# Tutorial 3. Deep Learning Toolbox

Xingyu ZENG(xyzeng@ee.cuhk.edu.hk)

- A open-source Matlab toolbox for Deep Learning
- You can download in

https://github.com/rasmusbergpalm/DeepLearnToolbox

If you use this toolbox in your research please cite
@MASTERSTHESIS\{IMM2012-06284, author = "R. B. Palm", title = "Prediction as a candidate for learning deep hierarchical models of data", year = "2012", }



- Advantage
  - Matlab, easy to use
  - Open-source
- Disadvantage
  - Only CPU version, slow

#### Install Steps

- 1. Download the toolbox,
- Addpath(genpath('DeepLearnToolbox'))



- A Matlab toolbox for Deep Learning
  - NN/ A library for Feedforward Backpropagation Neural Networks
  - CNN/ A library for Convolutional Neural Networks
  - DBN/ A library for Deep Belief Networks
  - SAE/ A library for Stacked Auto-Encoders
  - CAE/ A library for Convolutional Auto-Encoders
  - util/ Utility functions used by the libraries
  - data/ Data used by the examples
  - tests/ unit tests to verify toolbox is working



#### Feedforward Backpropagation Neural Networks

- Common Function
  - > nnsetup.m
    - > To setup one network
  - > nntrain.m
    - > To train one network
  - > nnpredict.m
    - > To test samples with one network



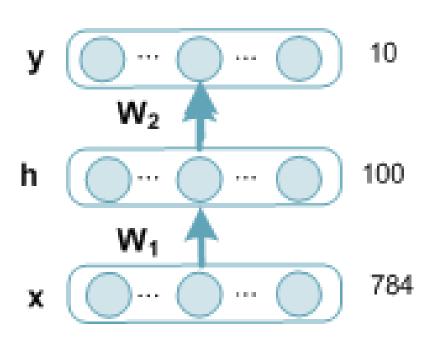
nnsetup.m

Usage example:

nn = nnsetup([784 100 10]); % to build up one three layers network

nn.activation\_function = 'sigm';
nn.output= 'softmax';

h=sigmoid( $W_1$ \*x); y=softmax( $W_2$ \*h);





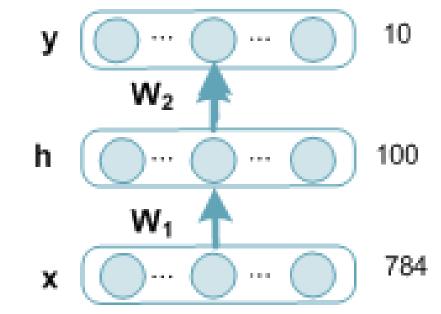
nn = nntrain(nn, train\_x, train\_y, opts);



nnpredict.m

Usage example:

labels = nnpredict(nn, test\_x);



#### Notes:

- 1. labels, the classes predicted by nn
- 2. nn.a{end}, the values of the output layer

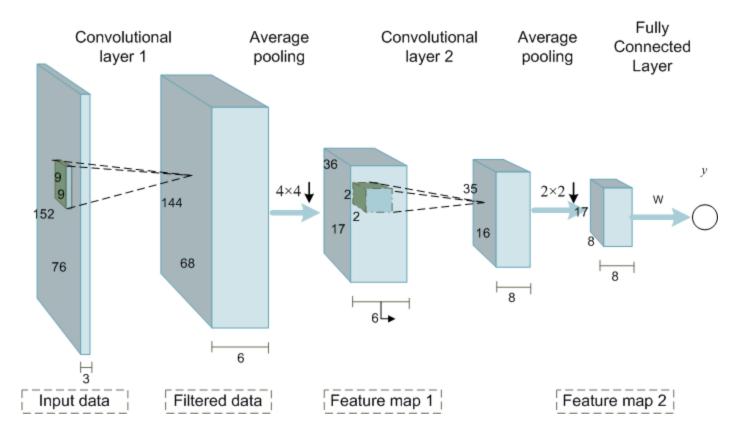
 $[\sim, labels] = max(nn.a{end}, [], 2);$ 

#### Convolutional Neural Networks

- Common Function
  - > cnnsetup.m
    - > To setup one convolutional network
  - > cnntrain.m
    - > To train one convolutional network
  - > cnnff.m
    - > Forward step with one convolutional network
  - > cnntest.m
    - > To test samples with one convlutional network



▶ Suppose input size: I 52\*76\*3





cnnsetup.mUsage example:

```
cnn.layers = {
    struct('type', 'i') %input layer
    struct('type', 'c', 'outputmaps', 6, 'kernelsize', 9) %convolution layer
    struct('type', 's', 'scale', 4) %sub sampling layer
    struct('type', 'c', 'outputmaps', 8, 'kernelsize', 2) %convolution layer
    struct('type', 's', 'scale', 2) %subsampling layer
    struct('type', 's', 'scale', 2) %subsampling layer
};
```

% the size of {train\_x, train\_y} is useful for setup cnn

cnn = cnnsetup(cnn, train\_x, train\_y);

Convolutional

layer 1

Convolutional

pooling

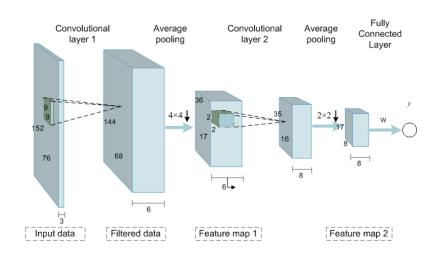
pooling



cnntrain.m

Usage example:

```
opts.alpha = I; % learning rate
opts.batchsize = 50;
opts.numepochs = I;
```

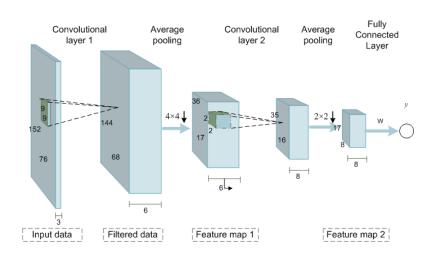


cnn = cnntrain(cnn, train\_x, train\_y, opts);

cnnff.m

Usage example:

cnn = cnnff(cnn, x);



#### Notes:

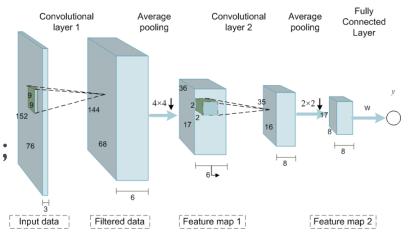
- 1. cnn.o, the output values of the output layer
- 2. cnn.fv, the feature of the input samples
- 3. cnn.o=sigm(cnn.ffW \* cnn.fv + repmat(cnn.ffb, I, size(cnn.fv, 2)));



cnntest.m

Usage example:

[er, bad] = cnntest(cnn, test\_x, test\_y);



#### Notes:

- er, error fraction value
- 2. bad, index of misclassified testing samples

More examples can be found in

https://github.com/rasmusbergpalm/DeepLearnToolbox/blob/master/README.md

More details about deep learning, Book, 'Learning Deep Architectures for Al'

