# LEARNING (AI)

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# Learning

It denotes changes in the system that are adaptive in the sense that they enable the system to do the same task or tasks drawn from the same population more efficiently and more effectively the next time.

# **Knowledge Acquisition**

It is the process of adding new knowledge to knowledge base and refining or otherwise improving knowledge that was previously acquired.

It is associated with some purpose such as expanding the capabilities of a system or improving its performance at some specified task.

Thus knowledge acquisition is a goal oriented creation and refinement of knowledge. Acquired knowledge may consist of facts, rules, concepts, procedures, heuristics, formulas, relationships, statistics or other useful information.

### **Rote Learning**

This is also known as learning by memorization. It requires the least amount of interface and is accomplished by simply coping the knowledge in the same form that it will be used directly into the knowledge base. When computation is more expensive than recall, this strategy saves a significant amount of time.

## **Capabilities of Rote Learning System**

In order to retrieve the stored information quickly, the system should have the following capabilities:

- Organized storage of information: As complexity of stored information increases, more sophisticated techniques are required in order to access the appropriate stored value quickly.
- Generalization: The number of distinct objects that might potentially be shared can be very large. In order to store lesser number of objects, some kind of generalization is required.

## **Explanation-Based Learning**

It uses an explicitly represented domain theory to construct an explanation of a training example which logically follows from the explanation of the instance, rather than from the instance itself.

Explanation-based learning filters noise, selects relevant aspects of experience and organizes training data into a systematic and coherent structure.

## **Inductive Learning**

Classification is the process of assigning to a particular input, the name of a class to which it belongs.

The classes from which the classification procedure can choose can be described in many ways:

- Each class can be defined by a weighted sum of the values of features, that are relevant to the task domain.
- Each class can be defined as a structure so that features that are relevant to that of the task domain may be compared.

This task of constructing class definitions is called concept learning or induction.

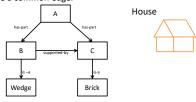
# **Explanation-Based Learning**

Explanation based learning begins with:

- Target concept: The learner's task is to determine an
  effective definition of this concept. Depending upon the
  specific application, the target concept may be a
  classification, a theorem, a plan for achieving a goal or a
  heuristic for a problem solver.
- 2. Training example: It is an instance of the target.
- Domain Theory: It is a set of rules and facts that are used to explain how the training example is an instance of the goal concept.
- Operation criteria: It is some means of describing the form that concept definitions may take.

## **Structural Description**

Marry link – Two objects marry if they have faces that touch and they have a common edge.

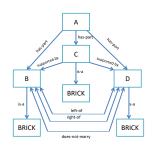


Node A→ Represents entire structure.

# **Structural Description**

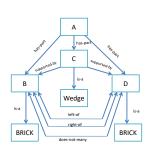
Arch





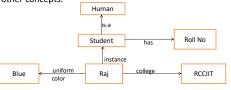
# **Structural Description**





#### **Semantic Nodes**

In this, information is represented as a set of nodes connected to each other by a set of labeled arcs, which represent relationship among the nodes. Thus the meaning of a concept comes from the ways in which it is connected to other concepts.



In this network, inheritance can be used to derive the additional relation has(Raj, Roll No)

#### **Intersection Search**

It is a process which is used to find relationships among objects by spreading out activation from each of the nodes and seeing where the activations meet. The kind of reasoning exploits the entity-based organization of knowledge to find relationships among objects.

#### **Representing Non-Binary predicates:**

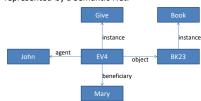
Three or more place predicates can be represented (converted) into binary form by creating one new binary object representing the entire predicate statement and then introducing binary predicates to describe the relations of this new object each of the original arguments.

Example: Score(INDIA, BANGLADESH, 10-5)

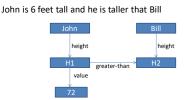


#### Semantic Network for an N place predicate

"John gave the book to Mary" – This sentence can be represented by a Semantic Net.



John is 72" tall .

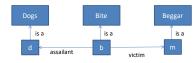


height 72"

The arc values may define entities or relationship

#### **Partitioned Semantic Nets**

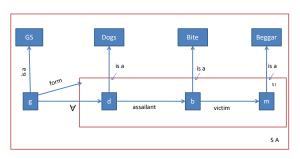
"The dog bit the beggar"



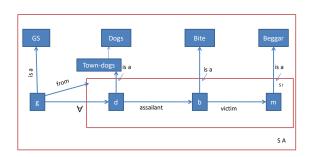
If we want to represent simple quantified expressions using semantic nets, then portioning of the semantic net is required. "Every dog has bitten a beggar".

 $\forall x : Dog(x) \rightarrow \exists y : Beggar(y) \land Bite(x,y)$ 

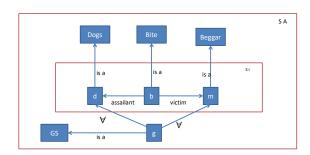
It is necessary to encode the scope of the universally qualified variable x.



## Every dog has bitten the beggar



## Every dog has bitten every beggar



#### Frame

A frame is a **collection** of **attributes**(slots) and **associated values** that describe **some entity** in the world. Sometimes a frame describes an entity in some absolute sense.

Frame **describes** a particular **node** and collection of frames are connected to each other by virtue of the fact that the value of an attribute of one frame may be another frame.

## Frames as Sets and Instance

Frames represent either a class(a set) or an instance (element of a class). The attributes of a set can be about the set itself or inherited by each element of the set.

Human

isa: Mammal Cardinality: 6,000,000,00

Student

isa: **Human** Cardinality:2,000,000,000

\*height: 4-5"

Instance: Student

Height:5-10

# **THANK YOU**