



# INT404 ARTIFICIAL INTELLIGENCE

Lecture 3





Artificial intelligence techniques can be divided into **two** types:

- 1) Symbolic computation (including sub-symbolic)
- 2) Non-symbolic computation

Example:- When a child is born, he possess only basic knowledge of how to survive in the brain, gradually he will learn the rest of the knowledge through learning by experiences.





#### Symbolic Computation has two branches

- Heuristic search Adjoining, Segment 1 of the World view.
   Heuristic A guide, an approximation, a thumb rule. Basically helps in pruning(Weed out unwanted or unnecessary things) the search tree.
- 2. Knowledge-based systems In the world view, **between** heuristic search and sub-symbolic computation (neural networks).





# **Knowledge Based Systems:**

Core Areas of Knowledge Based systems

- 1. Knowledge Base Representation
- 2. Inference Engine
- 3. User interface
- 4. Knowledge acquisition module





# **Knowledge Based Systems:**

# Representation techniques are primarily:

 production rules – sets of if-then rules, similar to production rules used to specify a grammar.

Example: If the car does not start check the battery, by pressing the horn.





#### **Knowledge Based Systems:**

Representation techniques are primarily:

2. Semantic Networks – Set of Nodes and Links between them. The links represent Relationships between the nodes

Example: Nodes – Man, Hands, Legs, Walk

Relationships – Has (between Man and hands and between Man and Legs) and Can (between Man and Walk).

A type of Semantic networks is Frames (Slot-filler notation). These encode default (commonly occurring) values (filler) for the attributes in a relation (slot).





## **Knowledge Based Systems:**

Inference Engine - Search on the knowledge base leads to Inferences.

Knowledge Acquisition module - The knowledge being incomplete will be dynamic. Provision to acquire knowledge is provided by using machine learning strategies.





#### **Knowledge Based Systems:**

#### **Machine Learning Strategies:**

- 1. Rote learning The system is told the actual knowledge. The system's work is to map the knowledge into its internal representation.
- 2. Learning by being told The system is given paragraphs that convey the knowledge. The system has to **glean** the knowledge and then store it.
- 3. Learning by being told and asking questions In addition to strategy 2, the system analyses the knowledge, finds discrepancies and asks questions to sort out the conflicts.
- 4. Learning by induction from positive examples The system is given examples of the concept. It generalizes the examples to arrive at the knowledge





## **Knowledge Based Systems:**

### **Machine Learning Strategies:**

- 5. Learning by Induction from Positive examples and Negative examples

  To avoid over generalization, negative examples are given, which are used to specialize the knowledge.
- 6. Learning by Induction through experimentation The system generates examples itself by designing experiments on the environment.
- 7. Learning by Analogy The system maps the knowledge it has to the new problem, using analogy.
- 8. Learning by Abduction The system creates new hypotheses and designs experiments to ratify them.



# **Knowledge Based Systems:**

## Genetic Programming:

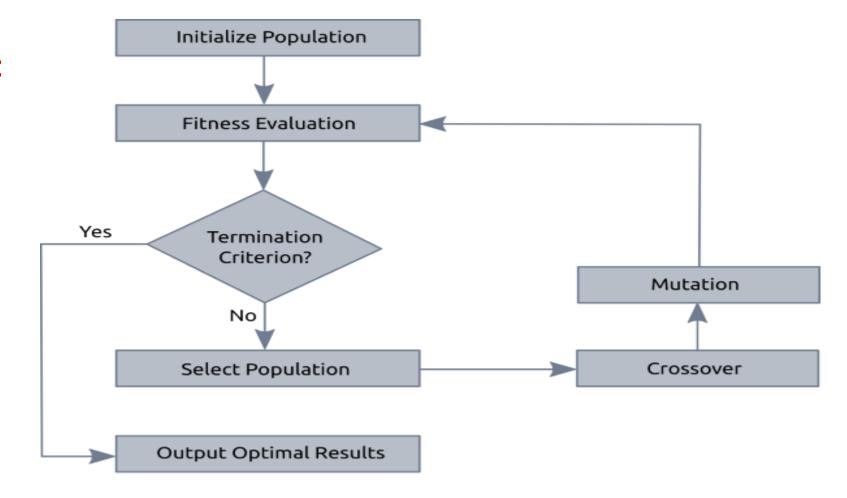


Figure 2: Basic structure of Genetic Algorithm





#### **Knowledge Based Systems:**

#### **Genetic Programming:**

This field lies at the extreme of Knowledge Based Systems (adjoining subsymbolic computation in the World view)— They model Human evolution methods.

#### Approach:

- 1. Create an initial population of entities
- 2. Each entity's characteristics are represented
- A fitness function evaluates the entities.
- 4. The best two of the population are chosen
- 5. These two are used to generate 'offspring's' -> new population. Process repeats.





## **Knowledge Based Systems:**

#### Genetic Programming:

Offspring generation operators:

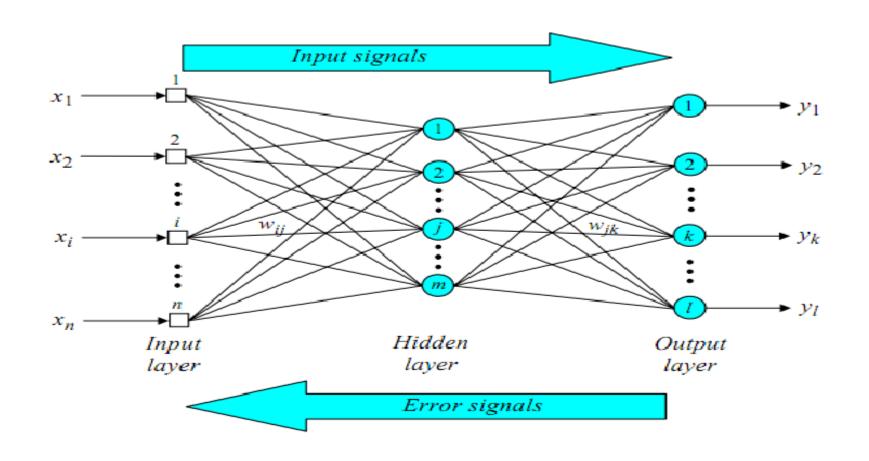
- Reproduction All characteristics of both parents are reproduced in the offspring.
- 2. Crossover A subset of characteristics of one parent are linked with the subset of characteristics of the other parent.
- 3. Mutation The characteristics of one parent are changed randomly to create the offspring. Handles the Local Maxima problem





# Sub Symbolic Computation (Neuro-computing)

- Adjoins Segment 3 of the world view.
- Deals with signal level computation required because a number of problems do not have explicit knowledge associated with them.
   Example – recognizing people or recognizing handwriting.
- This area deals with patterns that are more complex than the ones dealt with by symbolic computation.







#### Sub Symbolic Computation (Neuro-computing):

Core areas of Sub-symbolic computation are:

- 1. Architecture
- 2. Learning mechanism

In sub-symbolic computation all the knowledge is learnt by the system.

Neuro-computing attempts to mimic the structure of the human intelligence system, with its neurons and synapses.

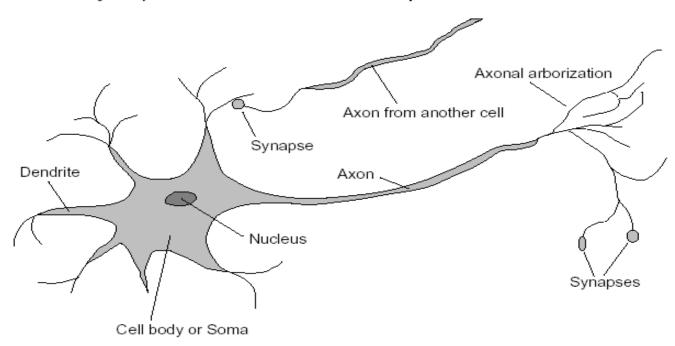
Neuron – receives input from many other neurons. Each input is magnified by a multiplication factor. (This multiplication factor represents the degree of interest, effect that the particular input has on the neuron.)



# P

#### Brains

 $10^{11}$  neurons of  $\,>20$  types,  $10^{14}$  synapses, 1ms–10ms cycle time Signals are noisy "spike trains" of electrical potential







#### Sub Symbolic Computation (Neuro-computing):

- All the multiplied values are summed up and compared to a 'threshold value'. If the threshold value is less then the neuron fires an output.
- Knowledge is acquired by learning the correct multiplication values.

#### Learning is done in one of two ways:

1. Supervised learning - Here the desired output for a given input is known. A simple method is Back Propagation network. Here the output is compared with the desired output. Differences are propagated backwards, to make changes to the multiplication factors.





#### Sub Symbolic Computation (Neuro-computing):

- 2. Unsupervised learning Here the desired output is not given to the system. The system uses Clustering to club similar input together. Example Kohonen
- 3. A third learning technique is Self-Supervised Learning Here the results of a previous iteration are used to bias the clustering results in the current iteration. Example – Adaptive Resonance Technique.