

Learning Deep Neural Networks (Deep Learning)

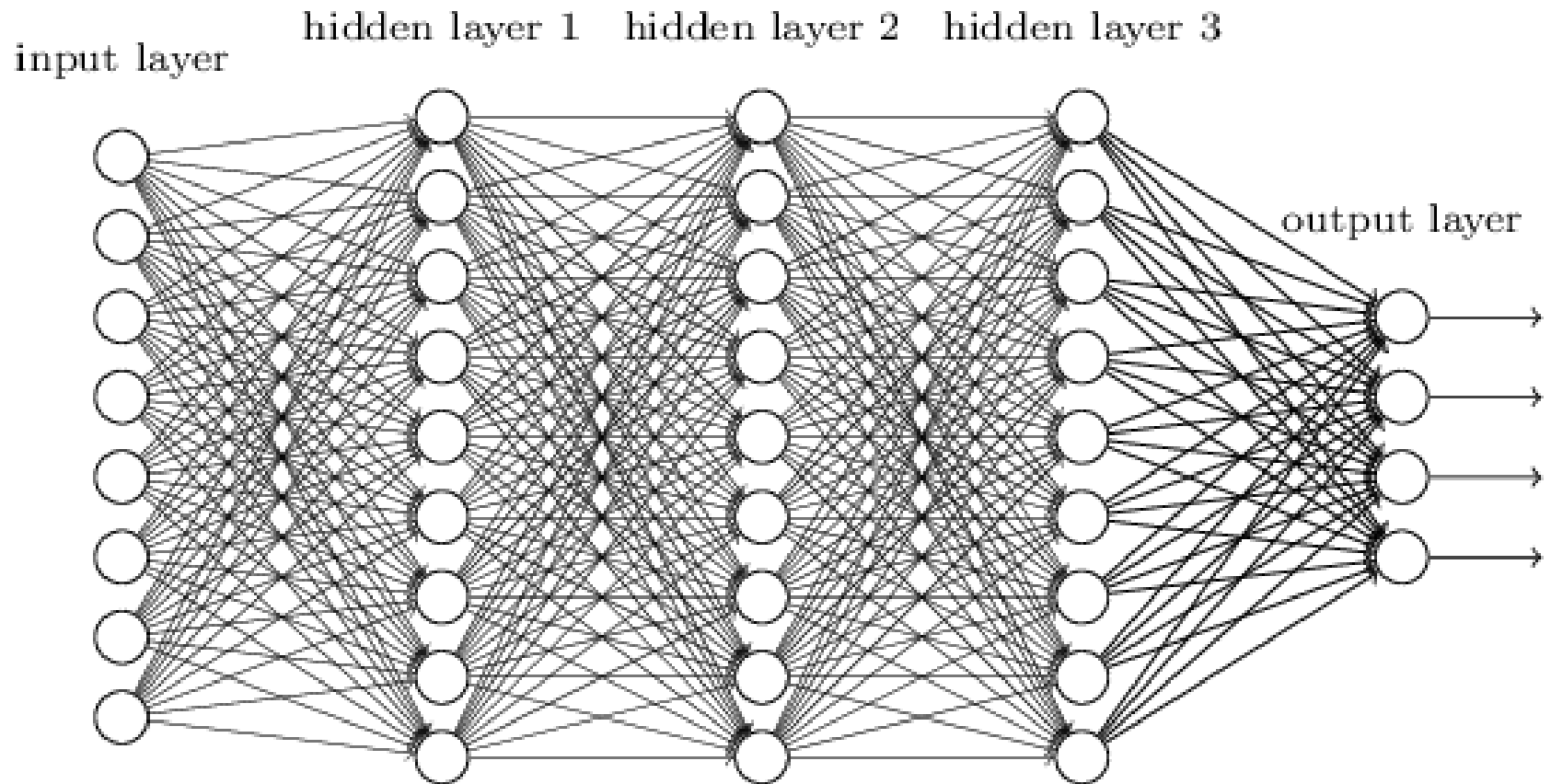
CONVOLUTIONAL NEURAL NETWORK

RECURRENT NEURAL NETWORK

What is deep learning? Why to use it?

- ▶ Deep Neural Networks
 - ▶ Having lots of hidden layer neurons
 - ▶ Hard to train
 - ▶ but much more powerful!!!
- ▶ Deep Learning
 - ▶ Techniques to use/train a deep neural net efficiently
 - ▶ CNN, RNN, GPU processing, ...

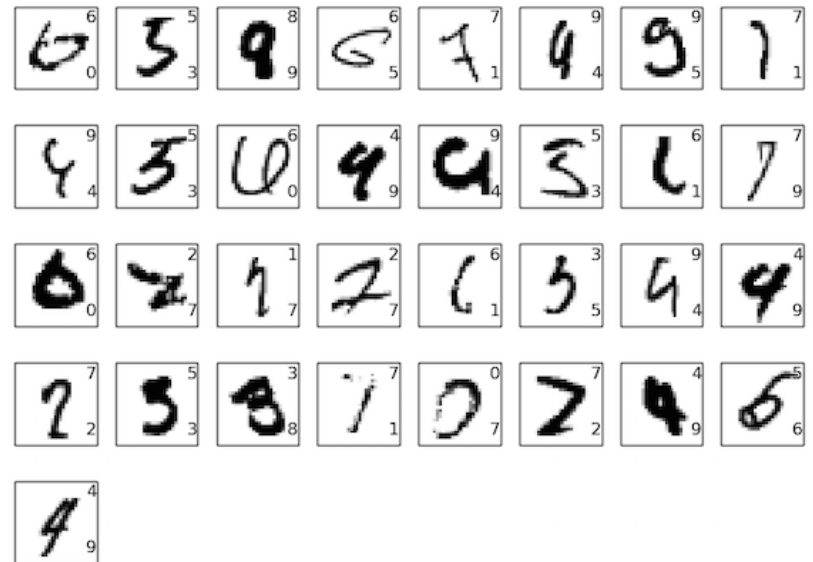
How deep are the deep neural nets



MNIST

- ▶ A famous database for image processing purposes
- ▶ MNIST database of handwritten digits
 - ▶ 28*28 pixel images

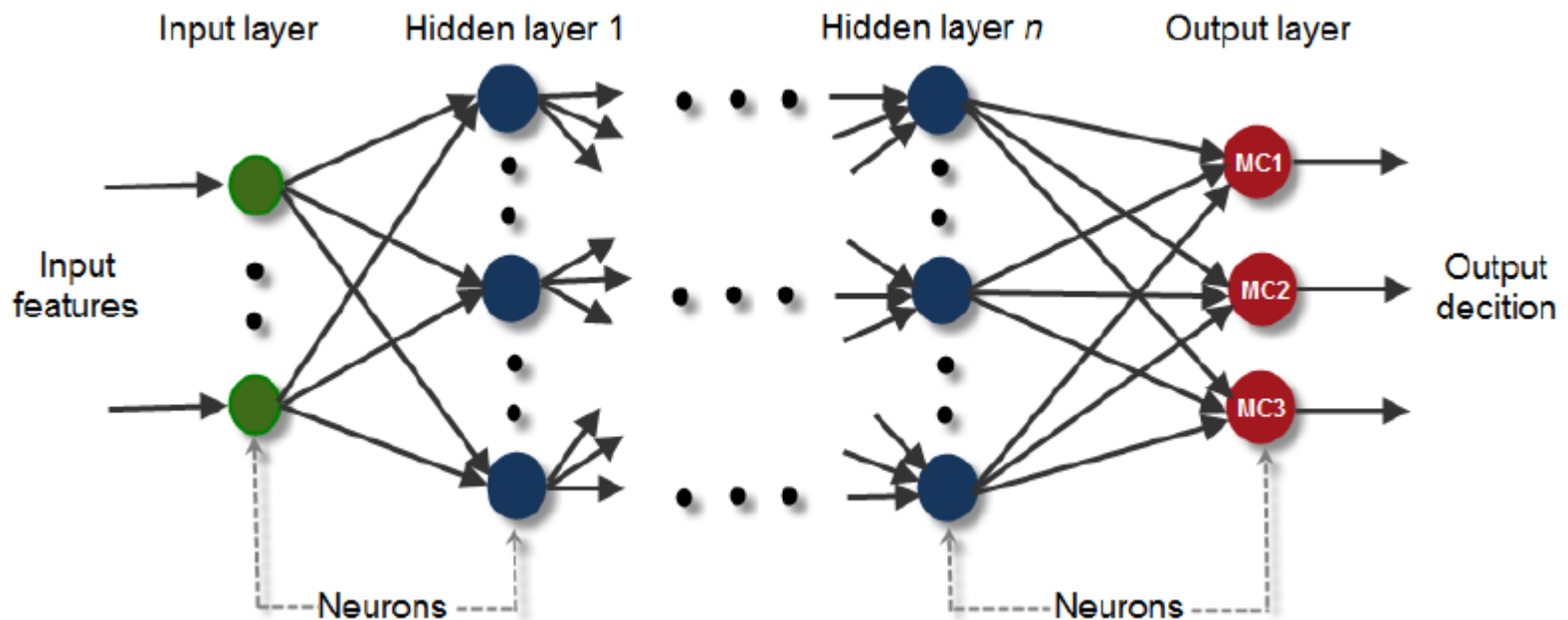
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<http://yann.lecun.com/exdb/mnist/>

Classic MLP Networks; a review

- ▶ Fully connected
- ▶ Input – hidden – output



Convolutional Neural Network (CNN)

- ▶ A powerful classifier
 - ▶ Recognition
 - ▶ Detection
 - ▶ Extracts useful features automatically!
- ▶ 3 Basic feature (To be discussed later)
 - ▶ Local Receptive Fields
 - ▶ Shared Weight and Biases
 - ▶ Pooling Layer

CNN structure

- ▶ 4 group of layers

- I. Input layer

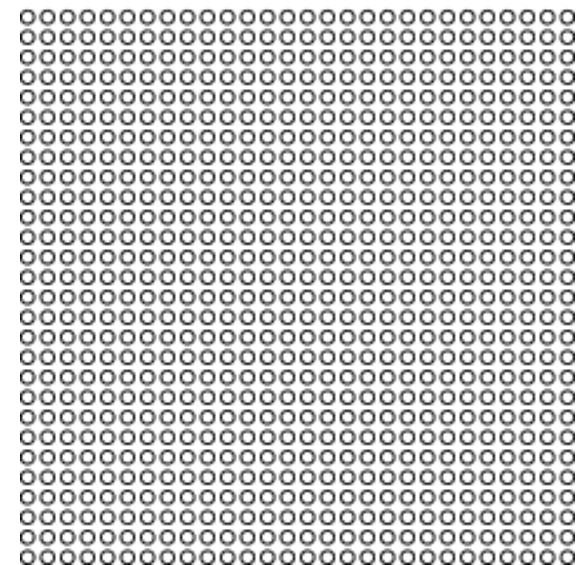
- ▶ Neurons are ordered in a matrix form not a vector

- II. Convolutional layer(s)

- III. Pooling layer

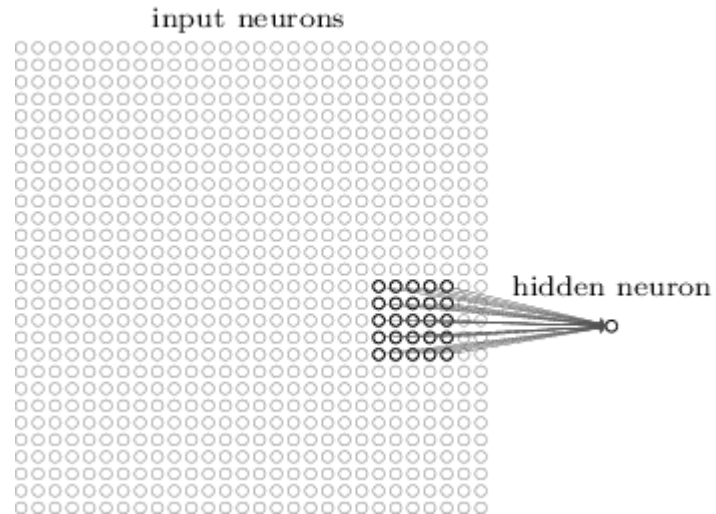
- IV. Output layer

input neurons



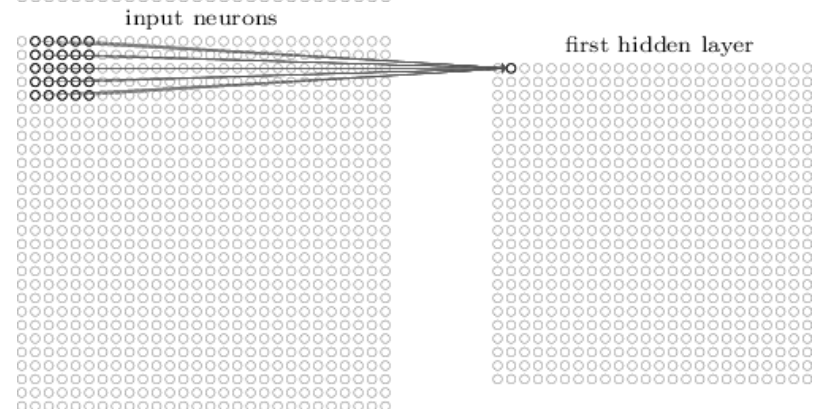
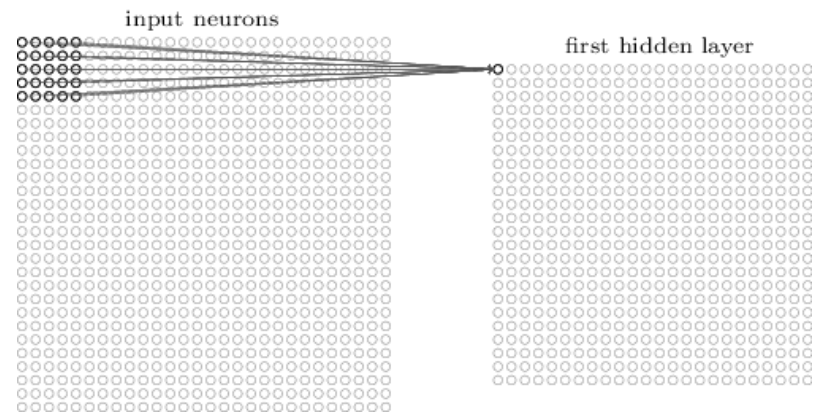
Local Receptive Field

- ▶ Unlike classic nets
 - ▶ Each hidden layer neuron is connected only to a small region of the input/previous layer
 - ▶ We call this region “Local Receptive Field” of that specific neuron



Local Receptive Field (cont.)

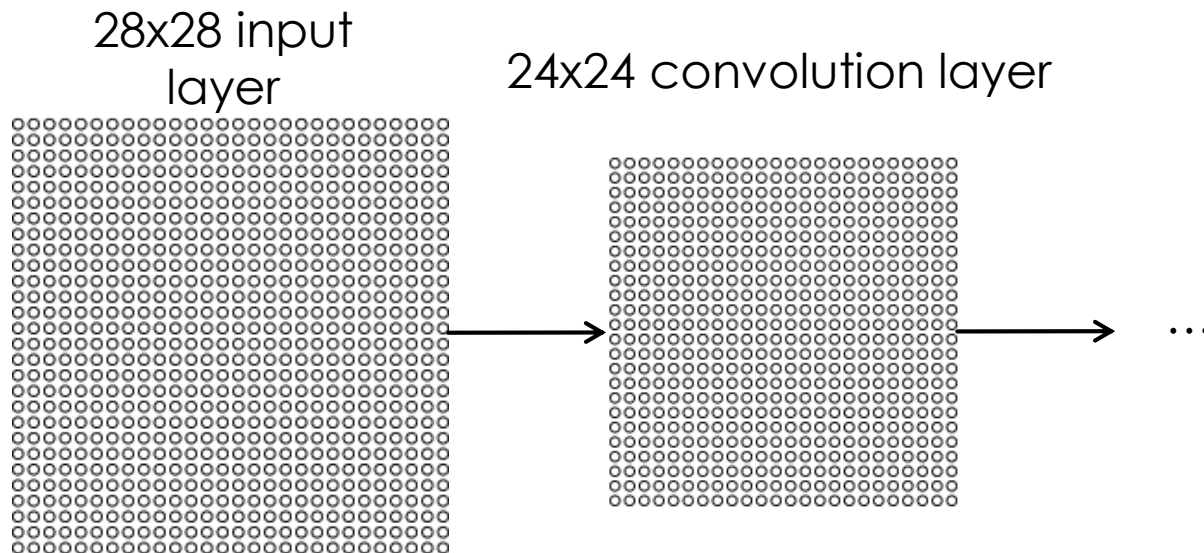
- ▶ By sliding this field (region) over the input layer we form the first layer of hidden layer.
- ▶ This is a convolutional layer. Guess why?!
- ▶ Convolution operation
- ▶ Stride length



Local Receptive Field (cont.)

- ▶ Consider the followings
 - ▶ input as 28x28 pixel image
 - ▶ Local receptive field as 5x5 window
- ▶ What is the size for the convolutional layer?
 - ▶ 24x24

CNN structure so far



Shared Weight and Biases

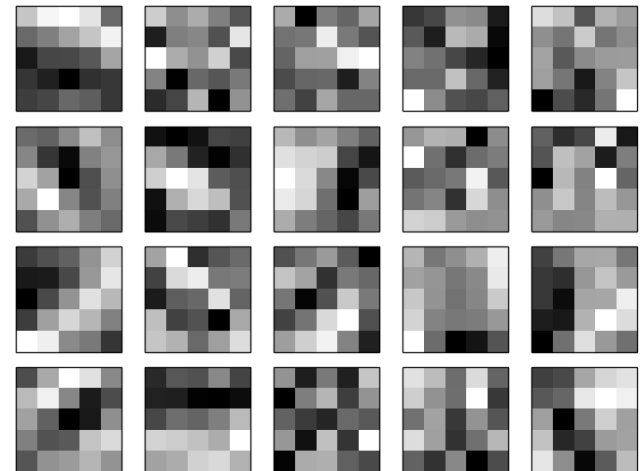
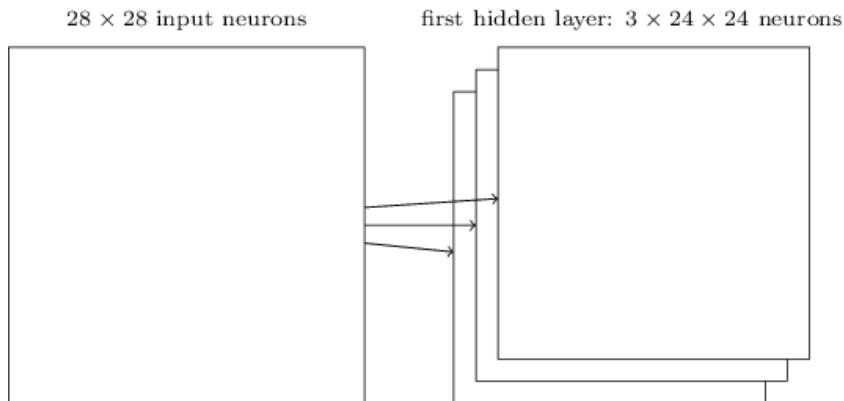
- ▶ hidden layer neurons
 - ▶ 1 Bias
 - ▶ $5 \times 5 = 25$ Weights
- ▶ Interesting that
 - ▶ All the hidden neuron in a single layer shares the weights and bias!
- ▶ How this helps?!
- ▶ Output for a hidden layer neuron



$$a_{j,k} = f \left(b + \sum_{l=0}^4 \sum_{m=0}^4 w_{l,m} \times a_{j+l,k+m} \right)$$

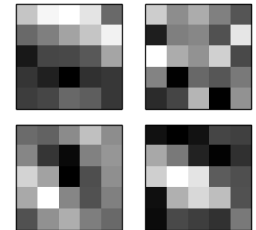
Shared Weight and Biases (cont.)

- ▶ So shared weights and biases means
 - ▶ All neurons in the first hidden layer detect exactly the same feature in different locations.
 - ▶ This feature can be a simple vertical edge
- ▶ Now if we have multiple convolutional layers
 - ▶ Detect different features



Shared Weight and Biases (cont.)

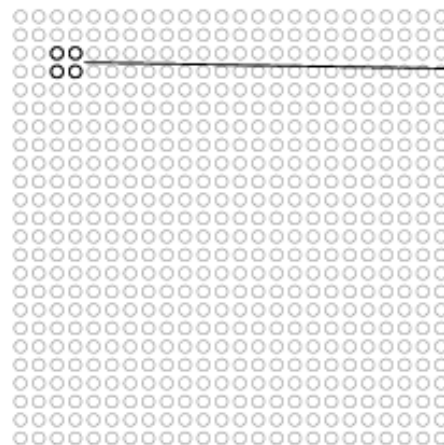
- ▶ Shared weights + bias defines a kernel/filter
- ▶ Feature map
 - ▶ Map from input layer to hidden layer
 - ▶ Containing shared weights and bias
 - ▶ Visualizing
 - ▶ Darker pixels: larger weights
 - ▶ Brighter pixels: smaller weights
- ▶ Considering all above so far we are able to
 - ▶ Detect multiple features at different locations



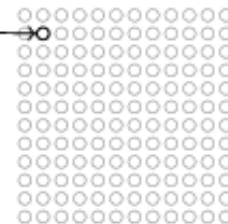
Pooling Layer

- ▶ Right after the convolutional layer
- ▶ Summarize results in convolution layer
 - ▶ Represents finding a feature in a region without any consideration of the exact location

hidden neurons (output from feature map)

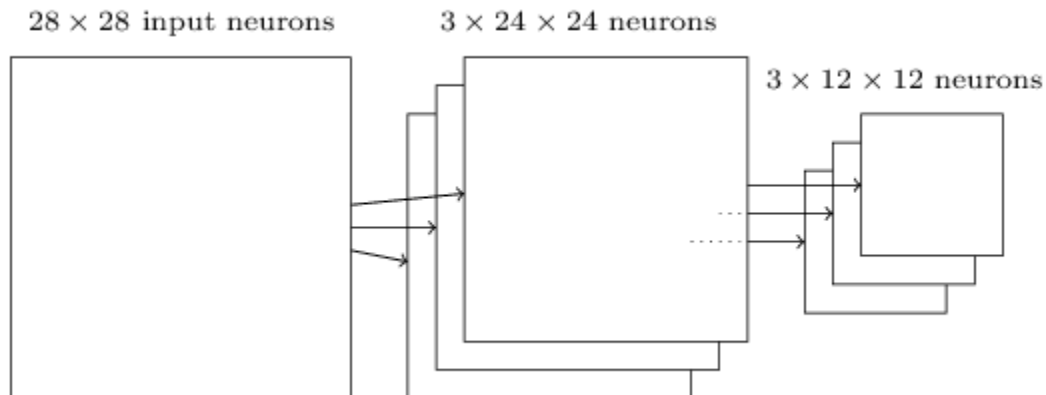


max-pooling units



Pooling Techniques

- ▶ Max Pooling
- ▶ L2 Pooling
- ▶ Each map (convolution layer) has a pooling layer



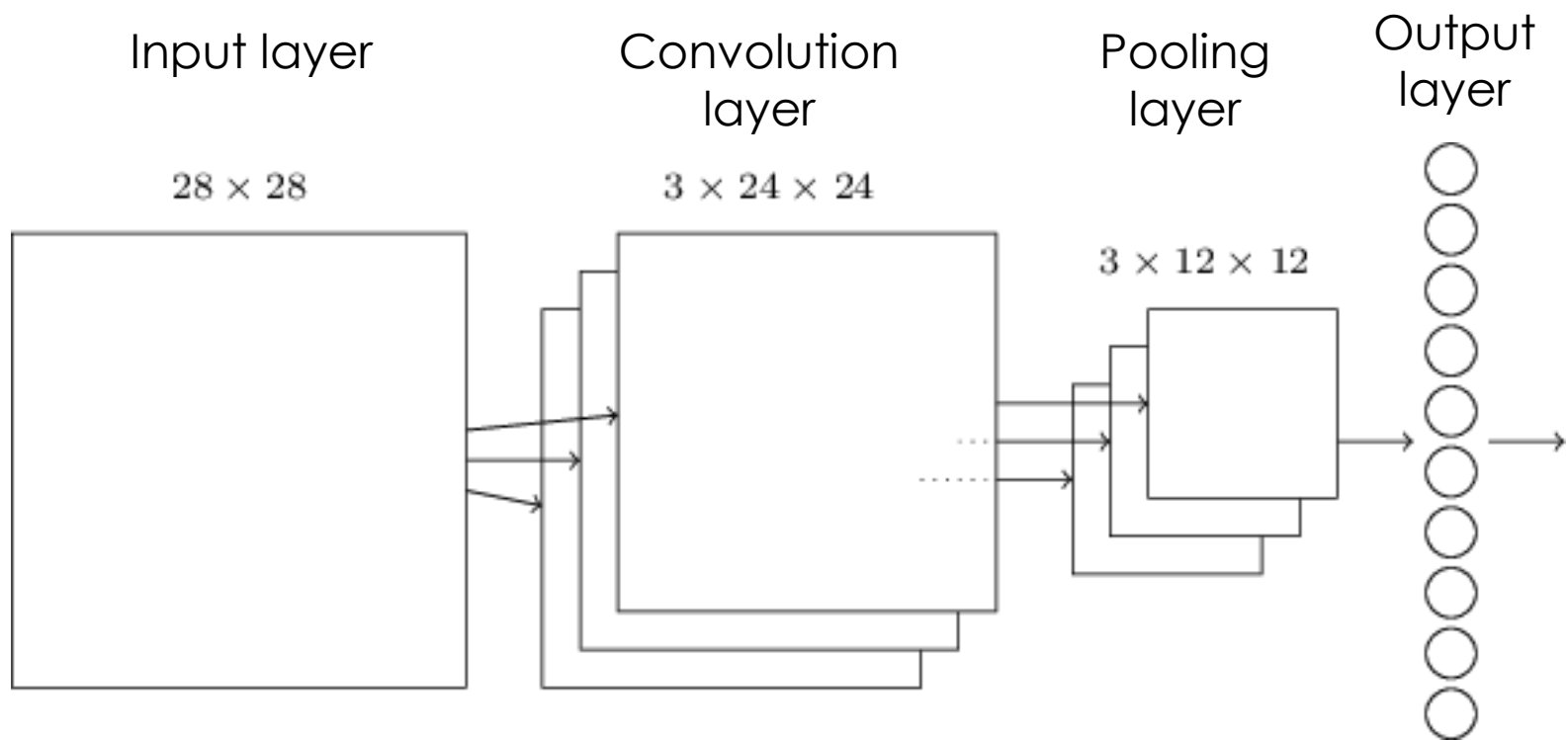
$$\max_{i=1..l} \max_{j=1..k} a_{i,j}$$

$$\sqrt{\sum_{i=1}^l \sum_{j=1}^k (a_{i,j})^2}$$

Output Layer

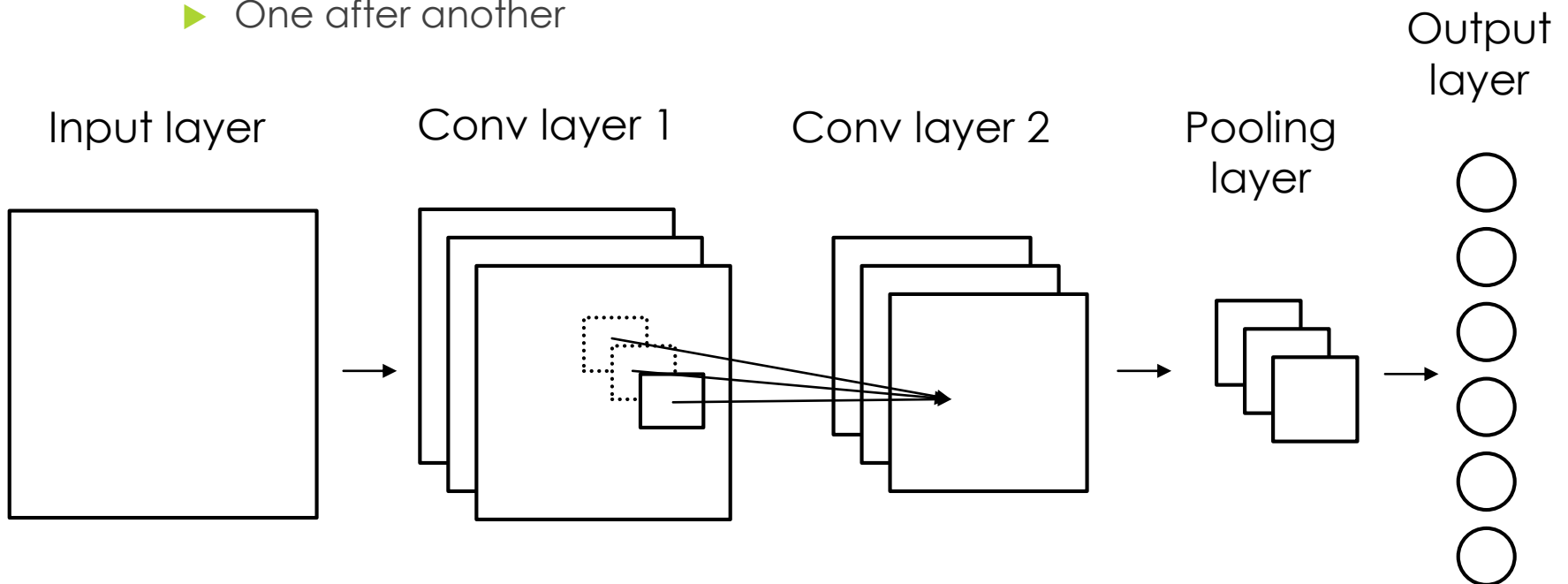
- ▶ Pulling all together
- ▶ Fully connected network
- ▶ Contains N neurons, number of all possible values

Final CNN Structure



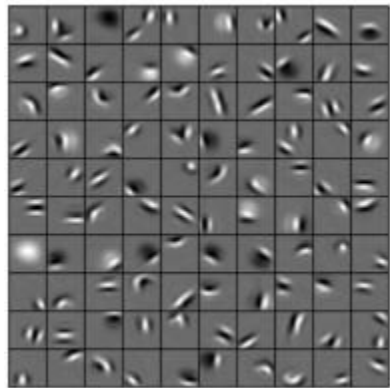
Supplementary Notes

- ▶ Training using a modified version of EBP
- ▶ We can add more convolutional layers to detect complicated features
 - ▶ One after another



Complicated Feature

- ▶ By visualizing the Weights matrix in each convolutional layer

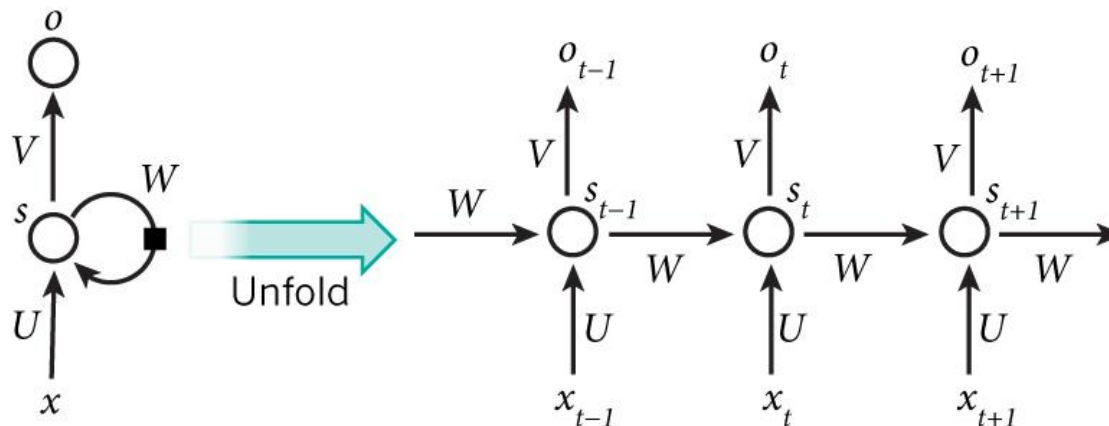


RNN; a short glance

- ▶ Recurrent Neural Networks
- ▶ Applications in
 - ▶ Character recognition
 - ▶ Speech recognition
 - ▶ NLP
- ▶ Use sequential information
- ▶ Inputs are not independent from each other
 - ▶ Predict next word in a sentence
 - ▶ Depends on previous words

RNN; a short glance

- ▶ X is the input for each step
- ▶ S is the memory unit at each step (Cumulative memory) or “hidden state”
- ▶ O is the output at each step
- ▶ They are all shared (W, U, V)
- ▶ To predict a 5 word sentence, 5 layer is needed.



Resources

- ▶ For more resources visit
 - ▶ <http://neuralnetworksanddeeplearning.com/chap6.html>
 - ▶ RNN, DL, NLP:
<http://www.wildml.com/2015/09/recurrent-neural-networks-tutorial-part-1-introduction-to-rnns/>



Thank you for your time