Neural Networks: II. Convolutional NN (Part 7)

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Motivation

• previous:

$$\underbrace{\mathsf{input}}_{1\times728} \longrightarrow \underbrace{x\cdot W}_{10\times728} \longrightarrow \underbrace{\mathsf{score}}_{1\times10}$$

• now: Keep structure and don't flatten it

$$\leadsto$$
 input is $28\times28\times1$

Convolutional Layer

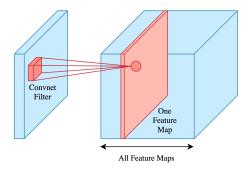


Figure: sliding and projecting (brilliant.org)

- use filter(s) and slide it over the image
- project the slices onto activation map(s) (onto one number):

$$x'w + b$$

• image-depth = filter-depth (foreshadowing)



Convolutional Layer: More elaborate

Example: Input $28 \times 28 \times 1$ and a filter of size $5 \times 5 \times 1$

- use filter(s) and slide it over the image
 - you can use different strides for sliding
- project the slices onto activation map(s) (onto one number):

$$x'w + b \rightsquigarrow 5 \cdot 5 \cdot 1$$
 dot product

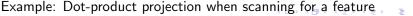
- image-depth = filter-depth
 - example: Coloured images given in RGB i.e. depth = 3



Why a filter?

Intuition:

- the filter has a broader "view" and can scan for particular shapes/features
- the deeper we stand in our network, the more particular are the shapes we are scanning for
- we use several filters, since every filter scans for different shapes
- example: Recognize handwritten digits
 - scanning for a cross
 - \Rightarrow 4, 7 or 8
 - scanning for a right angle
 - \Rightarrow 4 or 5
 - scanning for a kink
 - \Rightarrow 1, 2, 4, 5 or 7
 - scanning for a curve
- \Rightarrow 2, 3, 5, 6, 8, 9 or 0





Output Size

• Output Size:

$$N - F + 1$$

• Output Size (with stride > 1):

$$\frac{N-F}{\text{stride}}+1$$

Example: Image 5×5 and filter 3×3

• Running into an issue in here?



Pad the Border

- add a pad containing zeros
- stops dimension shrinking
- pad-size:

 $\mathit{F}=1\Rightarrow \mathsf{no} \mathsf{zero} \mathsf{pads}$

 $F=3\Rightarrow$ zero pads of 1

 $F=5\Rightarrow$ zero pads of 2

Example: Image 5×5 and filter 3×3 with stride 1

Convolutional NN Architecture

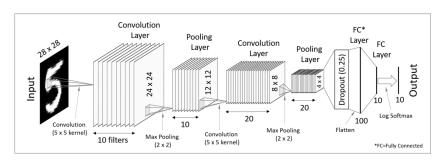


Figure: Architecture (codetolight.files.wordpress.com)

Pooling Layer

- used to reduce the dimensions
- intuition: Focus on the features that have been detected for certain
- apply pooling to every activation map

Example: Max Pooling

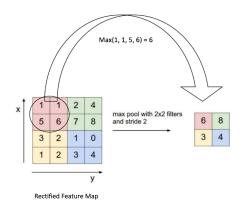


Figure: max pooling (blog.algorithmia.com)