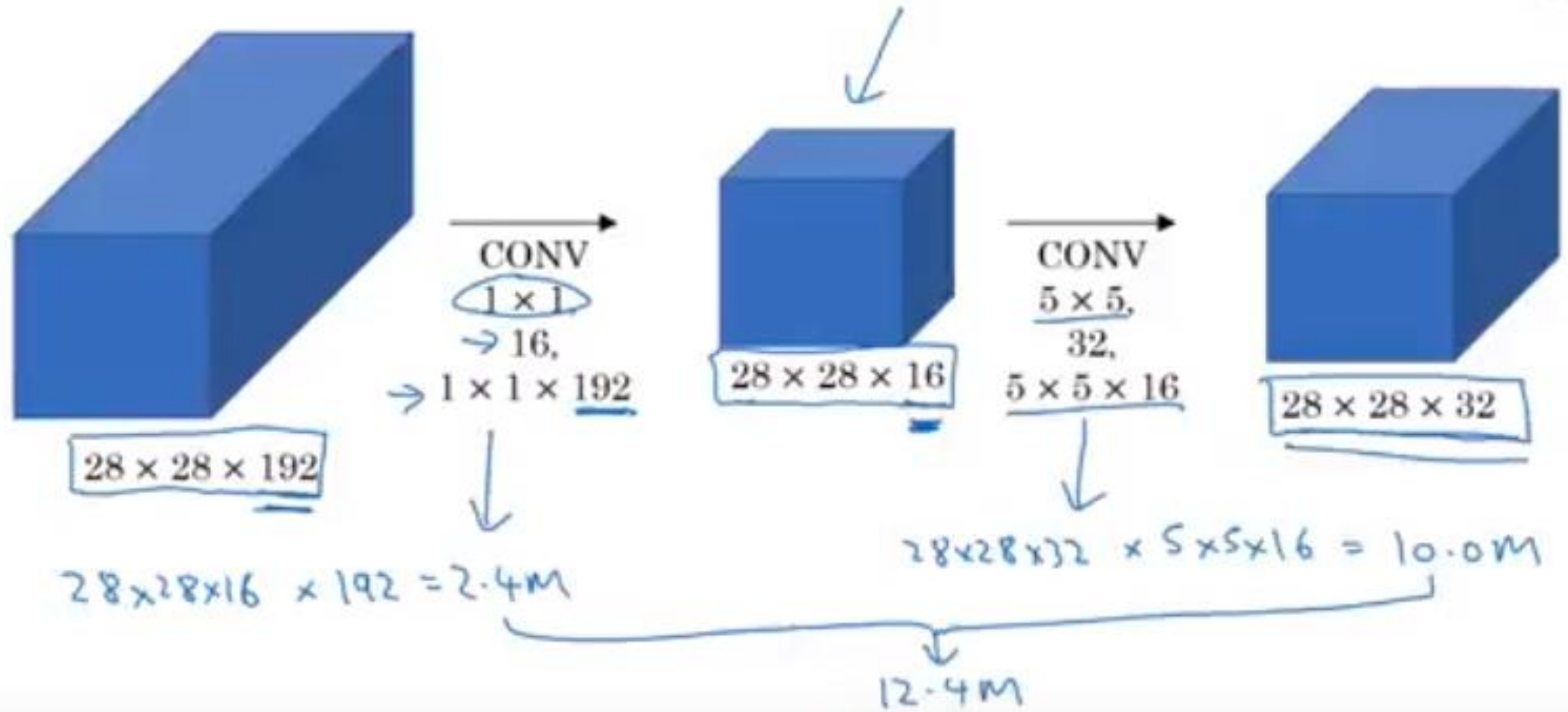
- Pooling layer Reduces the dimension (W and H)
- 1 x 1 Convolution reduces the number of channels

One by One [1 x 1] Convolution

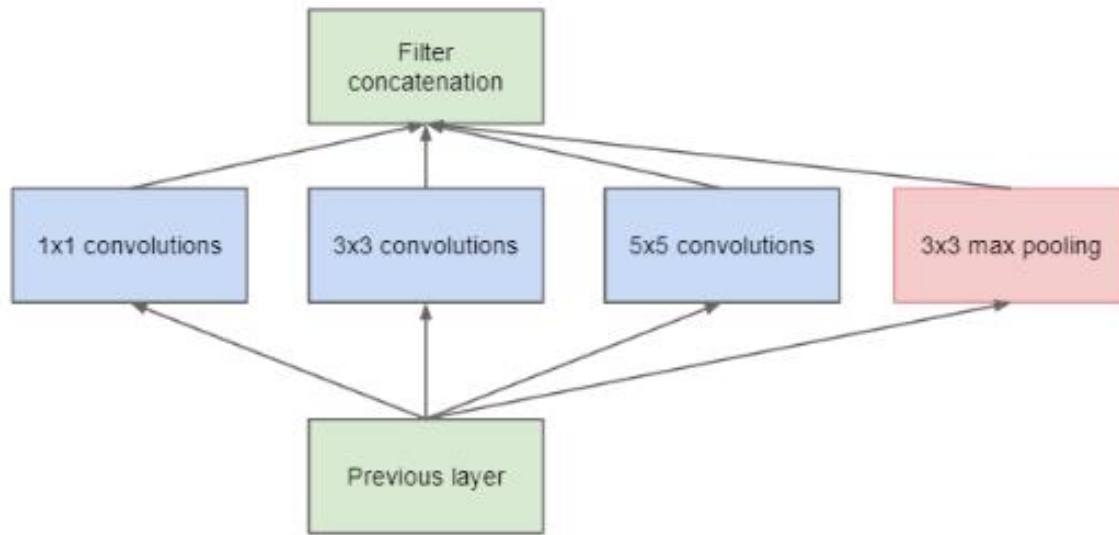
The problem of computational cost



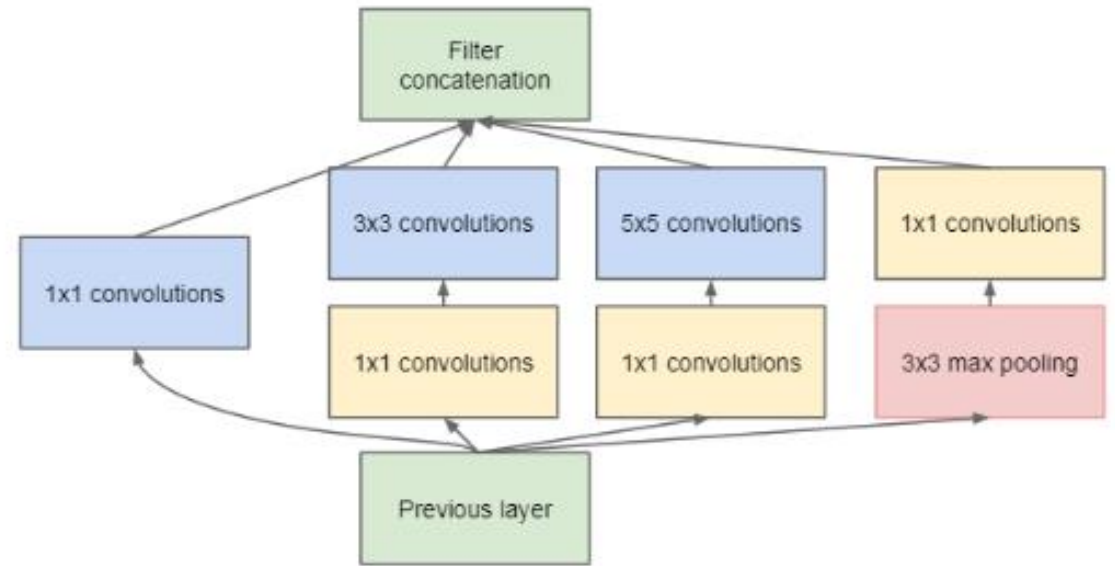
One by One [1 x 1] Convolution



One by One [1 x 1] Convolution in Inception



(a) Inception module, naïve version



(b) Inception module with dimension reductions

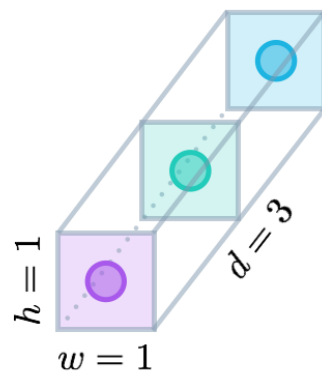
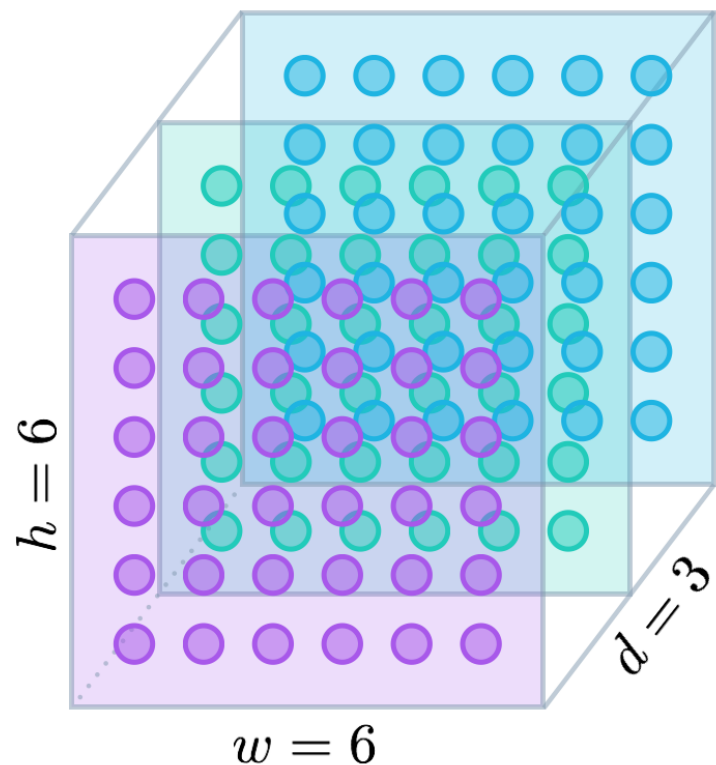
1x1 convolutions in GoogLeNet

Global Average Pooling

- Traditional we use fully connected layers in CNN as the final layer and apply softmax over it.
- The fully connected layers are prone to overfitting and heavily depend on dropout regularization
- global average pooling is itself a structural regularizer, which natively prevents overfitting for the overall structure.

Internal

Global Average Pooling



```
[convolutional]  
batch_normalize=1  
filters=1024  
size=3  
stride=1  
pad=1  
activation=leaky
```

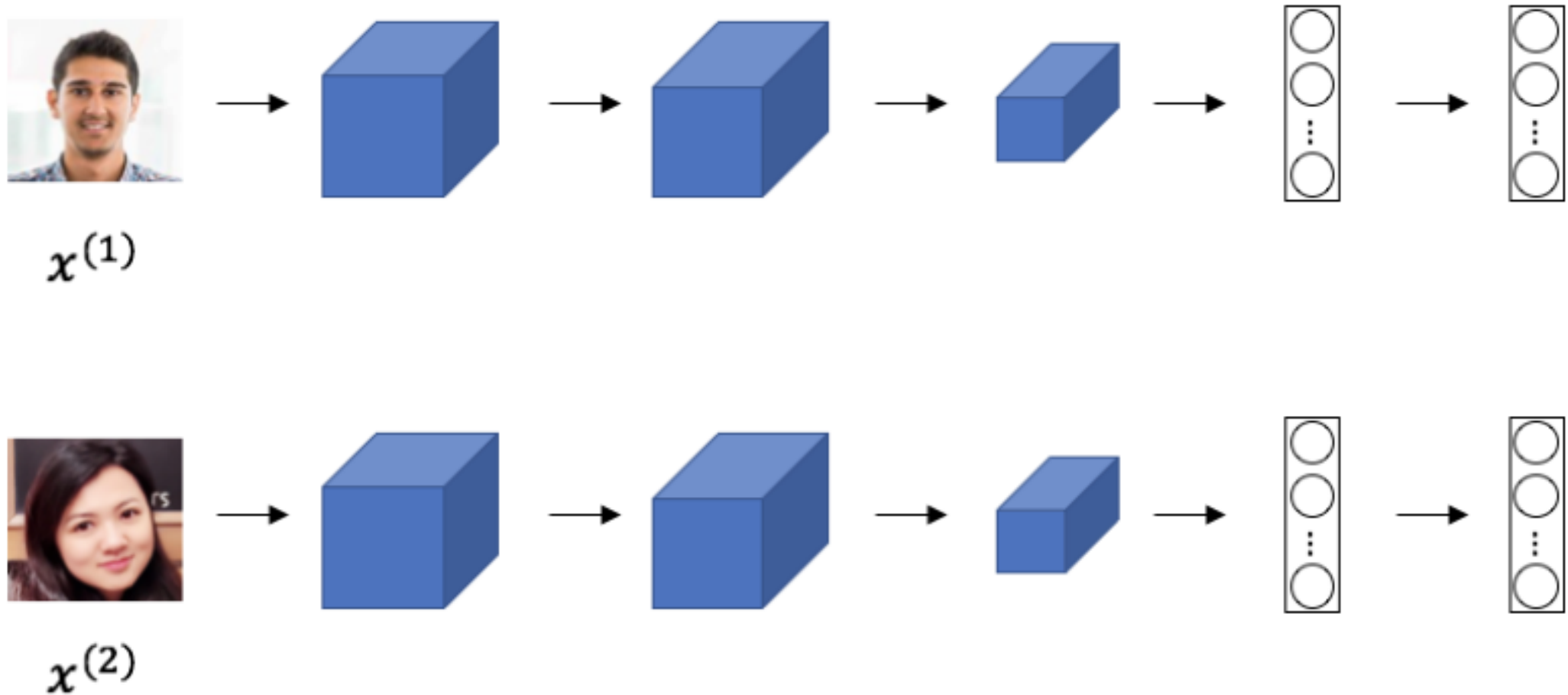
```
[shortcut]  
from=-3  
activation=linear
```

```
[avgpool]
```

```
[convolutional]  
filters=1000  
size=1  
stride=1  
pad=1  
activation=linear
```

```
[softmax]  
groups=1
```

Siamese Network(One-Shot Learning)



Siamese Network

Triplet Loss



Anchor



Positive



Anchor



Negative

$$\mathcal{L} = \max(d(a, p) - d(a, n) + \text{margin}, 0)$$

Reference

- DeepLearning.ai