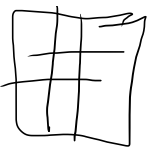
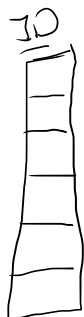


CNN-Coding



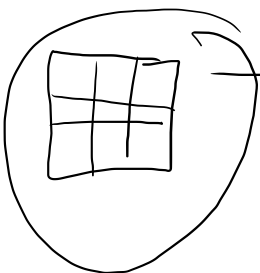
Dense →



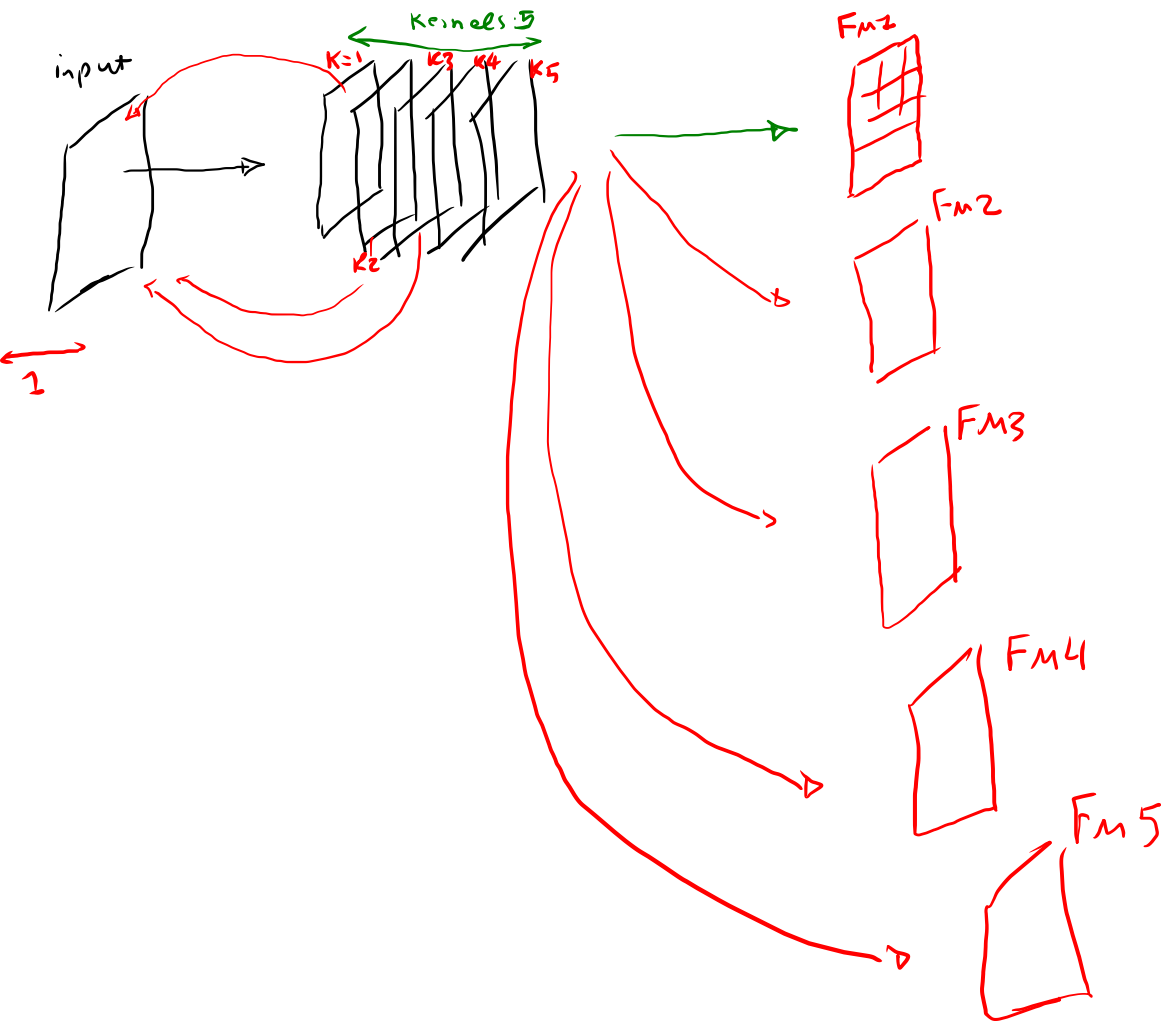
Dense
0
0
0
0

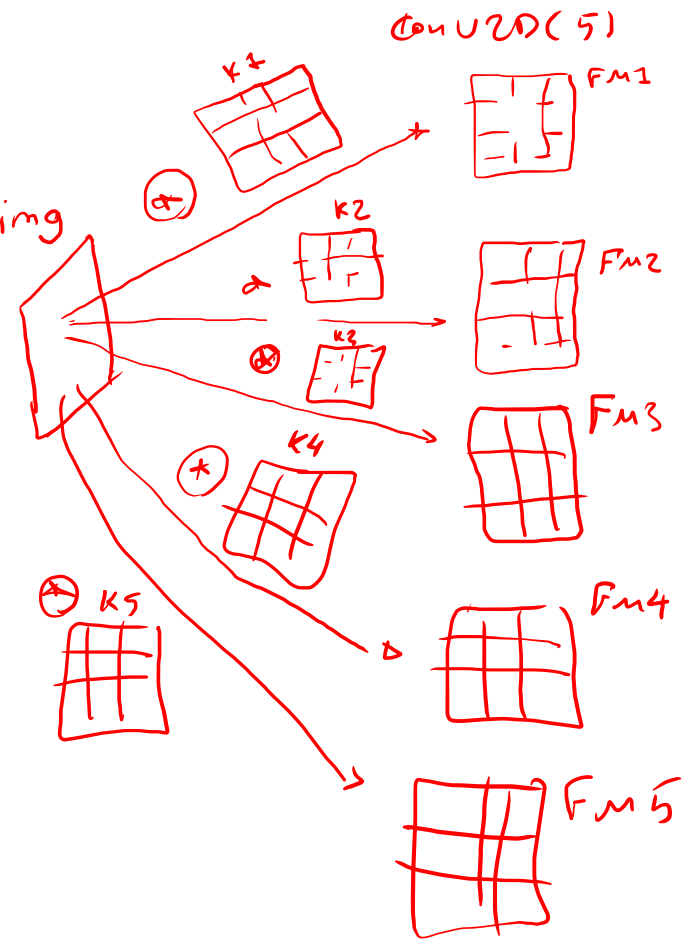
input

output
- 0 →
0 →
0 →

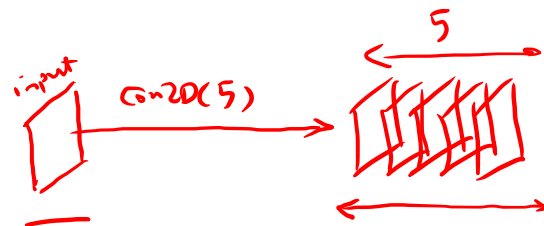


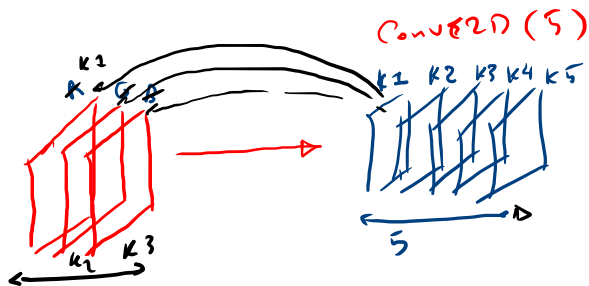
CNN





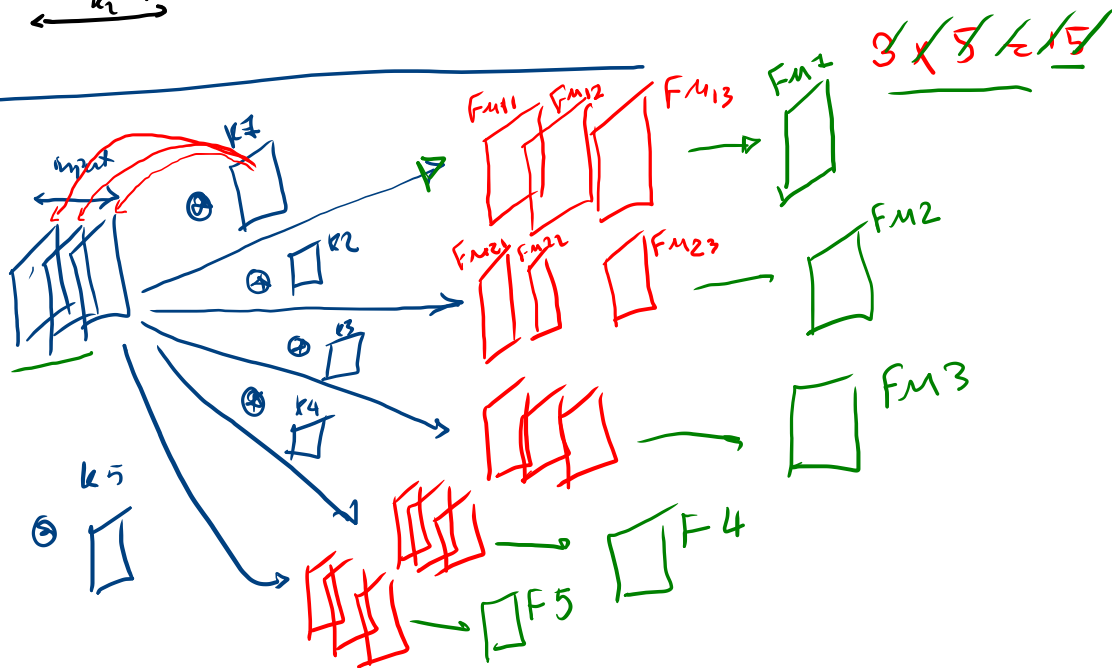
$$K=5 \longrightarrow F_M=5$$



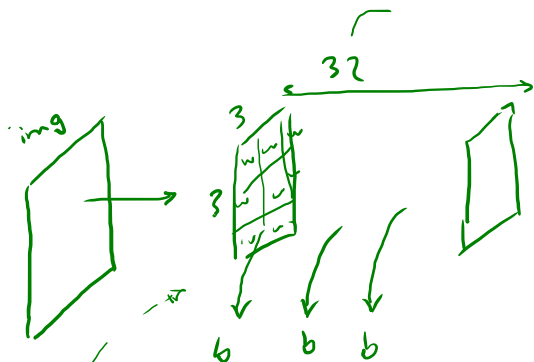


$$\text{net} \rightarrow 0$$

$$\text{net} = w_1 x_1 + w_2 x_2 + w_3 x_3 \dots$$



conv2D
layer 2

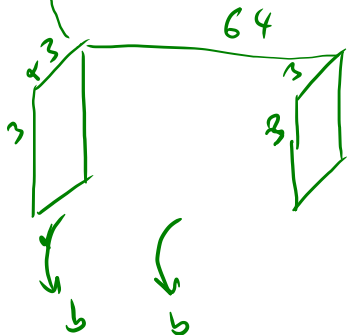


$$32 \times 3 \times 3 = 288$$

$$288 + 32 = \underline{320}$$

320

conv2D
2



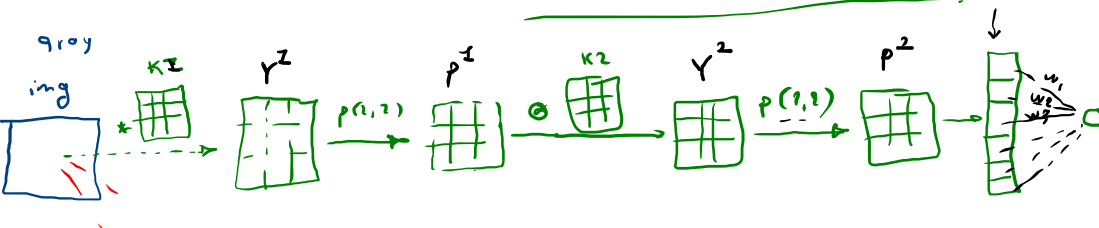
$$\rightarrow 64 \times 3 \times 3 = 567$$

$$\cdot 567 \times 32 = 18432$$

$$18432 + 64 =$$

18496

CNN Backpropagation



$\text{Conv2D}(\underline{1}, (3, 3), p: \text{same})$

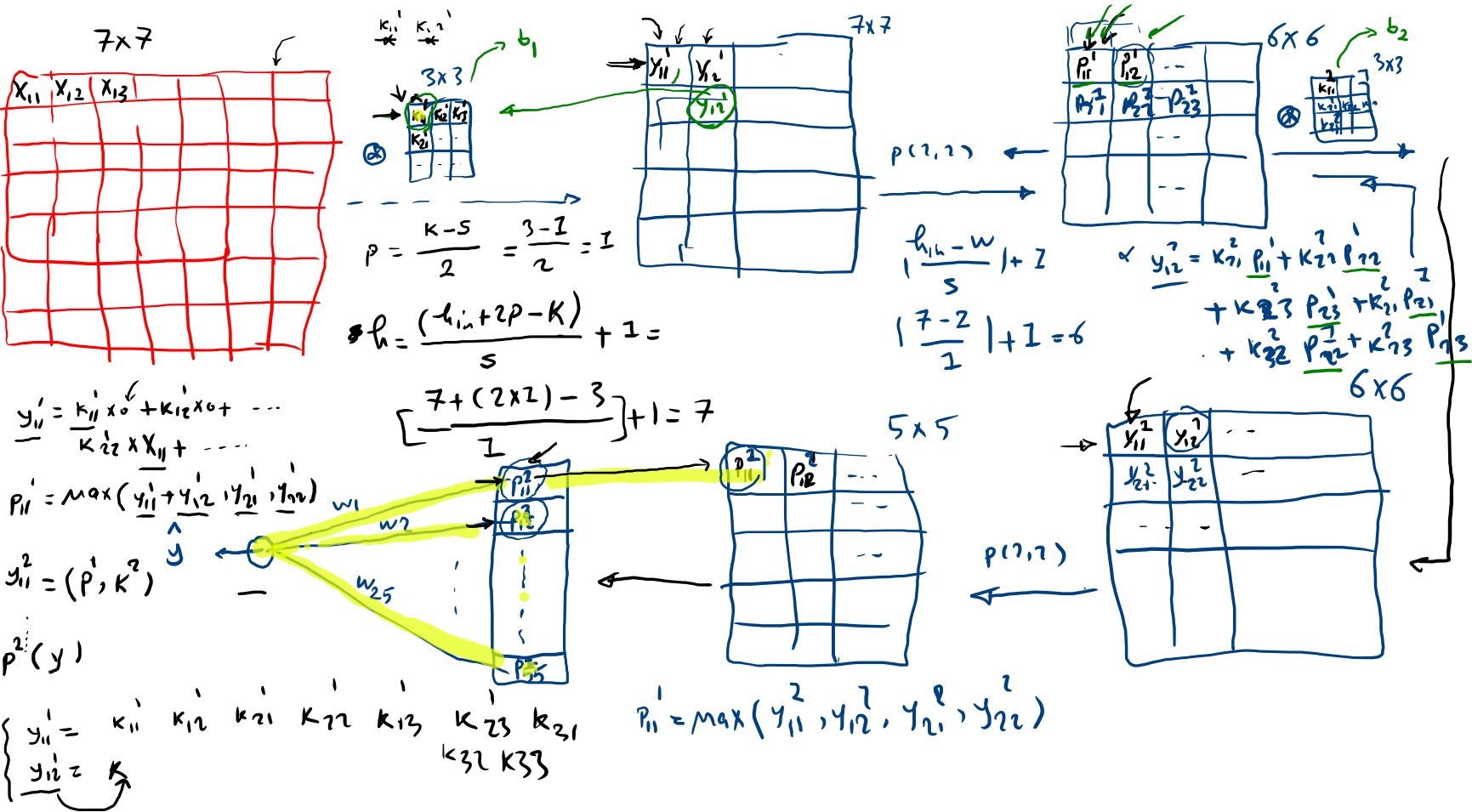
$\text{maxpool}((2, 2))$

→ $\text{Conv2D}(1, (3, 3), p: \text{same})$

$\text{maxpool}((2, 2))$

→ $\text{Flatten}()$

$\text{dense}(1, \text{sigmoid})$



$$\underline{K_{ii}^{\text{new}} = K_{ii}^{\text{old}} - \alpha \left(\frac{\partial L}{\partial K_{ii}^1} \right)}$$

$$0: w_1 p_{ii}^2 + w_2 p_{ii}^2 + \dots$$

\downarrow \downarrow
 K_{ii}^1 K_{ii}^1

$$\frac{\partial L}{\partial K_{ii}^1} = \frac{\partial L}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial o} \left[\frac{\partial o}{\partial p_{ii}^2} + \frac{\partial o}{\partial p_{ii}^2} + \dots \right]$$

$$p_{ii}^1 = \max(y_{ii}^1, y_{ii}^2, y_{ii}^3, y_{ii}^4)$$

$$p_{ii}^2 = \max(y_{ii}^2, y_{ii}^3, y_{ii}^4, y_{ii}^5)$$

\downarrow
 $\max: y_{ii}^2$

$$\frac{\partial o}{\partial p_{ii}^2} = \frac{\partial p_{ii}^2}{\partial y_{ii}^2}$$

$$\left[\frac{\partial y_{ii}^2}{\partial p_{ii}^1} + \frac{\partial y_{ii}^2}{\partial p_{ii}^2} + \frac{\partial y_{ii}^2}{\partial p_{ii}^3} + \dots \right]$$

$$\frac{\partial y_{ii}^2}{\partial p_{ii}^1} \left(\frac{\partial p_{ii}^1}{\partial y_{ii}^2} \right)$$

$$\frac{\partial y_{ii}^2}{\partial K_{ii}^1}$$

$$y_{ii}^2 = \underline{K_{ii}^1 p_{ii}^1}$$

$$\left(\underline{K_{ii}^1} \right) \times \dots$$

End