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# Neural Networks - Assignment 1

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## 1 OUTLINE CHAPTER I

- What is a neural network?
  - Motivated by human brain
  - complex, nonlinear, parallel
  - Benefits of neural networks
    - \* Nonlinearity
    - \* Input-output mapping
    - \* Adaptivity
    - \* Evidential Response
    - \* Contextual information
    - \* Fault tolerance
    - \* VLSI implementability
    - \* Uniformity of analysis and design
    - \* Neurobiological analogy
- Human brain
  - about 10 billion neurons
  - about 60 trillion synapses/connections
  - energetic efficient

- chemical synapses: electrical signal – > chemical signal – > electrical signal
- plasticity via new connections and modification of existing ones
- structural level of organization
- Models of a neuron
  - Neuron: Information-processing unit
  - Three basic elements:
    - \* Set of synapses
    - \* Adder
    - \* Activation function/squashing function
  - Types of activation function
    - \* Threshold function (McCulloch-Pitts model)
    - \* Piecewise-linear function
    - \* Sigmoid function
  - Stochastic model of a neuron
- Neural Networks viewed as directed graphs
  - Block diagram: functional description
  - Signal-flow graph: signal flow
    - \* Rule1: Signal flows in direction defined by arrow
    - \* Rule2: Node signal equals sum of entering signals
    - \* Rule3: Signal of node is transmitted to each outgoing link
  - Architectural graph: network layout
- Feedback
  - Feedback: Output influences input
  - Recurrent networks
  - closed-loop operator:  $\frac{A}{1-AB}$
  - open-loop operator:  $AB$
- Network architectures
  - Single-layer feedforward networks
    - \* No hidden layer
  - Multi-layer feedforward networks
    - \* Contains hidden layer
  - Recurrent Networks

- \* At least one feedback loop
- Knowledge representation
  - Training via input-output pairs
  - Learning: applying algorithm on a training set
  - Generalization: test learned model on test set
  - Rule1: Similar input from similar classes should be classified into the same category
  - Rule2: Items of different classes should be given different representations in the network.
  - Rule3: The more important a feature, the more neurons to represent it
  - Rule4: Prior information should be built into the network beforehand
  - How to build prior information into neural network design
    - \* No well-defined rules
    - \* receptive fields
    - \* weight-sharing
  - How to build invariances into neural network design
    - \* Invariance by structure
    - \* Invariance by training
    - \* Invariant feature space
- Artificial intelligence and neural networks
  - Representation
  - Reasoning
  - Learning
  - Level of explanation
  - Processing style
  - Representational structure
- Historical notes

## 2 FROM HAYKIN'S BOOK, CHAPTER 1 PROBLEMS - "MODELS OF A NEURON", SOLVE ANY 2 OUT OF 11 (1.1 TO 1.11).

### 2.1 EXERCISE 1.6

**Consider the pseudolinear activation function  $\phi(v)$  shown in figure P1.6.**

**(a) Formulate  $\phi(v)$  as a function of  $v$ .**

**(b) What happens to  $\phi(v)$  if  $\alpha$  is allowed to approach zero?**

(a)

$$\phi(v) = \begin{cases} 0 & \text{if } v < -0.5\alpha \\ b & \text{if } v > 0.5\alpha \\ \frac{b}{\alpha}v + 0.5b & \text{else} \end{cases}$$

(b)

The function will not be defined for  $\alpha = 0$ . The function becomes more and more similar to a step function with value 0 for  $v < 0$  and value  $b$  for  $v > 0$ .

## 2.2 EXERCISE 1.7

**Repeat Problem 1.6 for the pseudolinear activation function  $\phi(v)$  shown in Fig. P1.7.**

(a)

$$\phi(v) = \begin{cases} -b & \text{if } v < -\alpha \\ b & \text{if } v > \alpha \\ \frac{b}{\alpha}v & \text{else} \end{cases}$$

(b)

The function will not be defined for  $\alpha = 0$ . The function becomes more and more similar to a step function with value  $-b$  for  $v < 0$  and value  $b$  for  $v > 0$ .

## 3 FROM HAYKIN'S BOOK, CHAPTER 1 PROBLEMS - "NETWORK ARCHITECTURES", SOLVE ANY 2 OUT OF 7 (1.12 TO 1.19) INCLUDING 1.13.

### 3.1 EXERCISE 1.12

### 3.2 EXERCISE 1.16

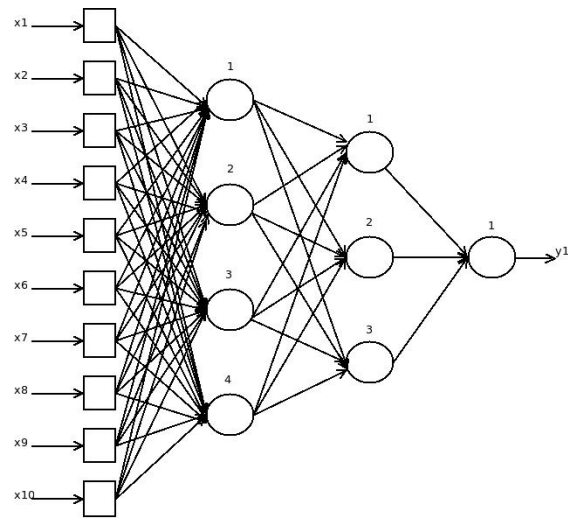


Figure 3.1: Fully recurrent network with five neurons, no self-feedback

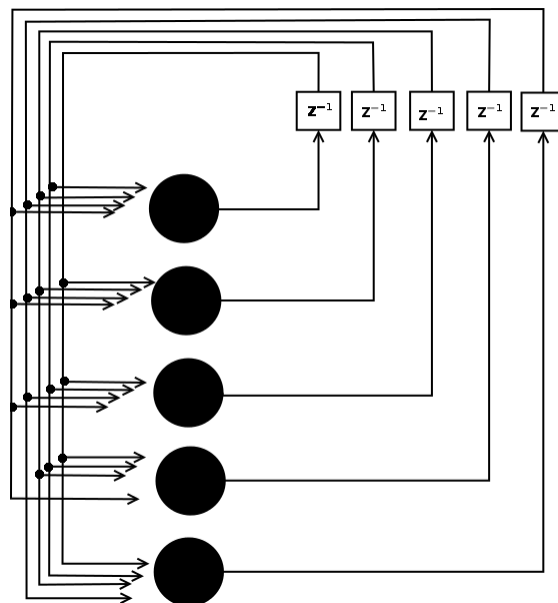


Figure 3.2: Fully connected 10-4-3-1 feedforward network