



# **SNS COLLEGE OF TECHNOLOGY**



**Coimbatore-35**

**An Autonomous Institution**

Accredited by NBA – AICTE and Accredited by NAAC – UGC with 'A++' Grade  
Approved by AICTE, New Delhi & Affiliated to Anna University, Chennai

## **DEPARTMENT OF INFORMATION TECHNOLOGY**

**19CSE303 – ARTIFICIAL INTELLIGENCE**

**III YEAR IV SEM**

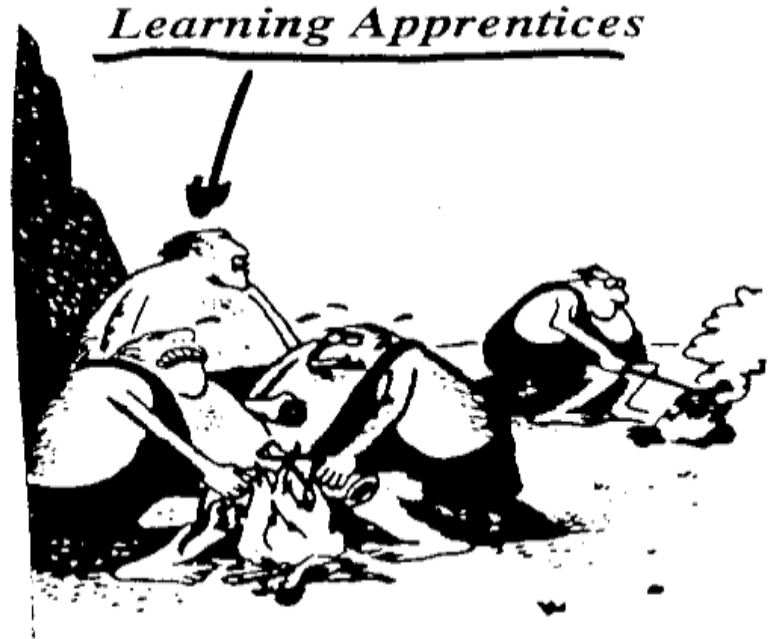
### **UNIT V – LEARNING**

**TOPIC – Explanation Based Learning**



# What is EBL ?

- Learning *general* problem-solving techniques by observing and analyzing human solutions to *specific* problems.
- EBL attempts to formulate a *generalization* after observing only a single example.
- Introduced by Gerald De Jong in 1981.



*“Hey! Look what Zog do!”*

(drawn by Gary Larson)



# The EBL Hypothesis



- EBL is based on the hypothesis that an intelligent system can learn a general concept after observing only a single example.
- By understanding why an example is a member of a concept, can learn the essential properties of the concept.
- EBL uses prior knowledge to analyze or explain each training example in order to infer what properties are relevant to the target function and which are irrelevant.



# Learning by Generalizing Explanations



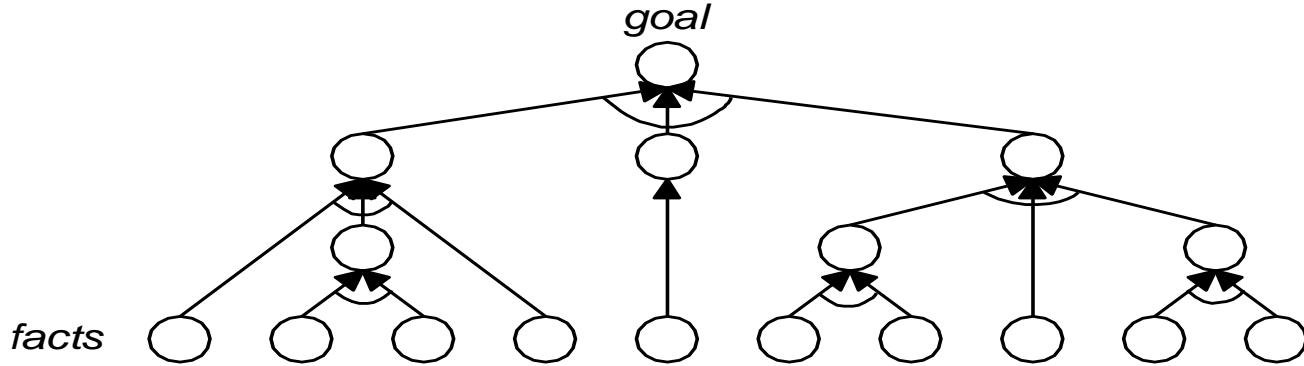
- Given
  - Goal concept (e.g., some predicate calculus statement)
  - Training example (facts)
  - Domain Theory (inference rules)
  - Operationality Criterion
- Given this four inputs, the task is to determine a generalization of the *training example* that is sufficient concept definition for the *goal concept* and that satisfies the *operationality criteria*.
- The operationality criterion requires that the final concept definition be described in terms of the predicates used to describe the training example.



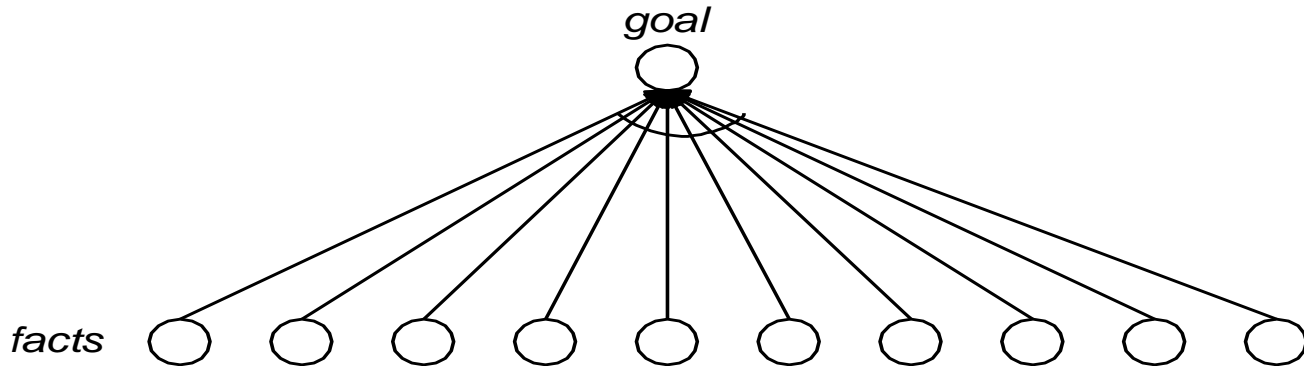
# Standard Approach to EBL



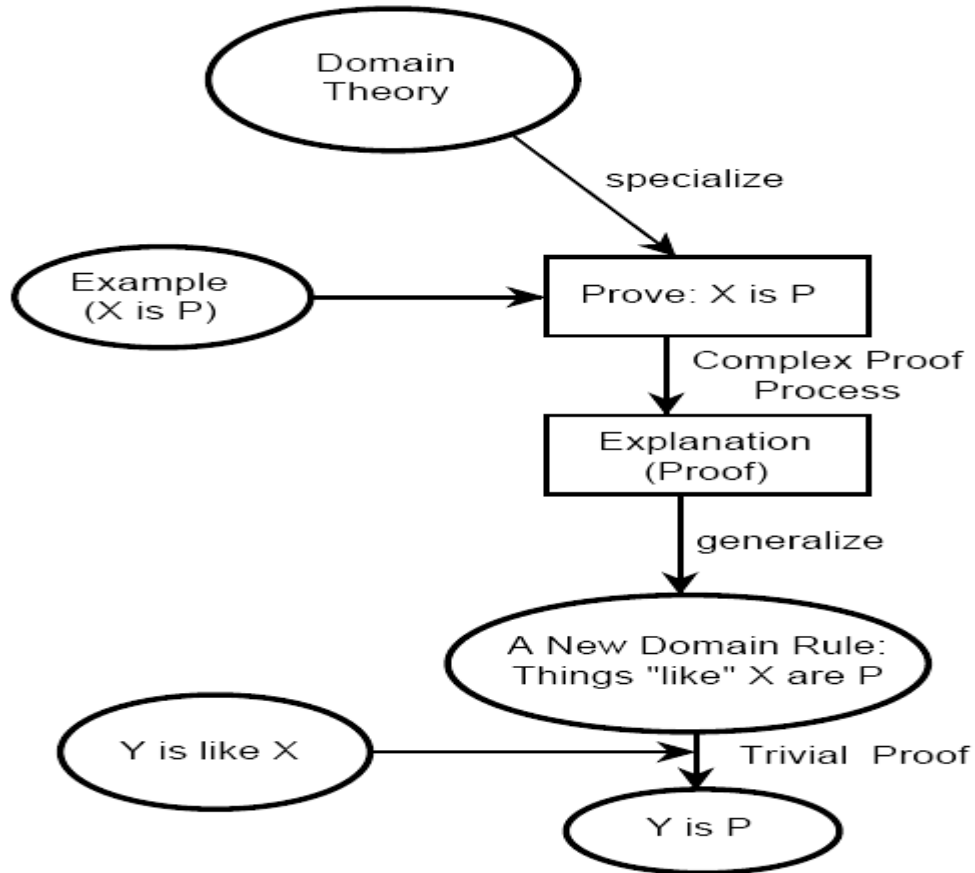
An Explanation (detailed proof of goal)



After Learning (go directly from facts to solution):



# The EBL Process



# An Example

- Domain theory:

$Fixes(u,u) \rightarrow Robust(u)$  // An individual that can fix itself is robust

$Sees(x,y) \wedge Habile(x) \rightarrow Fixes(x,y)$  // A habil individual that can see another entity can  
// fix that entity

$Robot(w) \rightarrow Sees(w,w)$  // All robots can see themselves

$R2D2(x) \rightarrow Habile(x)$  // R2D2-class in individuals are habil

... ..

- Facts:

$Robot(Num5)$

$R2D2(Num5)$

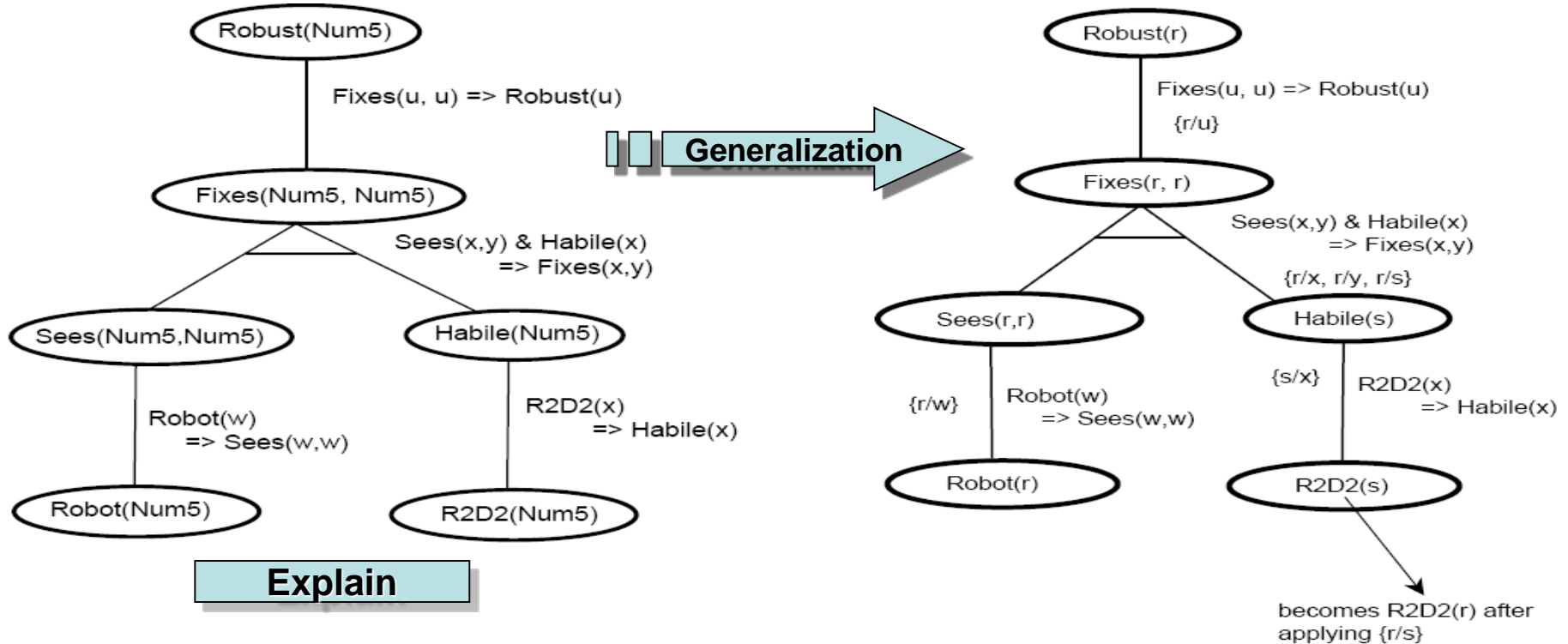
$Age(Num5,5)$

... ..

- Goal:



# An Example (continued...)



$$Robot(r) \wedge R2D2(r) \rightarrow Robust(r)$$





# History ??

- EBL may be viewed as a convergence of several distinct lines of research within machine learning.
- EBL has developed out of efforts to address each of the following problems:
  - Justified generalization.
  - Chunking.
  - Operationalization.
  - Justified analogy.



# Recommended Reading



- Mitchell T.M., Keller R.M., Kedar-Cabelli S.T., Explanation-Based Generalization: A Unifying View, *Machine Learning* 1, pp. 47-80, 1986, Kluwer Academic Publishing.
- DeJong G., Mooney R., Explanation-Based Learning: An Alternative View, *Machine Learning* 1, 1986, Kluwer Academic Publishing.
- Ellman, T, Explanation-Based Learning: A Survey of Programs and Perspectives, *ACM Computing Surveys*, Vol. 21, No. 2, 1989.



# Conclusions



- **Explanation Based Learning (EBL):**
  - Needs only one example.
  - Requires complete knowledge about the concept.
  - Shows the importance of prior knowledge in learning.