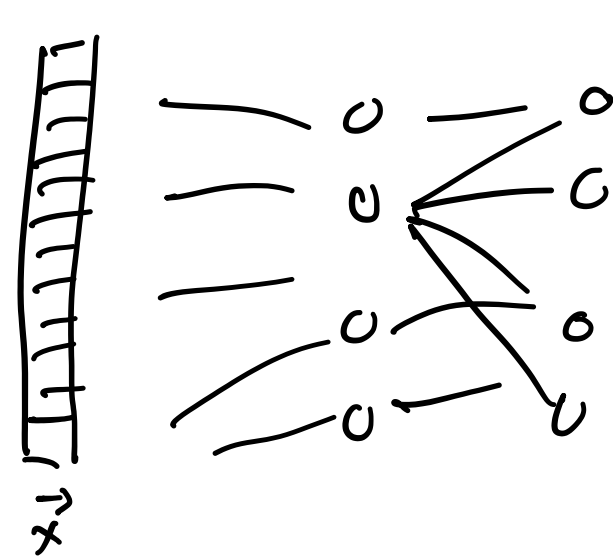
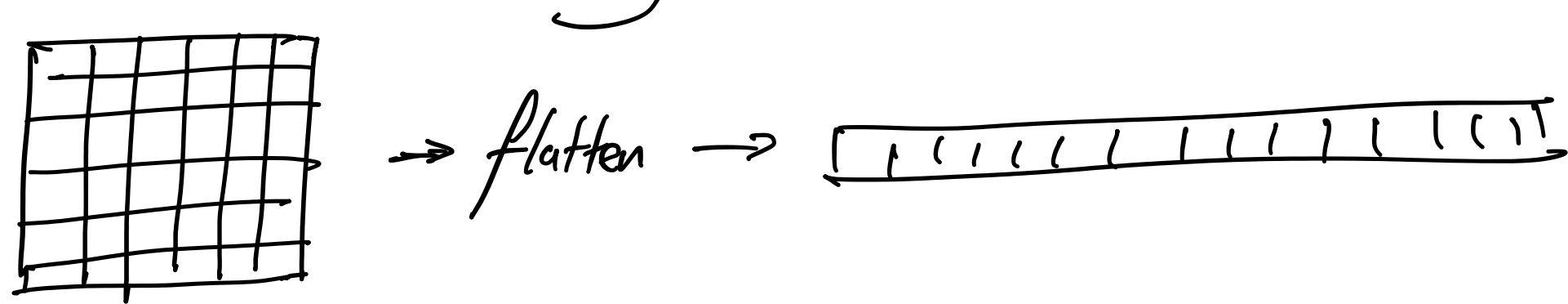
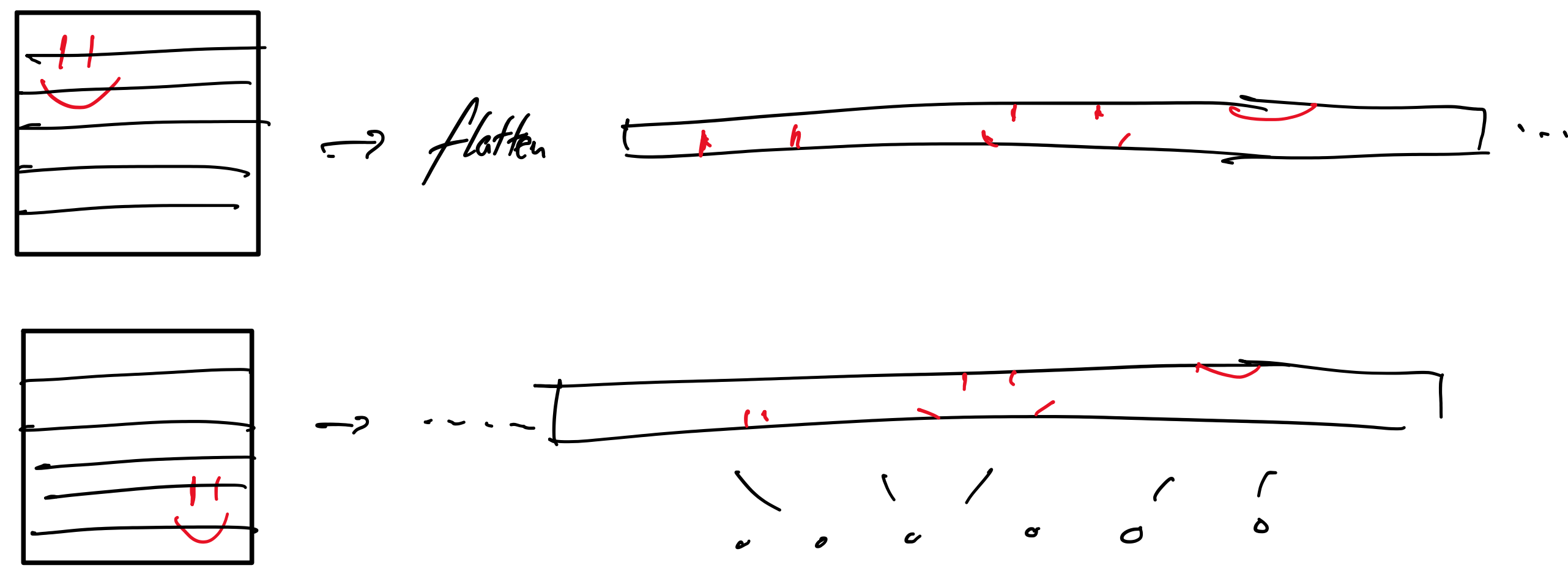


How can we handle images?



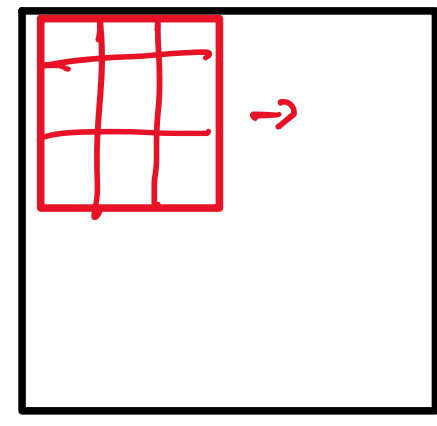
for a 10-class classification problem,
what is the size of the output?
10 - one for each class
if using softmax

Problems with images as vectors



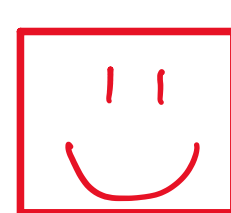
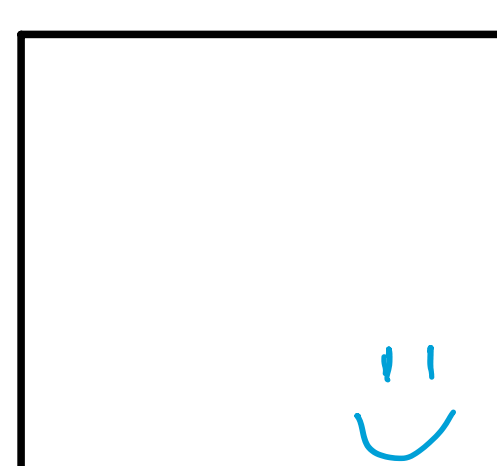
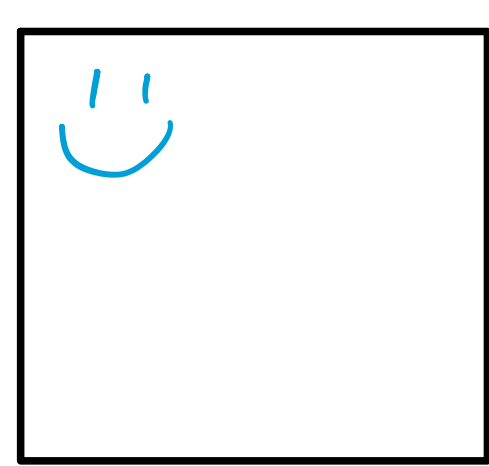
- different neurons have to deal with the same features, depending on where they are in the image
- redundant effort - very inefficient

Convolution

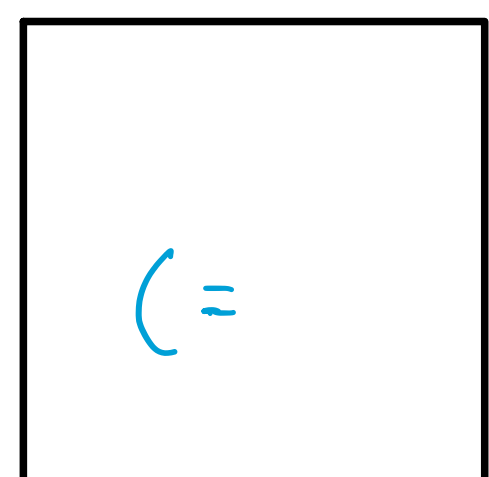


3x3 "filter"

- the filter slides across the whole image, "looking" at all of it
- recall: $w \cdot x$ in a neuron is a measure of similarity - how much does this filter "recognize" what it sees?



- "smile detector" activates regardless of the location of the smile in the image
- translational invariance



Q: is the smile detector going to activate?
translational invariance \neq rotational invariance

Grayscale Example

1	3	5	6
2	0	1	4
0	9	10	1
3	2	0	1

\vec{x}

0	1
2	1

\vec{w}

$$\begin{bmatrix} 1 & 3 \\ 2 & 0 \end{bmatrix} \cdot \begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix} = \begin{matrix} 1 \times 0 + 3 \times 1 + 2 \times 2 + 0 \times 1 \\ 0 + 3 + 4 + 0 \end{matrix} = \begin{matrix} 7 \\ 7 \end{matrix}$$

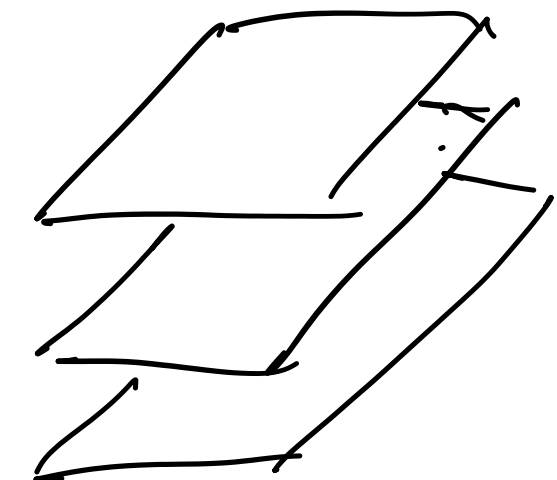
$\vec{x} \otimes \vec{w}$

stride - 2

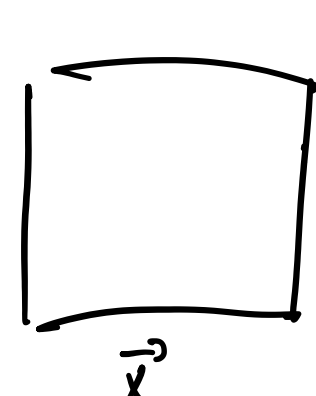
2x2 output

stride - 1

3x3



1x1



$$\vec{w}_0 \rightarrow \gamma(\vec{x} \otimes \vec{w}_0 + b_0)$$

stride - 1 (5, 7)

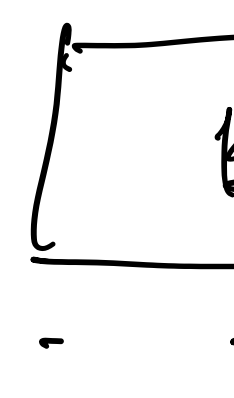
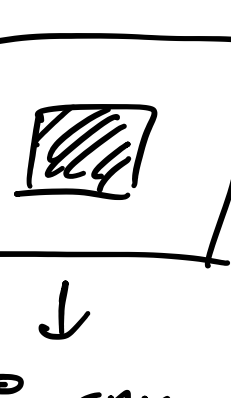
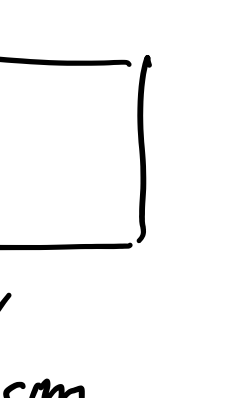
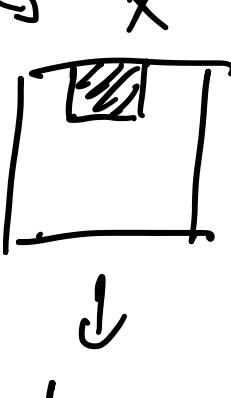
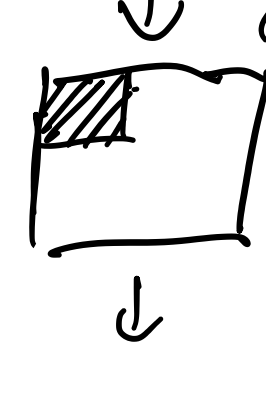
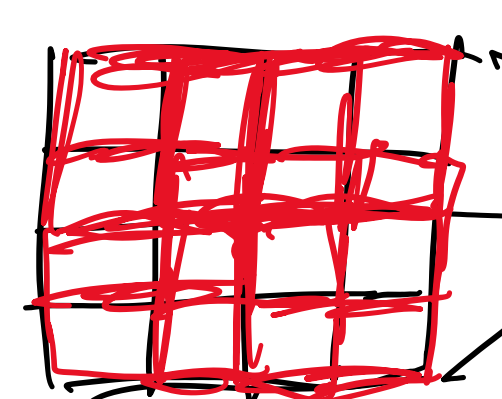
\vdots
 \vec{w}_{F-1}
 \vdots
 \vec{w}_F

$$\rightarrow \gamma(\vec{x} \otimes \vec{w}_i + b_i)$$

output: (F, 5, 7)

of filters

7
F
5
(sheet, rows, cols)

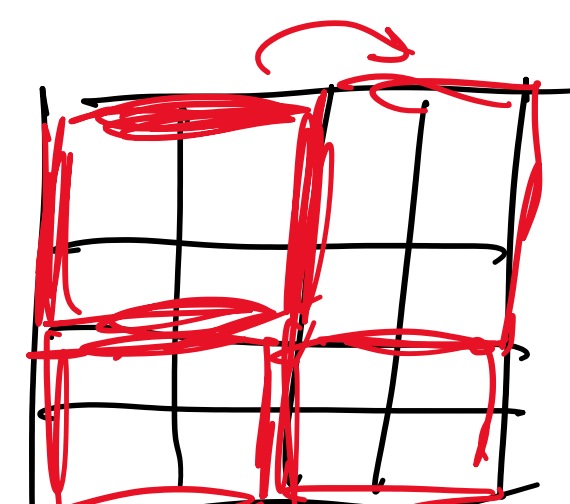


a	b	c
d	e	f
g	h	i

stride - 1

a	c
g	i

stride - 2



2x2

Max Pool

- operation that often follows one or more conv layers

- purely a discretization / downsampling process - no learnable parameters
- $f(\vec{x}) = \max(\vec{x})$

1	3	5	6
2	0	1	4
0	9	10	1
3	2	0	1

2x2 max pool
w/ stride 2

3	6
9	10

say this was the result of $3 \otimes \vec{x}$

high values come from high similarity

therefore \therefore max pool helps later parts of the network "focus on" high-similarity detections and reject low-similarity and redundant detections

9	8	1	1	9	0
9	0	2	1	0	1