

CS354R

DR SARAH ABRAHAM

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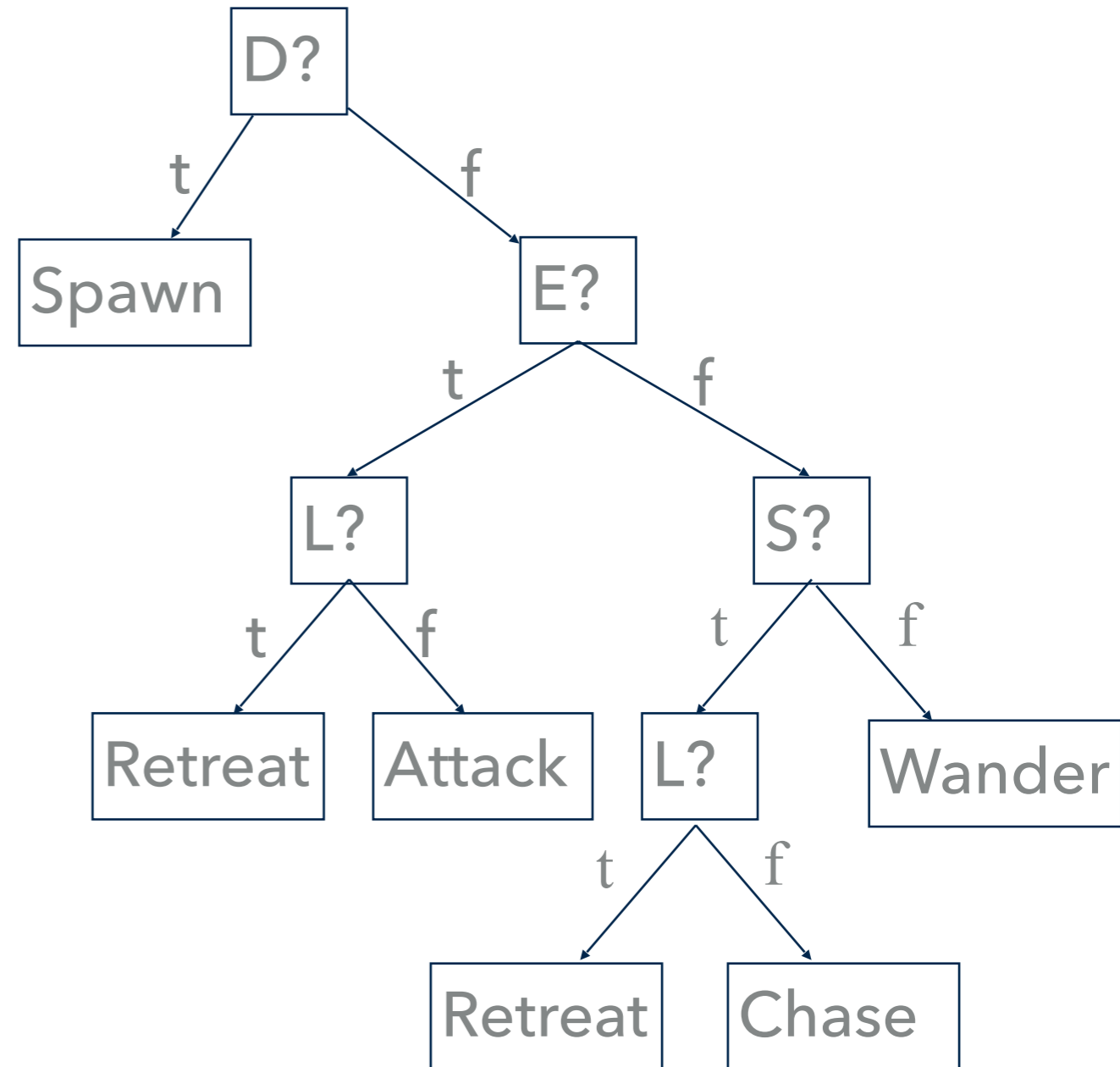
# AI DECISION TREES AND RULE SYSTEMS

# DECISION TREES

- ▶ Nodes represent attribute tests
  - ▶ One child for each outcome
- ▶ Leaves represent classifications
  - ▶ Can have same classification across leaves
- ▶ Classify by descending from root to a leaf
  - ▶ Perform test and descend
  - ▶ Return leaf's classification (action)
- ▶ Decision tree is a "disjunction of conjunctions of constraints on the attribute values of an instance"
  - ▶ Action if (A and B and C) or (A and  $\sim$ B and D) or ( ... ) ...
  - ▶ Retreat if (low health and see enemy) or (low health and hear enemy) or ( ... ) ...

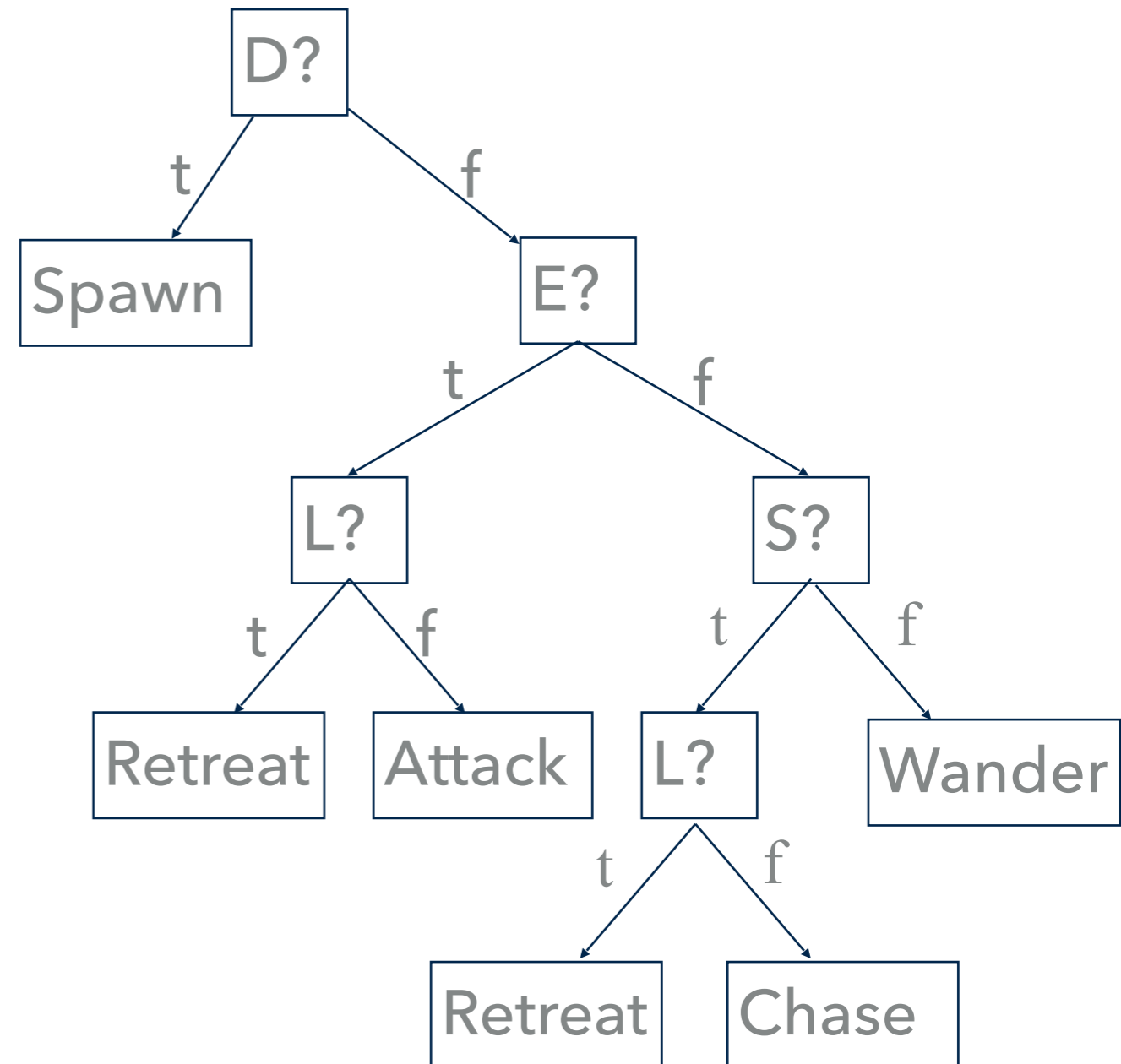
# DECISION TREE FOR QUAKE

- ▶ Just one tree
- ▶ Attributes:
  - Enemy= $\langle t, f \rangle$
  - Low= $\langle t, f \rangle$
  - Sound= $\langle t, f \rangle$
  - Death= $\langle t, f \rangle$
- ▶ Actions: Attack, Retreat, Chase, Spawn, Wander

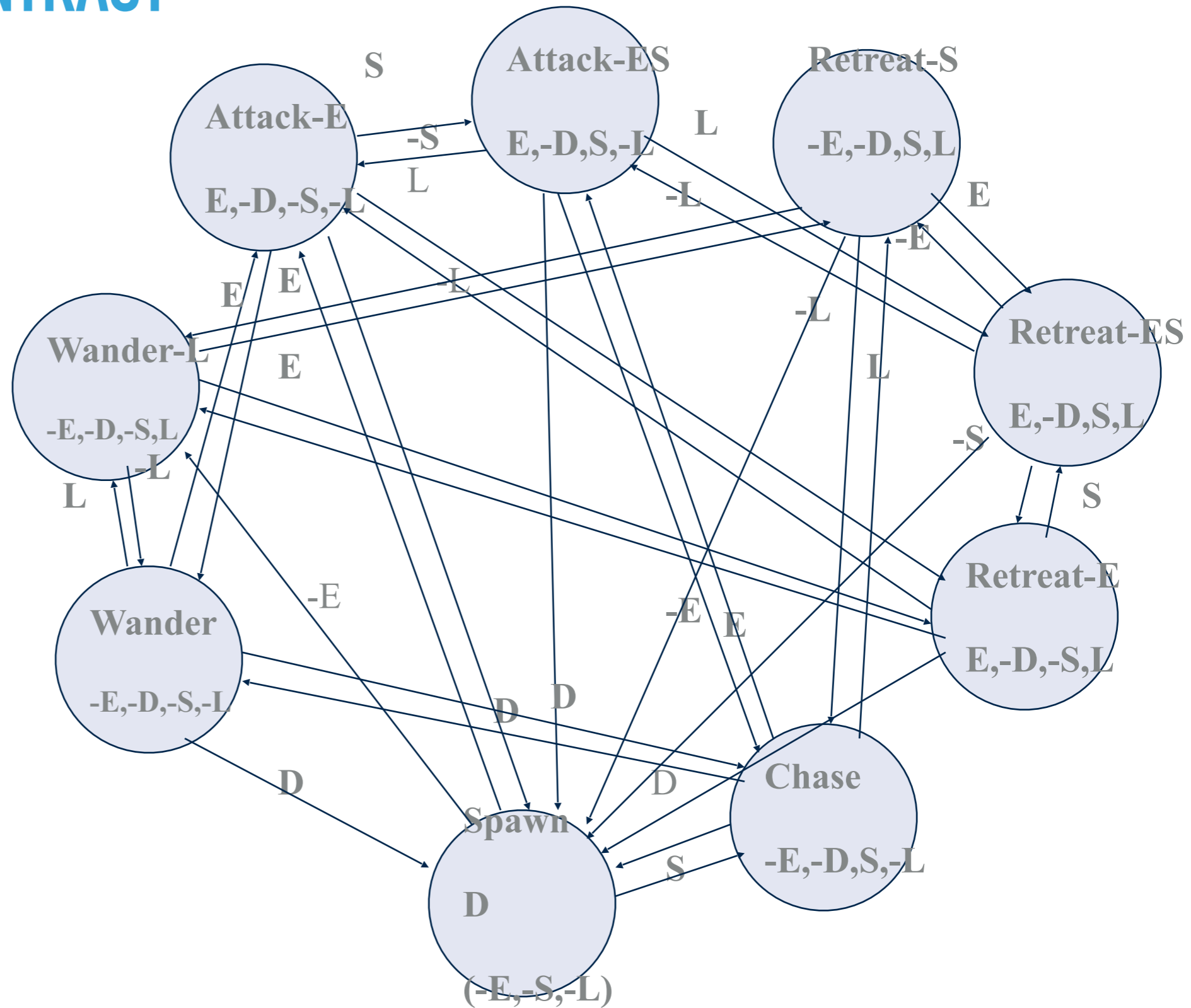


# DECISION TREE FOR QUAKE

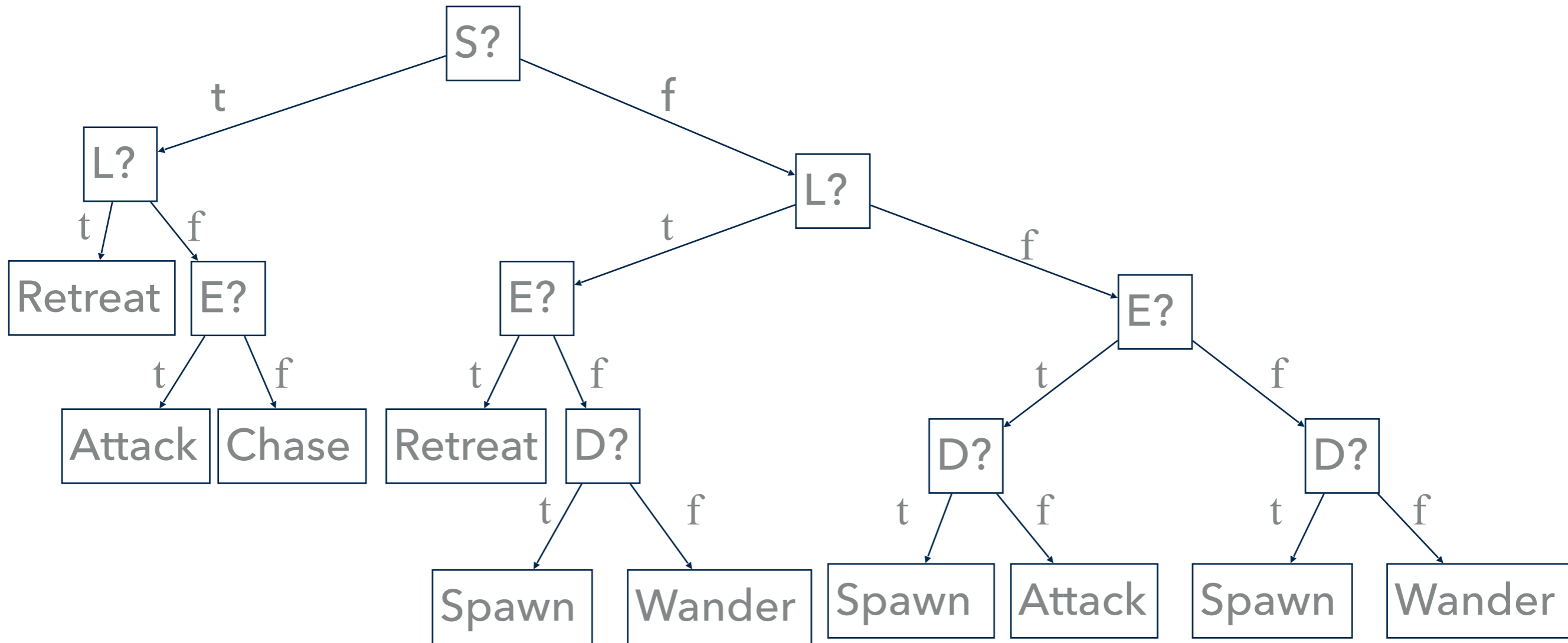
- ▶ Could add additional trees
  - ▶ If I'm attacking, which weapon should I use?
  - ▶ If I'm wandering, which way should I go?
  - ▶ Can be thought of as just extending given tree
  - ▶ Or, can share pieces of tree, such as a Retreat sub-tree



# COMPARE AND CONTRAST



# DIFFERENT TREES – SAME DECISION



# HANDLING SIMULTANEOUS ACTIONS

- ▶ Treat each output command as a separate classification problem
  - ▶ Given inputs should walk => <forward, backward, stop>
  - ▶ Given inputs should turn => <left, right, none>
  - ▶ Given inputs should run => <yes, no>
  - ▶ Given inputs should weapon => <blaster, shotgun...>
  - ▶ Given inputs should fire => <yes, no>
- ▶ Have a separate tree for each command
- ▶ If commands are not independent, two options:
  - ▶ Have a general conflict resolution strategy
  - ▶ Put dependent actions in one tree

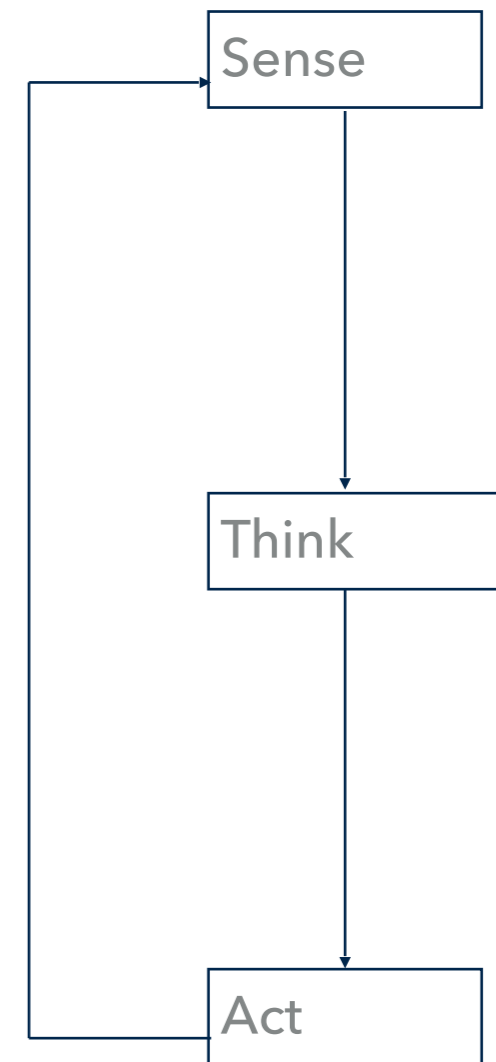
## DECIDING ON ACTIONS

- ▶ Each time the AI is called:
  - ▶ Poll each decision tree for current output
  - ▶ Event driven - only call when state changes
- ▶ Need current value of each input attribute
  - ▶ All sensor inputs describe the state of the world
- ▶ Store the state of the environment
  - ▶ Most recent values for all sensor inputs
  - ▶ Change state upon receipt of a message
  - ▶ Or, check validity when AI is updated
  - ▶ Or, a mix of both (polling and event driven)



# SENSE, THINK, ACT CYCLE

- ▶ Sense
  - ▶ Gather input sensor changes
  - ▶ Update state with new values
- ▶ Think
  - ▶ Poll each decision tree
- ▶ Act
  - ▶ Execute any changes to actions



# BUILDING DECISION TREES

- ▶ Decision trees can be constructed by hand
  - ▶ Think of the questions you would ask to decide what to do
  - ▶ For example: Tonight I can study, play games or sleep. How do I make my decision?
- ▶ But, decision trees in AI are typically *learned*:
  - ▶ Provide examples: many sets of attribute values and resulting actions
  - ▶ Algorithm then constructs a tree from the examples
  - ▶ Reasoning: We don't know how to decide on an action, so let the computer do the work

# LEARNING DECISION TREES

- ▶ Decision trees are usually learned by induction
  - ▶ Generalize from examples
  - ▶ Induction doesn't guarantee correct decision trees
- ▶ Bias towards smaller decision trees
  - ▶ Occam's Razor: Prefer simplest theory that fits the data
  - ▶ Too expensive to find the very smallest decision tree
- ▶ Learning is non-incremental
  - ▶ Need to store all the examples
- ▶ ID3 is the basic learning algorithm
  - ▶ C4.5 is an updated and extended version

# INDUCTION

- ▶ If  $X$  is true in every example that results in action  $A$ , then  $X$  must always be true for action  $A$ 
  - ▶ More examples are better
  - ▶ Errors in examples cause difficulty
    - ▶ If  $X$  is true in most examples  $X$  must always be true
    - ▶ D3 does a good job of handling errors (noise) in examples
  - ▶ Note that induction can result in errors
    - ▶ It may just be coincidence that  $X$  is true in all the examples
- ▶ Typical decision tree learning determines what tests are always true for each action
  - ▶ Assumes that if those things are true again, then the same action should result

# LEARNING ALGORITHMS

- ▶ Recursive algorithms
  - ▶ Find an attribute test that separates the actions
  - ▶ Divide the examples based on the test
  - ▶ Recurse on the subsets
- ▶ What does it mean to separate?
- ▶ Separation:
  - ▶ Ideally, there are no actions that have examples in both sets
  - ▶ Failing that, most actions have most examples in one set
  - ▶ The thing to measure is entropy - the degree of homogeneity (or lack of it) in a set
    - ▶ Entropy is also important for compression

## WHERE TO GET EXAMPLES?

- ▶ Generating examples:
  - ▶ Programmer/designer provides examples
  - ▶ Capture an expert player's actions, and the game state, while they play
- ▶ Number of examples needed depends on difficulty of concept
  - ▶ Difficulty: Number of tests needed to determine the action
  - ▶ More is always better
- ▶ Training set vs. Testing set
  - ▶ Train on most (75%) of the examples
  - ▶ Use the rest to validate the learned decision trees by estimating how well the tree does on examples it hasn't seen

## DECISION TREE ADVANTAGES

- ▶ Simpler, more compact representation
- ▶ State is recorded in a memory
  - ▶ Create "internal sensors" - Enemy-Recently-Sensed
- ▶ Easy to create and understand
- ▶ Decision trees can be learned

## DECISION TREE DISADVANTAGES

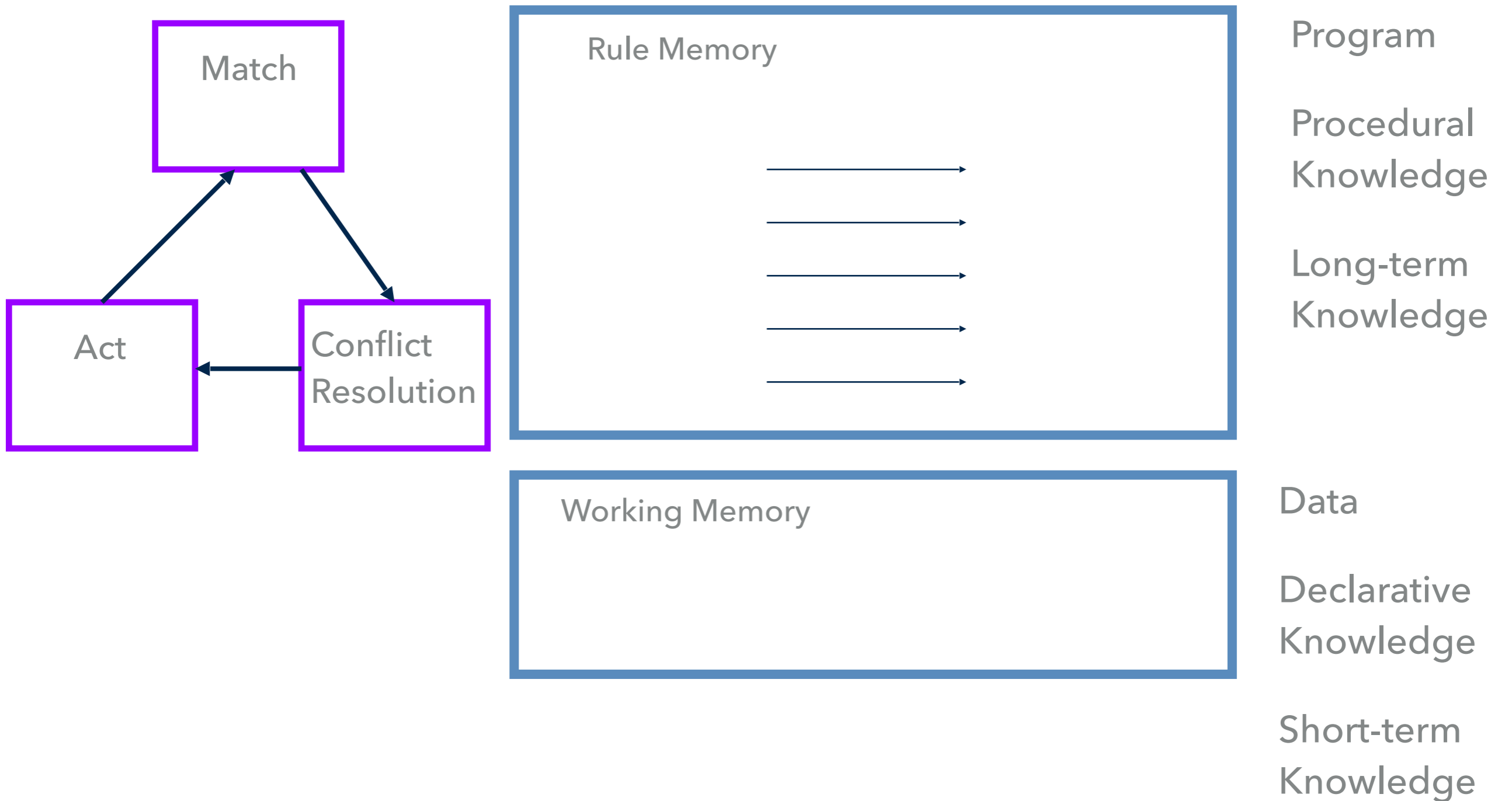
- ▶ Decision tree engine requires more coding than FSM
- ▶ Need as many examples as possible
- ▶ Higher CPU cost (but not much higher)
- ▶ Learned decision trees may contain errors



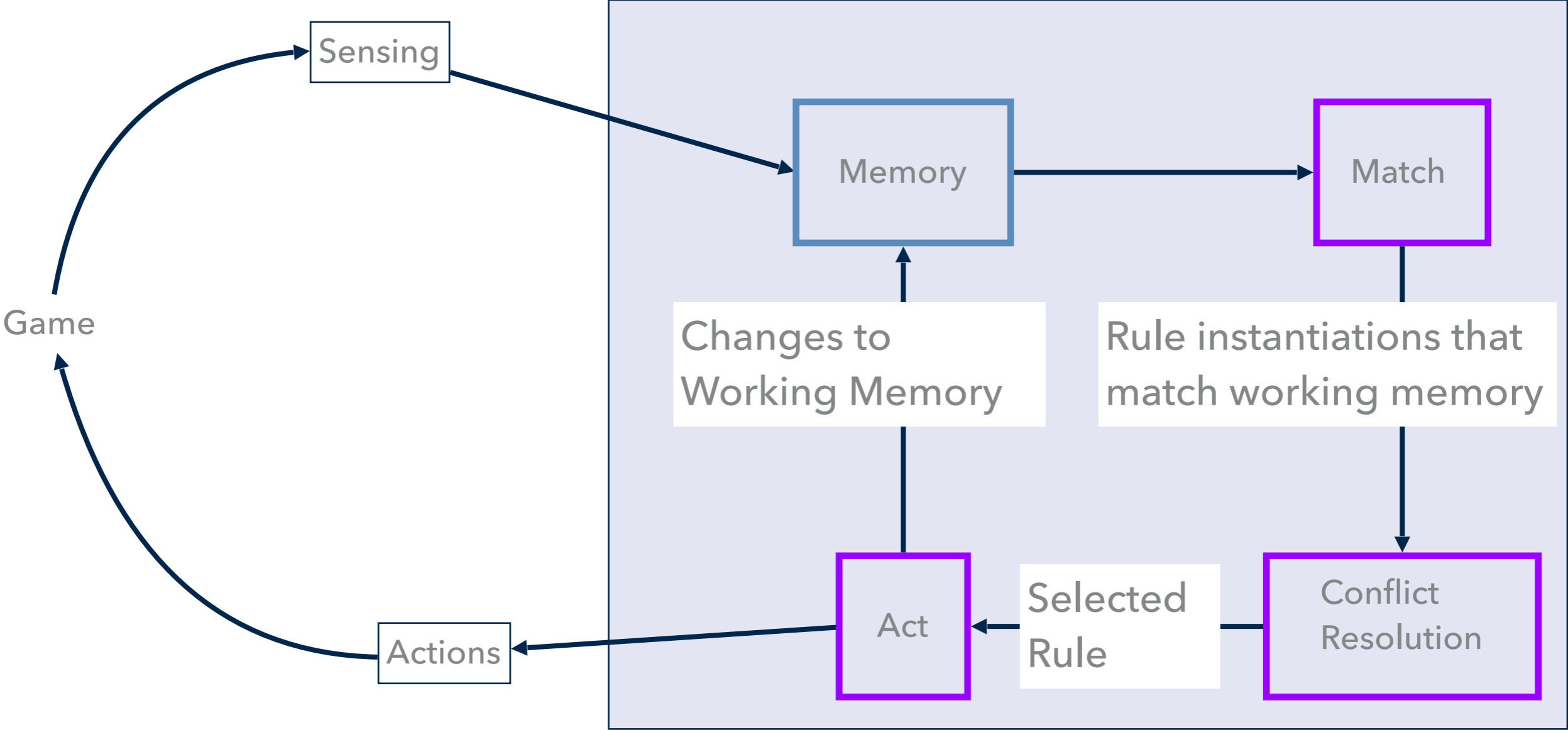
## RULE-BASED SYSTEMS

- ▶ Rule-based systems let you write the rules
  - ▶ Decision trees can be converted into rules
- ▶ System consists of:
  - ▶ A rule set - the rules to evaluate
  - ▶ A working memory - stores state
  - ▶ A matching scheme - decides which rules are applicable
  - ▶ A conflict resolution scheme - if more than one rule is applicable, decides how to proceed
- ▶ What types of games make the most extensive use of rules?

# RULE-BASED SYSTEMS STRUCTURE



# AI CYCLE



# AGE OF KINGS

; The AI will attack once at 1100 seconds and then again  
; every 1400 sec, provided it has enough defense soldiers.

```
(defrule  
  (game-time > 1100) ← Rule  
=>  
  (attack-now)  
  (enable-timer 7 1400)) } Action
```

```
(defrule  
  (timer-triggered 7)  
  (defend-soldier-count >= 12)  
=>  
  (attack-now)  
  (disable-timer 7)  
  (enable-timer 7 1400))
```

# AGE OF KINGS

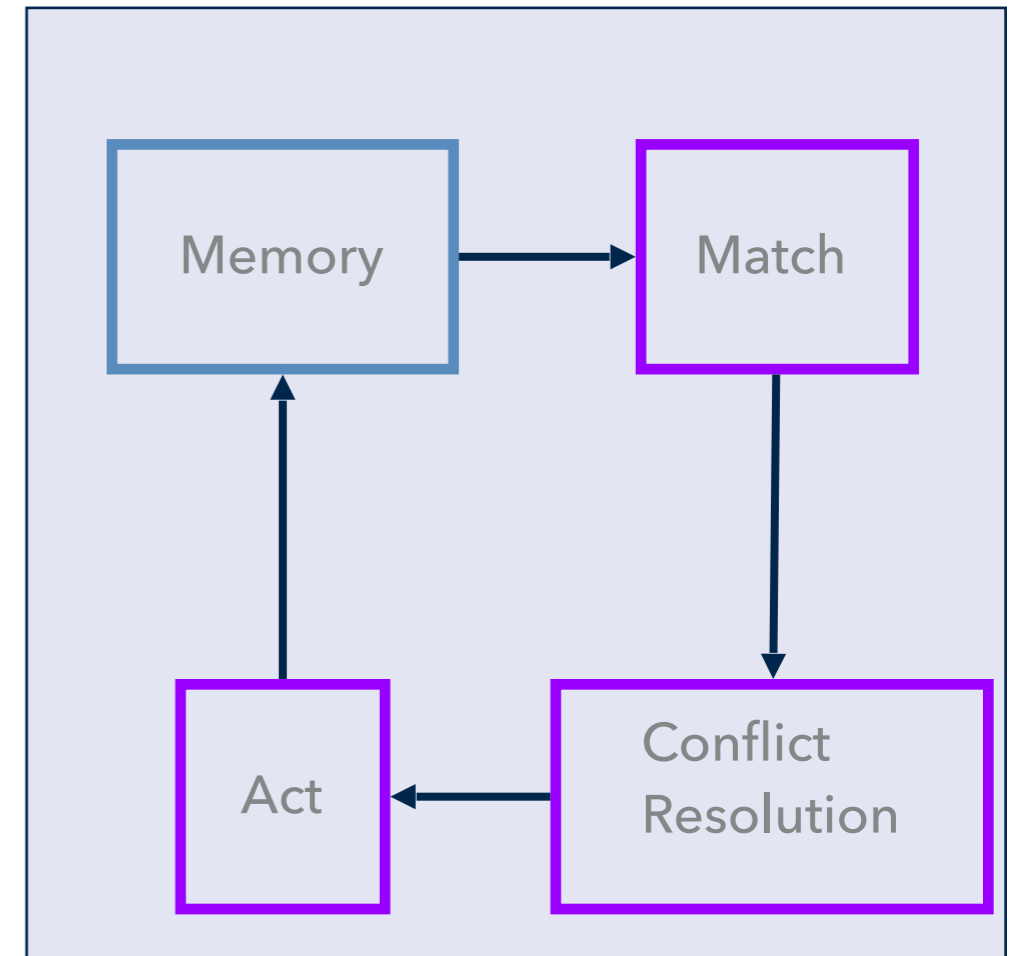
## ▶ What is it doing?

```
(defrule  
  (true)  
=>  
  (enable-timer 4 3600)  
  (disable-self))
```

```
(defrule  
  (timer-triggered 4)  
=>  
  (cc-add-resource food 700)  
  (cc-add-resource wood 700)  
  (cc-add-resource gold 700)  
  (disable-timer 4)  
  (enable-timer 4 2700)  
  (disable-self))
```

# IMPLEMENTING RULE-BASED SYSTEMS

- ▶ Where does the time go?
- ▶ 90-95% goes to Match
  - ▶ Matching all rules against all of working memory each cycle is way too slow
- ▶ Key observation
  - ▶ # of changes to working memory each cycle is small
  - ▶ If conditions, and hence rules, can be associated with changes, then we can make things fast (event-driven)



## GENERAL CASE

- ▶ Rules can be arbitrarily complex
  - ▶ In particular: function calls in conditions and actions
- ▶ If we have arbitrary function calls in conditions:
  - ▶ Run through rules one at a time and test conditions
  - ▶ Pick the first one that matches (or do something else)
- ▶ Time to match depends on:
  - ▶ Number of rules
  - ▶ Complexity of conditions
  - ▶ Number of rules that don't match

## RESOLVING MULTIPLE MATCHES?

- ▶ Rule order - pick the first rule that matches
  - ▶ Makes order of loading important - not good for big systems
- ▶ Rule specificity - pick the most specific rule
- ▶ Rule importance - pick rule with highest priority
  - ▶ When a rule is defined, give it a priority number
  - ▶ Forces a total order on the rules - is right 80% of the time
  - ▶ Decide Rule 4 [80] is better than Rule 7 [70]
  - ▶ Decide Rule 6 [85] is better than Rule 5 [75]
  - ▶ Enforces ordering between all of them



## REDUCING COST OF MATCHING

- ▶ Save intermediate match information (RETE)
  - ▶ Memory intensive
  - ▶ Fast search
  - ▶ DAGs that represent high-level rule sets
  - ▶ Tuples of facts matched against hierarchy of rules
  - ▶ Relevant facts asserted in working memory
- ▶ Recompute match for rules affected by change (TREAT)
  - ▶ Memory efficient
  - ▶ May be faster than RETE
- ▶ Make extensive use of hashing (mapping between memory and tests/rules)

## RULE-BASED SYSTEM: ADVANTAGES

- ▶ Corresponds to way people often think of knowledge
- ▶ Very expressive
- ▶ Modular knowledge
  - ▶ Easier to write and debug compared to decision trees
  - ▶ More concise than FSMs

## RULE-BASED SYSTEM: DISADVANTAGES

- ▶ Can be memory intensive
- ▶ Can be computationally intensive
- ▶ Can be difficult to debug

## FURTHER READING

- ▶ RETE:

- ▶ Forgy, C. L. Rete: A fast algorithm for the many pattern/  
many object pattern match problem. *Artificial  
Intelligence*, 19(1) 1982, pp. 17-37

- ▶ TREAT:

- ▶ Miranker, D. TREAT: A new and efficient match algorithm  
for AI production systems. Pittman/Morgan Kaufman,  
1989