



19ECS 773: Deep Learning

Topic: Feed Forward Neural Networks

Unit I

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Outline

- What are Neural Networks?
- Biological Neural Networks
- ANN – The basics
- Feed forward net
- Training
- Example – Voice recognition
- Applications – Feed forward nets



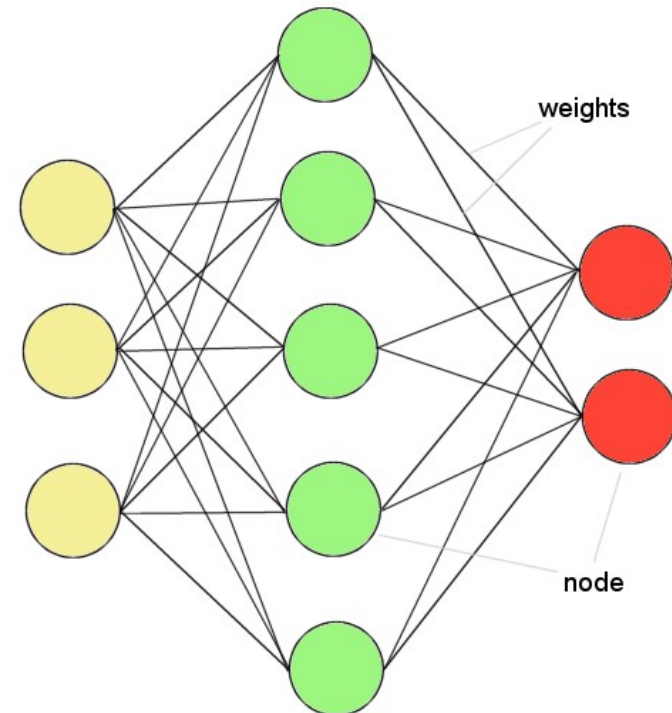
What are Neural Networks?

- Models of the brain and nervous system
- Highly parallel
 - Process information much more like the brain than a serial computer
- Learning
- Very simple principles
- Very complex behaviours
- Applications
 - As powerful problem solvers
 - As biological models

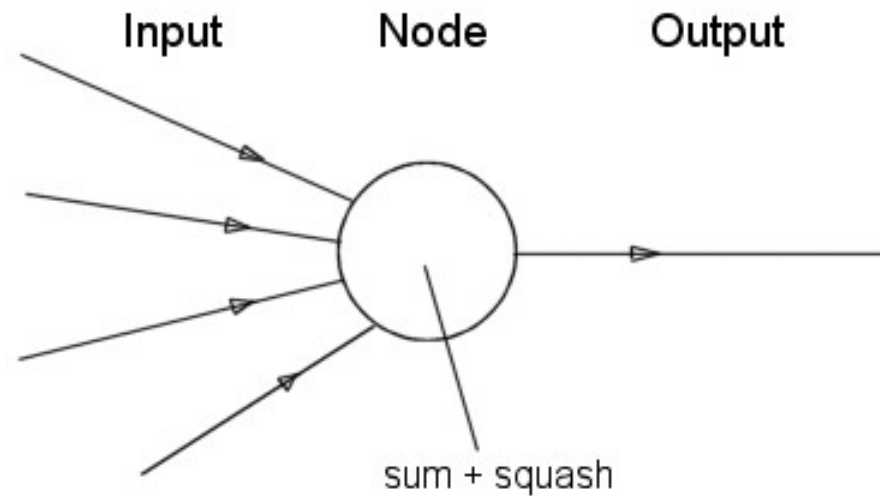
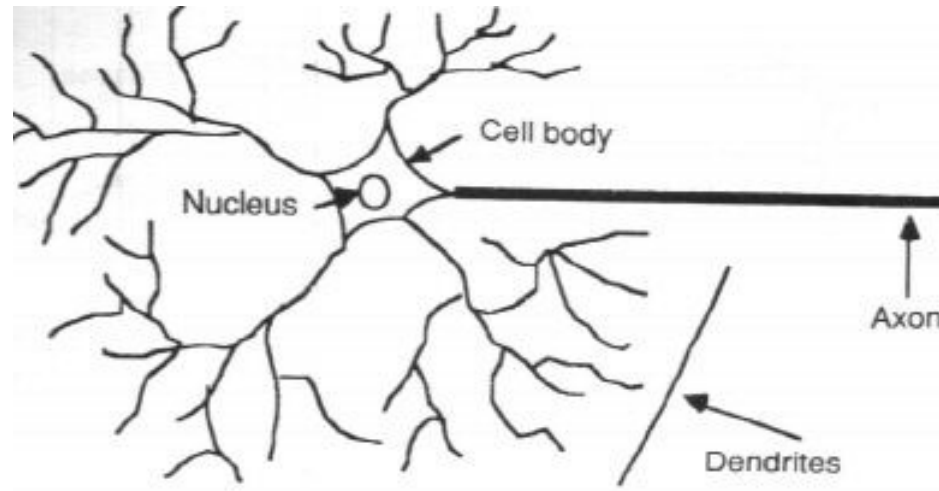
ANNs – The basics

- ANNs incorporate the two fundamental components of biological neural nets:

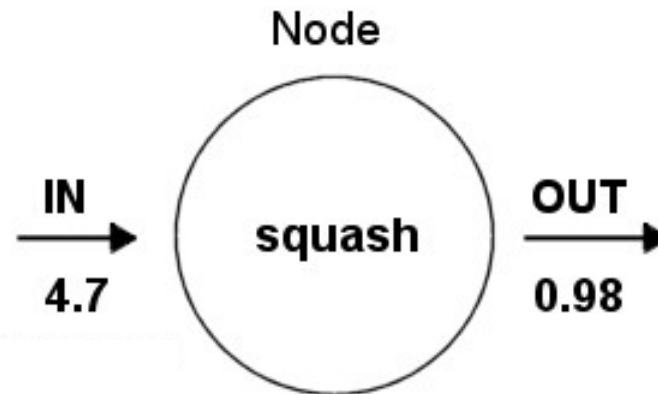
1. Neurons (nodes)
2. Synapses (weights)



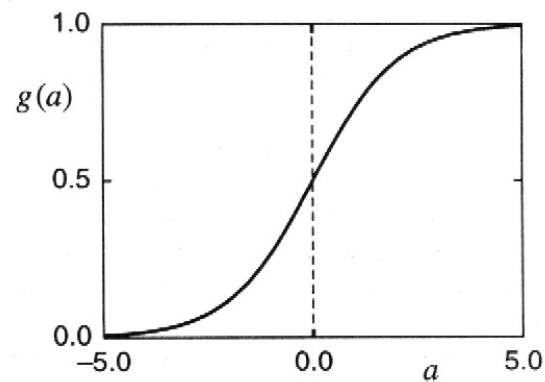
- Neurone vs. Node



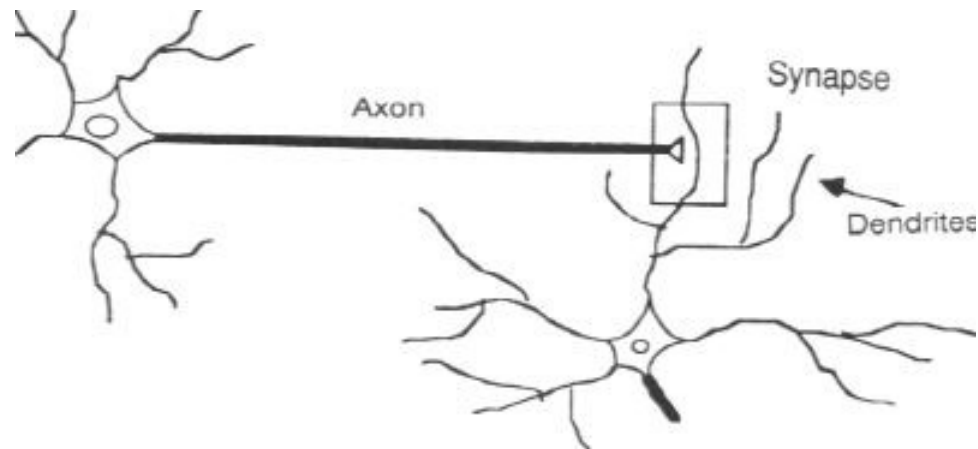
- Structure of a node:



- Squashing function limits node output:

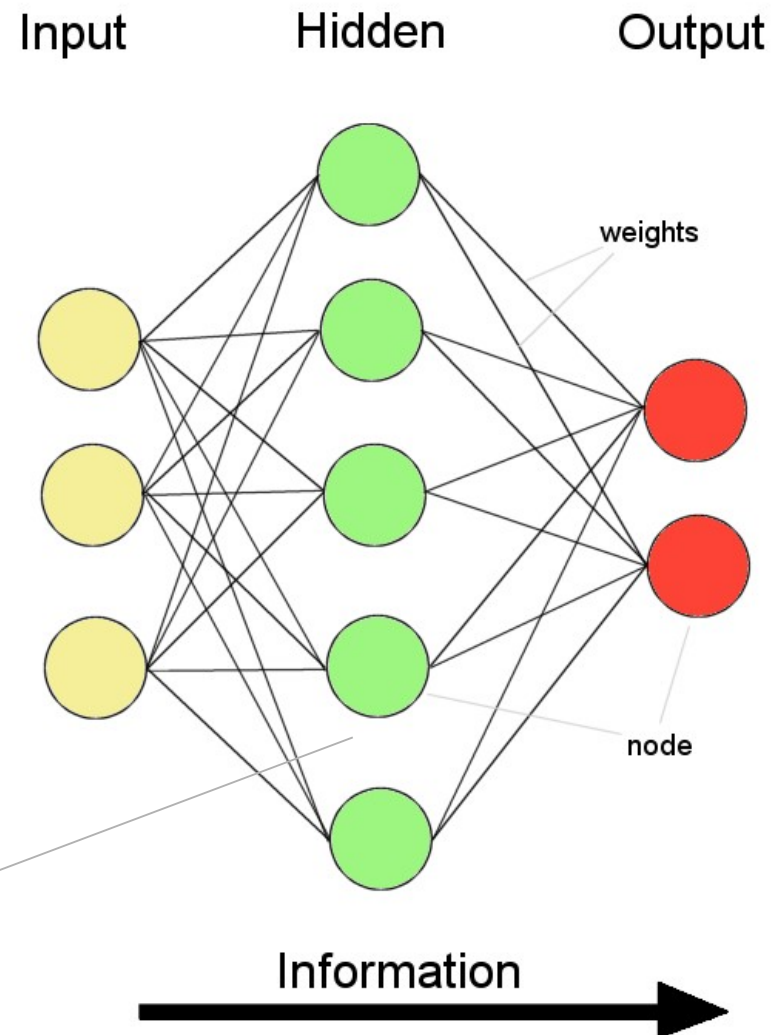


- Synapse vs. weight



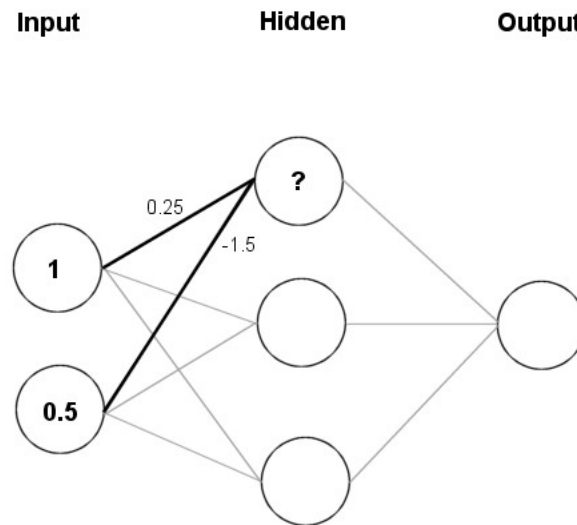
Feed-forward nets

- Information flow is unidirectional
 - Data is presented to *Input layer*
 - Passed on to *Hidden Layer*
 - Passed on to *Output layer*
- Information is distributed
- Information processing is parallel



Internal representation (interpretation) of data

- Feeding data through the net:



$$(1 \times 0.25) + (0.5 \times (-1.5)) = 0.25 + (-0.75) = -0.5$$

Squashing: $\frac{1}{1 + e^{0.5}} = 0.3775$

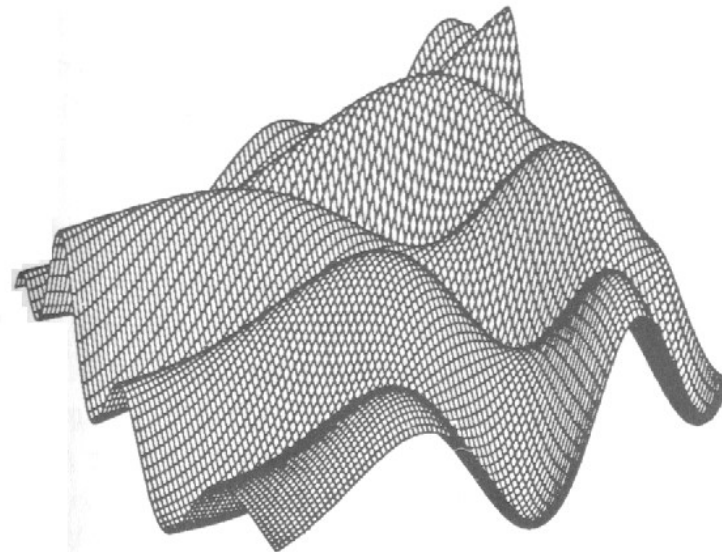
- Data is presented to the network in the form of activations in the input layer
- Examples
 - Pixel intensity (for pictures)
 - Molecule concentrations (for artificial nose)
 - Share prices (for stock market prediction)
- Data usually requires preprocessing
 - Analogous to senses in biology
- How to represent more abstract data, e.g. a name?
 - Choose a pattern, e.g.
 - 0-0-1 for “Chris”
 - 0-1-0 for “Becky”

- Weight settings determine the behaviour of a network
- How can we find the right weights?

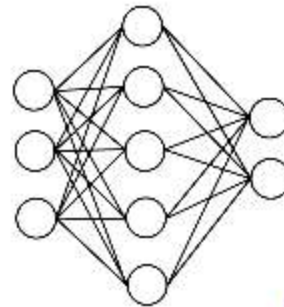


Training the Network - Learning

- Backpropagation
 - Requires training set (input / output pairs)
 - Starts with small random weights
 - Error is used to adjust weights (supervised learning)
 - Gradient descent on error landscape



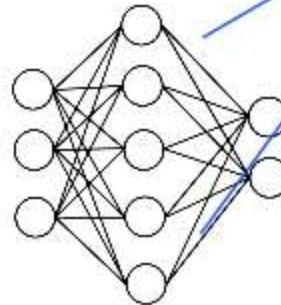
1.



Wallace

Wallace - Darwin (calculate error)

2.

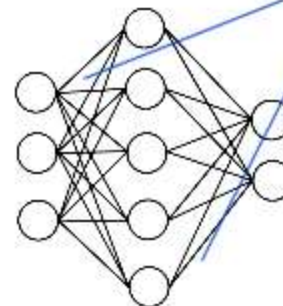


adjust weights

Wallace

Wallace - Darwin (calculate error)

3.





adjust weights

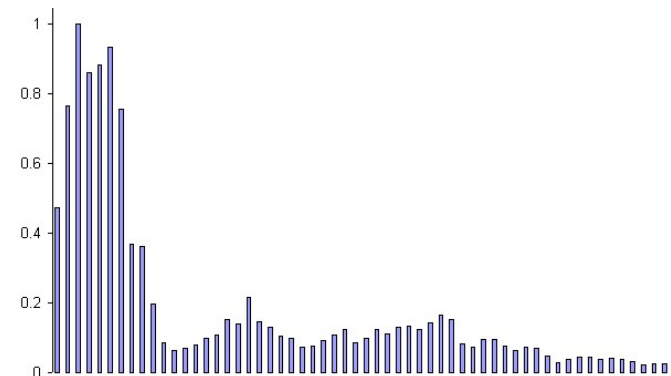
Darwin

- **Advantages**
 - It works!
 - Relatively fast
- **Downsides**
 - Requires a training set
 - Can be slow
 - Probably not biologically realistic
- **Alternatives to Backpropagation**
 - Hebbian learning
 - Not successful in feed-forward nets
 - Reinforcement learning
 - Only limited success
 - Artificial evolution
 - More general, but can be even slower than backprop

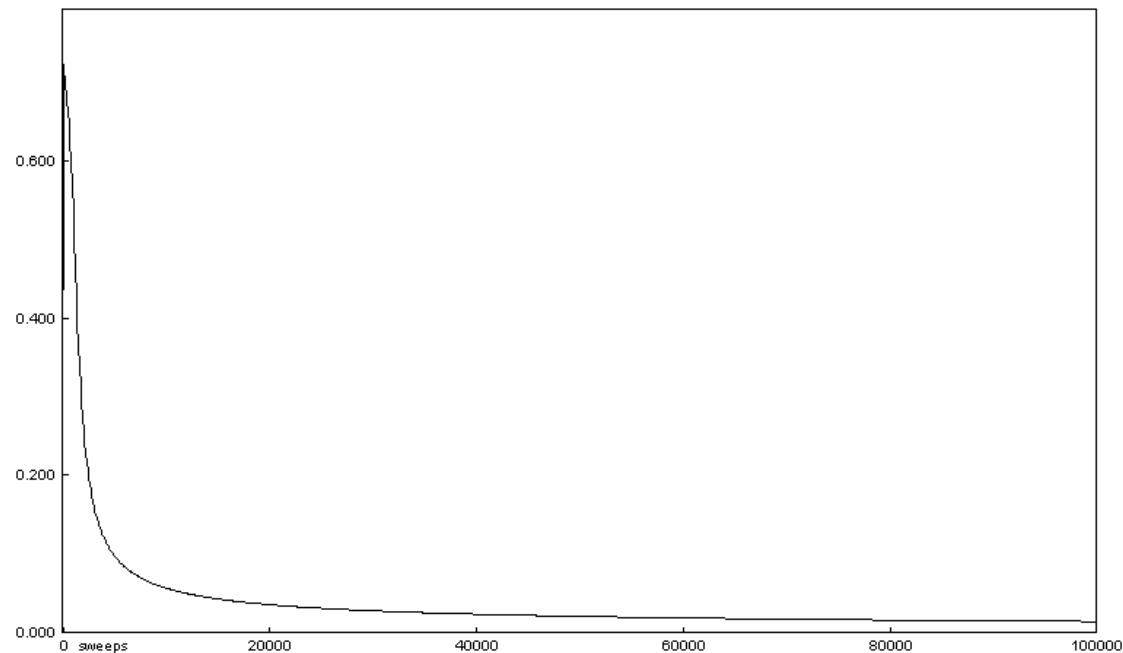


Example: Voice Recognition



- Task: Learn to discriminate between two different voices saying “Hello”
- Data
 - Sources
 - Steve Simpson 
 - David Raubenheimer 
 - Format
 - Frequency distribution (60 bins)
 - Analogy: cochlea

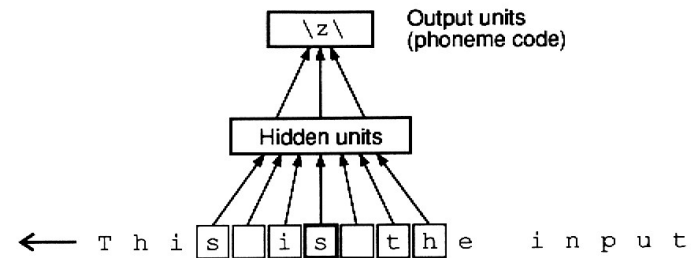


- Repeat process (sweep) for all training pairs
 - Present data
 - Calculate error
 - Backpropagate error
 - Adjust weights
- Repeat process multiple times



Applications of Feed-forward nets

- Pattern recognition
 - [Character recognition](#)
 - Face Recognition
- Sonar mine/rock recognition (Gorman & Sejnowksi, 1988)
- Navigation of a car (Pomerleau, 1989)
- Stock-market prediction
- Pronunciation (NETtalk)  
(Sejnowksi & Rosenberg, 1987)



Thank You!