DEEP LEARNING – PART 1

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ComputeFest 2015

What We Cover Today:

- Basic theory of deep neural networks (DNN)
- Introduction to Theano
- Training deep neural networks with Theano
- Some useful tips and tricks for training DNNs

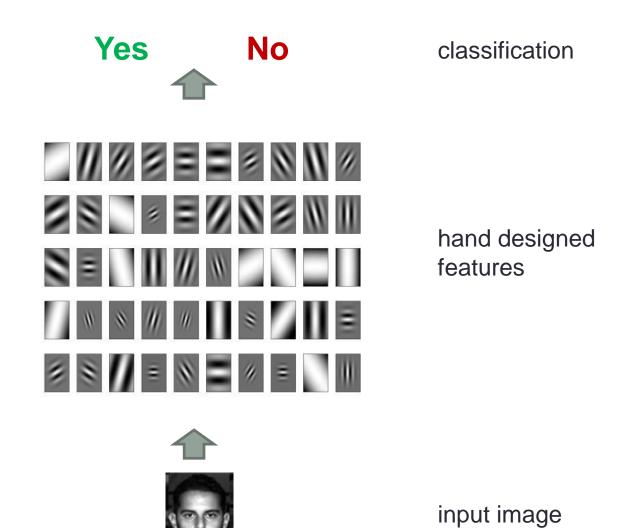
Workshop Goal

- Flatten the learning curve for the Theano tutorials
- http://deeplearning.net/tutorial/

Deep Learning

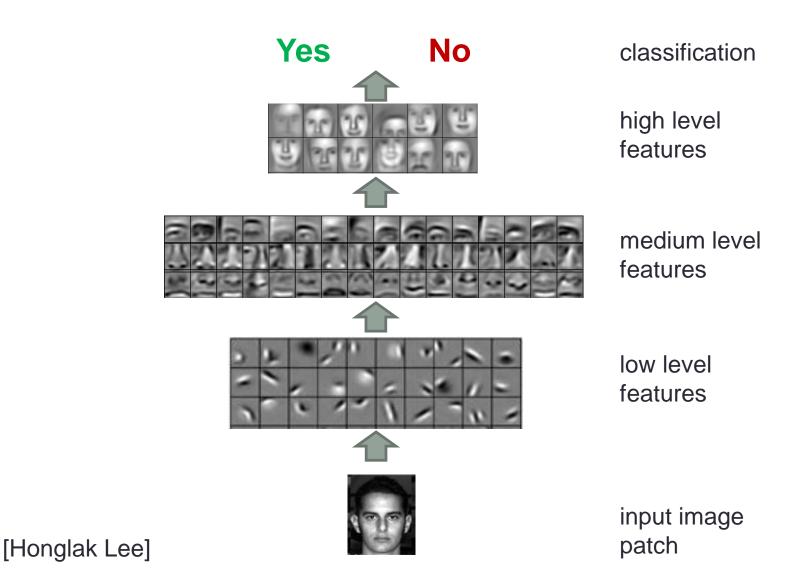


Why – Feature Design

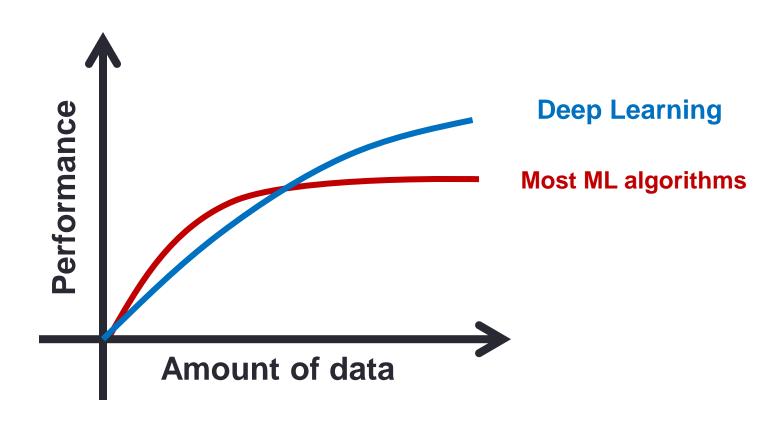


patch

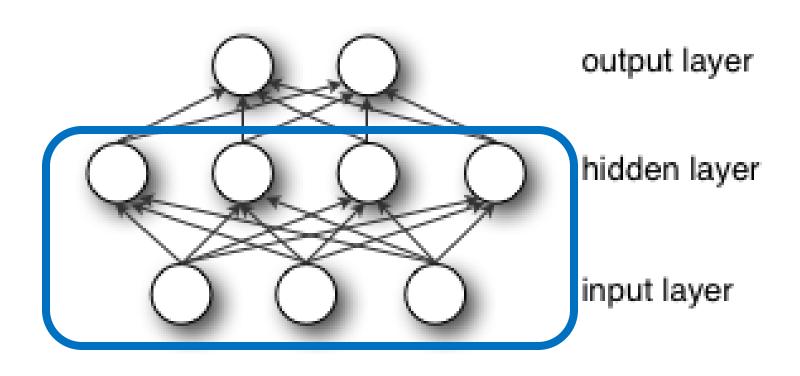
Learned Feature Hierarchy



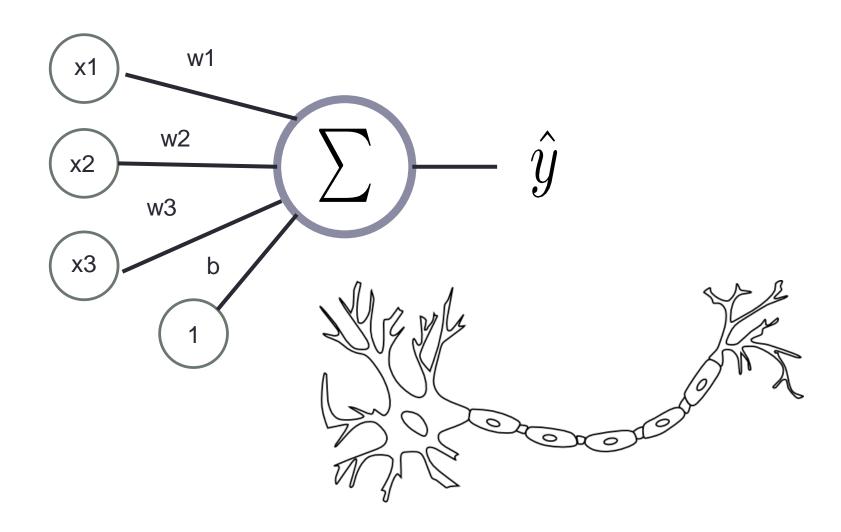
Scaling with Data Size



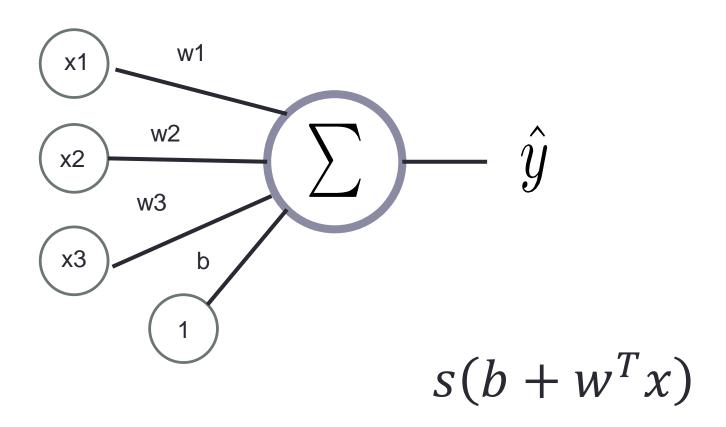
What – Neural Network Architecture



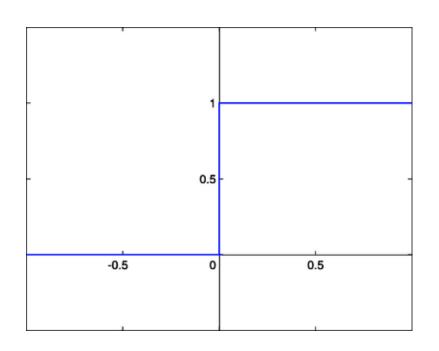
Artificial Neuron

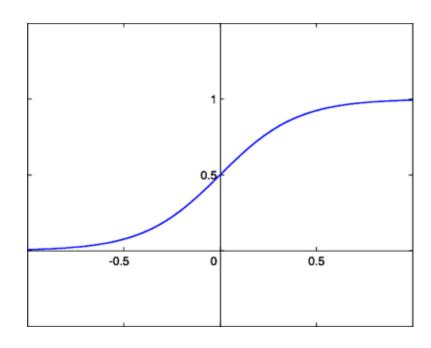


Artificial Neuron



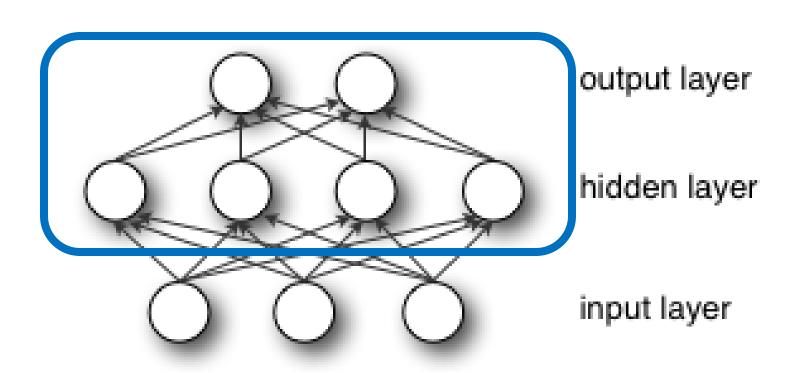
Step vs Sigmoid Activation





$$s(x) = \frac{1}{1 + e^{-cx}}$$

Neural Network Architecture



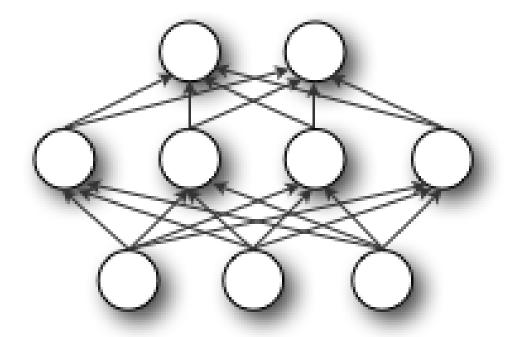
Logistic Regression (Mulitclass)

Probabilistic classifier

$$P(Y = i|x, W, b) = softmax_i(Wx + b)$$
$$= \frac{e^{W_i x + b_i}}{\sum_j e^{W_j x + b_j}}$$

$$y_{pred} = \operatorname{argmax}_{i} P(Y = i | x, W, b)$$

Neural Network Architecture



output layer

hidden layer

input layer

$$f(x) = G(b^{(2)} + W^{(2)}(s(b^{(1)} + W^{(1)}x)))$$

How - Optimization

 As so many machine learning methods, also DNNs are about optimizing a loss function:

$$\underset{\text{model parameter}}{\operatorname{arg\,min}} \frac{1}{T} \sum_{t}^{t} l(f(\mathbf{x}^{(t)}; \boldsymbol{\theta}), y^{(t)}) + \lambda \Omega(\boldsymbol{\theta})$$

Loss Function

- Classification error would be good, but it's not smooth
- Need to find smooth proxy
- Last layer is logistic regression
- The whole network estimates class probabilities

$$P(Y = i|x, W, b) = softmax_i(Wx + b)$$

Maximize the correct class assignment probabilities in the training set

Minimize Negative Log Likelihood

- Maximize the likelihood of the data set given the model parameters
- Take the log to simplify for numerical stability and math simplicity
- Take negative log likelihood to cast into minimization

training data
$$\bigvee_{I \in \mathcal{D}} \int_{i=0}^{|\mathcal{D}|} \log P(Y=y^{(i)}|x^{(i)},\theta)$$
 Parameter W, b

Stochastic Gradient Decent (SGD)

- Take small steps downward on the surface of the loss function
- For each training sample:

$$\Delta = -\nabla_{\boldsymbol{\theta}} l(f(\mathbf{x}^{(t)}; \boldsymbol{\theta}), y^{(t)})$$

- Estimate gradient
- Update Parameters

$$\boldsymbol{\theta} \leftarrow \boldsymbol{\theta} + \alpha \Delta$$

learning rate

- We minimize negative log likelihood
- Need to compute gradient Ugh
- Theano does it for us!!

Theano

- Python library for efficient evaluation of mathematical expressions
- tight integration with NumPy Use numpy.ndarray in Theano-compiled functions.
- transparent use of a GPU Perform data-intensive calculations up to 140x faster than with CPU.(float32 only)
- efficient symbolic differentiation Theano does your derivatives for function with one or many inputs.

Free and Open Source

- F. Bastien, P. Lamblin, R. Pascanu, J. Bergstra, I.
 Goodfellow, A. Bergeron, N. Bouchard, D. Warde-Farley and Y. Bengio
- Giving credit: http://deeplearning.net/software/theano/citation.html#citation

Exercise: clone github repository

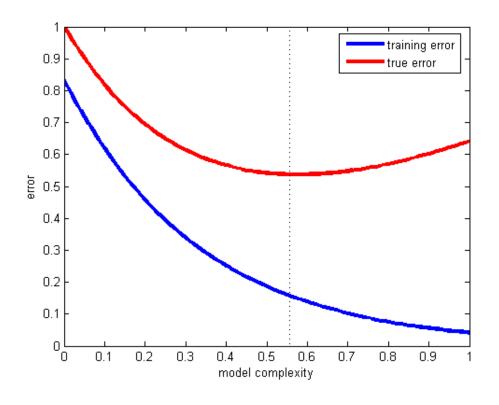
 git clone --recursive <u>https://github.com/vkaynig/ComputeFest2015_DeepLearning.git</u>

LET'S GET SOME EXPERIENCE

TIPS AND TRICKS

Number of Layers / Size of Layers

- If data is unlimited larger and deeper should be better
- Larger networks can over fit more easily
- Take computational cost into account



Learning Rate

- One of the most important parameters
- If network diverges most probably learning rate is too large
- Smaller works better
- Can slowly decay over time
- Can have one learning rate per layer

Other tips for SGD:

http://leon.bottou.org/publications/pdf/tricks-2012.pdf

Momentum

- Helps to escape local minima
- Crucial to achieve high performance

$$v_{t+1} = \mu v_t - \varepsilon \nabla f(\theta_t)$$

$$\theta_{t+1} = \theta_t + v_{t+1}$$

More about Momentum:

http://www.jmlr.org/proceedings/papers/v28/sutskever13.pdf

Convergence

- Monitor validation error
- Stop when it doesn't improve within n iterations
- If learning rate decays you might want to adjust number of iterations

Initialization of W

- Need randomization to break symmetry
- Bad initializations are untrainable
- Most heuristics depend on the number of input (and output) units
- Sometimes W is rescaled during training
 - Weight decay (L2 regularization)
 - Normalization

Data Augmentation

- Exploit invariances of the data
- Rotation, translation
- Nonlinear transformation
- Adding Noise

Type \$	Classifier +
Neural network	6-layer NN 784-2500-2000-1500-1000-500-10 (on GPU), with elastic distortions
Convolutional neural network	Committee of 35 conv. net, 1-20-P-40-P-150-10, with elastic distortions

Preprocessing +	Error rate (%) \$
None	0.35 ^[17]
Width normalizations	0.23 ^[8]

Data Normalization

We have seen std and mean normalization

- Whitening
 - Neighbored pixels often are redundant
 - Remove correlation between features

More about preprocessing: http://deeplearning.stanford.edu/wiki/index.php/Data_Preprocessing

Non-Linear Activation Function

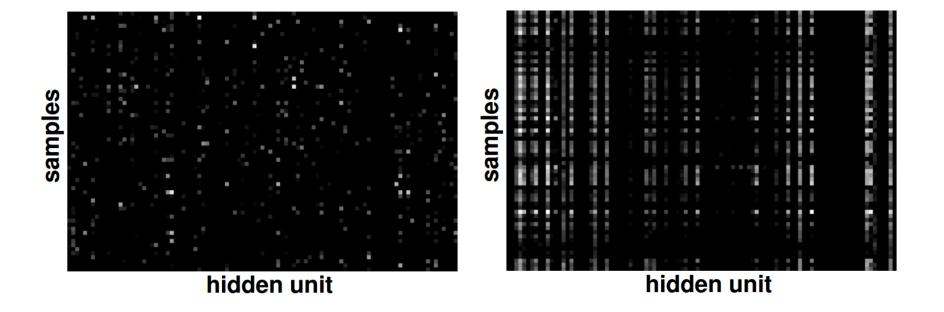
- Sigmoid
 - Traditional choice
- Tanh
 - Symmetric around the origin
 - Better gradient propagation than Sigmoid
- Rectified Linear (discussed tomorrow)

L1 and L2 Regularization

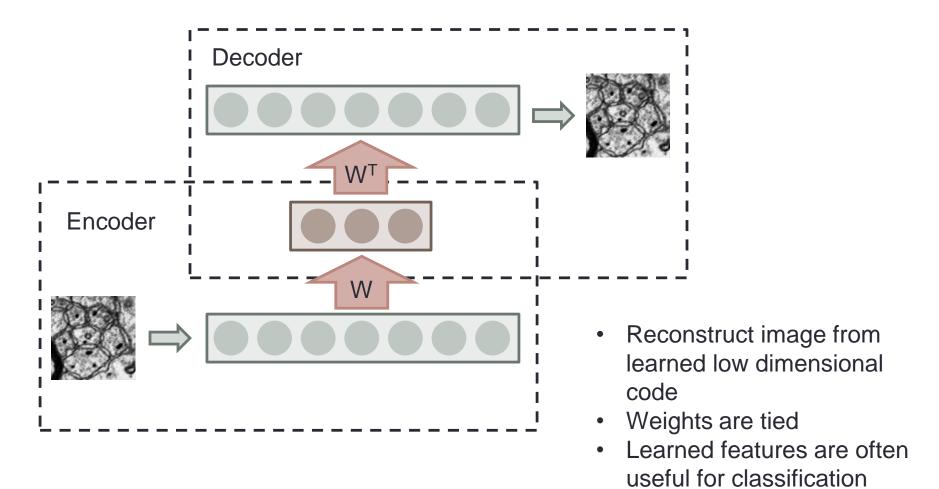
- Most pictures of nice filters involve some regularization
- L2 regularization corresponds to weight decay
- L2 and early stopping have similar effects
- L1 leads to sparsity
- Might not be needed anymore (more data, dropout)

Monitoring Training

- Monitor training and validation performance
- Can monitor hidden units
- Good: Uncorrelated and high variance

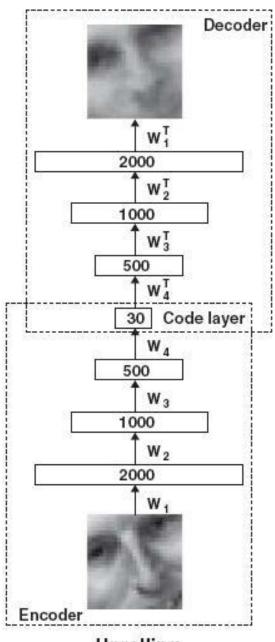


Autoencoder



Deep Autoencoder

- Each Layer trained by itself
- Higher layers build on top of lower layers
- Can do fine tuning in the end



Unrolling

Further Resources

- More about theory:
 - Yoshua Bengio's book:http://www.iro.umontreal.ca/~bengioy/dlbook/
 - Deep learning reading list: http://deeplearning.net/reading-list/
- More about Theano:
 - http://deeplearning.net/software/theano/
 - http://deeplearning.net/tutorial/