CNN Layers

Convolutional Layers

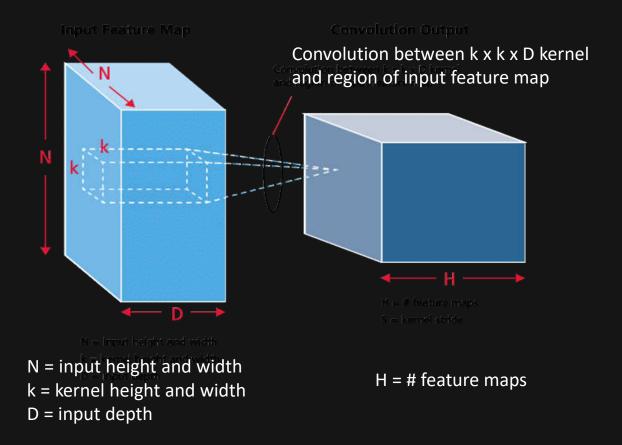


Image source (<u>link</u>)

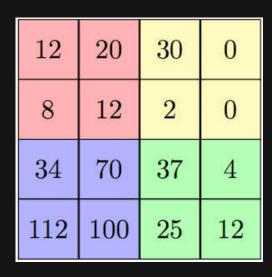
Convolutional Layer

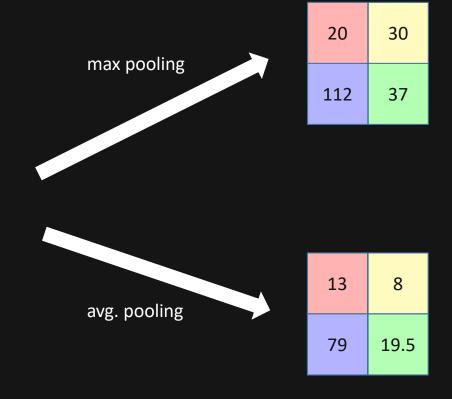
- Performs filtering on the image or result from the previous layers
- Saves weights by using the same weights for all filters of a feature map
- Multiple filters can be applied to the result of the previous layer resulting in multiple feature maps

• Params:

- nxn: the filter size (typically 3x3, 5x5, 7x7). Can be 1x1!
- depth: the number of feature maps
- stride: shift to apply next filter (1 applies to every pixel, 2 skips a row/column, ...)
- padding: can be used to create a border of zeros around the image(s) from the previous layer (preserves dimensionality)

Pooling



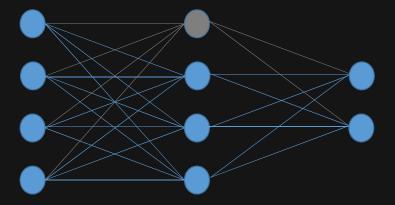


Pooling

- Enables dimensionality reduction
- Typical kernel = 2x2 => reduction by half in each dimension
- Typical stride for kernel 2x2 = 2 (no overlap)
- Most common is max pooling
- Considering an input of 28 x 28 x 256 the output will be 14 x 14 x 256
- Pooling layer is commonly placed after a convolutional layer
- There is no consensus about pooling. Some prefer to use Convolutional layer with stride: STRIVING FOR SIMPLICITY: THE ALL CONVOLUTIONAL NET (link)

Dropout

• Randomly selects neurons to discard during feedforward and backward passes.



Droupout

 Srivastava, et al. (2014) <u>Dropout: A Simple Way to Prevent Neural</u> <u>Networks from Overfitting</u>

• Should provide better generalization, as it is forcing the network to have multiple paths to achieve the same result.

Regularization

- Penalization for large weights
- CNNs with smaller weights tend to generalize better.

In Keras