

31YB: Lecture 4

Thu, 2nd Oct 2003

Lecture Overview

- Learning /Adaptation: Biological Analogy
- 2-term (Hebbian) & 3-term (supervised) learning rules
- Issues of: Locality, Inductive learning & generalization

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Learning /Adaptation & Biological Analogy

- What makes both real and artificial neural networks different is adaptation
 - that is, some aspect of the artificial neural network alters in response to
 - input
 - the result of the computation
 - some external force (or teacher)
- On the other hand, biological adaptation takes many forms
 - and runs over many simultaneous timescales
 - short-term: milliseconds, seconds
 - long term: many seconds to indefinite
 - and involves many different systems
 - alterations at synapses
 - hormones altering the whole neuron's behaviour
 - growth and decay
 - including growth of myelin sheaths (glial cells)

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Learning: Model neuron adaptation

- Generally is applied at the (model) synapse
- For simple model neurons this means altering the weight associated with each synapse
- There are many possible ways of doing this
 - exactly how the brain alters synaptic efficacies is not clear
 - although there is considerable evidence that it does alter synaptic efficacies
- We will consider 2-term and 3-term rules
- 2-term rules:
 - the alteration depends entirely on
 - the presynaptic neuron and the postsynaptic neuron
- 3-term rules
 - the alteration depends on
 - the presynaptic neuron and the postsynaptic neuron and some external (teacher) signal

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2-term (Hebbian) learning

- Hebbian learning rules are
 - Simple 2-term learning rules which depend on the output of the pre- and post-synaptic neurons alone
- Varieties
 - pure Hebbian
 - increase the weight when the pre-and post-synaptic neurons are co-active
 - variations on the above...
 - Stent-Singer rule
 - as pure Hebbian, and also decrease the weight when the pre-synaptic unit is inactive, but the post-synaptic unit is active
 - Pre-synaptic rule
 - as pure Hebbian, and also decrease the weight when the pre-synaptic unit is active, but the post-synaptic unit is inactive
 - There is biological evidence for forms of both of these rules
 - Hopfield rule
 - all three of the above conditions for altering the weight,
 - plus increase the weight when both units are inactive at the same time

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3-term learning rules

- Depend on the values of the pre-and post-synaptic neurons
- AND on some external signal
 - frequently this signal is the desired output
- Often the desired output is used to generate an error signal:
$$E = (D - Y)$$
 - where E is the error signal generated,
 - D is the desired output
 - Y is the actual output
- The learning rule aims to make E as small as possible
- More later on this ..

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Issue of Locality in Learning

- Locality of a learning rule means
 - all the information required to make the adaptation is local to the place where the adaptation occurs.
- Why locality?
 - Local adaptation rules are implementable
 - in the sense of being buildable
 - Non-local rules call for either
 - action at a distance
 - some mechanism for bringing the required information to the place where the adaptation occurs or is computed.
- Note that software implementations are not bound by locality!

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Concept of Inductive Learning

Deduction:

Given some general rules, apply them to specific examples to *deduce* outputs.

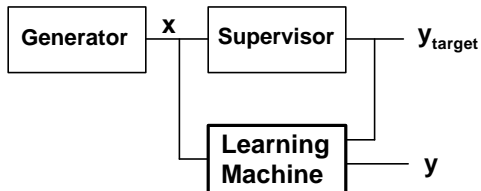
Induction:

Given specific examples, *induce* general 'rules' which can then be applied to novel examples.

Learning & Generalization

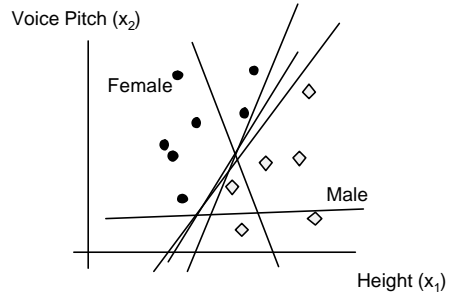
- Learn from a *representative* set of examples
- Extract the 'essence' or underlying form (e.g. in the shape/form of a decision boundary)
- Given a novel (previously unseen) example, produce the *expected* output.

3-term or Supervised Learning



Training: Learn from training pairs $(\mathbf{x}, \mathbf{y}_{\text{target}})$

Testing: Given any \mathbf{x} , output a value \mathbf{y} close to the supervisor's output $\mathbf{y}_{\text{target}}$



Learning Problem: What is a good decision boundary ?

Next Lecture: More on 3-term Learning