

The background of the slide is a dense, 3D-rendered field of numbers. The numbers are in various shades of light blue and white, creating a sense of depth and perspective. They are scattered across the entire frame, with some numbers appearing larger and more prominent than others. The overall effect is a complex, abstract pattern of digits.

Perceptrons and CNNs

Sugata Banerji

CSCI 450

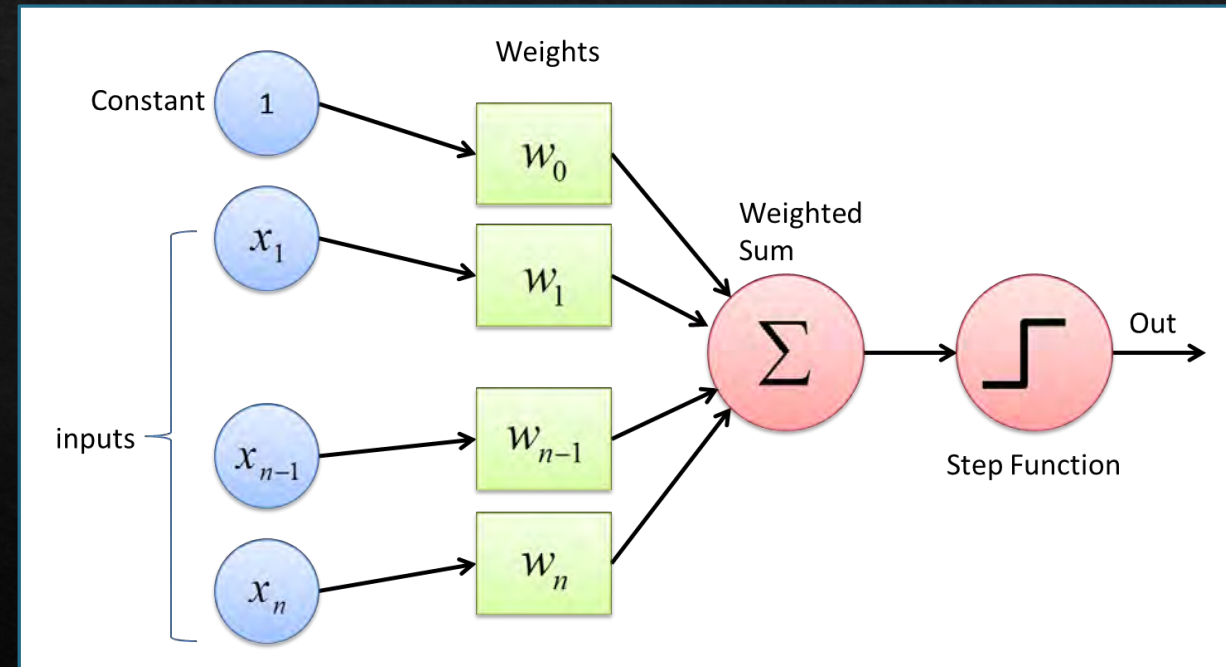
Lake Forest College

The Perceptron - History

- ◆ An algorithm for supervised learning of binary classifiers
- ◆ Invented in 1958 at the Cornell Aeronautical Laboratory by Frank Rosenblatt.
- ◆ Initially seemed promising, but it was quickly proved that they are only capable of learning linearly separable patterns.
 - ◆ This caused the field of neural network research to stagnate for many years
 - ◆ Later it was found that a multilayer network of perceptrons could solve many problems
- ◆ Neural network research experienced a resurgence in the 1980s.

The Perceptron - Structure

- ◆ The perceptron is a simplified model of a biological neuron.
- ◆ It takes an n -dimensional input
- ◆ Internally it has n weights and a *bias*
- ◆ It outputs 0 or 1
- ◆ The weights and bias start as random and get proper values during training
- ◆ Training is done by the perceptron algorithm (feedforward)



The Perceptron Algorithm (Feedforward)

$$f(x) = \begin{cases} 1 & \text{if } w.x + b > 0 \\ 0 & \text{otherwise} \end{cases} \quad \text{Where } w.x = \text{dot product of } w \text{ and } x, \text{ and } b \text{ is the } \textit{bias}$$

$$\diamond w_i(t + 1) = w_i(t) + r \cdot [d_j - y_j(t)] x_{j,i}$$

◇ j signifies the j-th training sample

◇ i signifies the i-th dimension

◇ r is the learning rate

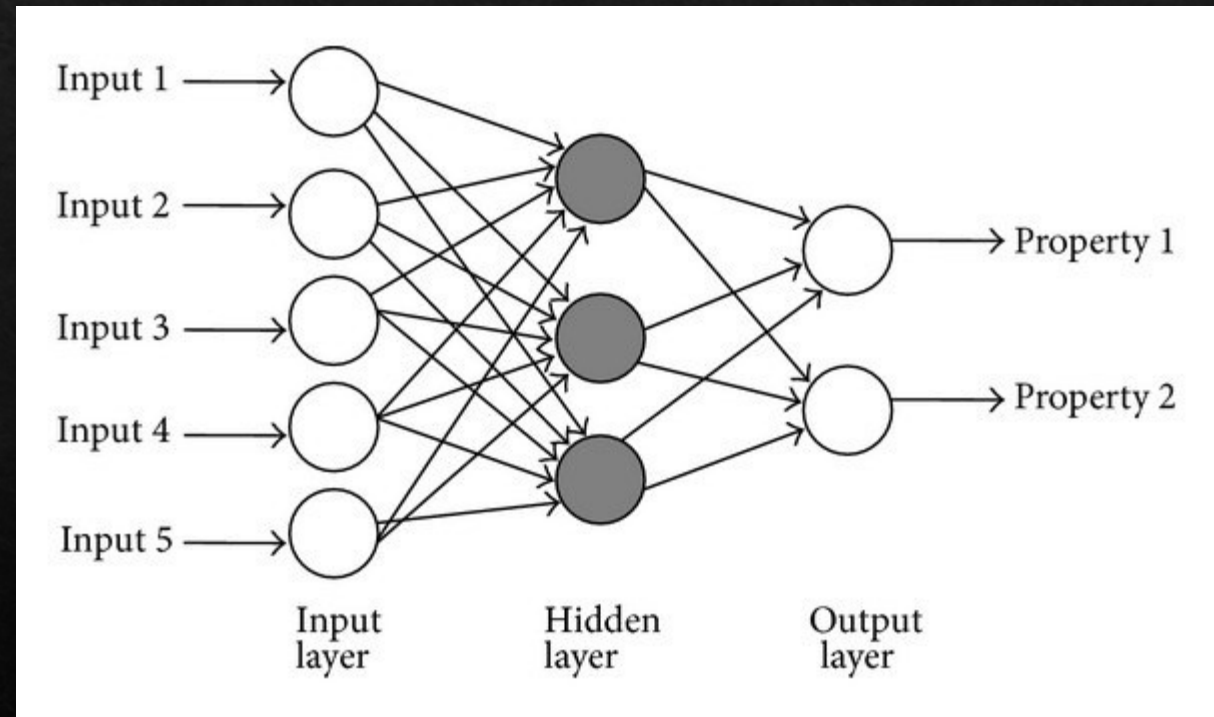
◇ d is the desired output

◇ y is the actual output

◇ t is the iteration

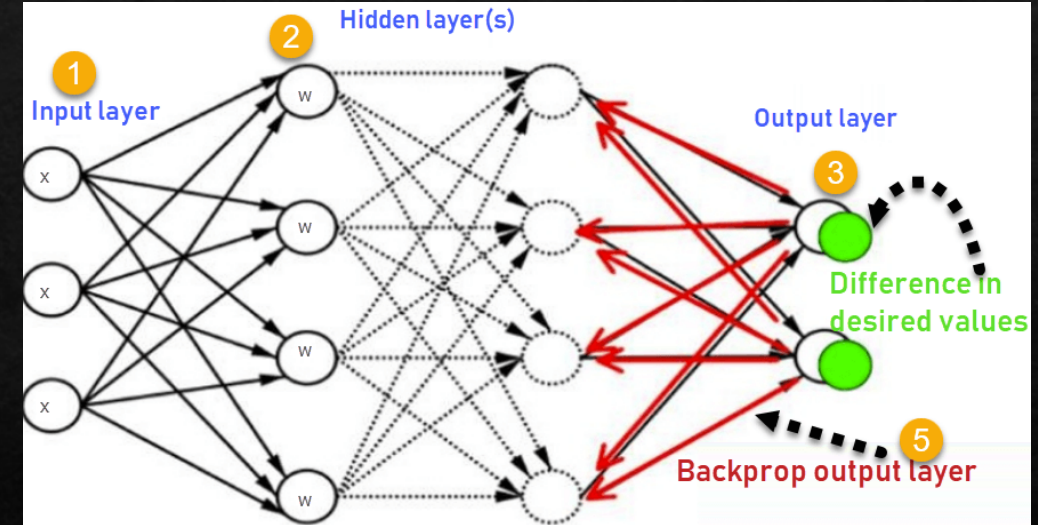
Multilayer Perceptrons

- ◆ Has more than one layer
- ◆ Outputs of one layer become inputs of the next
- ◆ Layers between input and output layers are called hidden layers
- ◆ Can detect more complex patterns
- ◆ May need *backpropagation* and sigmoid functions



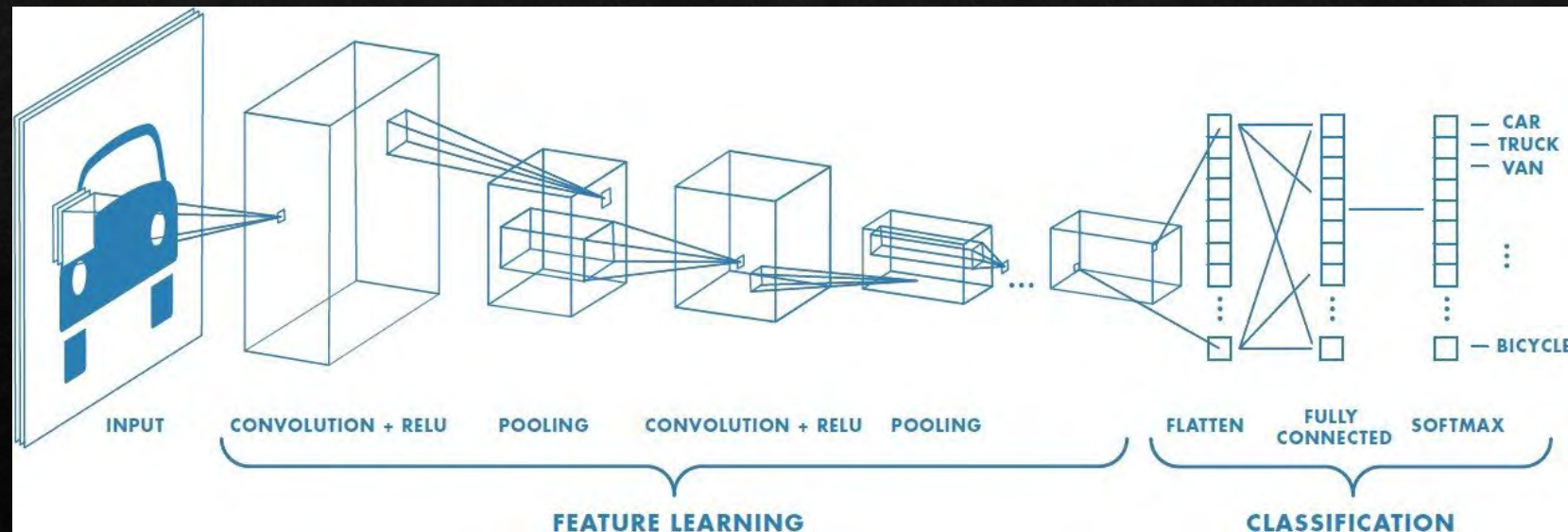
Backpropagation

- ◇ An algorithm to train multilayer neural networks (perceptrons)
- ◇ The error causes the weights to change backwards from the output to the input, hence the name
- ◇ The gradient of the error function with respect to the neural network's weights is calculated
- ◇ Much faster than forward-propagation
- ◇ Further Reading:
<https://brilliant.org/wiki/backpropagation/>



Convolutional Neural Networks (CNN)

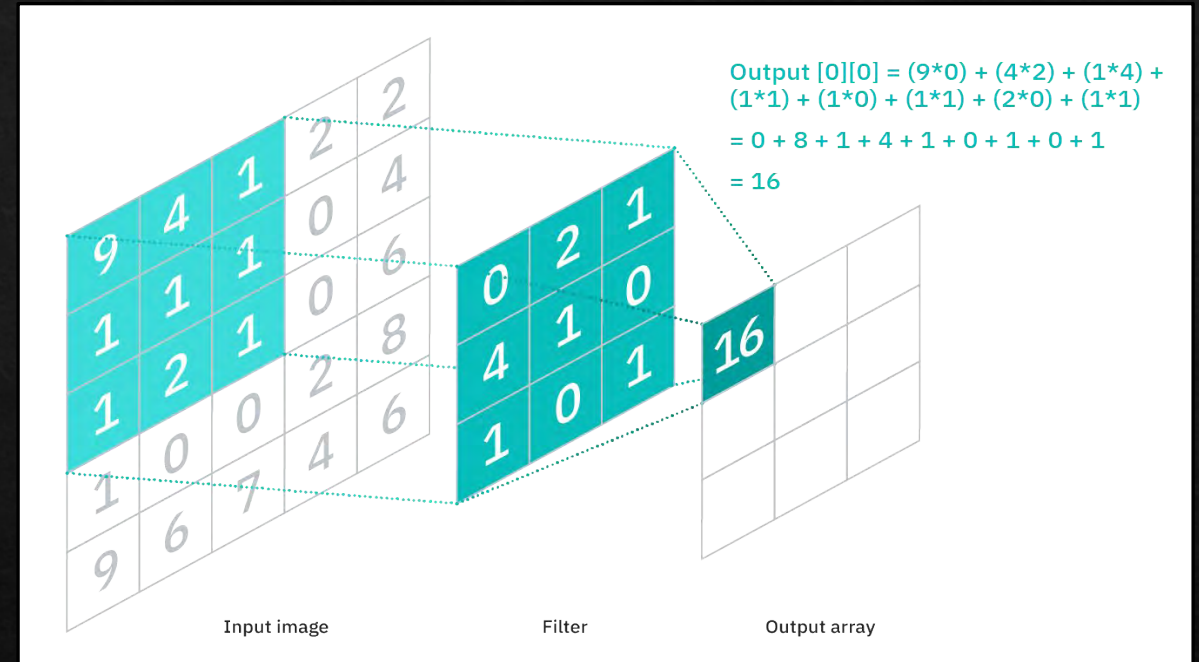
- ◆ A deep network (more than one hidden layer)
- ◆ Takes 2-D (or 3D) input
- ◆ Components
 - ◆ Convolution
 - ◆ Pooling
 - ◆ Dropout
 - ◆ ReLU
 - ◆ Fully Connected
 - ◆ Softmax
 - ◆ Sigmoid
 - ◆ Etc.



CNN Layers

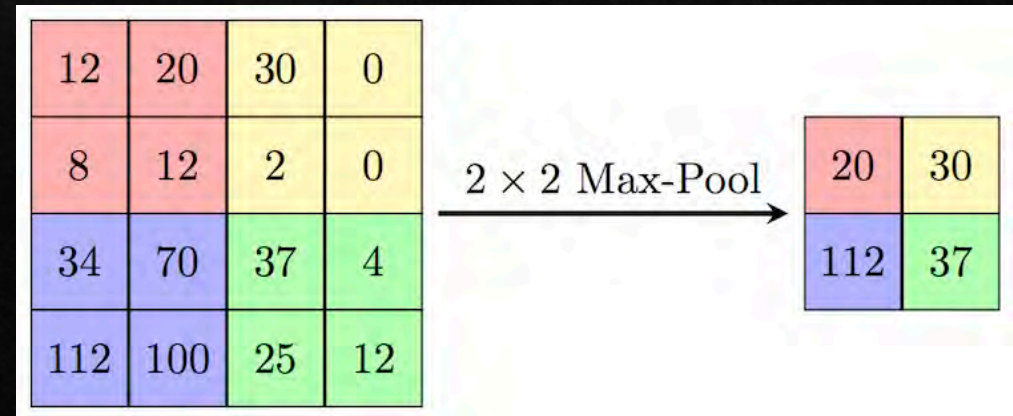
Convolution

- Filters to detect basic patterns in the image



Pooling

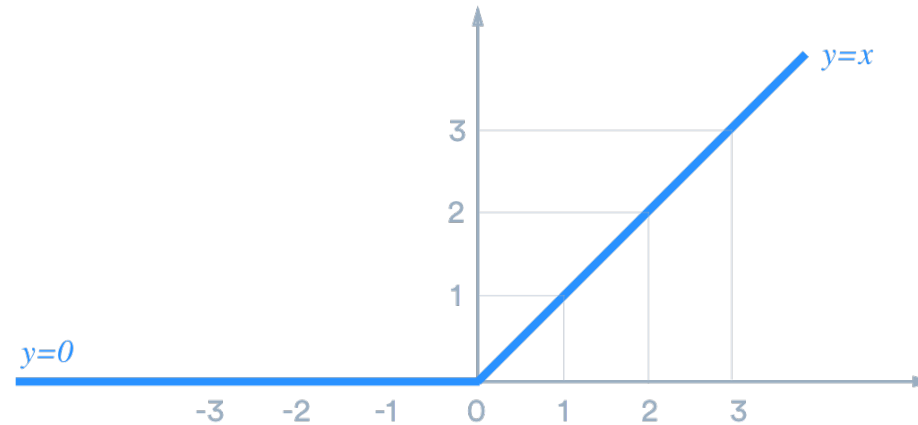
- Downscales the image while preserving approximate location of highest responses



CNN Layers

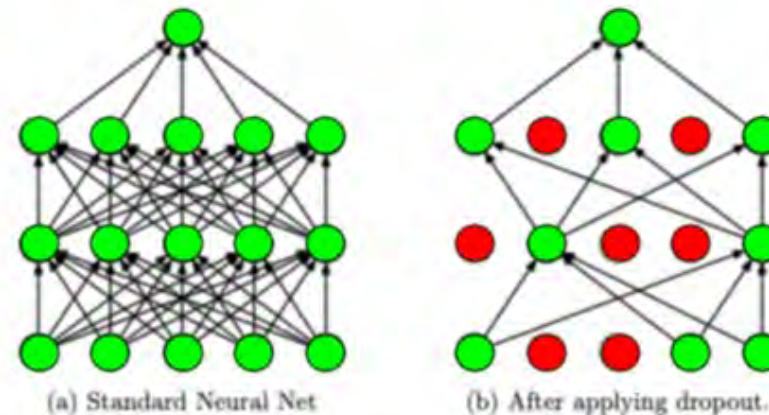
ReLU

- Rectified Linear Units or ReLU layers keep positive outputs intact while making negative outputs zero



Dropout

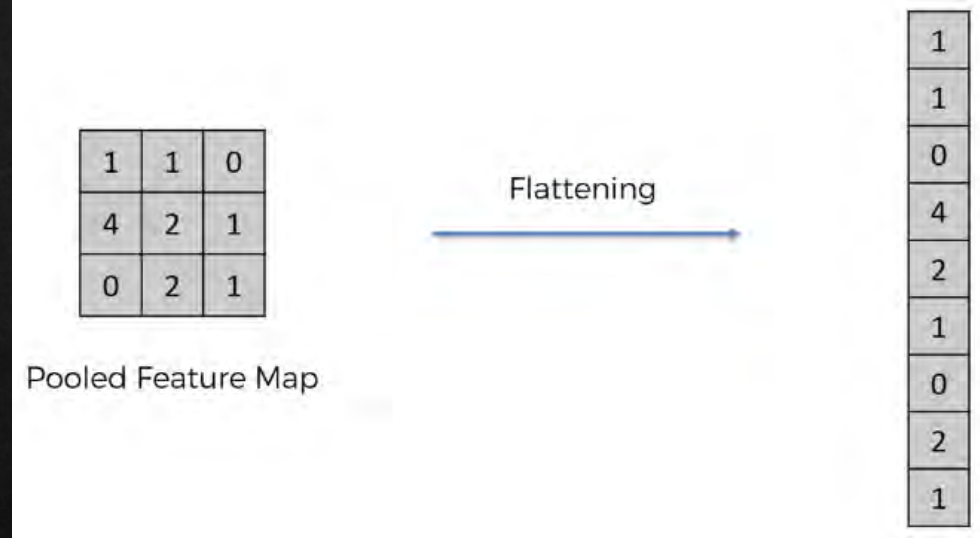
- Dropout layers “kill” some of the outputs from the previous layer to make the training more robust



CNN Layers

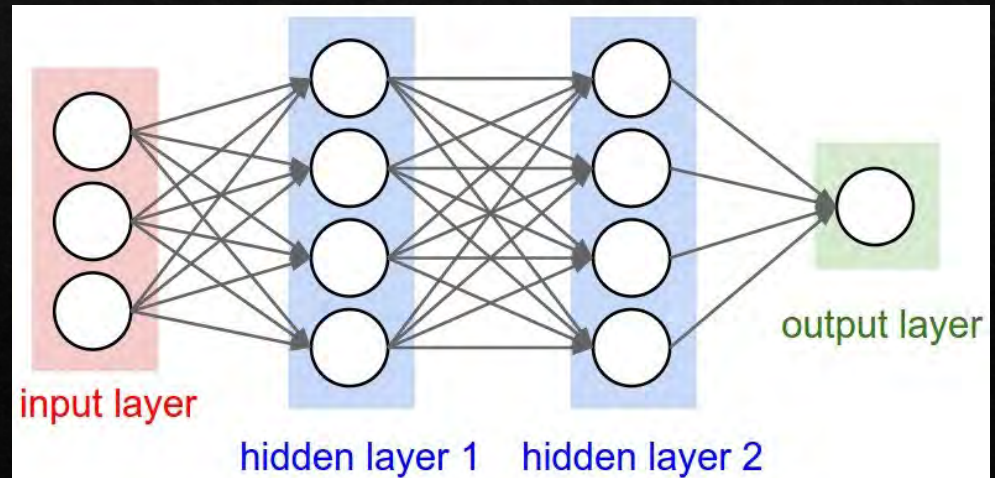
Flattening

- Changes the n-dimensional output from a layer into a 1-dimensional vector before the FC layers



Fully Connected

- A standard network of perceptrons all connected to each other that take 1-D inputs



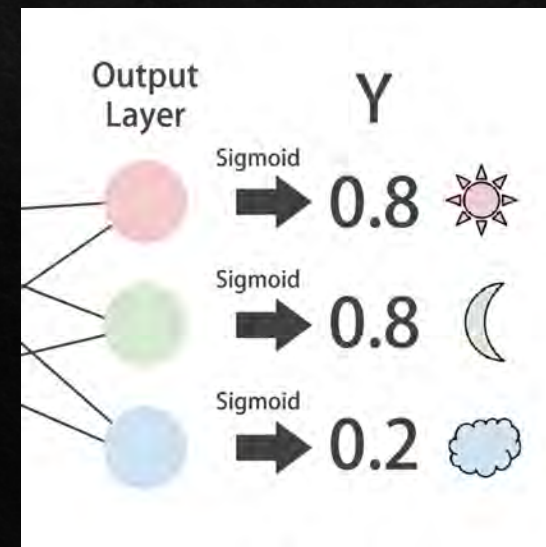
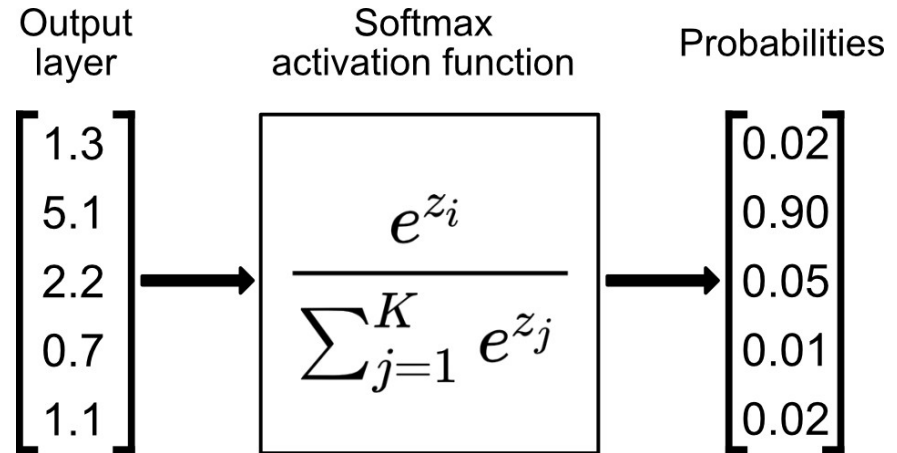
CNN Layers

Softmax

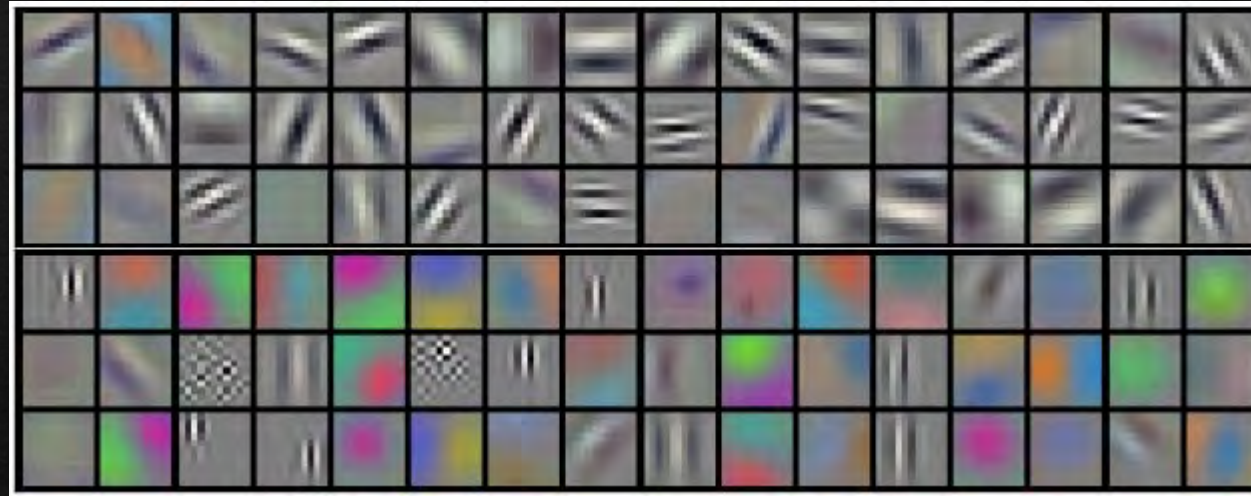
- Last layer in single-label classification networks to make the output scores sum to 1

Sigmoid

- Used in output layer of multi-label classification networks where the scores don't sum to 1



First Layer filters



First Layer filters learnt by AlexNet by Krizhevsky et al
(<https://papers.nips.cc/paper/2012/file/c399862d3b9d6b76c8436e924a68c45b-Paper.pdf>)