# Artificial Neural Networks

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# 1 Introduction

#### 1.1 Introduction

Artificial neural networks are one of the most popular and promising areas of artificial intelligence research. Artificial Neural Networks are abstract computational models, roughly based on the organizational structure of the human brain. There are a wide variety of network architectures and learning methods that can be combined to produce neural networks with different computational abilities.

### 1.2 What is This Book About?

This book is going to serve as a general-purpose overview of artificial neural networks, including network construction, use, and applications.

### 1.3 Who is This Book For?

This book is going to be aimed at advanced undergraduates and graduate students in the areas of computer science, mathematics, engineering, and the sciences.

# 1.4 What Are The Prerequisites?

Readers of this book are going to require a solid mathematical background that includes, but may not be limited to:

- Linear Algebra<sup>1</sup>
- Abstract Algebra<sup>2</sup>
- Calculus<sup>3</sup>

Students may also find some benefit in the following engineering texts:

- Signals and Systems<sup>4</sup>
- Engineering Analysis<sup>5</sup>

<sup>1</sup> http://en.wikibooks.org/wiki/Linear%20Algebra

<sup>2</sup> http://en.wikibooks.org/wiki/Abstract%20Algebra

<sup>3</sup> http://en.wikibooks.org/wiki/Calculus

<sup>4</sup> http://en.wikibooks.org/wiki/Signals%20and%20Systems

<sup>5</sup> http://en.wikibooks.org/wiki/Engineering%20Analysis

Students who wish to implement the lessons learned in this book should be familiar with at least one general-purpose programming language or have a background in:

- MATLAB Programming  $^6$
- Programmable  $Logic^7$

<sup>6</sup> http://en.wikibooks.org/wiki/MATLAB%20Programming

<sup>7</sup> http://en.wikibooks.org/wiki/Programmable%20Logic

# 2 Neural Network Basics

### 2.1 Artificial Neural Networks

Artificial Neural Networks, also known as "Artificial neural nets", "neural nets", or ANN for short, are a computational tool modeled on the interconnection of the neuron in the nervous systems of the human brain and that of other organisms. Biological Neural Nets (BNN) are the naturally occurring equivalent of the ANN. Both BNN and ANN are network systems constructed from atomic components known as "neurons". Artificial neural networks are very different from biological networks, although many of the concepts and characteristics of biological systems are faithfully reproduced in the artificial systems. Artificial neural nets are a type of non-linear processing system that is ideally suited for a wide range of tasks, especially tasks where there is no existing algorithm for task completion. ANN can be trained to solve certain problems using a teaching method and sample data. In this way, identically constructed ANN can be used to perform different tasks depending on the training received. With proper training, ANN are capable of generalization, the ability to recognize similarities among different input patterns, especially patterns that have been corrupted by noise.

#### 2.1.1 What Are Neural Nets?

The term "Neural Net" refers to both the biological and artificial variants, although typically the term is used to refer to artificial systems only. Mathematically, neural nets are nonlinear. Each layer represents a non-linear combination of non-linear functions from the previous layer. Each neuron is a multiple-input, multiple-output (MIMO) system that receives signals from the inputs, produces a resultant signal, and transmits that signal to all outputs. Practically, neurons in an ANN are arranged into layers. The first layer that interacts with the environment to receive input is known as the input layer. The final layer that interacts with the output to present the processed data is known as the output layer. Layers between the input and the output layer that do not have any interaction with the environment are known as hidden layers. Increasing the complexity of an ANN, and thus its computational capacity, requires the addition of more hidden layers, and more neurons per layer.

Biological neurons are connected in very complicated networks. Some regions of the human brain such as the cerebellum are composed of very regular patterns of neurons. Other regions of the brain, such as the cerebrum have less regular arrangements. A typical biological neural system has millions or billions of cells, each with thousands of interconnections with other neurons. Current artificial systems cannot achieve this level of complexity, and so cannot be used to reproduce the behavior of biological systems exactly.

## 2.2 Processing Elements

In an artificial neural network, neurons can take many forms and are typically referred to as **Processing Elements** (PE) to differentiate them from the biological equivalents. The PE are connected into a particular network pattern, with different patterns serving different functional purposes. Unlike biological neurons with chemical interconnections, the PE in artificial systems are electrical only, and may be either analog, digital, or a hybrid. However, to reproduce the effect of the synapse, the connections between PE are assigned multiplicative weights, which can be calibrated or "trained" to produce the proper system output.

#### 2.2.1 McCulloch-Pitts Model

Processing Elements are typically defined in terms of two equations that represent the McCulloch-Pitts model of a neuron:

[McCulloch-Pitts Model]

$$\zeta = \sum_{i} w_i x_i$$

$$y = \sigma(\zeta)$$

Where  $\zeta$  is the weighted sum of the inputs (the inner product of the input vector and the tap-weight vector), and  $\sigma(\zeta)$  is a function of the weighted sum. If we recognize that the weight and input elements form vectors  $\mathbf{w}$  and  $\mathbf{x}$ , the  $\zeta$  weighted sum becomes a simple dot product:

$$\zeta = \mathbf{w} \cdot \mathbf{x}$$

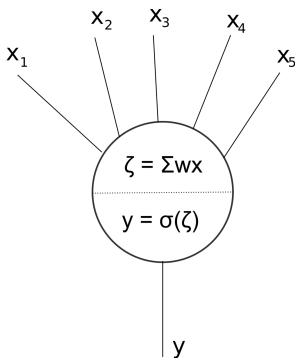


Figure 1

This may be called either the activation function (in the case of a threshold comparison) or a transfer function. The image to the right shows this relationship diagrammatically. The dotted line in the center of the neuron represents the division between the calculation of the input sum using the weight vector, and the calculation of the output value using the activation function. In an actual artificial neuron, this division may not be made explicitly.

The inputs to the network, x, come from an input space and the system outputs are part of the output space. For some networks, the output space Y may be as simple as  $\{0, 1\}$ , or it may be a complex multi-dimensional space. Neural networks tend to have one input per degree of freedom in the input space, and one output per degree of freedom in the output space.

The tap weight vector is updated during training by various algorithms. One of the more popular of which is the backpropagation algorithm which we will discuss in more detail later.

# 2.3 Why Use Neural Nets?

Artificial neural nets have a number of properties that make them an attractive alternative to traditional problem-solving techniques. The two main alternatives to using neural nets are to develop an algorithmic solution, and to use an expert system.

Algorithmic methods arise when there is sufficient information about the data and the underlying theory. By understanding the data and the theoretical relationship between the data, we can directly calculate unknown solutions from the problem space. Ordinary von

Neumann computers can be used to calculate these relationships quickly and efficiently from a numerical algorithm.

Expert systems, by contrast, are used in situations where there is insufficient data and theoretical background to create any kind of a reliable problem model. In these cases, the knowledge and rationale of human experts is codified into an expert system. Expert systems emulate the deduction processes of a human expert, by collecting information and traversing the solution space in a directed manner. Expert systems are typically able to perform very well in the absence of an accurate problem model and complete data. However, where sufficient data or an algorithmic solution is available, expert systems are a less than ideal choice.

Artificial neural nets are useful for situations where there is an abundance of data, but little underlying theory. The data, which typically arises through extensive experimentation may be non-linear, non-stationary, or chaotic, and so may not be easily modeled. Input-output spaces may be so complex that a reasonable traversal with an expert system is not a satisfactory option. Importantly, neural nets do not require any a priori assumptions about the problem space, not even information about statistical distribution. Though such assumptions are not required, it has been found that the addition of such a priori information as the statistical distribution of the input space can help to speed training. Many mathematical problem models tend to assume that data lies in a standard distribution pattern, such as Gaussian or Maxwell-Boltzmann distributions. Neural networks require no such assumption. During training, the neural network performs the necessary analytical work, which would require non-trivial effort on the part of the analyst if other methods were to be used.

## 2.4 Learning

Learning is a fundamental component to an intelligent system, although a precise definition of learning is hard to produce. In terms of an artificial neural network, learning typically happens during a specific training phase. Once the network has been trained, it enters a production phase where it produces results independently. Training can take on many different forms, using a combination of learning paradigms, learning rules, and learning algorithms. A system which has distinct learning and production phases is known as a static network. Networks which are able to continue learning during production use are known as dynamical systems.

A learning paradigm is supervised, unsupervised or a hybrid of the two, and reflects the method in which training data is presented to the neural network. A method that combines supervised and unsupervised training is known as a hybrid method. A learning rule is a model for the types of methods to be used to train the system, and also a goal for what types of results are to be produced. The learning algorithm is the specific mathematical method that is used to update the inter-neuronal synaptic weights during each training iteration. Under each learning rule, there are a variety of possible learning algorithms for use. Most algorithms can only be used with a single learning rule. Learning rules and learning algorithms can typically be used with either supervised or unsupervised learning paradigms, however, and each will produce a different effect.

Overtraining is a problem that arises when too many training examples are provided, and the system becomes incapable of useful generalization. This can also occur when there are too many neurons in the network and the capacity for computation exceeds the dimensionality of the input space. During training, care must be taken not to provide too many input examples and different numbers of training examples could produce very different results in the quality and robustness of the network.

### 2.5 Network Parameters

There are a number of different parameters that must be decided upon when designing a neural network. Among these parameters are the number of layers, the number of neurons per layer, the number of training iterations, et cetera. Some of the more important parameters in terms of training and network capacity are the number of hidden neurons, the learning rate and the momentum parameter.

#### 2.5.1 Number of neurons in the hidden layer

Hidden neurons are the neurons that are neither in the input layer nor the output layer. These neurons are essentially hidden from view, and their number and organization can typically be treated as a black box to people who are interfacing with the system. Using additional layers of hidden neurons enables greater processing power and system flexibility. This additional flexibility comes at the cost of additional complexity in the training algorithm. Having too many hidden neurons is analogous to a system of equations with more equations than there are free variables: the system is over specified, and is incapable of generalization. Having too few hidden neurons, conversely, can prevent the system from properly fitting the input data, and reduces the robustness of the system.

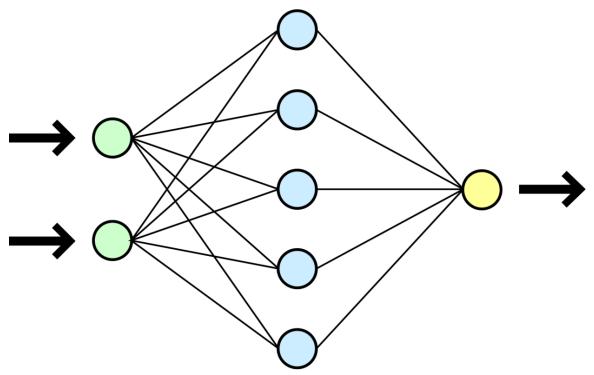


Figure 2

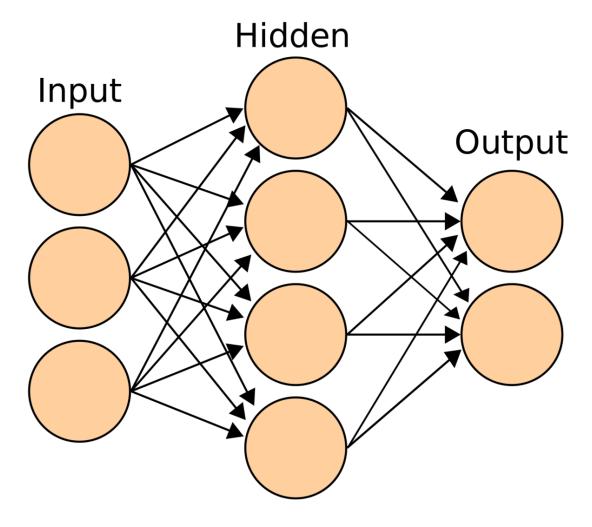


Figure 3

Data type: Integer Domain: [1, oo] Typical value: 8

Meaning: Number of neurons in the hidden layer (additional layer to the input and output layers, not connected externally).

### 2.5.2 Learning Rate

Data type: Real Domain: [0, 1] Typical value: 0.3

Meaning: Learning Rate. Training parameter that controls the size of weight and bias changes during learning.

### 2.5.3 Momentum

Data type: Real Domain: [0, 1] Typical value: 0.05

Meaning: Momentum simply adds a fraction m of the previous weight update to the current one. The momentum parameter is used to prevent the system from converging to a local minimum or saddle point. A high momentum parameter can also help to increase the speed of convergence of the system. However, setting the momentum parameter too high can create a risk of overshooting the minimum, which can cause the system to become unstable. A momentum coefficient that is too low cannot reliably avoid local minima, and can also slow down the training of the system.

### 2.5.4 Training type

Data type: Integer Domain: [0, 1] Typical value: 1

Meaning: 0 = train by epoch, 1 = train by minimum error

### 2.5.5 Epoch

Data type: Integer Domain: [1, oo] Typical value: 5000000

Meaning: Determines when training will stop once the number of iterations exceeds epochs. When training by minimum error, this represents the maximum number of iterations.

#### 2.5.6 Minimum Error

Data type: Real Domain: [0, 0.5] Typical value: 0.01

Meaning: Minimum mean square error of the epoch. Square root of the sum of squared differences between the network targets and actual outputs divided by number of patterns (only for training by minimum error).

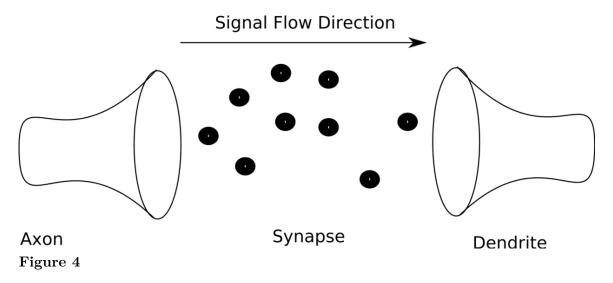
# 3 Biological Neural Networks

## 3.1 Biological Neural Nets

In the case of a biological neural net, neurons are living cells with axons and dendrites that form interconnections through electro-chemical synapses. Signals are transmitted through the cell body (soma), from the dendrite to the axon as an electrical impulse. In the pre-synaptic membrane of the axon, the electrical signal is converted into a chemical signal in the form of various neurotransmitters. These neurotransmitters, along with other chemicals present in the synapse form the message that is received by the post-synaptic membrane of the dendrite of the next cell, which in turn is converted to an electrical signal.

This page is going to provide a brief overview of biological neural networks, but the reader will have to find a better source for a more in-depth coverage of the subject.

### 3.1.1 Synapses



The figure above shows a model of the synapse showing the chemical messages of the synapse moving from the axon to the dendrite. Synapses are not simply a transmission medium for chemical signals, however. A synapse is capable of modifying itself based on the signal traffic that it receives. In this way, a synapse is able to "learn" from its past activity. This learning happens through the strengthening or weakening of the connection. External factors can also affect the chemical properties of the synapse, including body chemistry and medication.

### 3.2 Neurons

Cells have multiple dendrites, each receives a weighted input. Inputs are weighted by the strength of the synapse that the signal travels through. The total input to the cell is the sum of all such synaptic weighted inputs. Neurons utilize a threshold mechanism, so that signals below a certain threshold are ignored, but signals above the threshold cause the neuron to fire. Neurons follow an "all or nothing" firing scheme, and are similar in this respect to a digital component. Once a neuron has fired, a certain refraction period must pass before it can fire again.

## 3.3 Biological Networks

Biological neural systems are heterogeneous, in that there are many different types of cells with different characteristics. Biological systems are also characterized by macroscopic order, but nearly random interconnection on the microscopic layer. The random interconnection at the cellular level is rendered into a computational tool by the learning process of the synapse, and the formation of new synapses between nearby neurons.

# 4 History

## 4.1 Early History

The history of neural networking arguably started in the late 1800s with scientific attempts to study the workings of the human brain. In 1890, William James published the first work about brain activity patterns. In 1943, McCulloch and Pitts produced a model of the neuron that is still used today in artificial neural networking. This model is broken into two parts: a summation over weighted inputs and an output function of the sum.

## 4.2 Artificial Neural Networking

In 1949, Donald Hebb published *The Organization of Behavior*, which outlined a law for synaptic neuron learning. This law, later known as Hebbian Learning in honor of Donald Hebb is one of the simplest and most straight-forward learning rules for artificial neural networks.

In 1951, Marvin Minsky created the first ANN while working at Princeton.

In 1958 The Computer and the Brain was published posthumously, a year after John von Neumann's death. In that book, von Neumann proposed many radical changes to the way in which researchers had been modeling the brain.

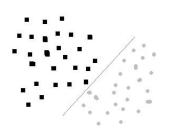
### 4.2.1 Mark I Perceptron

The Mark I Perceptron was also created in 1958, at Cornell University by Frank Rosenblatt. The Perceptron was an attempt to use neural network techniques for character recognition. The Mark I Perceptron was a linear system, and was useful for solving problems where the input classes were linearly separable in the input space. In 1960, Rosenblatt published the book *Principles of Neurodynamics*, containing much of his research and ideas about modeling the brain.

The Perceptron was a linear system with a simple input-output relationship defined as a McCulloch-Pitts neuron with a step activation function. In this model, the weighted inputs were compared to a threshold  $\theta$ . The output, y, was defined as a simple step function:

$$y = \begin{cases} 1 & \text{if } \zeta \ge \theta \\ 0 & \text{if } \zeta < \theta \end{cases}$$

Despite the early success of the Perceptron and artificial neural network research, there were many people who felt that there was limited promise in these techniques. Among these were Marvin Minsky and Seymore Papert, whose 1969 book *Perceptrons* was used to discredit ANN research and focus attention on the apparent limitations of ANN work. One of the limitations that Minsky and Papert pointed out most clearly was the fact that the Perceptron was not able to classify patterns that are not linearly separable in the input space. Below, the figure on the left shows an input space with a linearly separable classification problem. The figure on the right, in contrast, shows an input space where the classifications are not linearly separable.



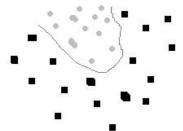


Figure 5

Figure 6

Despite the failure of the Mark I Perceptron to handle non-linearly separable data, it was not an inherent failure of the technology, but a matter of scale. The Mark I was a two layer Perceptron, Hecht-Nielsen showed in 1990 that a three layer machine (multi layer Perceptron, or MLP) was capable of solving nonlinear separation problems. *Perceptrons* ushered in what some call the "quiet years", where ANN research was at a minimum of interest. It wasn't until the rediscovery of the backpropagation algorithm in 1986 that the field gained widespread interest again.

#### 4.2.2 Backpropagation and Rebirth

The backpropagation algorithm, originally discovered by Werbos in 1974 was rediscovered in 1986 with the book *Learning Internal Representation by Error Propagation* by Rumelhart, Hinton and Williams. Backpropagation is a form of the gradient descent algorithm used with artificial neural networks for minimization and curve-fitting.

In 1987 the IEEE annual international ANN conference was started for ANN researchers. In 1987 the International Neural Network Society (INNS) was formed, along with the INNS Neural Networking journal in 1988.

# 5 MATLAB Neural Networking Toolbox

### 5.1 MATLAB

MATLAB ® is an ideal tool for working with artificial neural networks for a number of reasons. First, MATLAB is highly efficient in performing vector and matrix calculations. Second, MATLAB comes with a specialized Neural Network Toolbox ® which contains a number of useful tools for working with artificial neural networks.

This book is going to utilize the MATLAB programming environment and the Neural Network Toolbox to do examples and problems throughout the book.

### 5.2 Other Neural Network Software

Even though this book is going to focus on MATLAB for its problems and examples, there are a number of other tools that can be used for constructing, testing, and implementing neural networks.

# 6 Activation Functions

### 6.1 Activation Functions

There are a number of common activation functions in use with neural networks. This is not an exhaustive list.

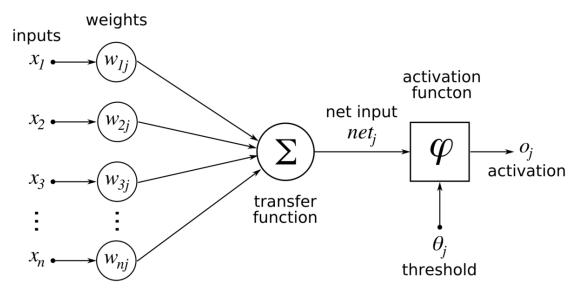


Figure 7

# 6.2 Step Function

A step function is a function like that used by the original Perceptron. The output is a certain value,  $A_1$ , if the input sum is above a certain threshold and  $A_0$  if the input sum is below a certain threshold. The values used by the Perceptron were  $A_1 = 1$  and  $A_0 = 0$ .

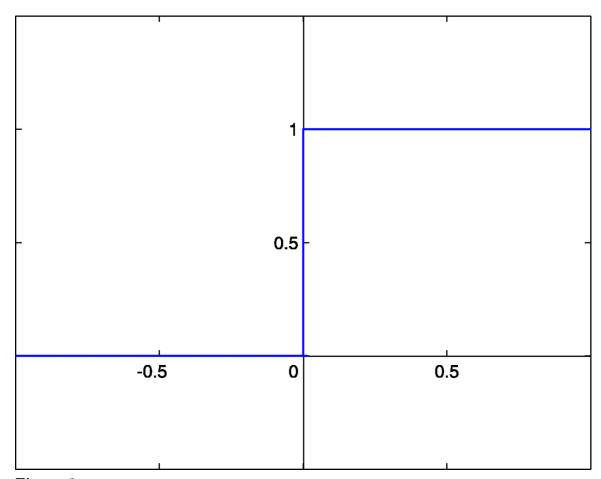


Figure 8

These kinds of step activation functions are useful for binary classification schemes. In other words, when we want to classify an input pattern into one of two groups, we can use a binary classifier with a step activation function. Another use for this would be to create a set of small **feature identifiers**. Each identifier would be a small network that would output a 1 if a particular input feature is present, and a 0 otherwise. Combining multiple feature detectors into a single network would allow a very complicated clustering or classification problem to be solved.

### 6.3 Linear combination

A linear combination is where the weighted sum input of the neuron plus a linearly dependant bias becomes the system output. Specifically:

$$y = \zeta + b$$

In these cases, the sign of the output is considered to be equivalent to the 1 or 0 of the step function systems, which enables the two methods be to equivalent if

$$\theta = -b$$

## 6.4 Continuous Log-Sigmoid Function

A log-sigmoid function, also known as a logistic function, is given by the relationship:

$$\sigma(t) = \frac{1}{1 + e^{-\beta t}}$$

Where  $\beta$  is a slope parameter. This is called the log-sigmoid because a sigmoid can also be constructed using the hyperbolic tangent function instead of this relation, in which case it would be called a tan-sigmoid. Here, we will refer to the log-sigmoid as simply "sigmoid". The sigmoid has the property of being similar to the step function, but with the addition of a region of uncertainty. Sigmoid functions in this respect are very similar to the input-output relationships of biological neurons, although not exactly the same. Below is the graph of a sigmoid function.

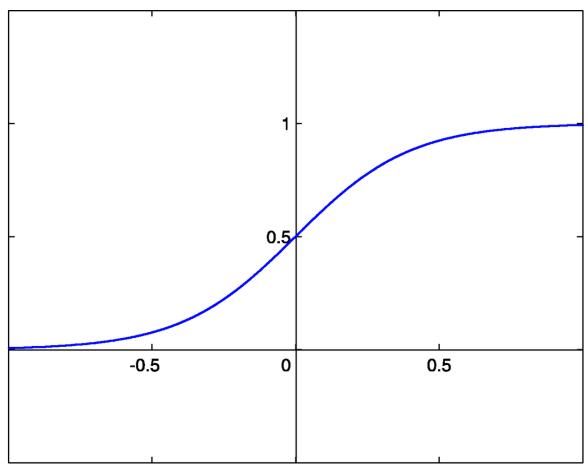


Figure 9

Sigmoid functions are also prized because their derivatives are easy to calculate, which is helpful for calculating the weight updates in certain training algorithms. The derivative when  $\beta = 1$  is given by:

$$\frac{d\sigma(t)}{dt} = \sigma(t)[1 - \sigma(t)]$$

When  $\beta \neq 1$ , using  $\sigma(\beta, t) = \frac{1}{1 + e^{-\beta t}}$ , the derivative is given by:

$$\frac{d\sigma(\beta,t)}{dt} = \beta[\sigma(\beta,t)[1-\sigma(\beta,t)]]$$

## 6.5 Continuous Tan-Sigmoid Function

$$\sigma(t) = tanh(t) = \frac{e^t - e^{-t}}{e^t + e^{-t}}$$

Its derivative is:

$$\frac{d\sigma(t)}{dt} = 1 - \tanh^2(t) = \operatorname{sech}^2(t) = 1 - \frac{(e^t - e^{-t})^2}{(e^t + e^{-t})^2}$$

### 6.6 Softmax Function

The softmax activation function is useful predominantly in the output layer of a clustering system. Softmax functions convert a raw value into a posterior probability. This provides a measure of certainty. The softmax activation function is given as:

$$y_i = \frac{e^{\zeta_i}}{\sum_{j \in L} e^{\zeta_j}}$$

L is the set of neurons in the output layer.

# 7 Feed-Forward Networks

# 7.1 Feedforward Systems

Feed-forward neural networks are the simplest form of ANN. Shown below, a feed-forward neural net contains only forward paths.

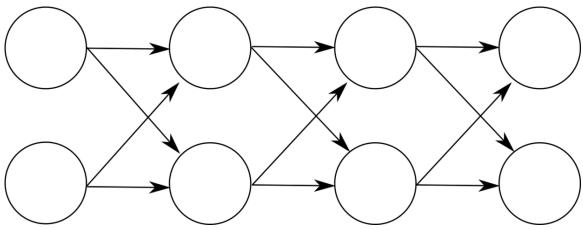


Figure 10

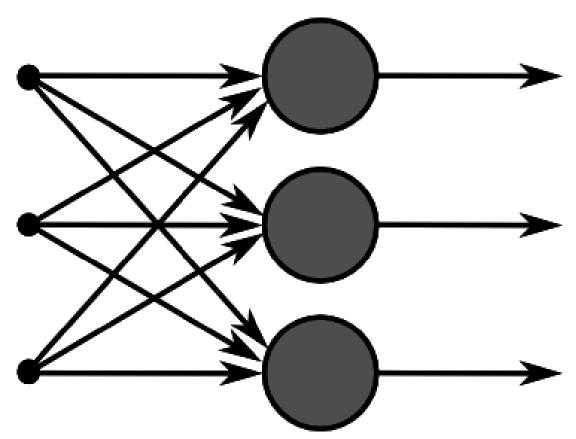


Figure 11

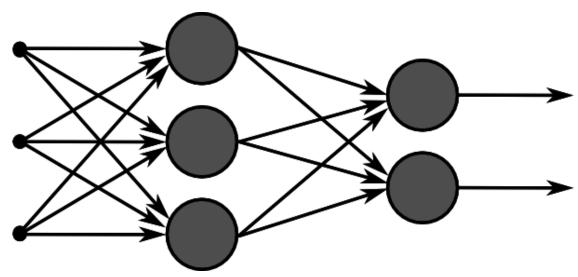


Figure 12

## 7.2 Connection Weights

In a feed-forward system PE are arranged into distinct layers with each layer receiving input from the previous layer and outputting to the next layer. There is no feedback. This means that signals from one layer are not transmitted to a previous layer. This can be stated mathematically as:

$$w_{ij} = 0$$
 if  $i = j$ 

$$w_{ij} = 0$$
 if  $layer(i) \le layer(j)$ 

Weights of direct feedback paths, from a neuron to itself, are zero. Weights from a neuron to a neuron in a previous layer are also zero. Notice that weights for the forward paths may also be zero depending on the specific network architecture, but they do not need to be. A network without all possible forward paths is known as a sparsely connected network, or a non-fully connected network. The percentage of available connections that are utilized is known as the connectivity of the network.

## 7.3 Mathematical Relationships

The weights from each neuron in layer l - 1 to the neurons in layer l are arranged into a matrix  $\mathbf{w}_l$ . Each row corresponds to a neuron in l - 1, and each column corresponds to a neuron in l. The input signal from l - 1 to l is the vector  $\mathbf{x}_l$ . If  $\rho_l$  is a vector of activation functions  $[\sigma 1 \ \sigma 2 \ \dots \ \sigma n]$  that acts on each row of input and  $\mathbf{b}_l$  is an arbitrary offset vector (for generalization) then the total output of layer l is given as:

$$\mathbf{y}_l = \rho_l(\mathbf{w}_l \mathbf{x}_l + \mathbf{b}_l)$$

Two layers of output can be calculated by substituting the output from the first layer into the input of the second layer:

$$\mathbf{y}_l = \rho_l(\mathbf{w}_l \rho_{l-1}(\mathbf{w}_{l-1} \mathbf{x}_{l-1} + \mathbf{b}_{l-1}) + \mathbf{b}_l)$$

This method can be continued to calculate the output of a network with an arbitrary number of layers. Notice that as the number of layers increases, so does the complexity of this calculation. Sufficiently large neural networks can quickly become too complex for direct mathematical analysis.

# 8 Radial Basis Function Networks

In a radial basis function (RBF) networks are neural nets with three layers. The first input layer feeds data to a hidden intermediate layer. The hidden layer processes the data and transports it to the output layer. Only the tap weights between the hidden layer and the output layer are modified during training. Each hidden layer neuron represents a basis function of the output space, with respect to a particular center in the input space. The activation function chosen is commonly a Gaussian kernel:

$$\sigma(\zeta) = e^{-\beta\zeta^2}$$

This kernel is centered at the point in the input space specified by the weight vector. The closer the input signal is to the current weight vector, the higher the output of the neuron will be. Radial basis function networks are used commonly in function approximation and series prediction.

# 9 Recurrent Networks

#### 9.1 Recurrent Networks

In a recurrent network, the weight matrix for each layer l contains input weights from all other neurons in the network, not just neurons from the previous layer. The additional complexity from these feedback paths can have a number of advantages and disadvantages in the network.

## 9.2 Simple Recurrent Networks

Recurrent networks, in contrast to feed-forward networks, do have feedback elements that enable signals from one layer to be fed back to a previous layer. A basic recurrent network is shown in figure 6. A simple recurrent network is one with three layers, an input, an output, and a hidden layer. A set of additional context units are added to the input layer that receive input from the hidden layer neurons. The feedback paths from the hidden layer to the context units have a fixed weight of unity.

A fully recurrent network is one where every neuron receives input from all other neurons in the system. Such networks cannot be easily arranged into layers. A small subset of neurons receives external input, and another small subset produce system output. A recurrent network is known as symmetrical network if:

$$w_{ij} = w_{ji} \forall i, j$$

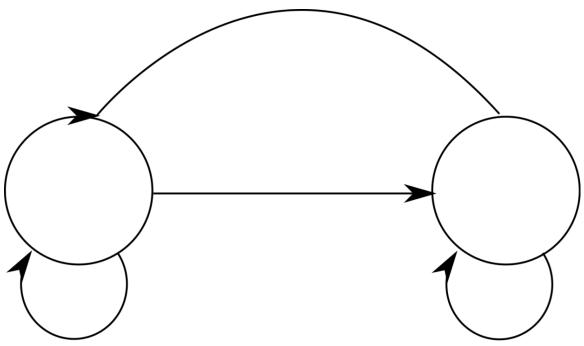


Figure 13

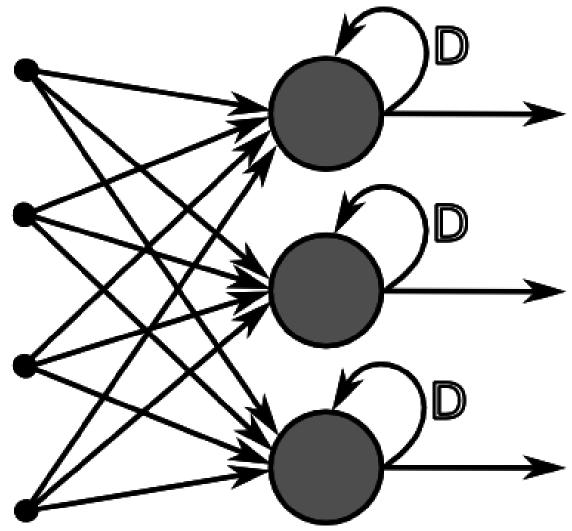


Figure 14

## 10 Echo State Networks

Echo state networks are recurrent networks where the hidden layer neurons are not completely connected to all input neurons. Such networks where all possible connections are not made are known as sparsely connected networks. Only the weights from the hidden layer to the output layer may be altered during training.

Echo state networks are useful for matching and reproducing specific input patterns. Because the only tap weights modified during training are the output layer tap weights, training is typically quick and computationally efficient in comparison to other multi-layer networks that are not sparsely connected.

# 11 Hopfield Networks

### 11.1 Hopfield Networks

Hopfield networks are one of the oldest and simplest networks. Hopfield networks utilize a network energy function. The activation function of a binary Hopfield network is given by the signum function of a biased weighted sum:

$$y_i = (\zeta_i - \theta_i)$$

Hopfield networks are frequently binary-valued, although continuous variants do exist. Binary networks are useful for classification and clustering purposes.

### 11.2 Energy Function

The energy function for the network is given as:

$$E = -\frac{1}{2} \sum_{i} \sum_{j} w_{ij} y_i y_j$$

Here, the y parameters are the outputs of the ith and jth units. During training the network energy should decrease until it reaches a minimum. This minimum is known as the attractor of the network. As a Hopfield network progresses, the energy minimizes itself. This means that mathematical minimization or optimization problems can be solved automatically by the Hopfield network if that problem can be formulated in terms of the network energy.

### 11.3 Associative Memory

Hopfield networks can be used as an associative memory network for data storage purposes. Each attractor represents a different data value that is stored in the network, and a range of associated patterns can be used to retrieve the data pattern. The number of distinct patterns p that can be stored in such a network is given approximately as:

$$p_{max} \approx 0.15n$$

Where n is the number of neurons in the network.

## 12 Self-Organizing Maps

### 12.1 Self-Organizing Maps

**Self-organizing maps** (SOM), sometimes called Kohonen SOM after their creator<sup>1</sup>, are used with unsupervised learning. SOM are modeled on biological neural networks, where groups of neurons appear to self organize into specific regions with common functionality.

### 12.2 Neuron Regions

Different regions of the SOM network are trained to be detectors for distinct features from the input set. Initial network weights are either set randomly, or are based off the eigenvectors of the input space. The Euclidean distance from each input sample to the weight vector of each neuron is computed, and the neuron whose weight vector is most similar to the input is declared the best match unit (BMU). The update formula is given as:

$$w_{j}[n+1] = w_{j}[n] + \Theta[j,n]\alpha[n](x[n] - w_{j}[n])$$

Here, w is the weight vector at time n.  $\alpha$  is a monotonically decreasing function that ensures the learning rate will decrease over time. x is the input vector, and  $\Theta[j, n]$  is a measure of the distance between the BMU and neuron j at iteration n. As can be seen from this algorithm, the amount by which the neuron weight vectors change is based on the distance from the BMU, and the amount of time. Decreasing the potential for change over time helps to reduce volatility during training, and helps to ensure that the network converges.

<sup>1</sup> http://en.wikipedia.org/wiki/Teuvo%20Kohonen

# 13 Competitive Models

## 13.1 Competitive Networks

Competitive networks are networks where neurons compete with one another. The weight vector is treated as a "prototype", and is matched against the input vector. The "winner" of each training session is the neuron whose weight vector is most similar to the input vector.

An example of a competitive network is shown below.

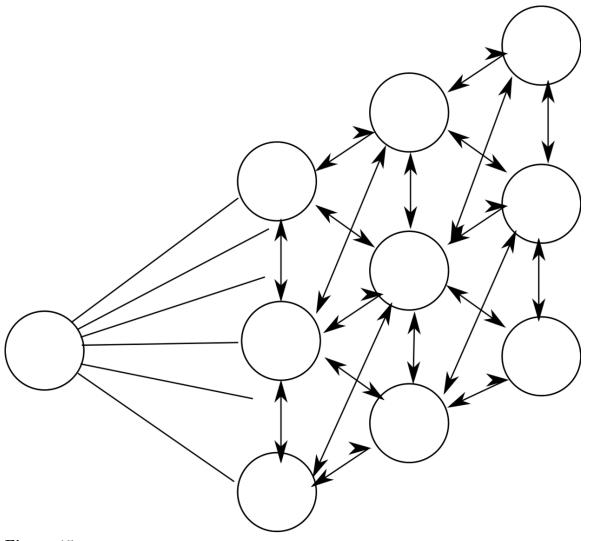


Figure 15

## 14 ART Models

In adaptive resonance theory (ART) networks, an overabundance of neurons leads some neurons to be committed (active) and others to be uncommitted (inactive). The weight vector, also known as the prototype, is said to resonate with the input vector if the two are sufficiently similar. Weights are only updated if they are resonating in the current iteration. ART networks commit an uncommitted neuron when a new input pattern is detected that does not resonate with any of the existing committed neurons. ART networks are fully-connected networks, in that all possible connections are made between all nodes.

## 15 Boltzmann Machines

Boltzmann learning compares the input data distribution P with the output data distribution of the machine, Q [24]. The distance between these distributions is given by the Kullback-Leibler distance:

$$w_{ij}[n+1] = w_{ij}[n] - \frac{\partial G}{\partial w_{ij}}$$

Where:

$$\frac{\partial G}{\partial w_{ij}} = -\frac{1}{T}[p_{ij} - q_{ij}]$$

Here, pij is the probability that elements i and j will both be on when the system is in its training phase (positive phase), and qij is the probability that both elements i and j will be on during the production phase (negative phase). The probability that element j will be on, pi, is given by:

$$p_i = \frac{1}{1 + e^{\frac{-\Delta E_i}{T}}}$$

T is a scalar constant known as the temperature of the system. Boltzmann learning is very powerful, but the complexity of the algorithm increases exponentially as more neurons are added to the network. To reduce this effect, a restricted Boltzmann machine (RBM) can be used. The hidden nodes in an RBM are not interconnected as they are in regular Boltzmann networks. Once trained on a particular feature set, these RBM can be combined together into larger, more diverse machines.

Because Boltzmann machine weight updates only require looking at the expected distributions of surrounding neurons, it is a plausible model for how actual biological neural networks learn.

## 16 Committee of Machines

Artificial neural networks can have very different properties depending on how they are constructed and how they are trained. Even in the case where two networks are trained on the same set of input data, different training algorithms can produce different systems with different characteristics. By combining multiple ANN into a single system, a committee of machines is formed. The result of a committee system is a combination of the results of the various component systems. For instance, the most common answer among a discrete set of answers in the committee can be taken as the overall answer, or the average answer can be taken.

Committee of machines (COM) systems tend to be more robust then the individual component systems, but they can also lose some of the "expertise" of the individual systems when answers are averaged out.

## 17 Learning Paradigms

### 17.1 Learning Paradigms

There are three different learning paradigms that can be used to train a neural network. Supervised and unsupervised learning are the most common, with hybrid approaches between the two becoming increasingly common as well.

### 17.2 Supervised Learning

Supervised learning is a technique where the input and expected output of the system are provided, and the ANN is used to model the relationship between the two. Given an input set x, and a corresponding output set y, an optimal rule is to be determined such that:

$$y = f(x) + e$$

Here, e is an approximation error that needs to be minimized. The input values are provided to the network which produces a result. This result is compared to the desired result, and this error signal is used to update the network weight vectors. Supervised learning is useful when we want the network to reproduce the characteristics of a certain relationship

### 17.3 Unsupervised Learning

In unsupervised learning, the data and a cost function are provided that is a function of the system input and output. The ANN is trained to minimize the cost function by finding a suitable input-output relationship.

Given an input set x, and a cost function g(x, y) of the input and output sets, the goal is to minimize the cost function through a proper selection of f (the relationship between x, and y). At each training iteration, the trainer provides the input to the network, and the network produces a result. This result is put into the cost function, and the total cost is used to update the weights. Weights are continually updated until the system output produces a minimal cost. Unsupervised learning is useful in situations where a cost function is known, but a data set is not know that minimizes that cost function over a particular input space.

# 18 Error-Correction Learning

### 18.1 Error-Correction Learning

Error-Correction Learning, used with supervised learning, is the technique of comparing the system output to the desired output value, and using that error to direct the training. In the most direct route, the error values can be used to directly adjust the tap weights, using an algorithm such as the backpropagation algorithm. If the system output is y, and the desired system output is known to be d, the error signal can be defined as:

$$e = d - y$$

Error correction learning algorithms attempt to minimize this error signal at each training iteration. The most popular learning algorithm for use with error-correction learning is the backpropagation algorithm, discussed below.

#### 18.2 Gradient Descent

The gradient descent algorithm is not specifically an ANN learning algorithm. It has a large variety of uses in various fields of science, engineering, and mathematics. However, we need to discuss the gradient descent algorithm in order to fully understand the backpropagation algorithm. The gradient descent algorithm is used to minimize an error function g(y), through the manipulation of a weight vector w. The cost function should be a linear combination of the weight vector and an input vector x. The algorithm is:

$$w_{ij}[n+1] = w_{ij}[n] + \eta g(w_{ij}[n])$$

Here,  $\eta$  is known as the step-size parameter, and affects the rate of convergence of the algorithm. If the step size is too small, the algorithm will take a long time to converge. If the step size is too large the algorithm might oscillate or diverge.

The gradient descent algorithm works by taking the gradient of the weight space to find the path of steepest descent. By following the path of steepest descent at each iteration, we will either find a minimum, or the algorithm could diverge if the weight space is infinitely decreasing. When a minimum is found, there is no guarantee that it is a global minimum, however.

### 18.3 Backpropagation

The **backpropagation** algorithm, in combination with a supervised error-correction learning rule, is one of the most popular and robust tools in the training of artificial neural networks. Back propagation passes error signals backwards through the network during training to update the weights of the network. Because of this dependence on bidirectional data flow during training, backpropagation is not a plausible reproduction of biological learning mechanisms. When talking about backpropagation, it is useful to define the term **interlayer** to be a layer of neurons, and the corresponding input tap weights to that layer. We use a superscript to denote a specific interlayer, and a subscript to denote the specific neuron from within that layer. For instance:

$$\zeta_{j}^{l} = \sum_{i=1}^{N^{l-1}} w_{ij}^{l} x_{i}^{l-1}$$
 (1) 
$$x_{i}^{l} = \sigma(\zeta_{i}^{l})$$
 (2)

Where  $x_i^{l-1}$  are the outputs from the previous interlayer (the inputs to the current interlayer),  $w_{ij}^l$  is the tap weight from the *i* input from the previous interlayer to the *j* element of the current interlayer.  $N^{l-1}$  is the total number of neurons in the previous interlayer.

The backpropagation algorithm specifies that the tap weights of the network are updated iteratively during training to approach the minimum of the error function. This is done through the following equation:

$$\begin{split} w_{ij}^{l}[n] &= w_{ij}^{l}[n-1] + \delta w_{ij}^{l}[n] \ (3) \\ w_{ij}^{l-1}[n] &= \eta \delta_{j}^{l} x_{i}^{l-1}[n] + \mu \Delta w_{ij}^{l}[n-1] \ (3) \end{split}$$

The relationship between this algorithm and the gradient descent algorithm should be immediately apparent. Here,  $\eta$  is known as the learning rate, not the step-size, because it affects the speed at which the system learns (converges). The parameter  $\mu$  is known as the momentum parameter. The momentum parameter forces the search to take into account its movement from the previous iteration. By doing so, the system will tend to avoid local minima or saddle points, and approach the global minimum. We will discuss these terms in greater detail in the next section.

The parameter  $\delta$  is what makes this algorithm a "back propagation" algorithm. We calculate it as follows:

$$\delta_{j}^{l} = \frac{dx_{j}^{l}}{dt} \sum_{k=1}^{r} \delta_{k}^{l+1} w_{kj}^{l+1}$$
 (4)

The  $\delta$  function for each layer depends on the  $\delta$  from the previous layer. For the special case of the output layer (the highest layer), we use this equation instead:

$$\delta_j^l = \frac{dx_j^l}{dt}(x_j^l - y_j) \ (5)$$

In this way, the signals propagate backwards through the system from the output layer to the input layer. This is why the algorithm is called the backpropagation algorithm.

#### 18.3.1 Log-Sigmoid Backpropagation

If we use log-sigmoid activation functions for our neurons, the derivatives simplify, and our backpropagation algorithm becomes:

$$\delta_j^l = x_j^l (1 - x_j^l) (x_j^l - y_j)$$

For the output layer, and

$$\delta_{j}^{l} = x_{j}^{l}(1 - x_{j}^{l}) \sum_{k=1}^{r} \delta_{k}^{l+1} w_{kj}^{l+1}$$

for all the hidden inner layers. This property makes the sigmoid function desirable for systems with a limited ability to calculate derivatives.

#### 18.3.2 Learning Rate

The **learning rate** is a common parameter in many of the learning algorithms, and affects the speed at which the ANN arrives at the minimum solution. In backpropagation, the learning rate is analogous to the step-size parameter from the gradient-descent algorithm. If the step-size is too high, the system will either oscillate about the true solution, or it will diverge completely. If the step-size is too low, the system will take a long time to converge on the final solution.

#### 18.3.3 Momentum Parameter

The **momentum parameter** is used to prevent the system from converging to a local minimum or saddle point. A high momentum parameter can also help to increase the speed of convergence of the system. However, setting the momentum parameter too high can create a risk of overshooting the minimum, which can cause the system to become unstable. A momentum coefficient that is too low cannot reliably avoid local minima, and also can slow the training of the system.

## 19 Hebbian Learning

### 19.1 Hebbian Learning

Hebbian learning is one of the oldest learning algorithms, and is based in large part on the dynamics of biological systems. A synapse between two neurons is strengthened when the neurons on either side of the synapse (input and output) have highly correlated outputs. In essence, when an input neuron fires, if it frequently leads to the firing of the output neuron, the synapse is strengthened. Following the analogy to an artificial system, the tap weight is increased with high correlation between two sequential neurons.

#### 19.2 Mathematical Formulation

Mathematically, we can describe Hebbian learning as:

$$w_{ij}[n+1] = w_{ij}[n] + \eta x_i[n]x_j[n]$$

Here,  $\eta$  is a learning rate coefficient, and x are the outputs of the ith and jth elements.

### 19.3 Plausibility

The Hebbian learning algorithm is performed locally, and doesn't take into account the overall system input-output characteristic. This makes it a plausible theory for biological learning methods, and also makes Hebbian learning processes ideal in VLSI hardware implementations where local signals are easier to obtain.

## 20 Competitive Learning

### 20.1 Competitive Learning

Competitive learning is a rule based on the idea that only one neuron from a given iteration in a given layer will fire at a time. Weights are adjusted such that only one neuron in a layer, for instance the output layer, fires. Competitive learning is useful for classification of input patterns into a discrete set of output classes. The "winner" of each iteration, element  $i^*$ , is the element whose total weighted input is the largest. Using this notation, one example of a competitive learning rule can be defined mathematically as:

$$w_{ij}[n+1] = w_{ij}[n] + \Delta w_{ij}[n]$$

$$\Delta w_{ij} = \begin{cases} \eta(x_i - w_{ij}) & \text{if } i = j\\ 0 & \text{otherwise} \end{cases}$$

### 20.2 Linear Vector Quantization

In a learning vector quantization (LVQ) machine, the input values are compared to the weight vector of each neuron. Neurons who most closely match the input are known as the best match unit (BMU) of the system. The weight vector of the BMU and those of nearby neurons are adjusted to be closer to the input vector by a certain step size. Neurons become trained to be individual feature detectors, and a combination of feature detectors can be used to identify large classes of features from the input space. The LVQ algorithm is a simplified precursor to more advanced learning algorithms, such as the self-organizing map. LVQ training is a type of competitive learning rule.

## 21 Boltzmann Learning

Boltzmann learning is statistical in nature, and is derived from the field of thermodynamics. It is similar to error-correction learning and is used during supervised training. In this algorithm, the state of each individual neuron, in addition to the system output, are taken into account. In this respect, the Boltzmann learning rule is significantly slower than the error-correction learning rule. Neural networks that use Boltzmann learning are called Boltzmann machines.

Boltzmann learning is similar to an error-correction learning rule, in that an error signal is used to train the system in each iteration. However, instead of a direct difference between the result value and the desired value, we take the difference between the probability distributions of the system.

# 22 ART Learning

### 22.1 ART Learning

Adaptive Resonance Theory (ART) learning algorithms compare the weight vector, known as the prototype, to the current input vector to produce a distance, r. The distance is compared to a specified scalar, the vigilance parameter p. All output nodes start off in the uncommitted state. When a new input sequence is detected that does not resonate with any committed nodes, an uncommitted node is committed, and it's prototype vector is set to the current input vector.

## 23 Self-Organizing Maps

### 23.1 Self-Organizing Maps

**Self-organizing maps** (SOM), sometimes called Kohonen SOM after their creator<sup>1</sup>, are used with unsupervised learning. SOM are modeled on biological neural networks, where groups of neurons appear to self organize into specific regions with common functionality.

### 23.2 Neuron Regions

Different regions of the SOM network are trained to be detectors for distinct features from the input set. Initial network weights are either set randomly, or are based off the eigenvectors of the input space. The Euclidean distance from each input sample to the weight vector of each neuron is computed, and the neuron whose weight vector is most similar to the input is declared the best match unit (BMU). The update formula is given as:

$$w_{j}[n+1] = w_{j}[n] + \Theta[j,n]\alpha[n](x[n] - w_{j}[n])$$

Here, w is the weight vector at time n.  $\alpha$  is a monotonically decreasing function that ensures the learning rate will decrease over time. x is the input vector, and  $\Theta[j, n]$  is a measure of the distance between the BMU and neuron j at iteration n. As can be seen from this algorithm, the amount by which the neuron weight vectors change is based on the distance from the BMU, and the amount of time. Decreasing the potential for change over time helps to reduce volatility during training, and helps to ensure that the network converges.

<sup>1</sup> http://en.wikipedia.org/wiki/Teuvo%20Kohonen

## 24 Pattern Recognition

Artificial neural networks are useful for pattern matching applications. Pattern matching consists of the ability to identify the class of input signals or patterns. Pattern matching ANN are typically trained using supervised learning techniques. One application where artificial neural nets have been applied extensively is optical character recognition (OCR). OCR has been a very successful area of research involving artificial neural networks.

An example of a pattern matching neural network is that used by VISA for identifying suspicious transactions and fraudulent purchases. When input symbols do not match an accepted pattern, the system raises a warning flag that indicates a potential problem.

# 25 Clustering

Similar to pattern matching, clustering is the ability to associate similar input patterns together, based on a measurement of their similarity or dissimilarity. An example of a clustering problem is the Netflix Prize, a competition to improve the Netflix recommendation system. Competitors are attempting to produce software systems that can suggest movies based on movies the viewer has already rated. Essentially, clustering movies based on the predicted like or dislike of the viewer.

## 26 Feature Detection

Feature detection or "association" networks are trained using non-noisy data, in order to recognize similar patterns in noisy or incomplete data. Correctly detecting features in the presence of noise can be used as an important tool for noise reduction and filtering.

For example, neural nets have been used successfully in a number of medical applications for detection of disease. One particular example is the use of ANN to detect breast cancer in mammography images.

## 27 Series Prediction

ANN can be trained to match the statistical properties of a particular input signal, and can even be used to predict future values of time series. There are a number of applications of feature prediction that have received significant research attention. Among these are financial prediction, meteorological prediction, and electric load prediction. ANN have shown themselves to be very robust in predicting complicated series, including non-linear non-stationary chaotic systems.

Financial prediction is useful for anticipating events in the economic system, which is considered to be a chaotic system. ANN have been used to predict the performance and failure rate of companies, changes in the exchange rate, and other economic metrics.

Meteorological prediction is a difficult process because current atmospheric models rely on highly recursive sets of differential equations which can be difficult to calculate, and which propagate errors through the successive iterations. Using neural nets for meteorological prediction is a non-trivial task, and there have been many conflicting reports of the efficacy of the technique.

# 28 Data Compression

## 29 Curve Fitting

#### 29.1 Curve Fitting

Curve-fitting problems represent an attempt for the neural network to identify and approximate an arbitrary input-output relation. Once the relation has been modeled to the necessary accuracy by the network, it can be used for a variety of tasks, such as series prediction, function approximation, and function optimization.

#### 29.2 Function Approximation

Function approximation or modeling is the act of training a neural network using a given set of input-output data (typically through supervised learning) in order to deduce the relationship between the input and the output. After training, such an ANN can be used as a black box with an input-output characteristic approximately equal to the relationship of the training problems. Because of the modular and non-linear nature of artificial neural nets, they are considered to be able to approximate any arbitrary function to an arbitrary degree of accuracy. More accuracy in this case represents a tradeoff with system complexity, and the ability to generalize.

## 30 Optimization

Optimization ANNs are concerned with the minimization of a particular cost function with respect to certain constraints. ANN are shown to be capable of highly efficient optimization.

## 31 Control

### 31.1 Control Systems

Artificial neural networks have been employed for use in control systems<sup>1</sup> because of their ability to identify patterns, and to match arbitrary non-linear response curves.

#### 31.2 See Also

• Control Systems $^2$ 

<sup>1</sup> http://en.wikibooks.org/wiki/Control%20Systems

<sup>2</sup> http://en.wikibooks.org/wiki/Control%20Systems

## 32 Criticisms and Problems

## 33 Artificial Intelligence

### 34 Resources

#### 34.1 Wikibooks

- Linear Algebra<sup>2</sup>
- Abstract Algebra<sup>3</sup>
- Calculus<sup>4</sup>
- Engineering Analysis<sup>6</sup>
- MATLAB Programming<sup>7</sup>

#### 34.2 Commons

- Commons: Artificial neural network  $^8$
- Commons:Category:Neural networks<sup>9</sup>

<sup>1</sup> http://en.wikibooks.org/wiki/Artificial%20Intelligence

<sup>2</sup> http://en.wikibooks.org/wiki/Linear%20Algebra

<sup>3</sup> http://en.wikibooks.org/wiki/Abstract%20Algebra

<sup>4</sup> http://en.wikibooks.org/wiki/Calculus

 $<sup>5 \</sup>qquad {\tt http://en.wikibooks.org/wiki/Signals\%20and\%20Systems}$ 

<sup>6</sup> http://en.wikibooks.org/wiki/Engineering%20Analysis

<sup>7</sup> http://en.wikibooks.org/wiki/MATLAB%20Programming

<sup>8</sup> http://en.commons.org/wiki/Artificial%20neural%20network

<sup>9</sup> http://en.commons.org/wiki/Category%3ANeural%2Onetworks

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 $<sup>8 \</sup>qquad {\tt http://en.wikibooks.org/w/index.php?title=User:Xania}$ 

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<sup>9</sup> Chapter 36 on page 91

1		GFDL
2	Dake <sup>10</sup> , Mysid <sup>11</sup>	
3	en:User:Cburnett <sup>12</sup>	GFDL
4		GFDL
5		PD
6		PD
7	Chrislb <sup>13</sup>	GFDL
8	Chrislb <sup>14</sup>	GFDL
9	Chrislb <sup>15</sup>	GFDL
10		GFDL
11	Chrislb <sup>16</sup>	GFDL
12	Chrislb <sup>17</sup>	GFDL
13		GFDL
14	Chrislb <sup>18</sup>	GFDL
15		GFDL

<sup>10</sup> http://en.wikibooks.org/wiki/User%3ADake

<sup>11</sup> http://en.wikipedia.org/wiki/User%3AMysid

<sup>12</sup> http://en.wikibooks.org/wiki/%3Aen%3AUser%3ACburnett

 $<sup>13 \</sup>quad \mathtt{http://en.wikibooks.org/wiki/User\%3AChrislb}$ 

 $<sup>14 \</sup>quad {\tt http://en.wikibooks.org/wiki/User\%3AChrislb}$ 

<sup>15</sup> http://en.wikibooks.org/wiki/User%3AChrislb

<sup>16</sup> http://en.wikibooks.org/wiki/User%3AChrislb

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Version 3, 29 June 2007

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