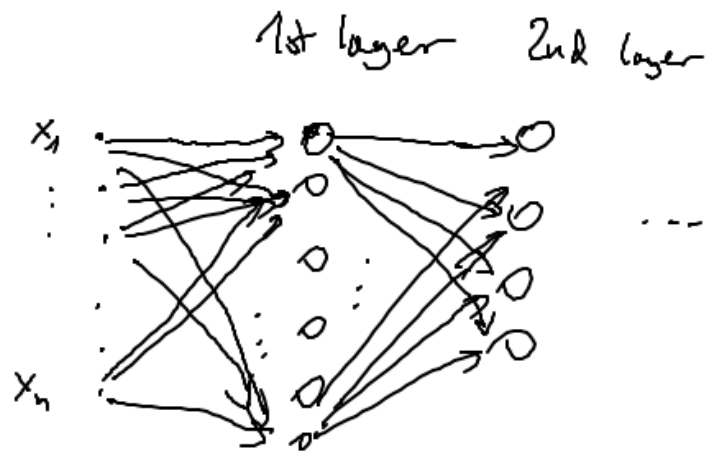


Neural nets



So far we have discussed neural nets with dense layers,
i.e., layers in which all neurons are connected to all inputs

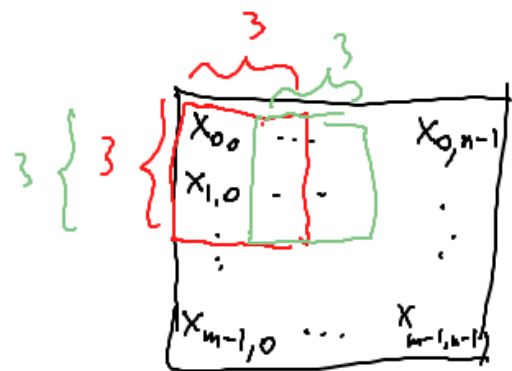
Convolutional neural networks

1980 Yann LeCun

Idea: exploit the 2-dim. nature of the input data

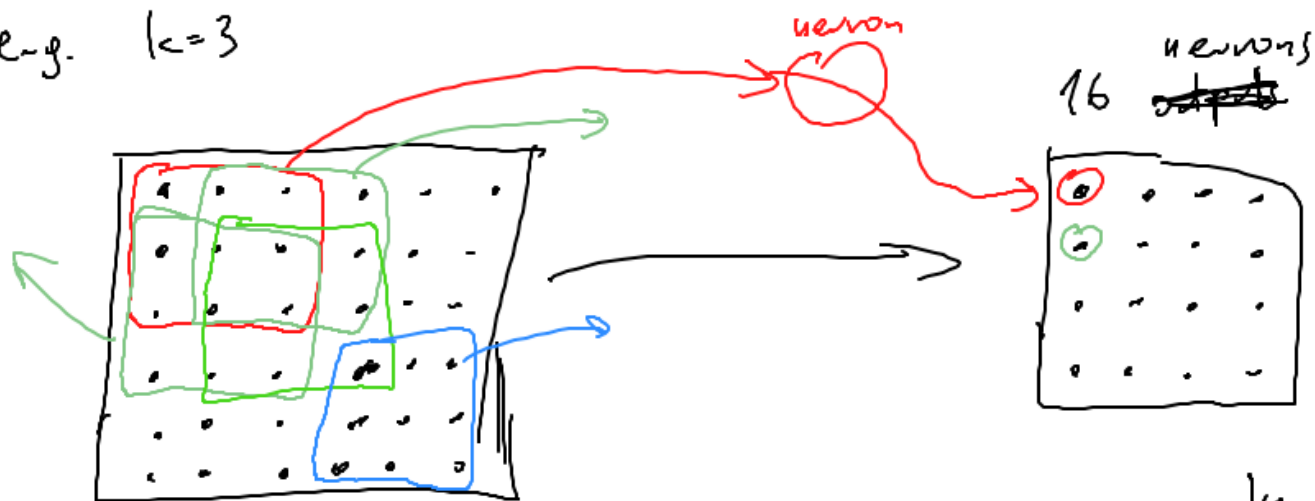
Usage: image recognition, pattern recognition





Dense layer would treat this image data as a sequence $(x_{ij})_{i=0, j=0}^{m-1, n-1}$

Convolution layer with filter size $k \times k$.
e.g. $k=3$



The weights are the same for each of the neurons

In this example we would have just 10 weights (9 for the input + 1 for the bias)



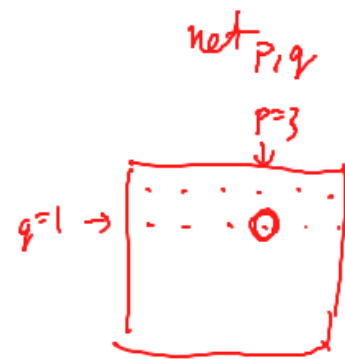
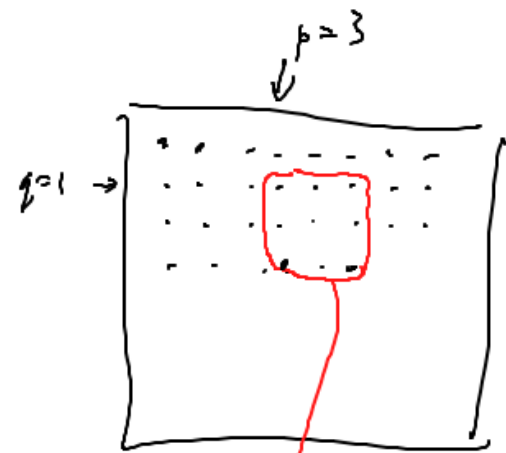
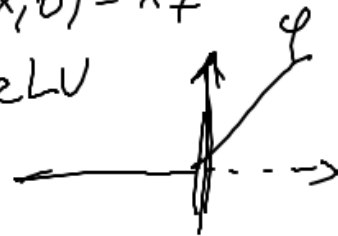
Weights: $(a_{ij})_{i,j=0}^{k-1}$, b - bias

For $p=0, \dots, m-k$; $q=0, \dots, n-k$ we put

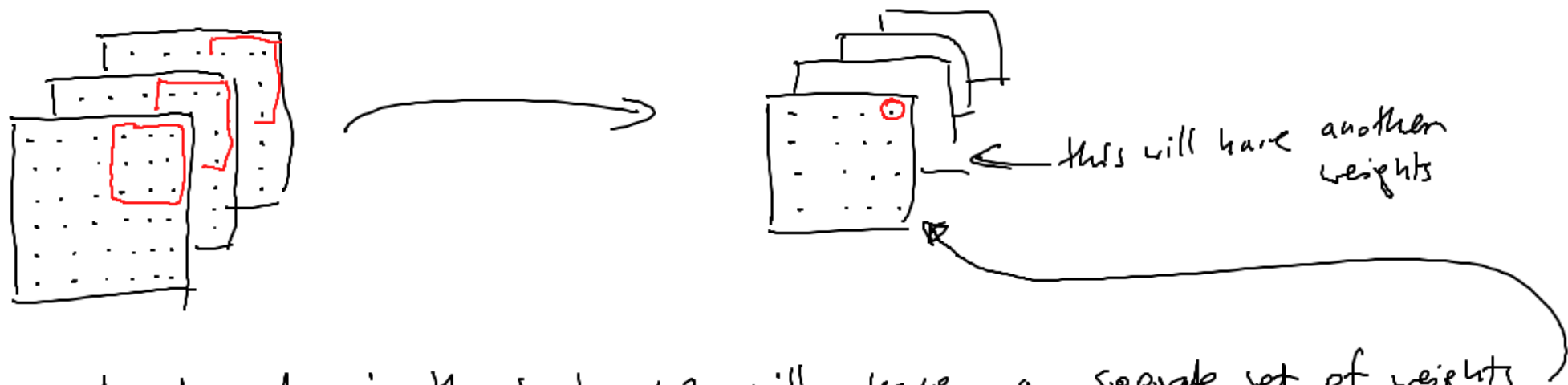
$$\text{net}_{p,q} = \sum_{i=0}^{k-1} \sum_{j=0}^{k-1} a_{ij} X_{i+p, j+q} + b$$

$$\text{output}_{p,q} = \varphi(\text{net}_{p,q})$$

In CNN one usually takes $\varphi(x) = \max(x, 0) = x_+$
ReLU



We have discussed a setup with just 1 input channel and 1 output channel
 If the input consists of several matrices (channels):



For each channel in the input we will have a separate set of weights

$$\left(a_{ij}^{(f)} \right)_{i,j=0}^{k-1}, b$$

$$f=1, \dots, F$$

↑
number of input channels

$$\text{net}_{p,q} = \sum_{f=1}^F \sum_{i=0}^{k-1} \sum_{j=0}^{k-1} a_{ij}^{(f)} \cdot X_{i+p, j+q}^{(f)} + b$$

$$\text{output}_{p,q} = \varphi(\text{net}_{p,q})$$

then



Example: input: 3 planes 100×100
(channels)

kernel size: 5×5

output: 64 planes (channels) 96×96 - size \rightarrow

in the ^{layer} ~~input~~ we have
 $64 \cdot 96 \cdot 96$ neurons

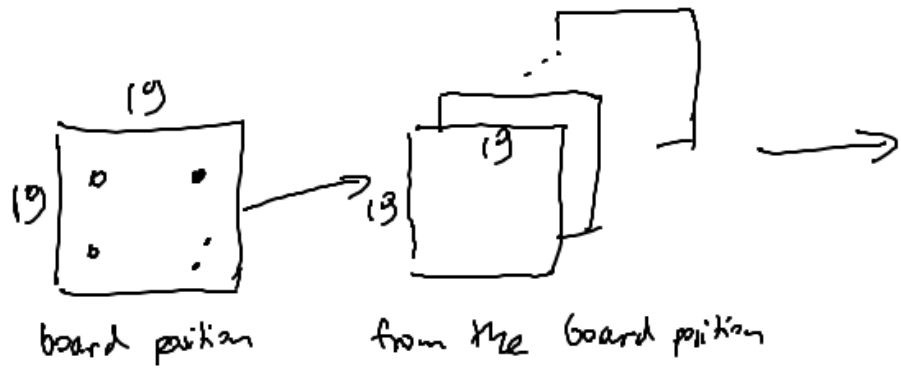
$$\text{number of weights} = 64 \cdot (5 \cdot 5 \cdot 3 + 1) = 64 \cdot 76 = 4864$$

{ dense layer with N neurons
would have

$$: (10000 + 1) \cdot N = 10001 \cdot N$$

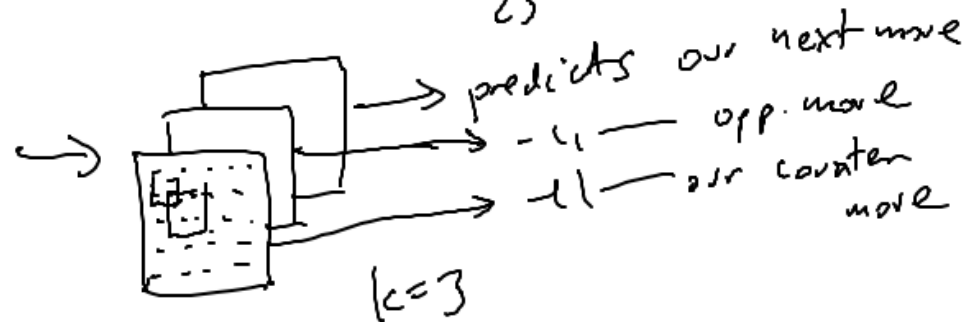
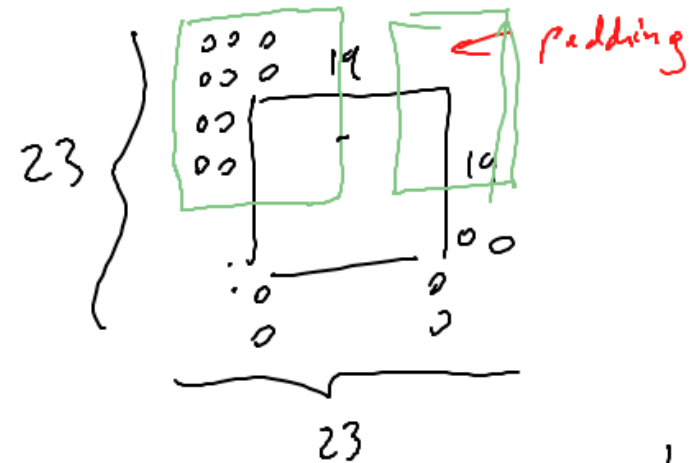
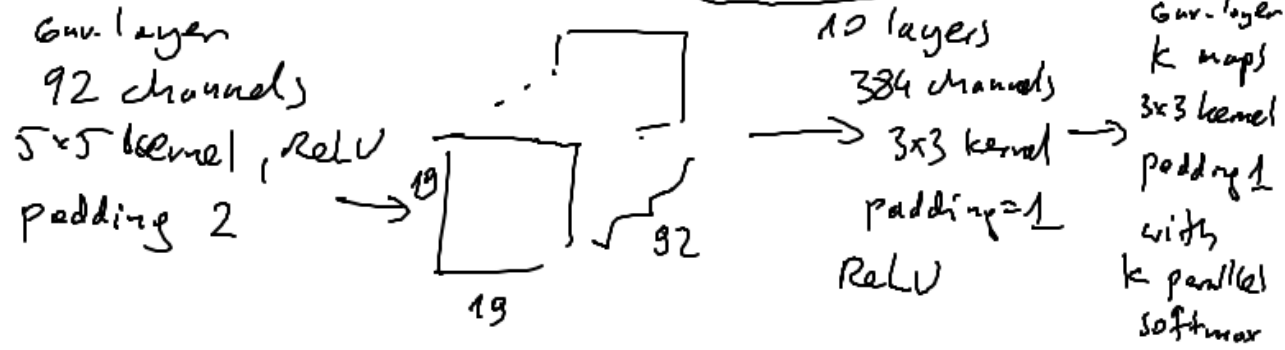
The training for
conv. layer is
much faster.

Example (almost pure CNN) - DarkForest, used for move prediction in the game Go



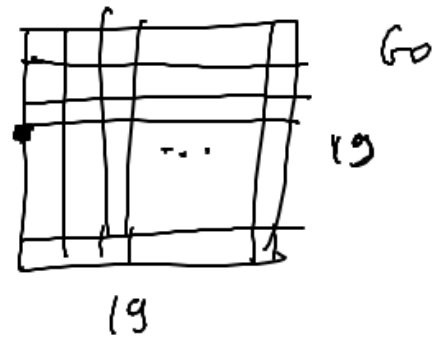
for example:

- 1) the position of white stones
- 2) -1 - black stones
- 3) -1 - empty places
- ...) the white groups with 1 liberty
- ...) -1 - 2 liberties
- ...) -1 - 33 -1 -



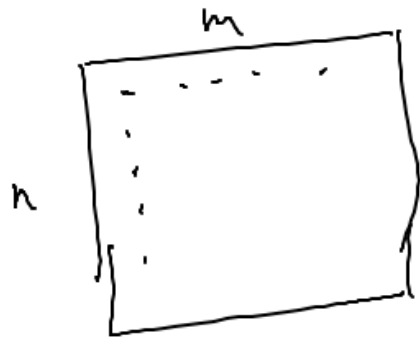
Dark Forest - CNN developed around 2015
Alpha Go 2015-16 ~~was~~ the first program
which won with the
best humans of Go on (9x9)

1996-97 Deep Blue
↑ ↑ (IBM)
Kasparov won

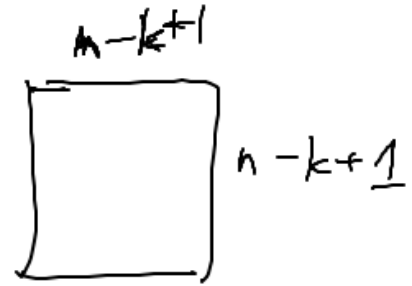


361 of possibilities
for the first
move

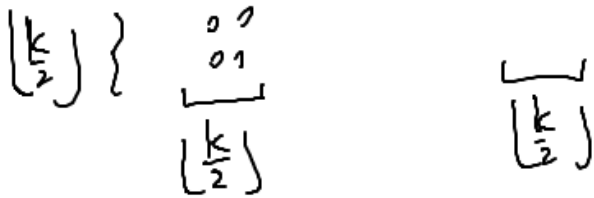
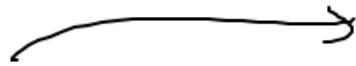
without padding



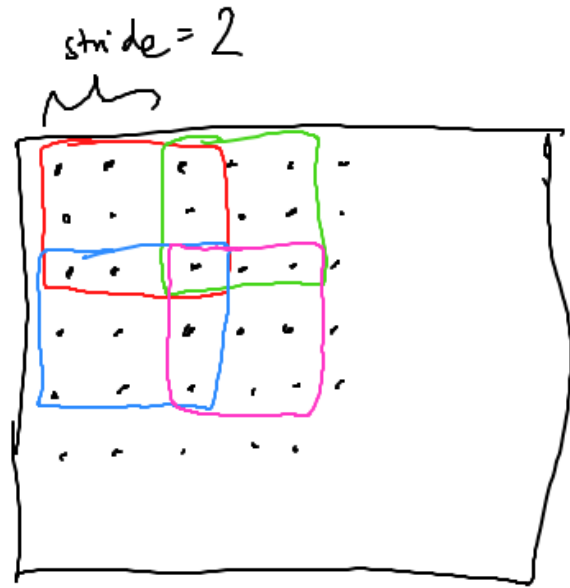
kernel size $k \times k$



with padding one can obtain the same size k -odd



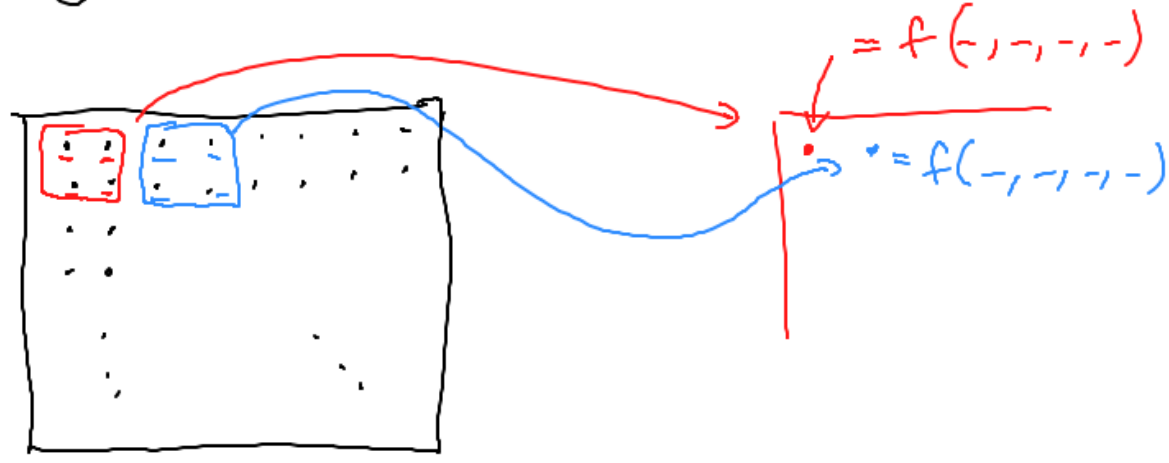
- strides: so far we considered only stride = 1



stride = 2 ~~the~~ more-less halves the width and the height of the image

reducing the dimensions by a factor ≈ 4

• Pooling layers - reduce the dimensions in a different way



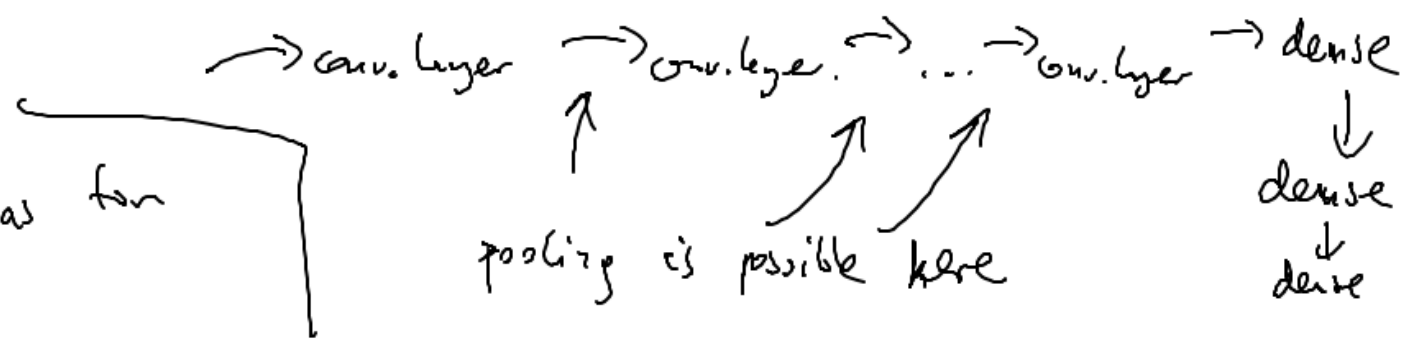
e.g. $f = \max$
 $f = \text{average}$

usually max pooling is used

• often a dense layer is used as the last one

CNN

• Training: back-propagation as for the classical NN

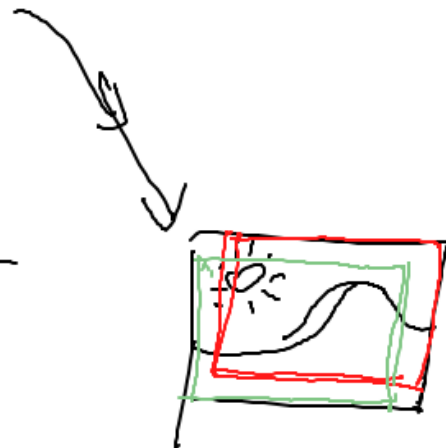


- Data augmentation
for example, use reflections

- (slight) rotations

- (slight) changes in the pixel intensities

- expand slightly and then crop some part

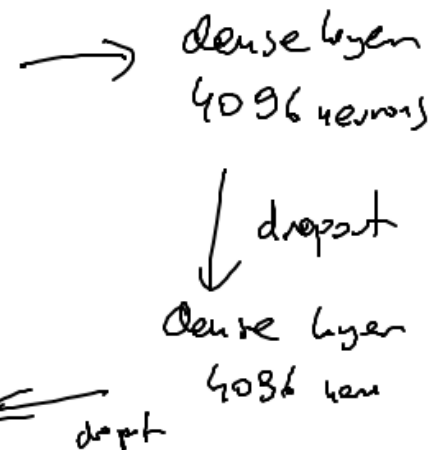
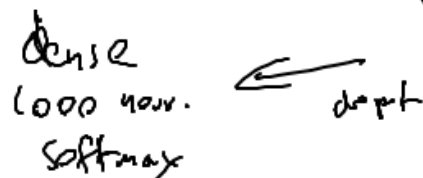
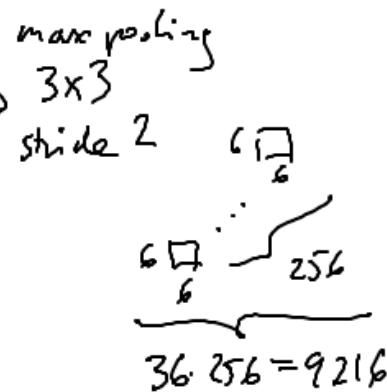
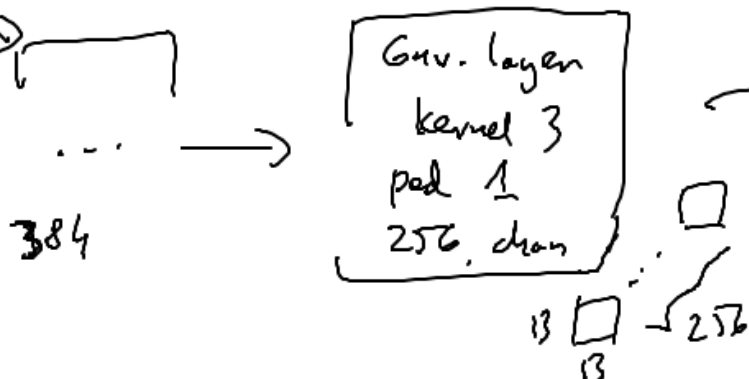
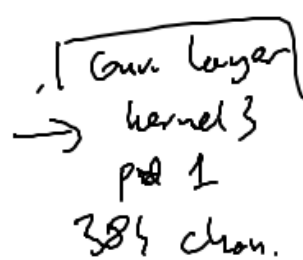
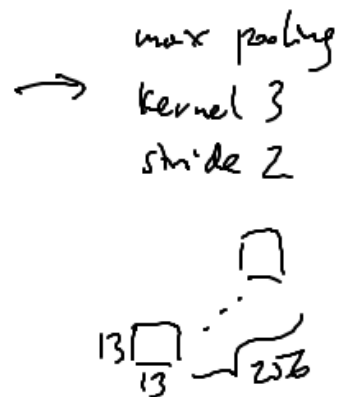
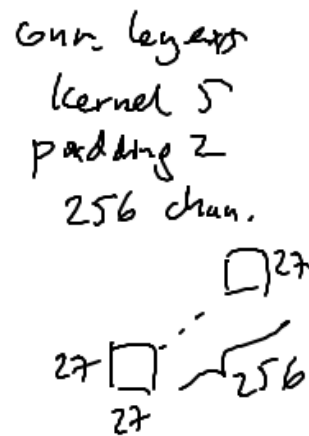
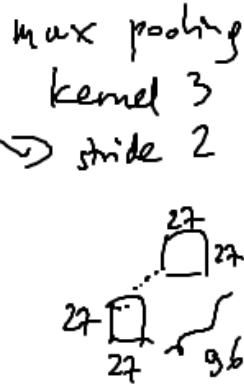
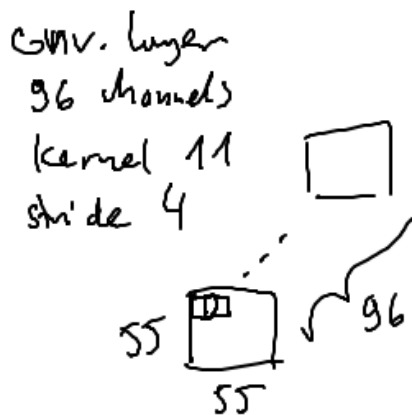


Remark: convolutional layers can be applied to inputs of different sizes

Example
AlexNet

2012

ReLU in all conv. / dense layers, except the last one, in which softmax was used



depth