→ Classical Architectures.

LeNet, AlexNet, VGG.

[ھایدى ئەملا بىرتىپ ئۇرايلايگوندا!]

Inception   Google   2014

↓

Inception module

Inception Module

Filter concatenation

32
3 ⊡ ▯ → (256)

32x32x64

smaller feature →

32x37x128

32x32x32

32x37x32

1x1 convolayer | 3x3 convlayer | 5x5 convlayer | 3x3 maxpool

x

160million

dimentionality reduction

previouslayer

32x32x200

32x32x200

32
31
200

$\rightarrow$ Conv (1 × 1 × 64)
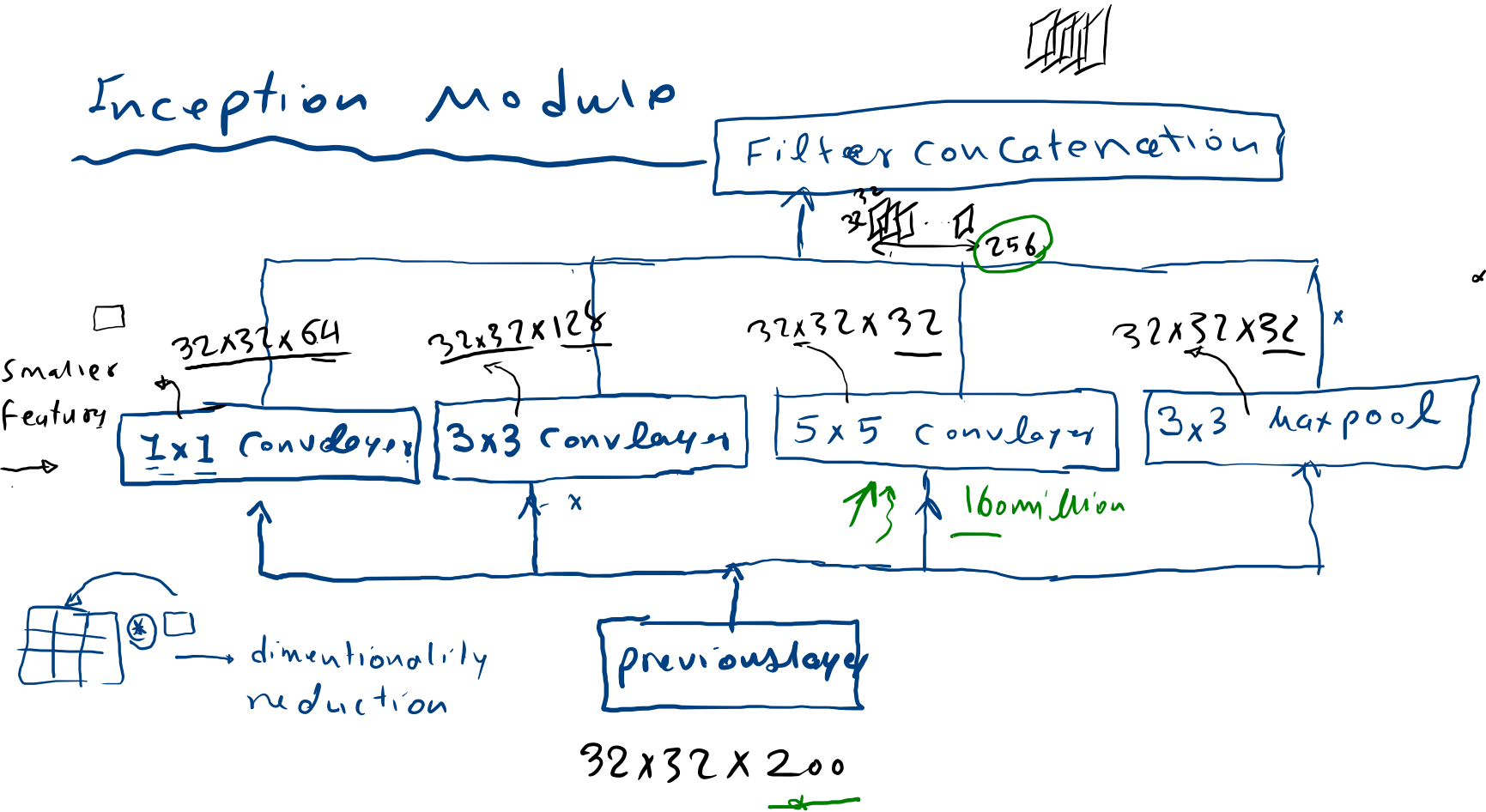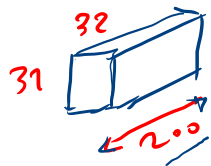
32
32
64
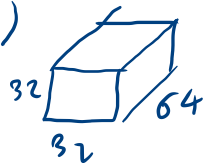
maxpool

down sampling
(W, y)

Con (1×1) $\longrightarrow$ depth

Con (64)
200

64
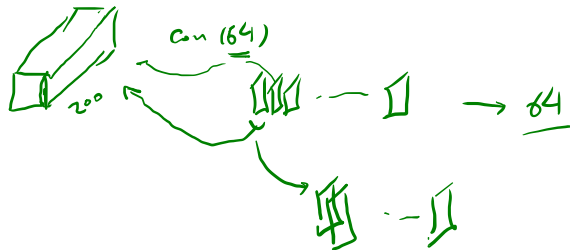
Inception module ~~plus~~ disadvantage.

high computational cost ⟿
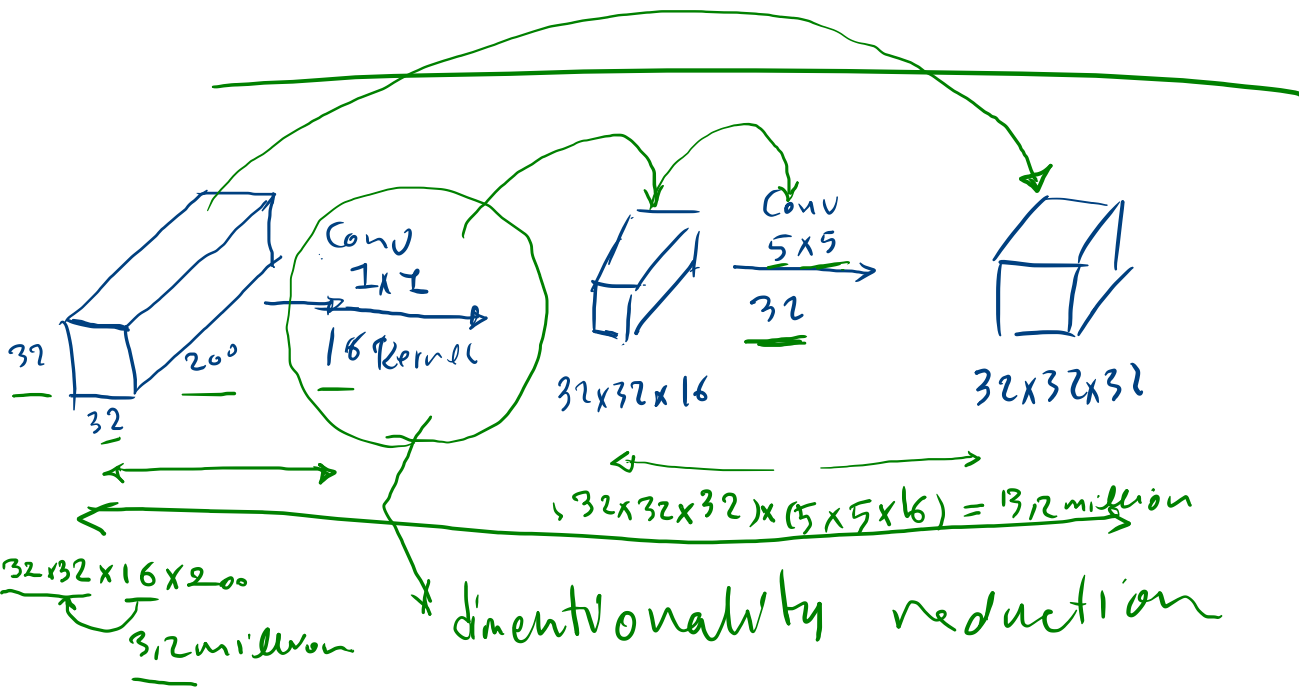
≈ $\underline{16}$ million

Conv
1×1
16 Kernel
32×32×16

Conv
5×5
32

32×32×32

32
32
200

32×32×16×200
3,2 million

(32×32×32)×(5×5×16) = 13,2 million

* dimentionality reduction   Conv (1×1)

Inception with Dimentianality reduction

Depth concatenation.

3x3 conv
5x5 con
1x1 conv

1x1 conv
1x1 conv
1x1 conv
3x3 Max pool

X previous layer

Google Net

inp
conv
maxp
3x3 conv
1x1 conv
maxp

part A

x2

maxp
3

x5

maxp

x2

avgpool
Dense
(softmax)
Flatten

part C: classifier

part B: 9 inception module
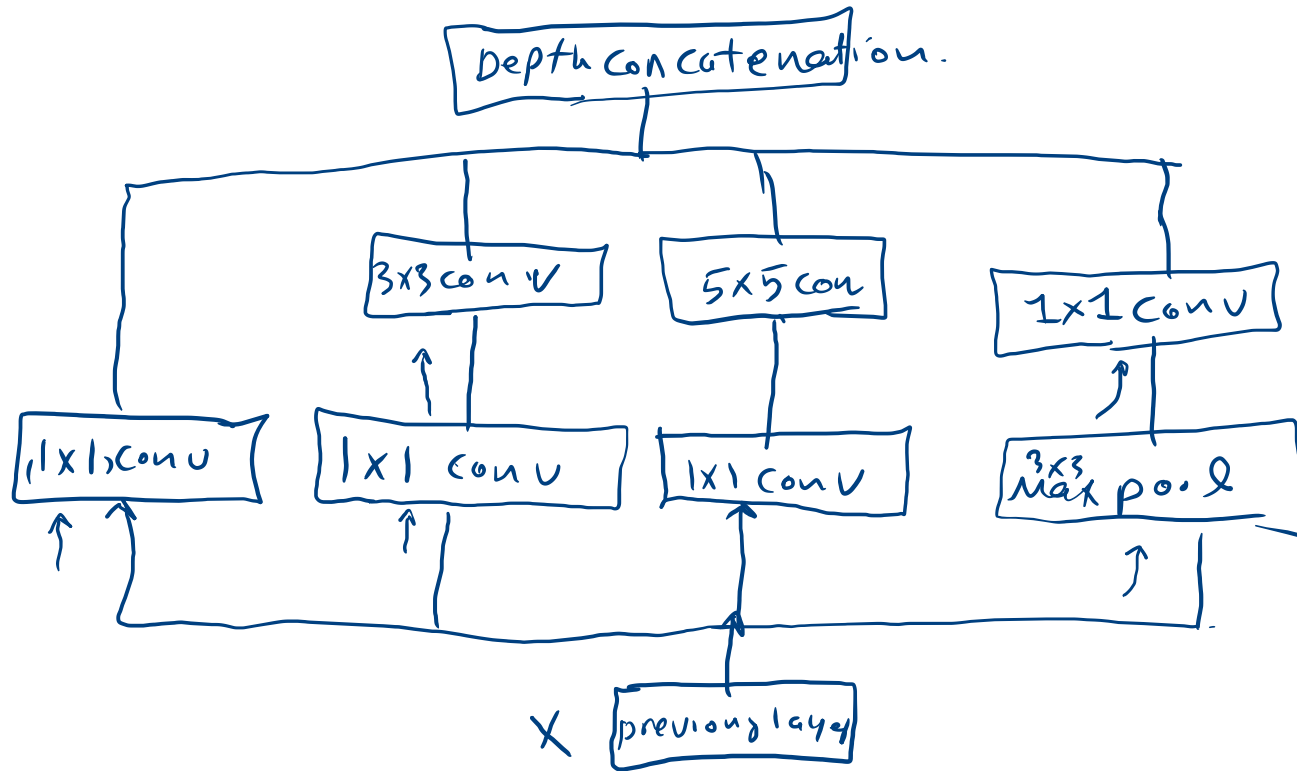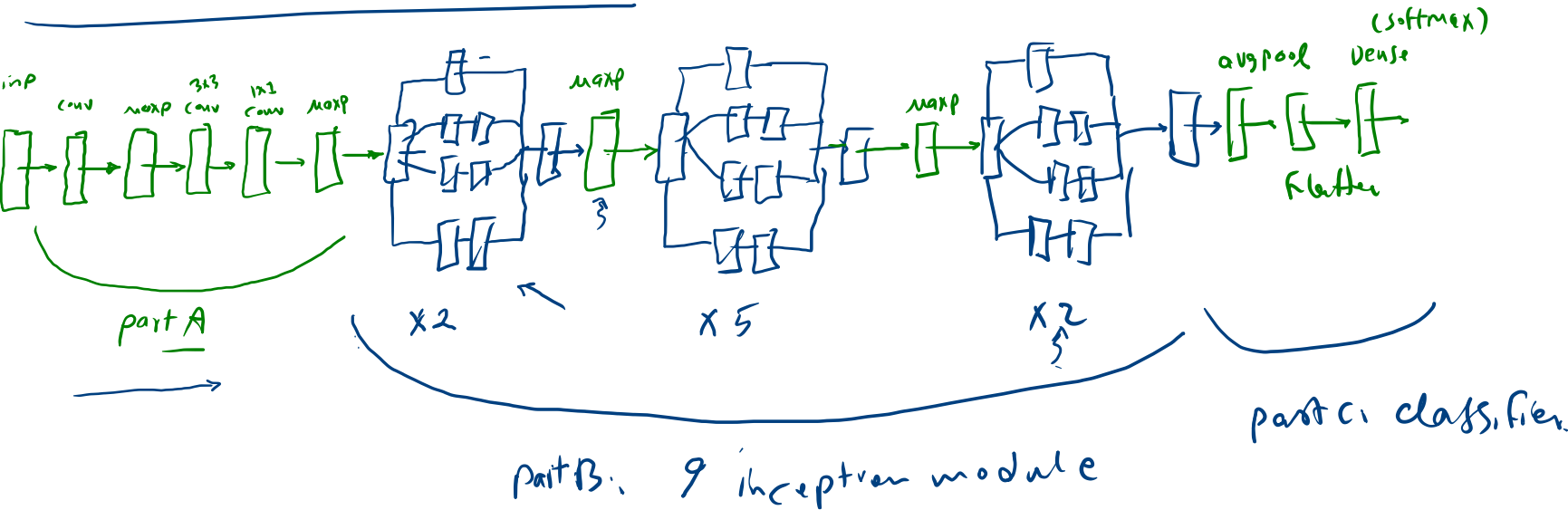
ResNet

Vanishing gradient

اگر گیریم مدل نظر از مقدار معمول عمق شود می شکی کنی؟

Shortcut path

x

| Conv | Relu | → | Conv | → | relu | . . . .   loss

$0/x$
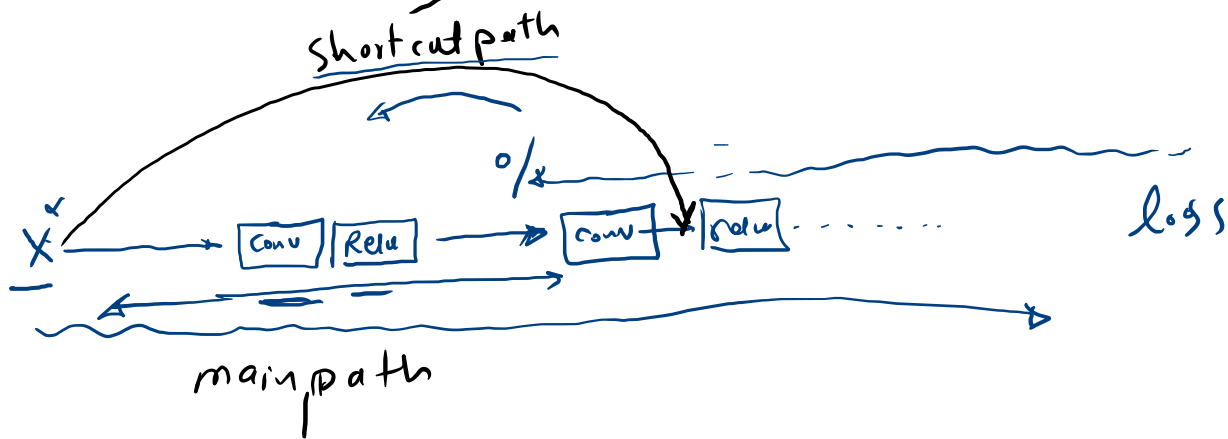
main path

ResNet $\longrightarrow$ Super deep Neural Networks.

shortcut path

X

X

X + f(x)

$\boxed{conv}$ | $\boxed{Relu}$ $\longrightarrow$ $\boxed{conv}$ $\xrightarrow{f(x)}$ $\oplus$ $\longrightarrow$ $\boxed{relu}$ $\longrightarrow$ Relu (f(x)+X)

main path

Residual Block

shortcut path

Conv 1×2 × depth · BN

$X_{shortcut}$

X → Conv2D | BN | Relu → Conv2D | BN | Relu → Conv2D | BN → ⊕ → Relu

1

2

3
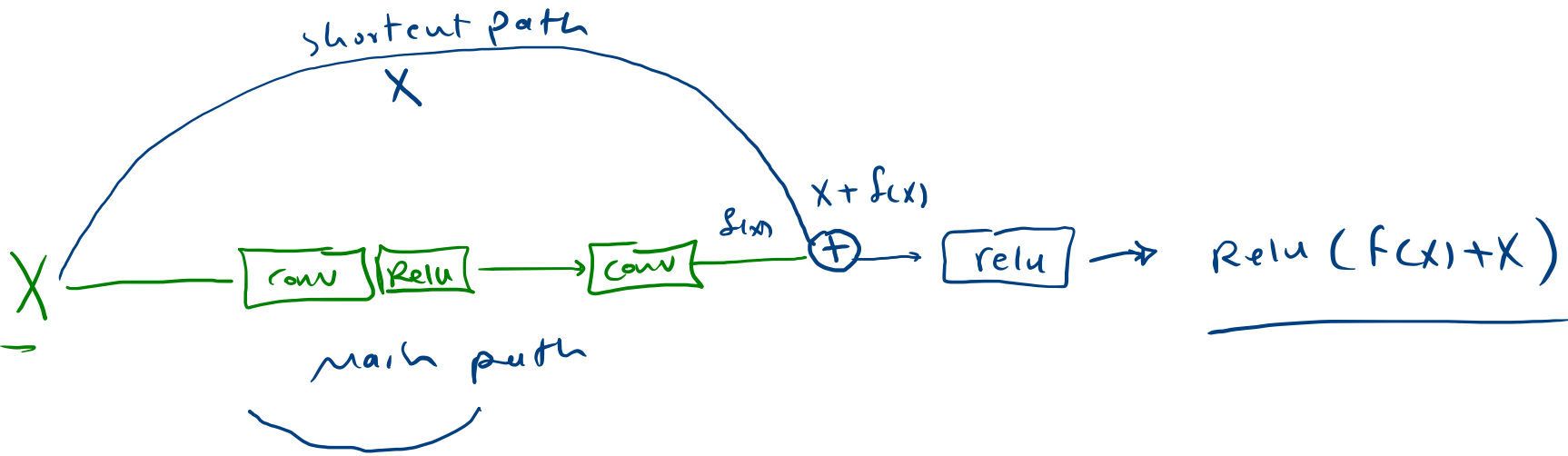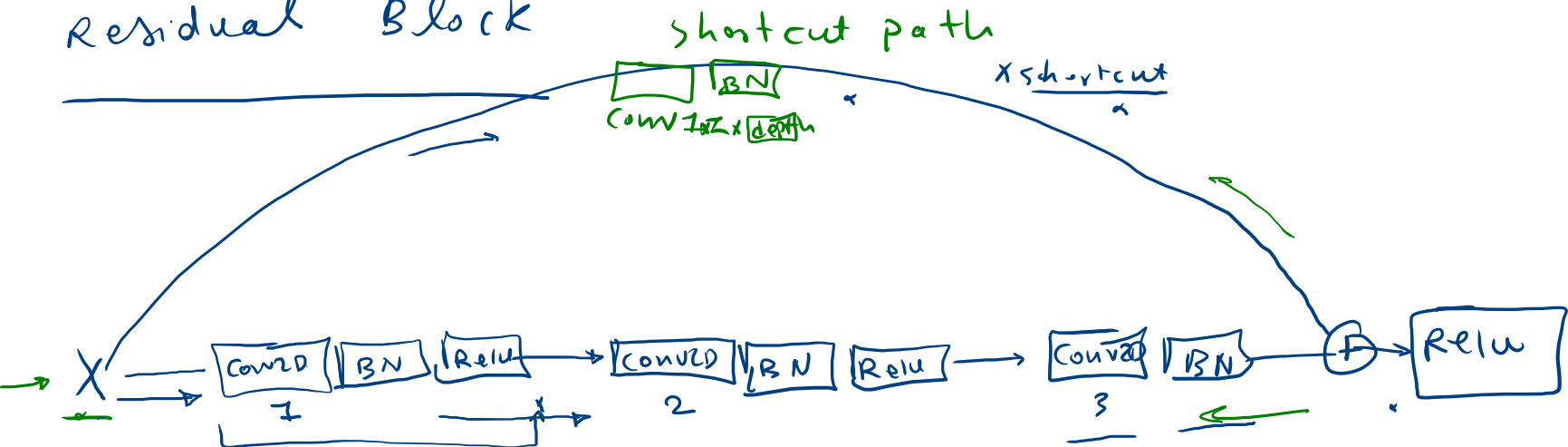
if res-block == True

res block

else.

Main path.

ResNet 50

Stage 1 $\longrightarrow$ 7x7 conv layer ~~7x7~~

Stage 2 $\longrightarrow$ 3 ~~3~~³ residual block (9 conv layers)

Stage 3 $\longrightarrow$ 4 res block (12 conv layers)

Stage 4 $\longrightarrow$ 6 RB (18 conv layers)

Stage 5 $\longrightarrow$ 3 RB (9 conv layers)

Fully connected layer

The End