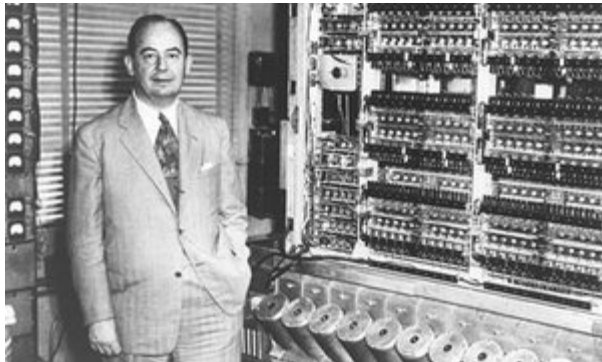
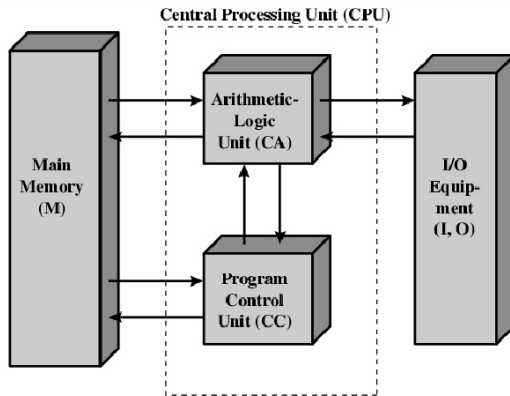


# Introduction to Soft Computing and Intelligent Systems

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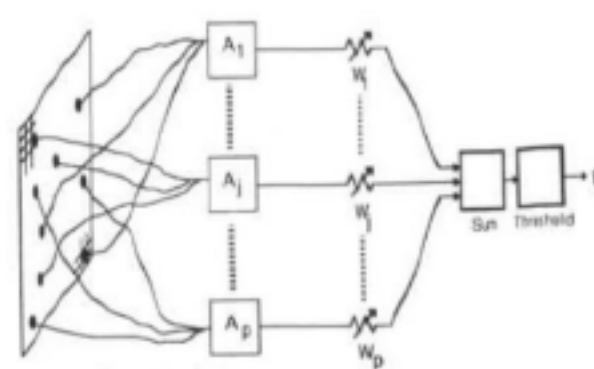
# Origins of Modern Computing and Neural Networks

## Von Neuman Computer



Father of Modern Computing

## Rosenblatt Perceptron

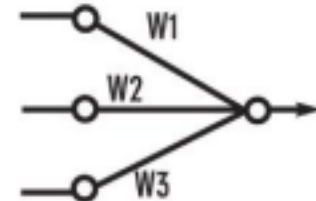


Frank Rosenblatt  
(1928-1971)

### Original Perceptron

(From *Perceptrons* by M. L. Minsky and S. Papert, 1969, Cambridge, MA: MIT Press. Copyright 1969 by MIT Press.)

### Simplified model:



Father of Connectionist Computing

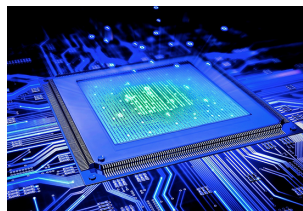
# Introduction

- Humans have developed sensory systems to perceive their environments and make decision based on what they observe, for example:
  - Understanding speech
  - Recognizing faces and objects
  - Discerning taste
- This is greatly influenced by how we perceive the patterns in our environment and act on this perception.

Humans have always strived to impart machines with similar capabilities as theirs for faster task execution, better predictability and higher accuracy

# Recent Breakthroughs

- IBM's **Cat-brain** project is a chip like brain (**SyNAPSE**) to capture surrounding information in real time. Simulated the cat brain cortex with 147,456 cores and 144TB of memory: basic synaptic circuit for the brain chip.



*Current work on  $10^{12}$  artificial neurons and  $10^{14}$  synapses*

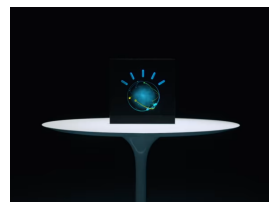
- Google's **DeepMind** project, aims at formalizing intelligence to implement in machines and to understand the human brain. Uses **deep learning** technologies.



*On March 2016, AlphaGo beat the top player in the world (Lee Sedol) in the Go game (Dan 9)*

# Recent Breakthroughs

- IBM's **Watson**, QA system is used as knowledge repository, to answer virtually any inquiry (in Wikipedia and other knowledge repositories). Uses NLU and ML technologies. Has access to more than  $10^9$  pages content.



*Won Jeopardy! Contest against top players in the world. Project Intu will equip devices and machines with Watson's powerful AI tools, hence providing them with cognitive capabilities and make them aware of their environments (ZDNET, Nov. 2016)*

- **Amazon, Google and Netflix** are using Big Data analytics based on **Deep Learning** to target consumer behavior in various areas including entertainment, shopping and provide adequate recommendations to the user

# Intelligent Systems: What Are They?

## Definition

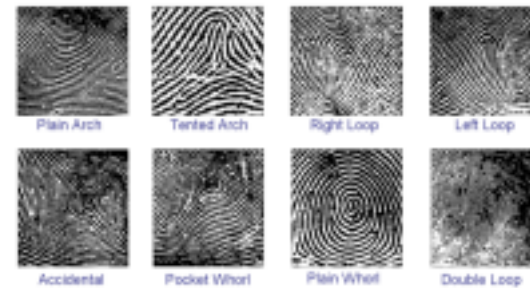
Intelligent systems are artificial entities involving a mix of software and hardware which have a capacity to acquire and apply knowledge in an "intelligent" manner and have the capabilities of perception, reasoning, learning, and making inferences (or, decisions) from incomplete information.

# Intelligent Systems: Examples

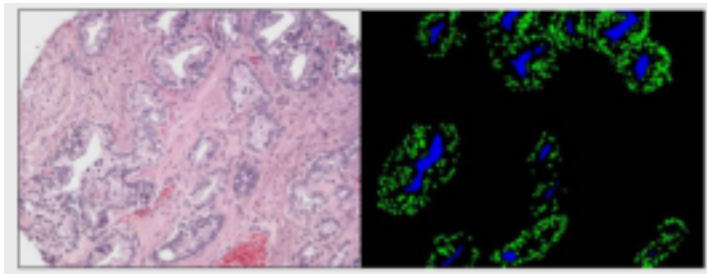
## License Plate Recognition



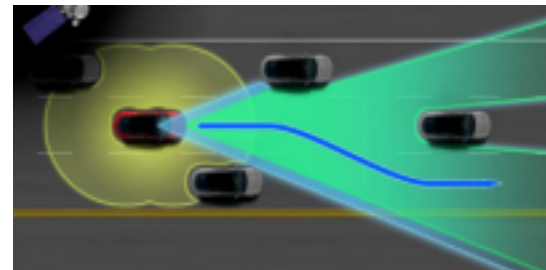
## Fingerprint Recognition



## Cancer Detection



## Autonomous Navigation



# Intelligent Systems: Characteristics

## Main Characteristics

A feature that is indispensable in these systems is the generation of outputs, based on some inputs and the nature of the system itself. The inputs to a system may include information as well as tangible items, and the outputs may include decisions as well as physical products.



# Intelligent Systems: Capabilities

It is commonly accepted that an intelligent system possesses one or more of the following characteristics and capabilities:

## Capabilities

Sensory perception; Pattern recognition; Learning and knowledge acquisition; Inference from incomplete information; Inference from qualitative or approximate information; Ability to deal with unfamiliar situations; Adaptability to new, yet related situations (through expectational knowledge; Inductive reasoning.

# Example

A typical input variable is identified for each of the following examples of dynamic systems:

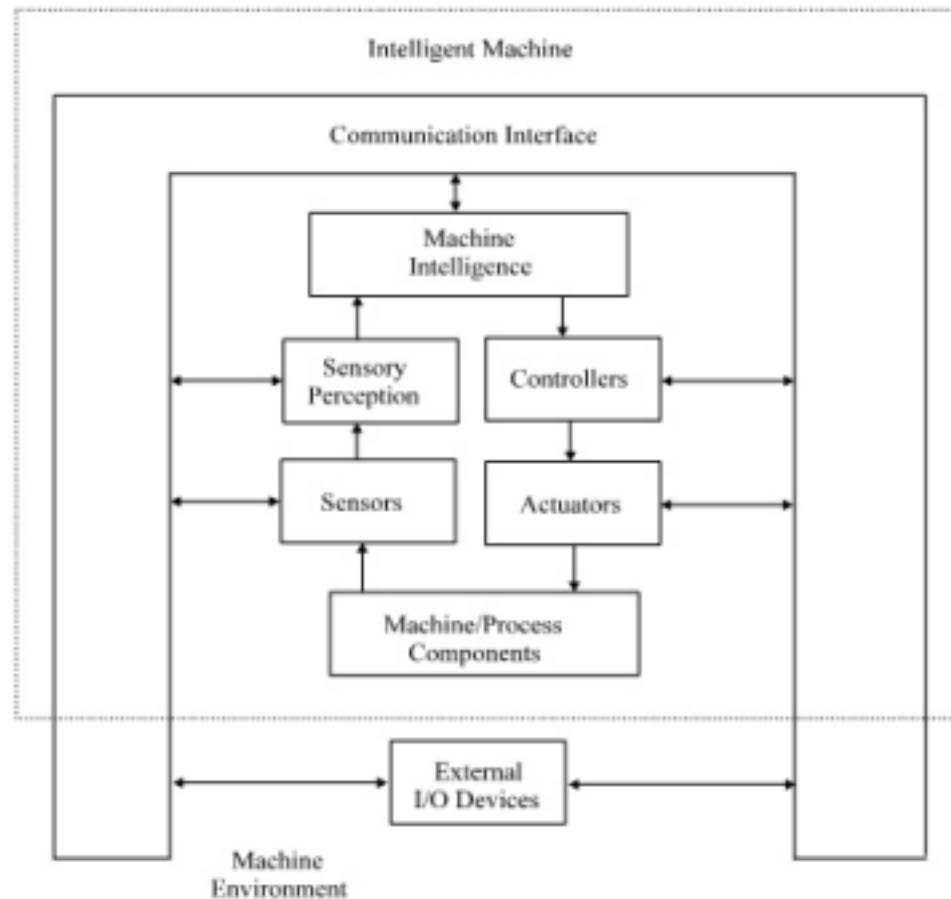
- Human body: neuroelectric pulses
- Company: information
- Power plant: fuel rate
- Automobile: steering wheel movement
- Robot: voltage to joint motor

# Example (Cont.)

Possible output variables for each of these systems are:

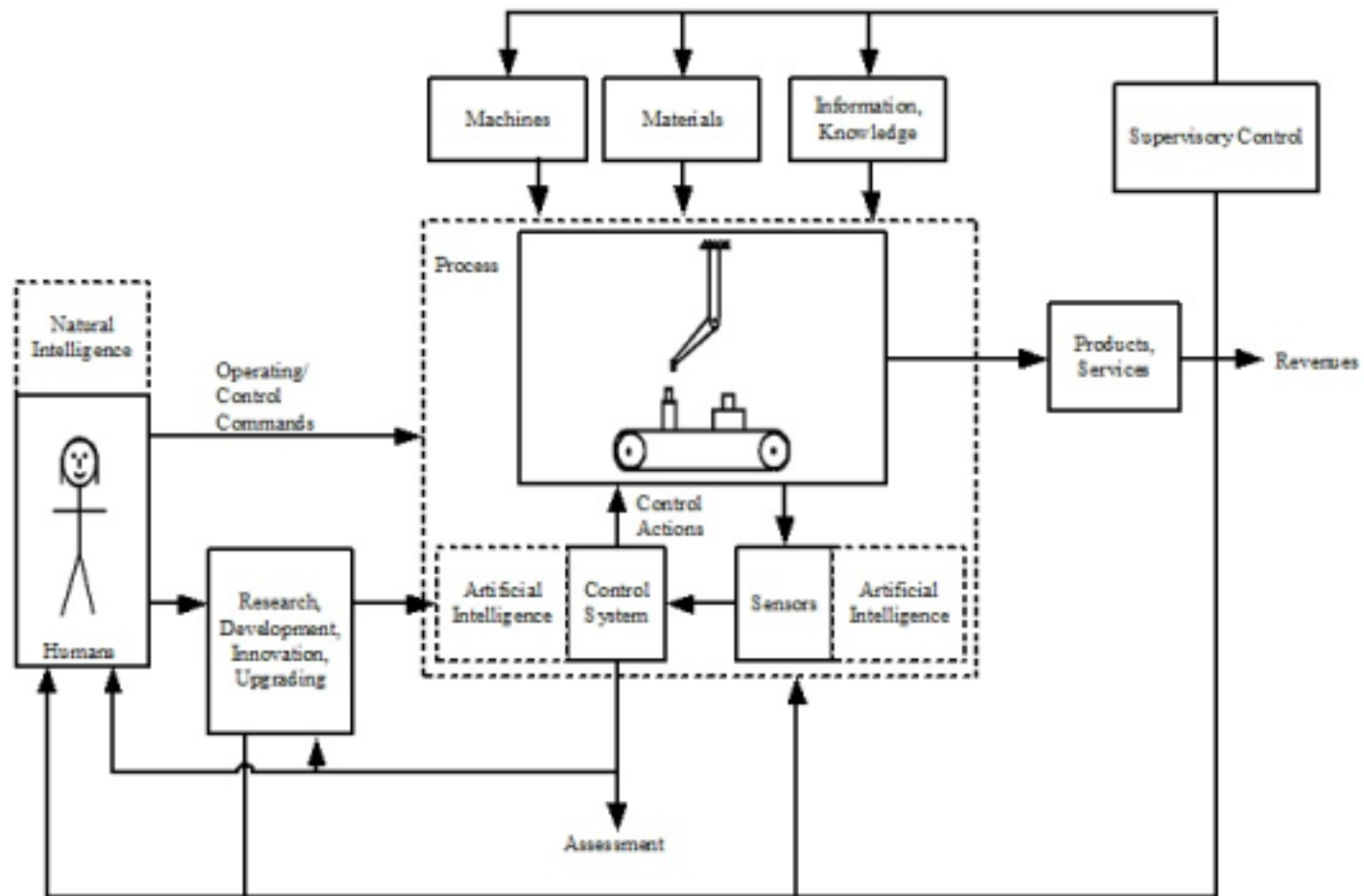
- Human body :Muscle contraction, body movements
- Company: Decisions, finished products
- Power plant: Electric power, pollution rate
- Automobile: Front wheel turn, direction of heading
- Robot: Joint motions, effector motion

# Intelligent Machines: Architecture



An intelligent Machine

# Intelligent Machines: Schematics and Modules



# Knowledge-Based Systems

## Definition

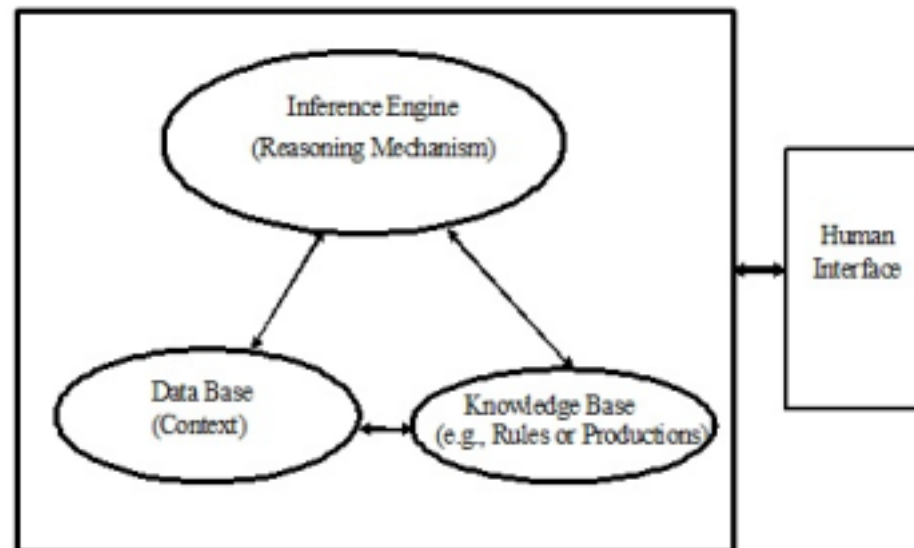
A knowledge-based system is able to make perceptions (e.g., sensory perception) and new inferences or decisions using its reasoning mechanism (inference engine).

This is done by interpreting the meaning and implications of the new information within the capabilities of the existing knowledge base. These inferences may form the outputs of the knowledge-based systems.

# Knowledge-Based Systems

## Decision Making

The associated decision-making task is an intelligent processing activity, which in turn may lead to enhancement, refinement, and updating of the knowledge base itself.



The Structure of a Knowledge-Based System

# Example

Consider a knowledge base for selecting a control technique, as given by the following set of rules:

- If the plant is linear and uncoupled then use Control Category A.
- If the plant is linear and coupled then use Control Category B.
- If the plant is nonlinear use Control Category C.
- If Category A and a plant model is known then use Subgroup 1.
- If Category B and a plant model is known then use Subgroup 2.
- If Subgroup 1 and high model uncertainty then use H-infinity control.



# Knowledge Representation

An appropriate representation of knowledge, including intuition and heuristic knowledge, is central to the development of machine intelligence and of knowledge-based systems.

In a knowledge-based system two types of knowledge are needed: knowledge of the problem (problem representation or modelling), and knowledge regarding methods for solving the problem.

Ways of representing and processing knowledge include: Logic, Semantic Networks, Frames, Production Systems, and Fuzzy Logic.

# Logic of Inferencing

## Definition

Logic is a useful technique of representing and processing knowledge, and is applicable in knowledge-based systems. In logic, knowledge is represented by statements called propositions, which may be joined together using connectives.

The knowledge may be processed through reasoning, by the application of various laws of logic including an appropriate rule of inference. There are two types of logics: crisp (binary) and fuzzy (multivalued).

# Crisp Logic

## Definition

Conventional (crisp) logic deals with statements called "propositions". In binary (or, two-valued) logic, a proposition can assume one of only two truth values: true (T), false (F). An example of a proposition would be "John is over 50 years old".

# Crisp Logic (Cont.d)

Now consider the following propositions:

1. Charcoal is white
2. Snow is cold
3. Temperature is above 60C

Here proposition 1 has the truth value F, and proposition 2 has the truth value T. But, for proposition 3 the truth value depends on the actual value of the temperature. Specifically, if the temperature is above 60C the truth value is T and otherwise it is F.

# Knowledge Processing

## Reasoning and Inferencing

Knowledge may be processed through reasoning. This is done by the application of various laws of logic including an appropriate rule of inference, subjected to a given set of data (measurements, observations, external commands, previous decisions, etc.) to arrive at new inferences or decisions.

# Knowledge Processing (Cont.d)

In intelligent control, for example, a knowledge base is processed through reasoning, subjected to a given set of data (measurements, observations, external commands, previous decisions, etc.) to arrive at new control decisions.

# Knowledge Processing (Cont.d)

The typical end objective of knowledge processing is to make inferences.

This may involve the following steps:

- Simplify the knowledge base by applying various laws of logic.
- Substitute into the knowledge base any new information (including data and previous inferences).
- Apply an appropriate rule of inference.

These steps may be followed in any order and repeated any number of times, depending on the problem.

# Propositional Calculus and Predicate Calculus

## Definition

Propositional calculus is the branch of logic where propositions (i.e., statements that are either true or false) are used in logic "calculations". Note that the term "calculus" in the present terminology denotes "the approach of calculation" and has nothing to do with differential or integral calculus.



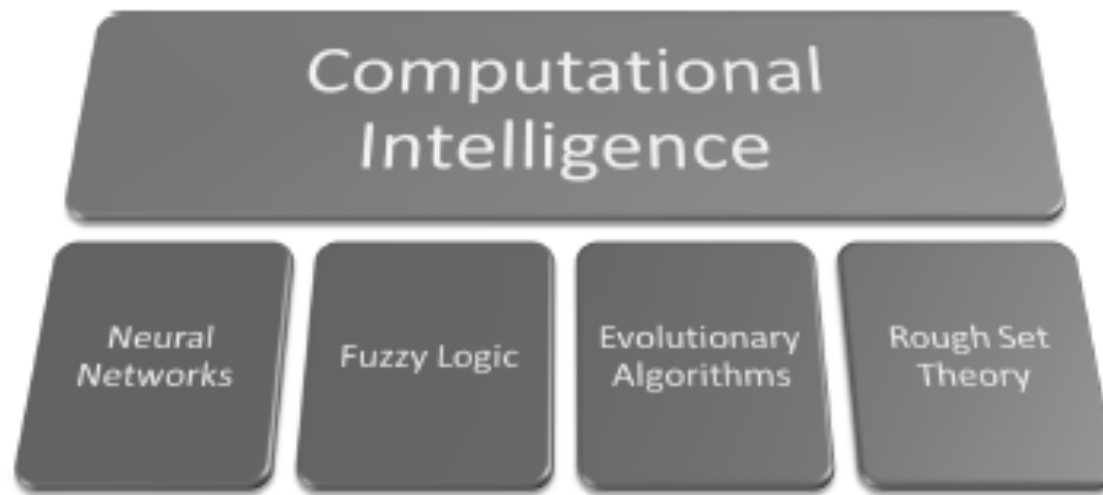
# Soft Computing/Computational Intelligence

## Definition

Soft computing is an important branch of study in the area of intelligent and knowledge-based systems. It has effectively complemented conventional AI in the area of machine intelligence (computational intelligence).

Human reasoning is predominantly approximated, qualitative, and "soft". Humans can effectively handle incomplete, imprecise, and fuzzy information in making intelligent decisions.

# Soft Computing/Computational Intelligence



# Soft Computing/Computational Intelligence

- Fuzzy logic, probability theory, neural networks, and genetic algorithms are cooperatively used in soft computing for knowledge representation and for mimicking the reasoning and decision-making processes of a human.
- Quite effective are the mixed or hybrid techniques, which synergistically exploit the advantages of two or more of these areas.

# Fuzzy Logic

- Fuzzy logic is useful in representing human knowledge in a specific domain of application and in reasoning with that knowledge to make useful inferences or actions.
- The conventional binary (bivalent) logic is crisp and allows for only two states. This logic cannot handle fuzzy descriptors, examples of which are "fast" which is a fuzzy quantifier and weak which is a fuzzy predicate.

# Fuzzy Logic

- Fuzzy logic (FL) allows for a realistic extension of binary, crisp logic to qualitative, subjective, and approximate situations, which often exist in problems of intelligent machines.
- In (FL), the knowledge base is represented by if-then rules of fuzzy descriptors. An example of a fuzzy rule would be, "If the speed is slow and the target is far, then moderately increase the power", which contains the fuzzy descriptors slow, far, and moderate.

# Neural Networks

## Definition

Artificial neural networks (NN) are massively connected networks of computational neurons, and represent parallel-distributed processing structures. The inspiration for NN has come from the biological architecture of neurons in human brain.

A key characteristic of neural networks is their ability to approximate arbitrary nonlinear functions. Since machine intelligence involves a special class of highly nonlinear decision-making, neural networks would be effective there.

# Neural Networks

- A neural network consists of a set of nodes, usually organized into layers, and connected through weight elements called synapses. The analogy to the operations in a biological neuron is interesting.
- Specifically, in a biological neuron, the dendrites receive information from other neurons. The soma (cell body) collects and combines this information, which is transmitted to other neurons using a channel (tubular structure) called axon.

# Neural Networks

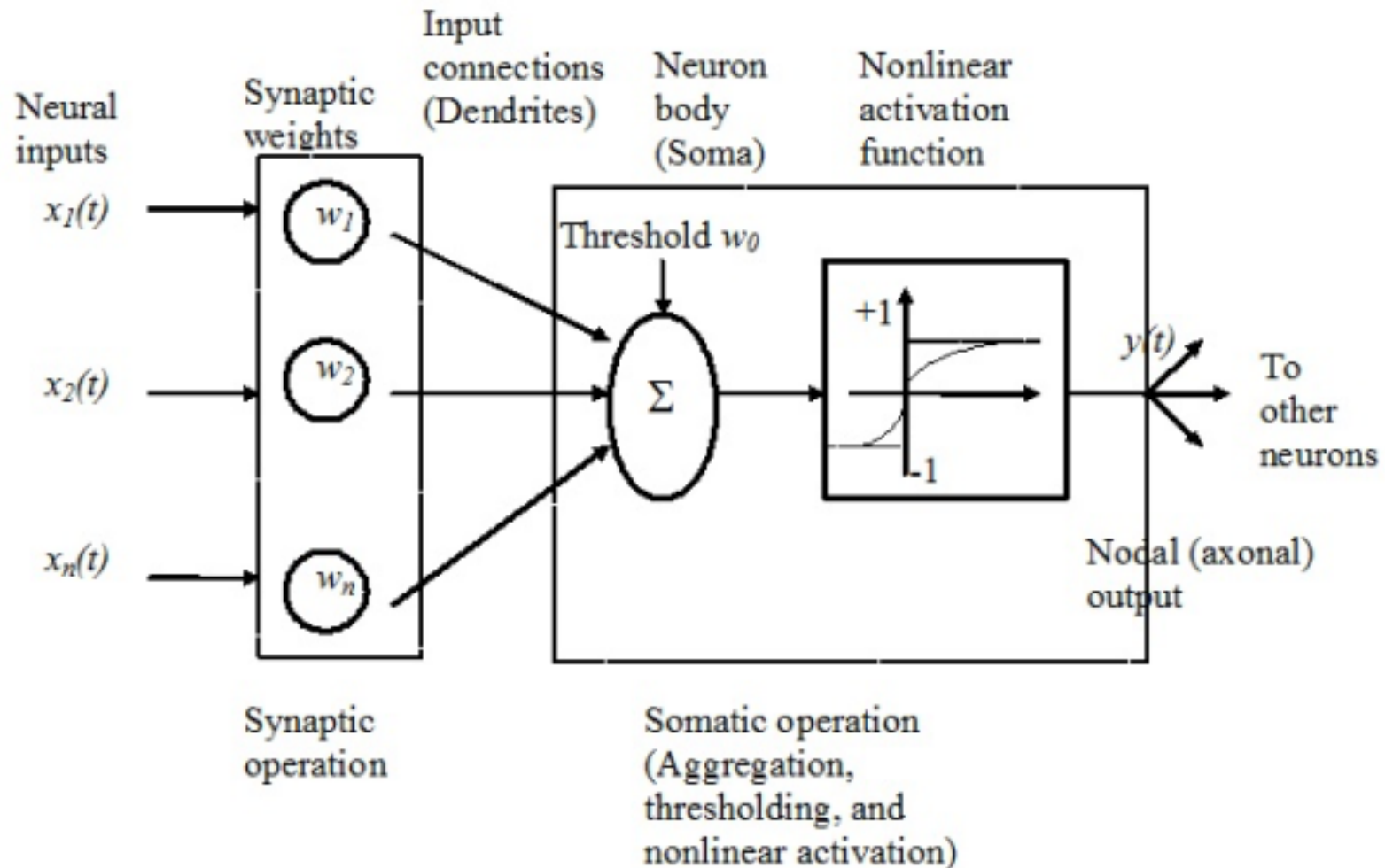
This biological analogy, apart from the abilities to learn by example, approximation of highly nonlinear functions, massive computing power, and memory, may be a root reason for inherent "intelligence" in a neural network. If the weighted sum of the inputs to a node (neuron) exceeds a threshold value  $w_0$ , then the neuron is fired and an output  $y(t)$  is generated according to:

$$y(t) = f(\sum_t^N w_i x_i - w_0)$$

where  $x_i$  are neuron inputs,  $w_i$  are the synaptic weights, and  $f(.)$  is the activation function.



# Neural Networks Basic Component



# Genetic Algorithms

## Definition

Genetic algorithms (GA) are derivative-free optimisation techniques, which can evolve through procedures analogous to biological evolution.

Genetic algorithms belong to the area of evolutionary computing. They represent an optimisation approach where a search is made to "evolve" a solution algorithm, which will retain the "most fit" components in a procedure that is analogous to biological evolution through natural selection, crossover, and mutation.

# Genetic Algorithms

Evolutionary computing can play an important role in the development of an optimal and self-improving intelligent machine. Evolutionary computing has the following characteristics:

- It is based on multiple searching points or solution candidates (population based search)
- It uses evolutionary operations such as crossover and mutation
- It is based on probabilistic operations

# Probabilistic Reasoning

- Probabilistic reasoning may be viewed in an analogous manner to fuzzy-logic reasoning, considering uncertainty in place of fuzziness as the concept of approximation that is applicable.
- Probability distribution/density functions are employed in place of membership functions.

# Uncertainty vs Fuzziness

- It is important to compare the concepts of uncertainty and fuzziness. Uncertainty is statistical inexactness due to random future events.
- Fuzziness arises when the decision of whether a particular object belongs to a given set is a matter of perception, which can be subjective. Probabilistic and statistical techniques may be employed in conjunction with fuzzy logic in a complementary manner.

# Fuzziness vs Uncertainty

	Fuzziness	Uncertainty
<b>Advantages</b>	<ul style="list-style-type: none"><li>• Incomplete information can be handled</li><li>• Particularly useful in representing and processing human-oriented knowledge</li><li>• Approximate reasoning is possible, with qualitative and linguistic knowledge</li><li>• It is a technique of soft computing</li></ul>	<ul style="list-style-type: none"><li>• Useful in situations having random influences with known probability distributions</li><li>• Governs many practical situations</li><li>• Mathematical procedures are well established</li><li>• System parameters can be determined using crisp experiments</li></ul>
<b>Disadvantages</b>	<ul style="list-style-type: none"><li>• May introduce a degree of inaccuracy</li><li>• Needs prior knowledge and experience of the problem in generating the knowledge base</li><li>• Can be slow</li></ul>	<ul style="list-style-type: none"><li>• Not related to fuzzy sets</li><li>• May fail under incomplete information</li><li>• Results are directly affected by the type and accuracy of the probability distributions</li></ul>

# Approximation

## Definition

Approximation is a "soft" concept and is related to intelligence. The capability of approximation for the purpose of comparison, pattern recognition, reasoning and decision-making is a manifestation of intelligence (e.g., dealing with incomplete and subjective information, unfamiliar situations, comparison, and judgement).

# Approximation (Cont.d)

## Concepts

There are many concepts of approximation, examples of which include imprecision, uncertainty, fuzziness, and belief. These concepts are not identical even though the associated methods of information representation and processing may display analytical similarities.



# Concepts of Approximation and their Characteristics

Concept	Property	Example
Ambiguity	The condition has a several different possibilities, and it is not determined which one is valid.	The machine response "may or may not" satisfy the specification.
Vagueness	The condition is not precisely (clearly) defined.	The machine response "may have" met the specification.
Generality	The condition may apply to many (finite or infinite) situations depending on the specific context.	The machine response is "x" times the specification.
Imprecision	Condition can assume a state within a clearly defined (crisp) tolerance interval.	The machine response is "within $\pm 5\%$ " of the specification.
Uncertainty	There is a degree of probability associated with occurrence of the condition.	There is a "90% probability" that the machine response meets the specification.
Fuzziness	The membership of the condition is not crisply defined (set boundary of the condition is not crisp).	The machine response is "close to" the specification.
Belief (Subjective Probability)	The level of belief on the condition (membership of a crisp or fuzzy set) is through knowledge and evidence.	It is believed at a level of 90% that the machine response meets the specification.
Plausibility	The plausibility of nonmembership fully complements the belief of membership (dual condition of belief). $Bl(x \in A) + Pl(x \notin A) = 1$	It is plausible at a level of 95% that the machine response meets the specification.

# Technology Needs

Significant advances have been made, in machine implementation of characteristics of intelligence such as sensory perception, pattern recognition, knowledge acquisition and learning, inference from incomplete information, inference from qualitative or approximate information, ability to handle unfamiliar situations, adaptability to deal with new yet related situations, and inductive reasoning.

# Technology Needs (Cont.)

Much research and development would be needed in these areas, pertaining to techniques, hardware, and software before a machine could reliably and consistently possess the level of intelligence of say, a dog.

# Example

Consider a handwriting recognition system, which is a practical example of an intelligent system. The underlying problem cannot be solved through simple template matching, which does not require intelligence.

Handwriting of the same person can vary temporally, due to various practical shortcomings such as missing characters, errors, non-repeatability, physiological variations, sensory limitations, and noise.

5	0	4	1
3	5	3	6
4	0	9	1
3	8	6	9

## Example (Cont.)

It should be clear from this observation that a handwriting recognition system has to deal with incomplete information and unfamiliar objects (characters), and should possess capabilities of learning, pattern recognition, and approximate reasoning, which will assist in carrying out intelligent functions of the system. Techniques of soft computing are able to challenge such needs of intelligent machines.

## Example 2 (Imagenet Library)



# Summary

Techniques of soft computing are powerful by themselves in achieving the goals of machine intelligence. They have a particular appeal in view of the biological analogies that exist within humans.

Fuzzy techniques attempt to approximate human knowledge and the associated reasoning process; neural networks are a simplified representation of the neuron structure of a brain; genetic algorithms follow procedures that are crudely similar to the process of evolution in biological species; and probabilistic techniques can analyze random future action of a human.



# Techniques of Computational Intelligence

Technique	Characteristic	A Popular Analogy
Fuzzy Logic	Uses fuzzy rules and approximate reasoning	Human knowledge
Neural Networks	Network of massively connected nodes	Neuron structure in brain
Genetic Algorithms	Derivative-free optimisation	Biological evolution
Probability	Incorporates uncertainty in predicting future events	Random action of a human
Conventional AI	Symbolic processing of information	Symbolic languages



# University of Waterloo

