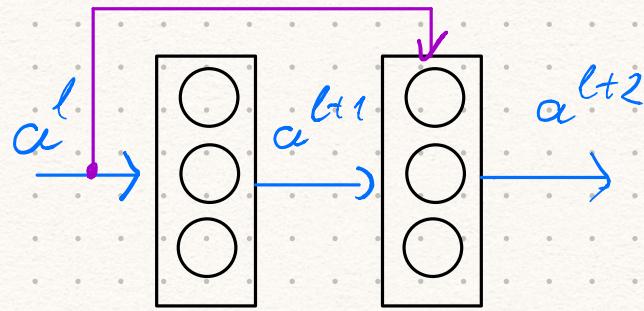
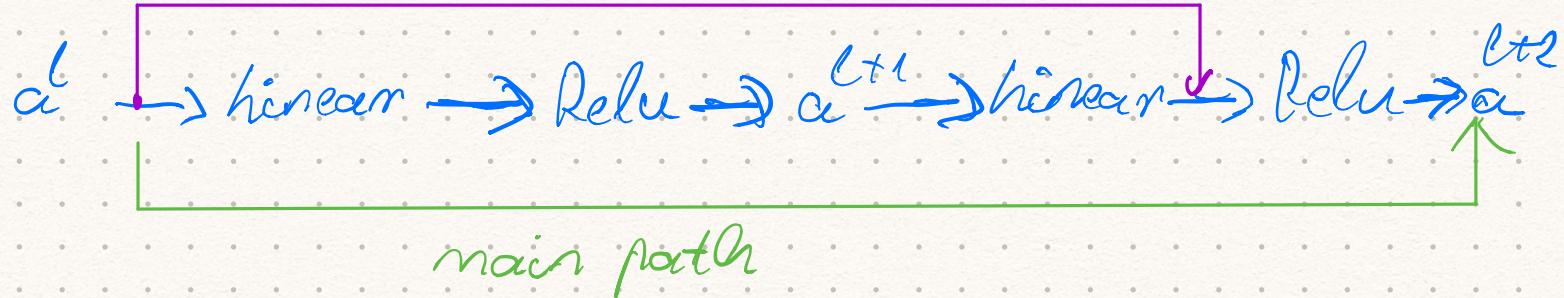


## Residual block



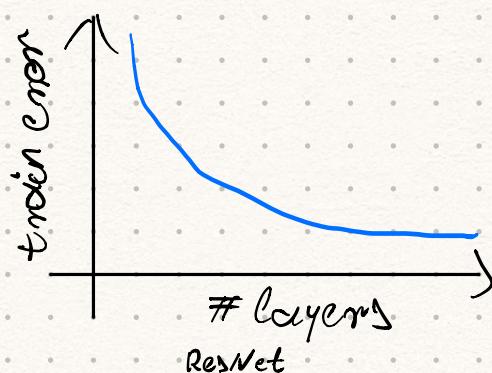
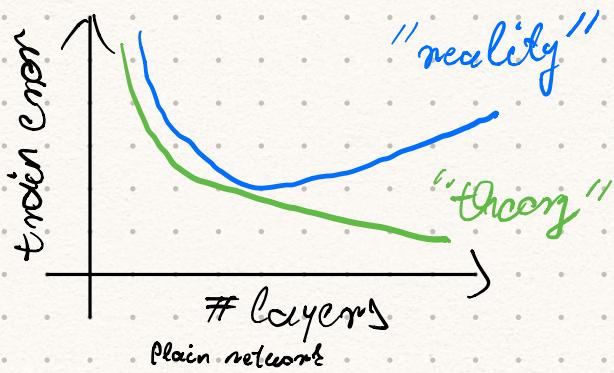
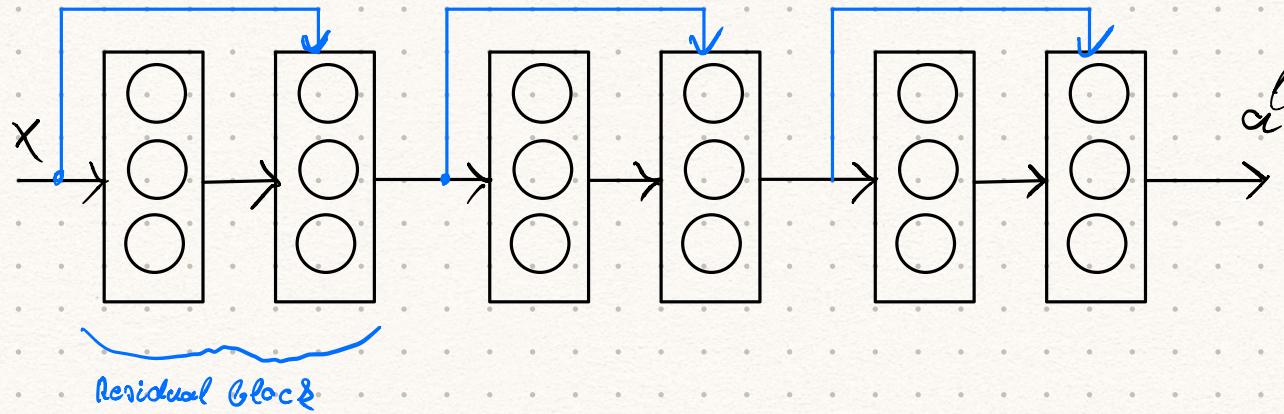
Short cut / skip connection



$$z^{l+1} = w^{l+1}\alpha^l + b^{l+1}, \quad \alpha^{l+1} = g(z^{l+1})$$

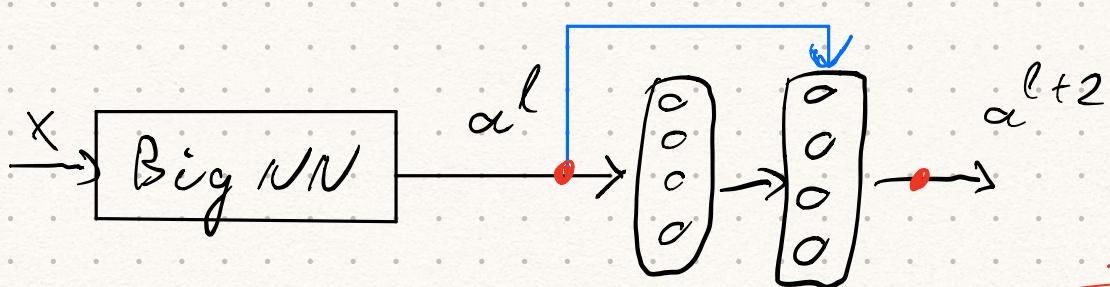
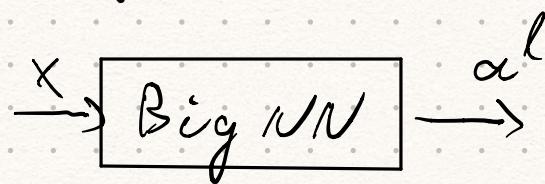
$$z^{l+2} = w^{l+2}\alpha^{l+1} + b^{l+2}, \quad \alpha^{l+2} = g(z^{l+2}) \Rightarrow \alpha^{l+2} = g(z^{l+2} + \alpha^l)$$

## Residual Network (ResNet)



skip connections  $\rightarrow$  allows to train much deeper networks

## Why ResNets work



$$\text{ReLU} \Rightarrow a \geq 0$$

$$a^{l+2} = g(z^{l+2} + a^l) = g(w^{l+2}a^{l+1} + b^{l+2} + a^l)$$

use "same" convolutions to get the same size

If we use L2 regularization or weight decay  
 $\Rightarrow$  shrink the value of  $w^{l+2}$  and  $b^{l+2}$

If  $w^{l+2} > 0$  and  $b^{l+2} = 0$

$$\hookrightarrow a^{l+2} = g(a^l) = a^l \Leftrightarrow \text{because of ReLU}$$

Because of skip connection  $\Rightarrow$  identity function is easy to learn  $\Rightarrow$  adding more layers do not change the original structures

# Network in network - $1 \times 1$ convolutions

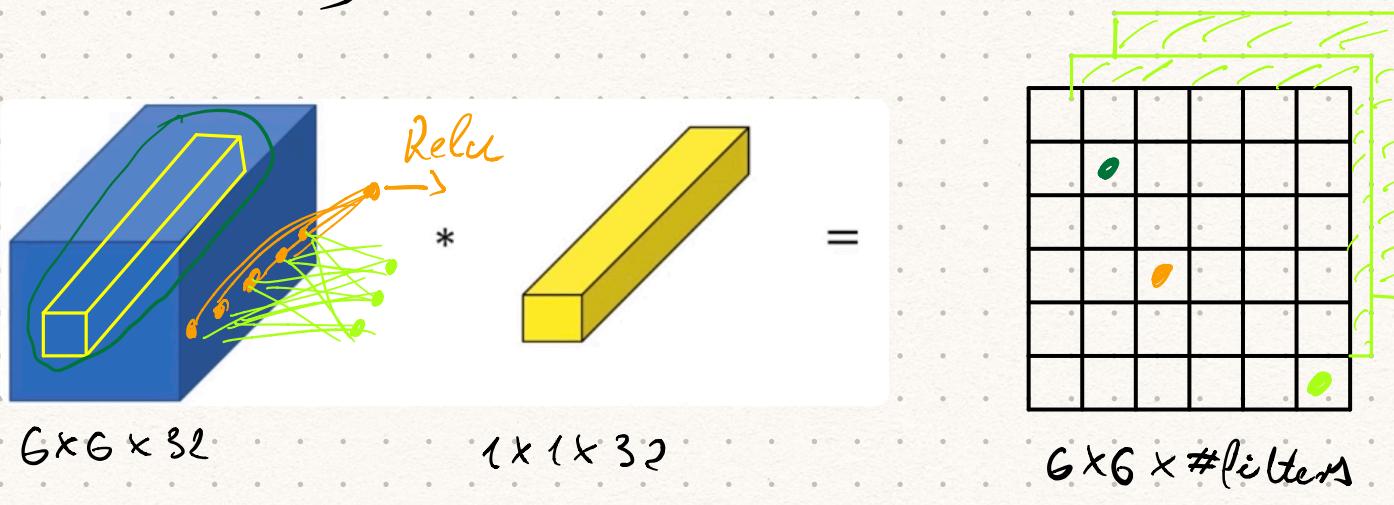
|   |   |   |   |   |   |
|---|---|---|---|---|---|
| 1 | 2 | 3 | 6 | 5 | 8 |
| 9 | 4 | 1 | 9 | 2 | 0 |
| 0 | 2 | 1 | 5 | 6 | 2 |
| 1 | 2 | 3 | 2 | 4 | 6 |
| 1 | 2 | 3 | 0 | 7 | 8 |
| 0 | 5 | 4 | 1 | 2 | 9 |

$$\ast \quad 2 \quad =$$

|   |   |   |    |    |
|---|---|---|----|----|
| 2 | 4 | 6 | .. | .. |
| - | - | - | -  | -  |
| - | - | - | -  | -  |
| - | - | - | -  | -  |
| - | - | - | -  | -  |

with one filter  $\Rightarrow$  just a multiplication

But  $\Rightarrow$



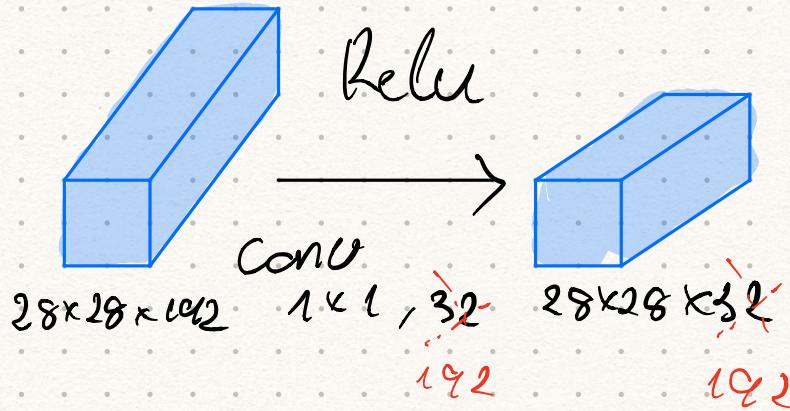
take element-wise product and apply ReLU  $\Rightarrow$  single real number

- if # filters = 1
  - if # filters > 1  $\Rightarrow$  fully connected network
- 32 input filters  $\rightarrow$  # filters output values  
 $\hookrightarrow$  network in network

## Using $1 \times 1$ convolution

Pooling  $\rightarrow$  shrinking width and height

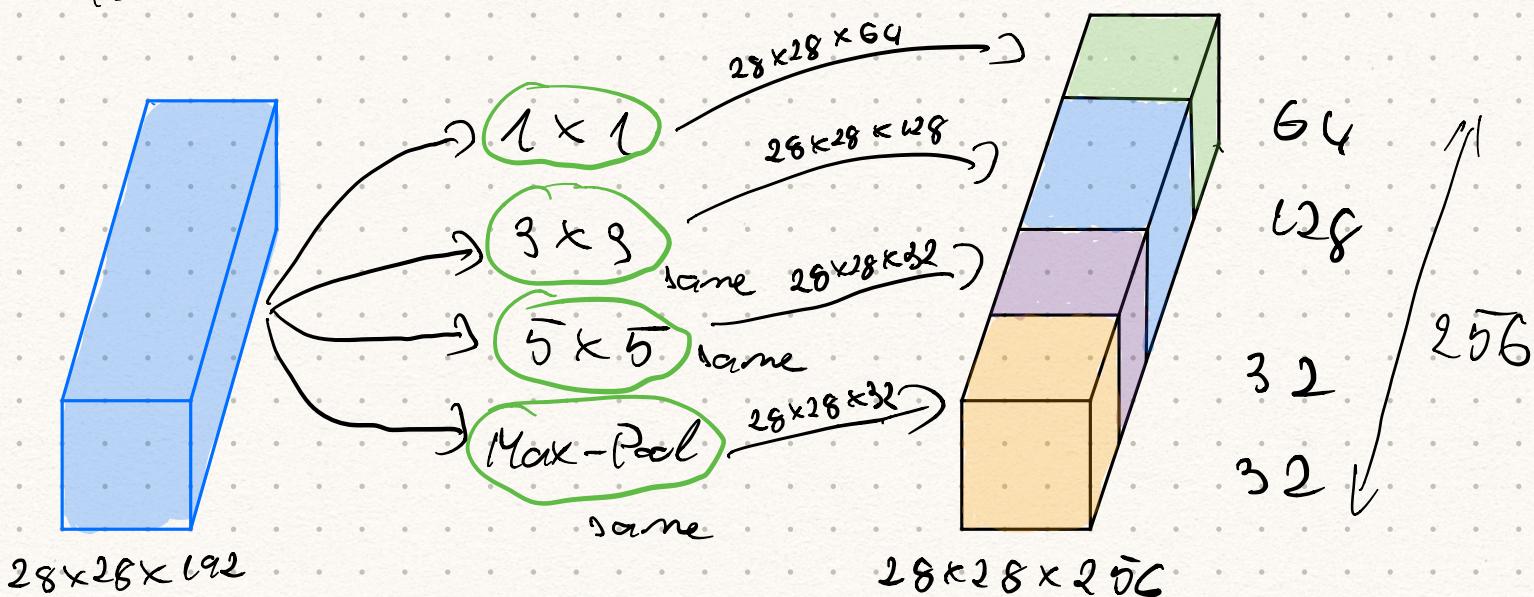
To shrink # filters  $\rightarrow$  use  $1 \times 1$  convolutions



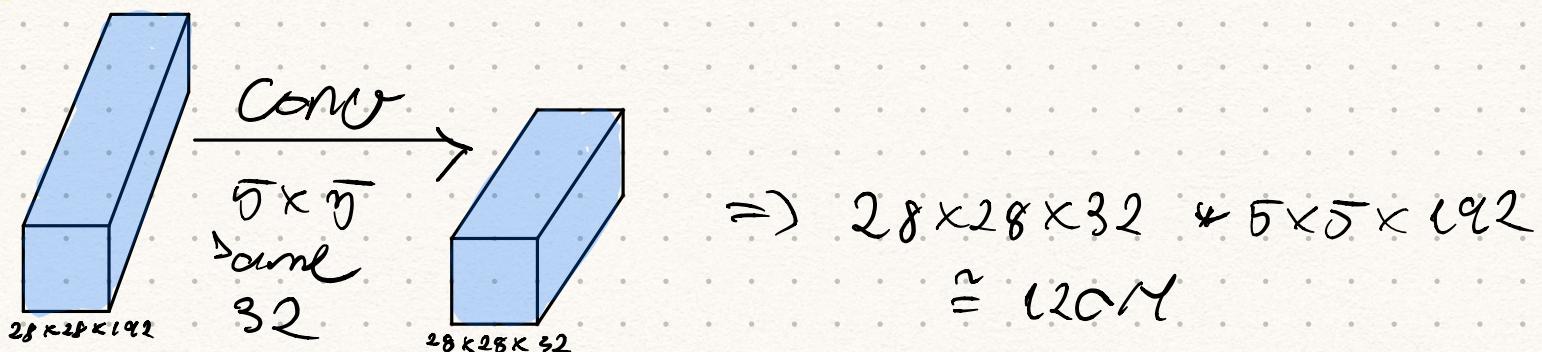
If # filters is the same  $\Rightarrow$  to learn more complex functions

## Inception network

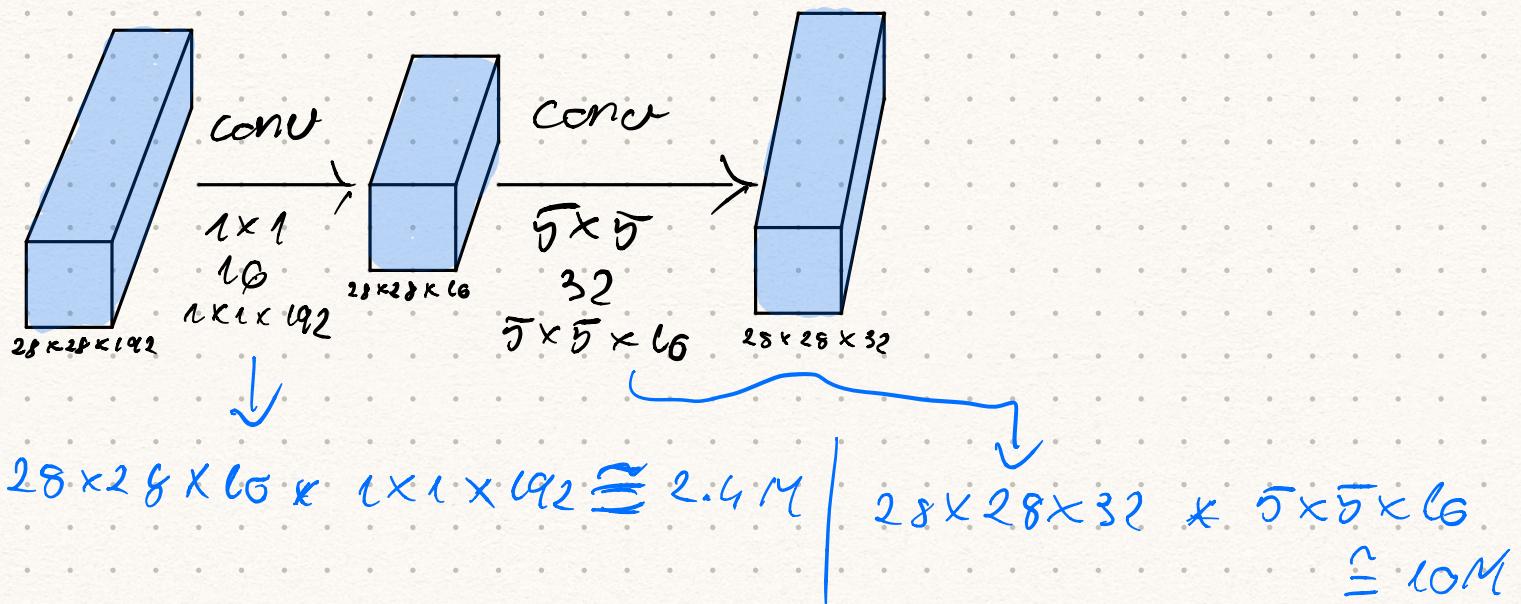
Choosing the conv., and pooling filter size is difficult  $\rightarrow$  don't choose use more



## The problem of computational cost

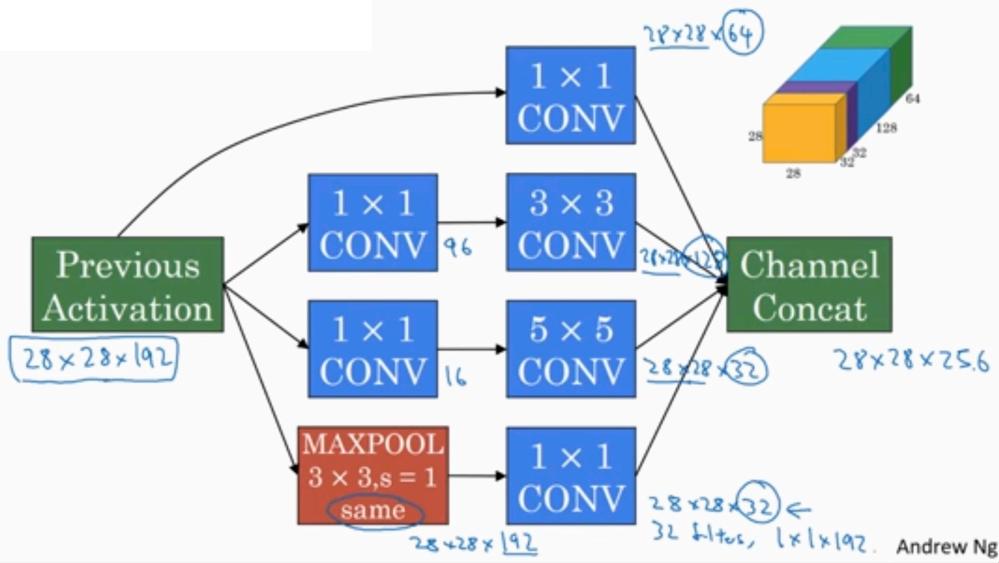


$\Rightarrow$  reduce computation cost

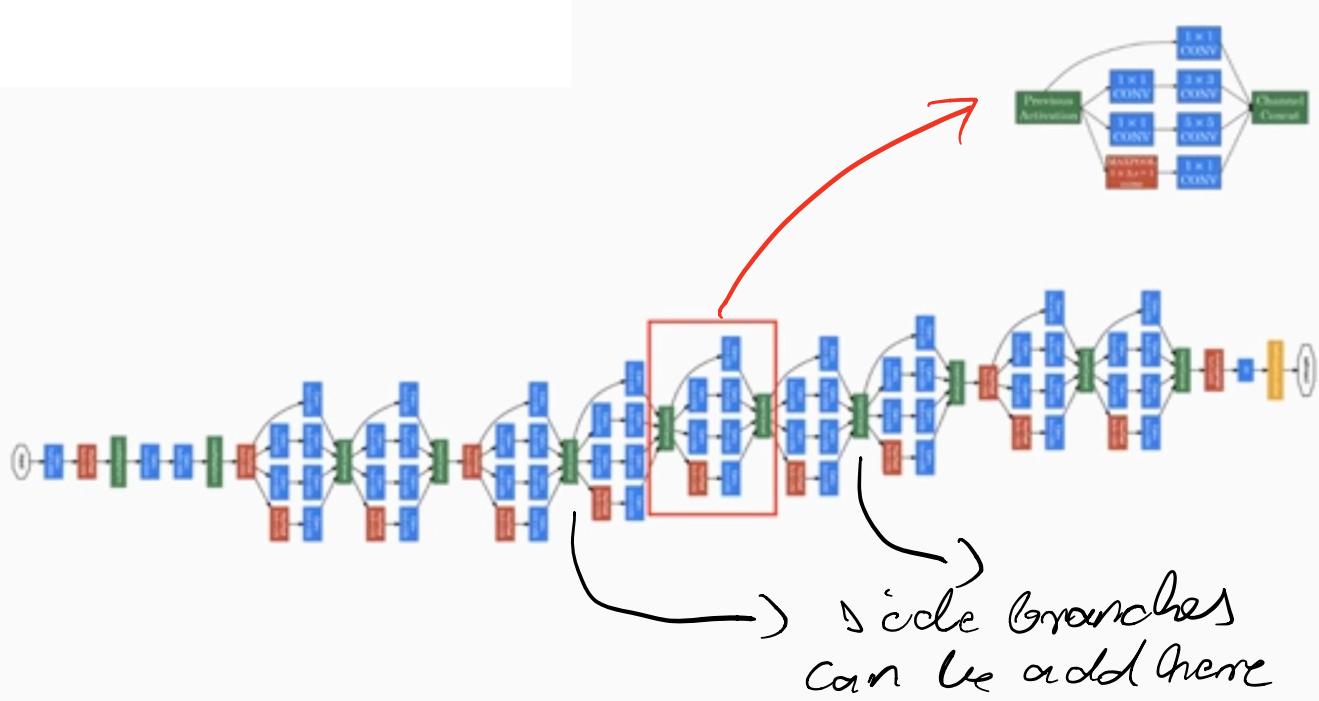


$\approx 12.4M \Rightarrow 10$  times smaller

## Inception module



# Inception network ~ google net



[Szegedy et al., 2014, Going Deeper with Convolutions]

Andrew Ng