



Basic Game AI

Technical Game Development II

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With material from: Millington and Funge, *Artificial Intelligence for Games*, Morgan Kaufmann 2009. (Chapter 5)

Definitions?

- What is artificial intelligence (AI) ?
 - subfield of computer science ?
 - subfield of cognitive science ?
- What is “AI for Games” ?
 - versus “academic AI” ?

In games, *everything* (including the AI) is in service of the *player's* experience (“fun”)

- What does it mean for a game AI to “cheat”?

Resources: introduction to Buckland, www.gameai.com,
aigamedev.com, www.aiwisdom.com, www.ai4games.org

IMGD 4100, B? Term 2013



What's the AI part of a game?

- Everything that isn't graphics (sound) or networking... 😊
 - or physics (though sometimes lumped in)
 - usually via the non-player characters
 - but sometimes operates more broadly, e.g.,
 - Civilization-style games (sophisticated simulations)
 - interactive storytelling (drama control)

“Levels” of Game AI

- *Basic*
 - decision-making techniques commonly used in almost all games
- *Advanced*
 - used in practice, but in more sophisticated games
- *Future*
 - not yet used, but explored in research

This course

- **Basic** game AI

- decision-making techniques commonly used in almost all games

- basic pathfinding (A*) *(IMGD 3000)*
- decision trees *(today)*
- (hierarchical) state machines *(today)*

- **Advanced** game AI

- used in practice, but in more sophisticated games
 - advanced pathfinding *(next Thurs)*
 - behavior trees (in Halo 3) *(next Fri)*

Future Game AI ?

- Take IMGD 4100 in 2013 (B?) [alt yr course]
“AI for Interactive Media and Games”
 - fuzzy logic
 - more goal-driven agent behavior
- Take CS 4341 “Artificial Intelligence”
 - machine learning
 - planning



Two Fundamental Types of AI Algorithms

- *Non-Search* vs. *Search*
 - *Non-Search*: amount of computation is predictable
 - e.g., decision trees, state machines
 - *Search*: upper bound depends on size of search space (often large)
 - e.g., minimax, planning
 - scary for real-time games
 - need to otherwise limit computation (e.g., threshold)
- Where's the “knowledge”?
 - *Non-Search*: in the code logic (or external tables)
 - *Search*: in state evaluation and search order functions

How about AI Middleware (“AI Engines”)?

- Rercent panel at GDC AI Summit: “Why so wary of AI middleware?”
- Only one panelist reported completely positive experience
 - Steve Gargolinski, Blue Fang (Zoo Tycoon, etc.)
 - Used Havok Behavior (with Physics)
- Most industry AI programmers still mostly write their own AI from scratch (or reuse their own code)
- So we are going to look at coding details

AI Coding Theme (for Basic AI)

- Use *object-oriented* paradigm

instead of...

- A tangle of *if-then-else* statements

First Basic AI Technique:

Decision Trees

See code at:

<https://github.com/idmillington/aicore>

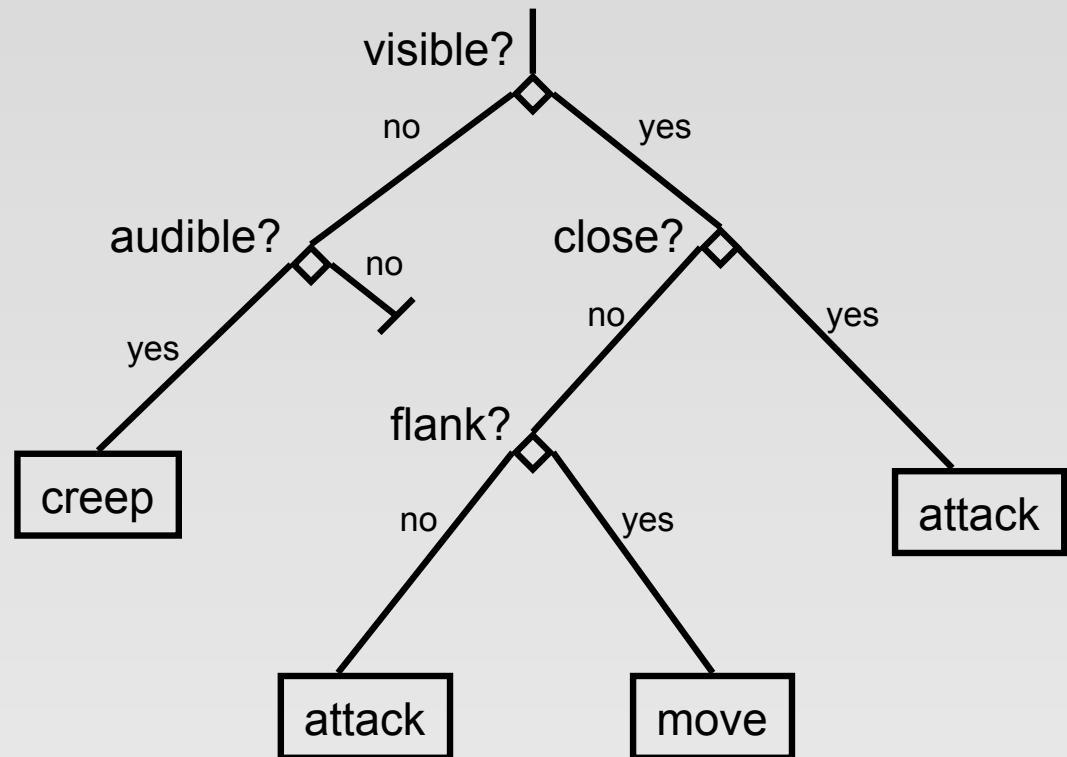
src/dectree.cpp and src/demos/c05-dectree

Decision Trees

- The most basic of the basic AI techniques
- Easy to implement
- Fast execution
- Simple to understand

Deciding how to respond to an enemy

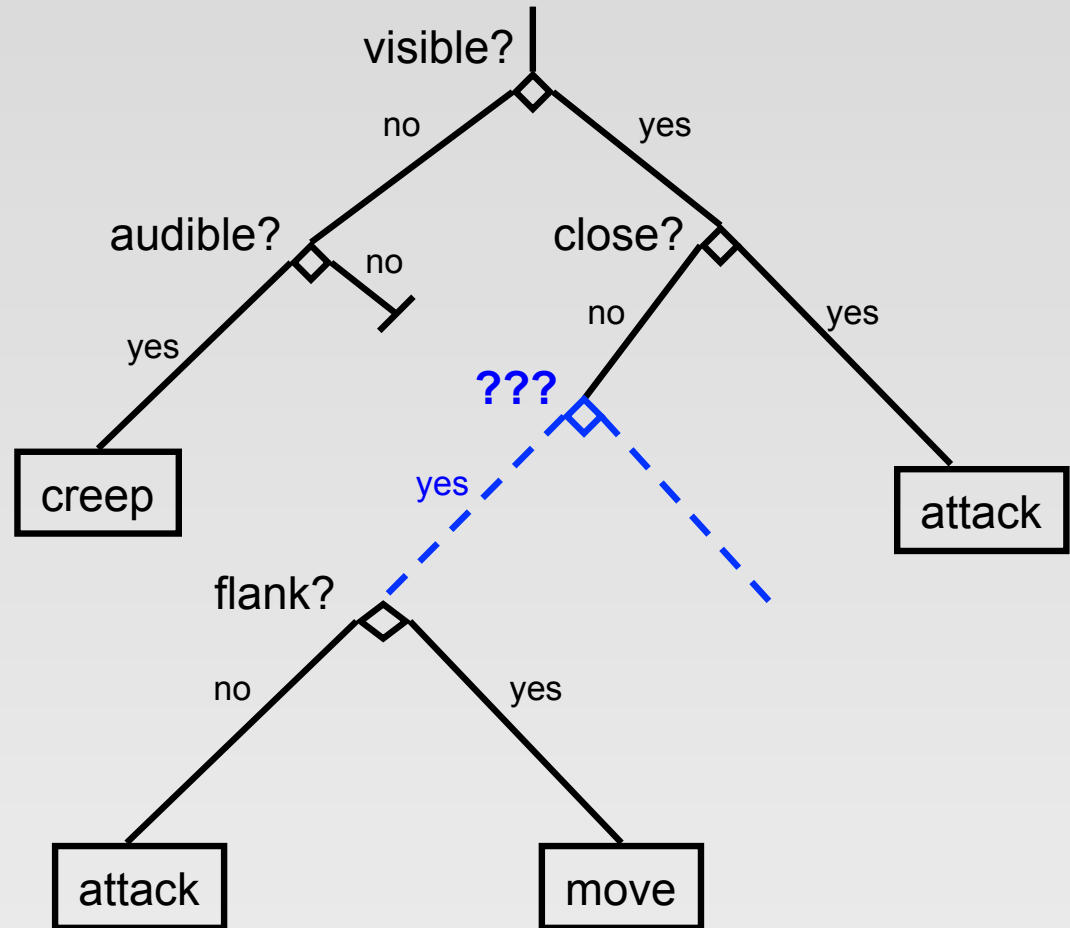
```
if visible? {  
  if close? {  
    attack;  
  } else {  
    if flank? {  
      move;  
    } else {  
      attack;  
    }  
  }  
} else {  
  if audible? {  
    creep;  
  }  
}
```



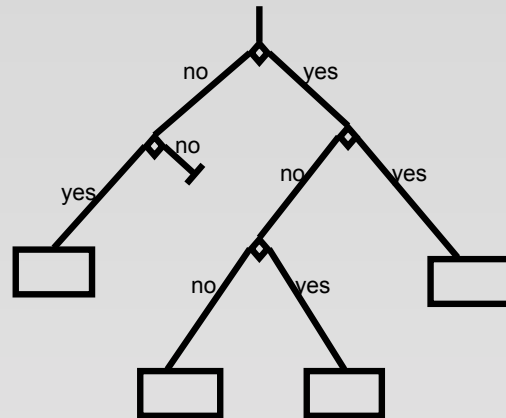
Which would you rather modify?

```
if visible? {  
  if close? {  
    attack;  
  } else if flank? {  
    move;  
  } else {  
    attack;  
  }  
}  
else if audible? {  
  creep;  
}
```

???



O-O Decision Trees (Pseudo-Code)



```
class Node
  def decide() //return action
```

```
class Decision : Node
  def getBranch() //return node
  def decide()
    return getBranch().decide()
```

```
class Action : Node
  def decide() return this
```

```
class Boolean : Decision
  yesNode
  noNode
```

```
class MinMax : Boolean
  minValue
  maxValue
  testValue
```

```
def getBranch()
  if maxValue >= testValue >= minValue
    return yesNode
  else return noNode
```



Building an O-O Decision Tree

```
visible = new Boolean...  
audible = new Boolean...  
close = new MinMax...  
flank = new Boolean...
```

```
attack = new Move...  
move = new Move...  
creep = new Move...
```

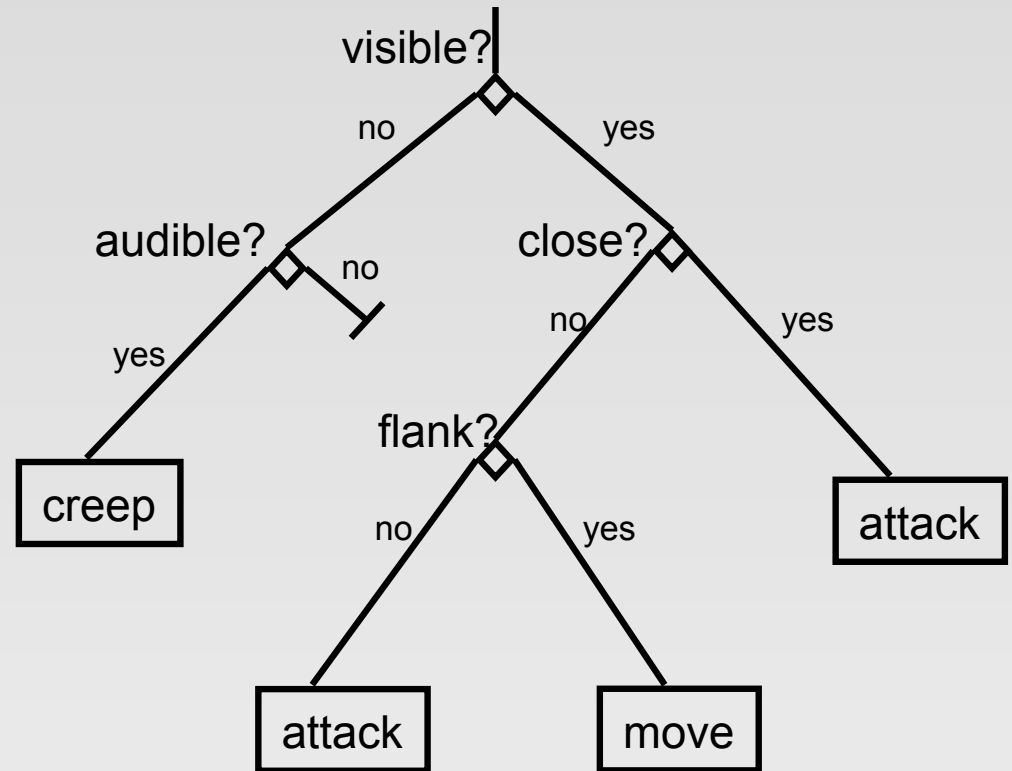
```
visible.yesNode = close  
visible.noNode = audible
```

```
audible.yesNode = creep
```

```
close.yesNode = attack  
close.noNode = flank
```

```
flank.yesNode = move  
flank.noNode = attack
```

...



...or a graphical editor

Modifying an O-O Decision Tree

visible = new Boolean...
audible = new Boolean...
close = new MinMax...
flank = new Boolean...
??? = new Boolean...

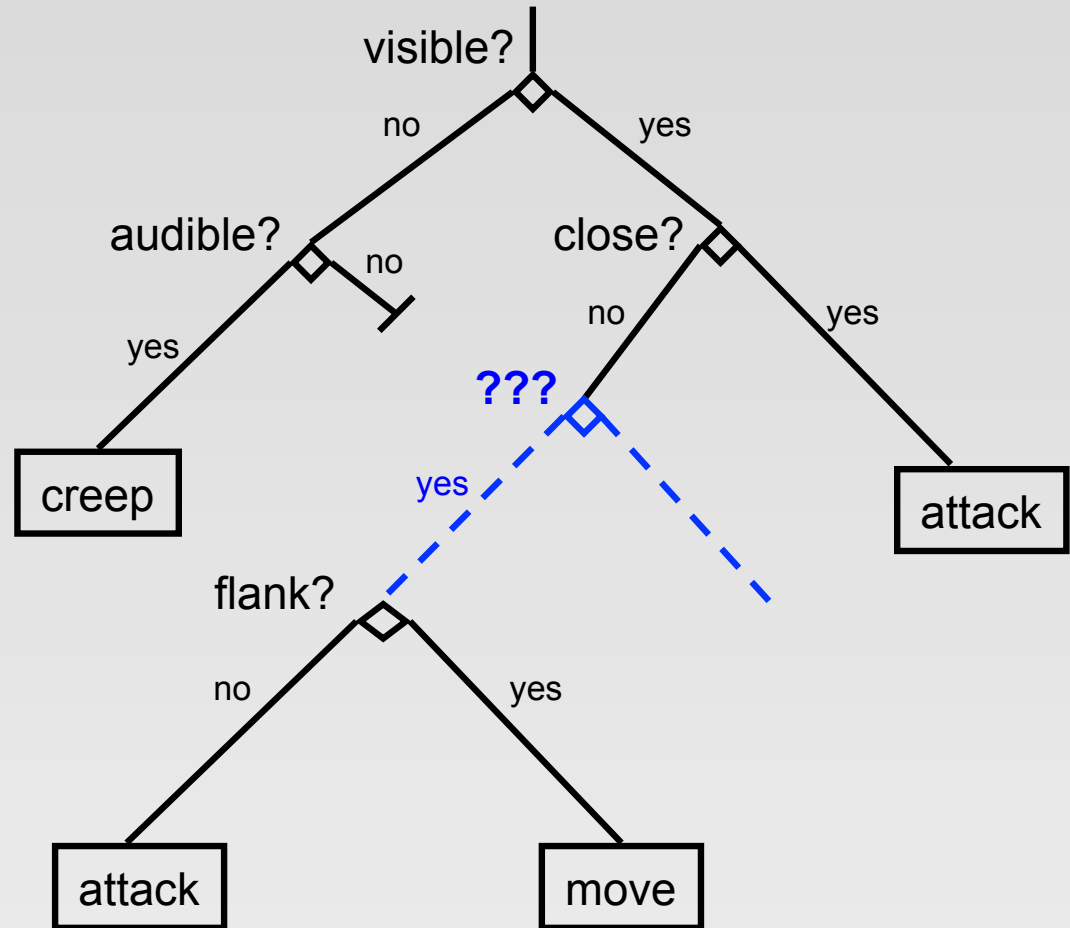
attack = new Move...
move = new Move...
creep = new Creep...

visible.yesNode = close
visible.noNode = audible

audible.yesNode = creep

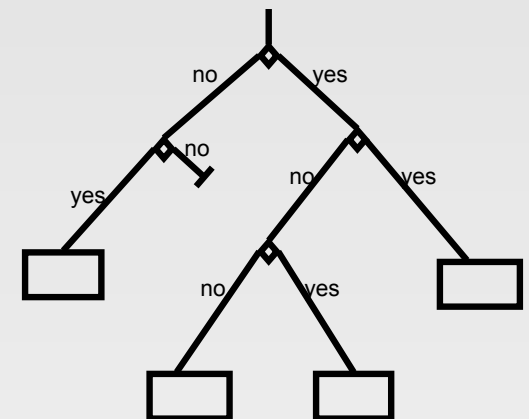
close.yesNode = attack
close.noNode = ???
???.yesNode = flank

flank.yesNode = move
flank.noNode = attack



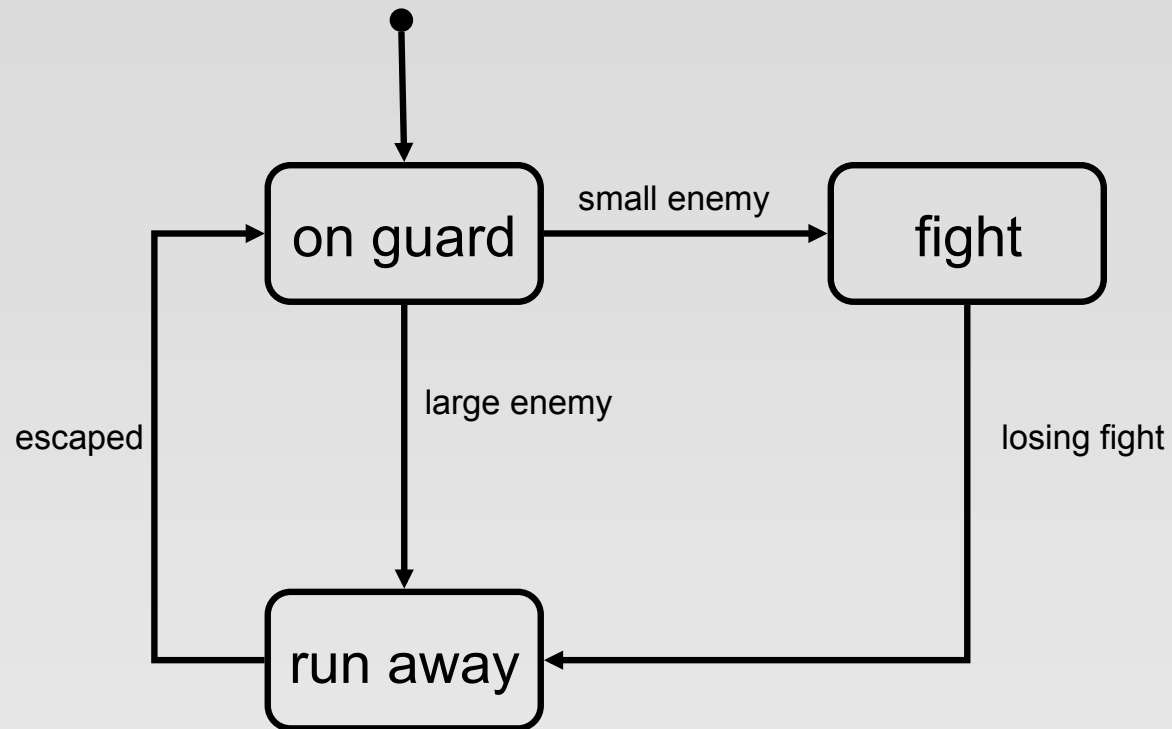
Decision Tree Performance Issues

- individual node tests (getBranch) typically constant time (and *fast*)
- worst case behavior depends on *depth* of tree
 - longest path from root to action
- roughly “balance” tree (when possible)
 - not too deep, not too wide
 - make commonly used paths shorter
 - put most *expensive* decisions late



Second Basic AI Technique: (Hierarchical) State Machines

State Machines

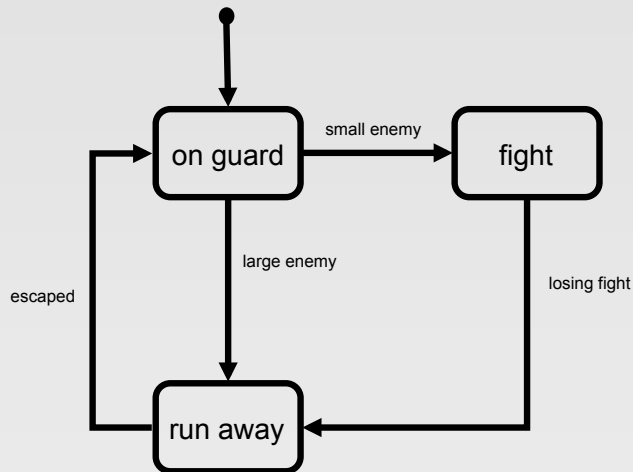


Hard-Coded Implementation

```
class Soldier
```

```
    enum State  
        ON_GUARD  
        FIGHT  
        RUN_AWAY
```

```
    currentState
```



```
def update()
```

```
    if currentState == ON_GUARD {  
        if small enemy {  
            currentState = FIGHT  
            start Fighting  
        } else if big enemy {  
            currentState = RUN_AWAY  
            start RunningAway  
        }  
    } else if currentState == FIGHT {  
        if losing fight {  
            currentState = RUN_AWAY  
            start RunningAway  
        }  
    } else if currentState == RUN_AWAY {  
        if escaped {  
            currentState = ON_GUARD  
            start Guarding  
        }  
    }  
}
```

Hard-Coded State Machines

- Easy to write (at the start)
- Very efficient
- Notoriously hard to maintain (e.g., debug)

Cleaner & More Flexible O-O Implementation

```
class State
  def getAction()
  def getEntryAction()
  def getExitAction()
  def getTransitions()
```

```
class Transition
  def isTriggered()
  def getTargetState()
  def getAction()
```

...add tracing

```
class StateMachine
```

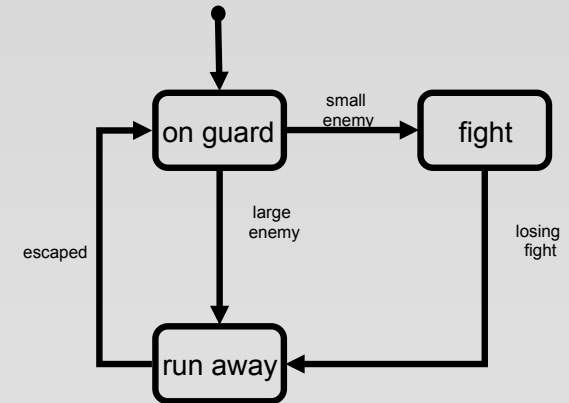
```
  states
  initialState
  currentState = initialState
```

```
  def update()
```

```
    triggeredTransition = null
```

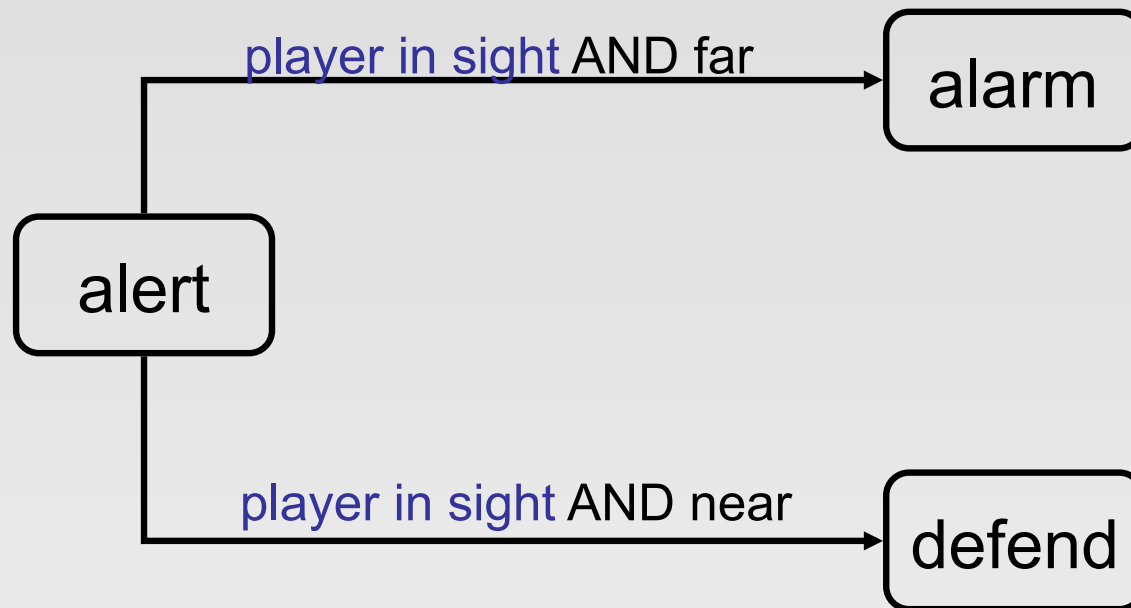
```
    for transition in currentState.getTransitions() {
      if transition.isTriggered() {
        triggeredTransition = transition
        break
      }
    }
```

```
    if triggeredTransition != null {
      targetState = triggeredTransition.getTargetState()
      actions = currentState.getExitAction()
      actions += triggeredTransition.getAction()
      actions += targetState.getEntryAction()
      currentState = targetState
      return actions
    } else return currentState.getAction()
```

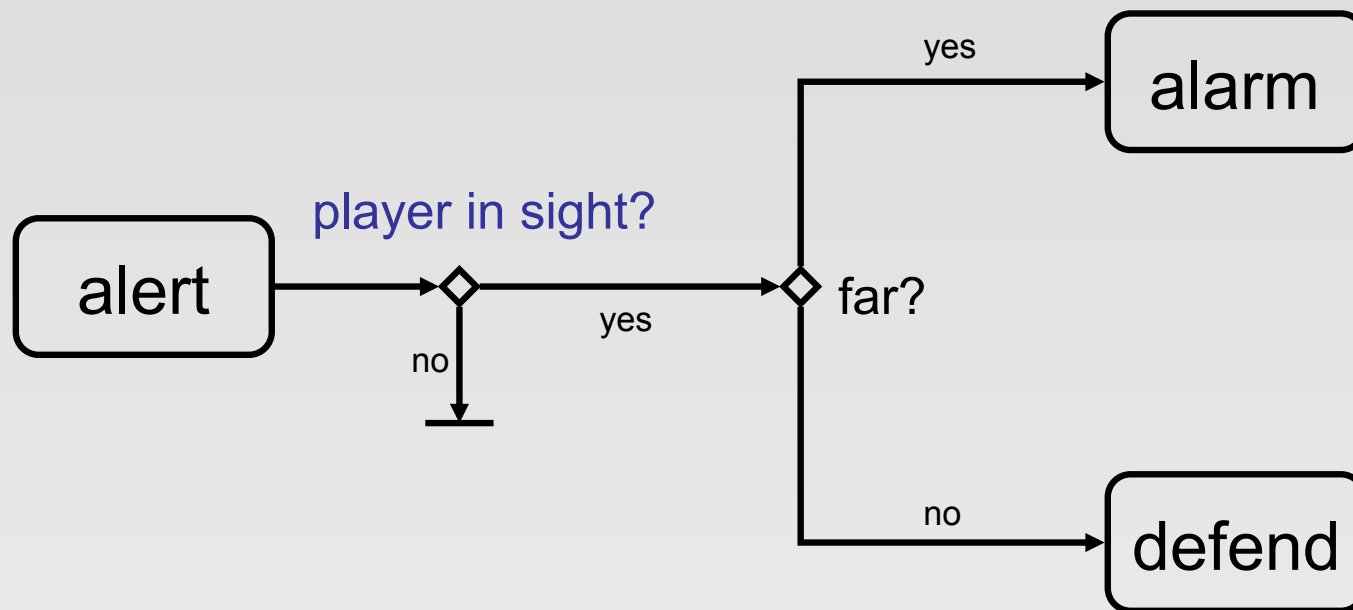


Combining Decision Trees & State Machines

- Why?
 - to avoid duplicating **expensive** tests in state machine:

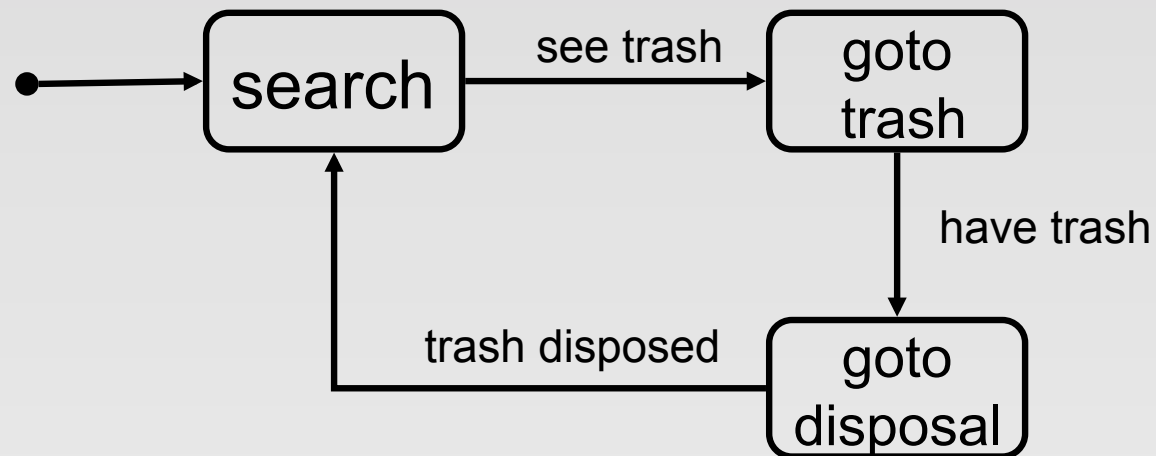


Combining Decision Trees & State Machines

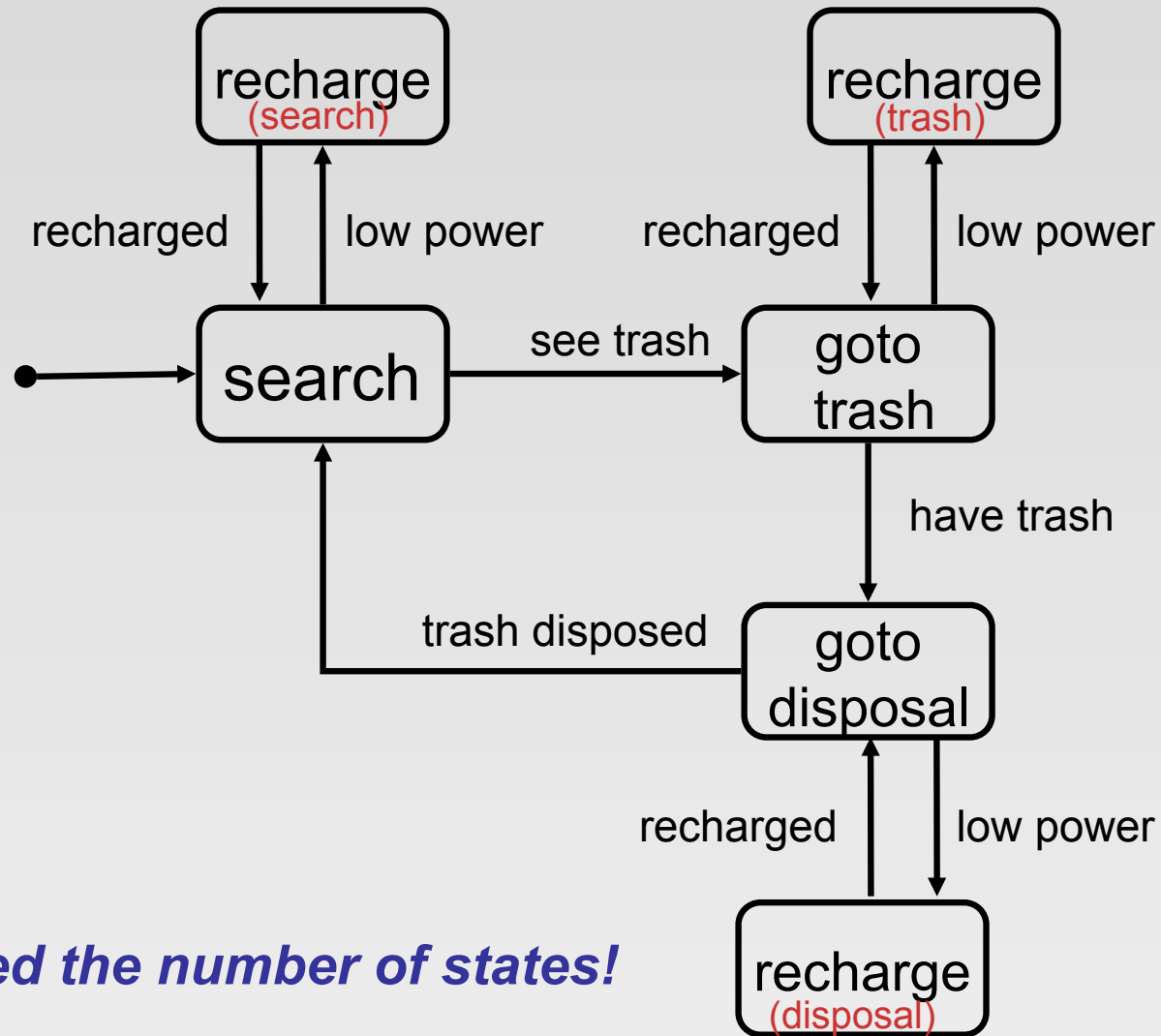


Hierarchical State Machines

- Why?

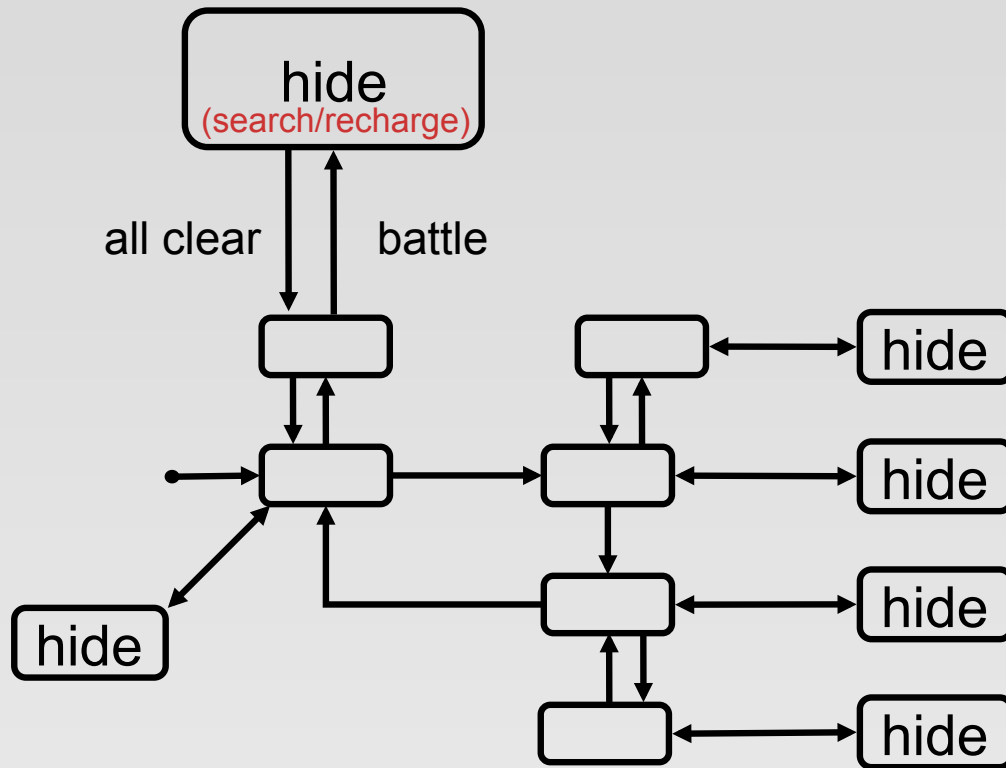


Interruptions (Alarms), e.g., Recharging?



6 - doubled the number of states!

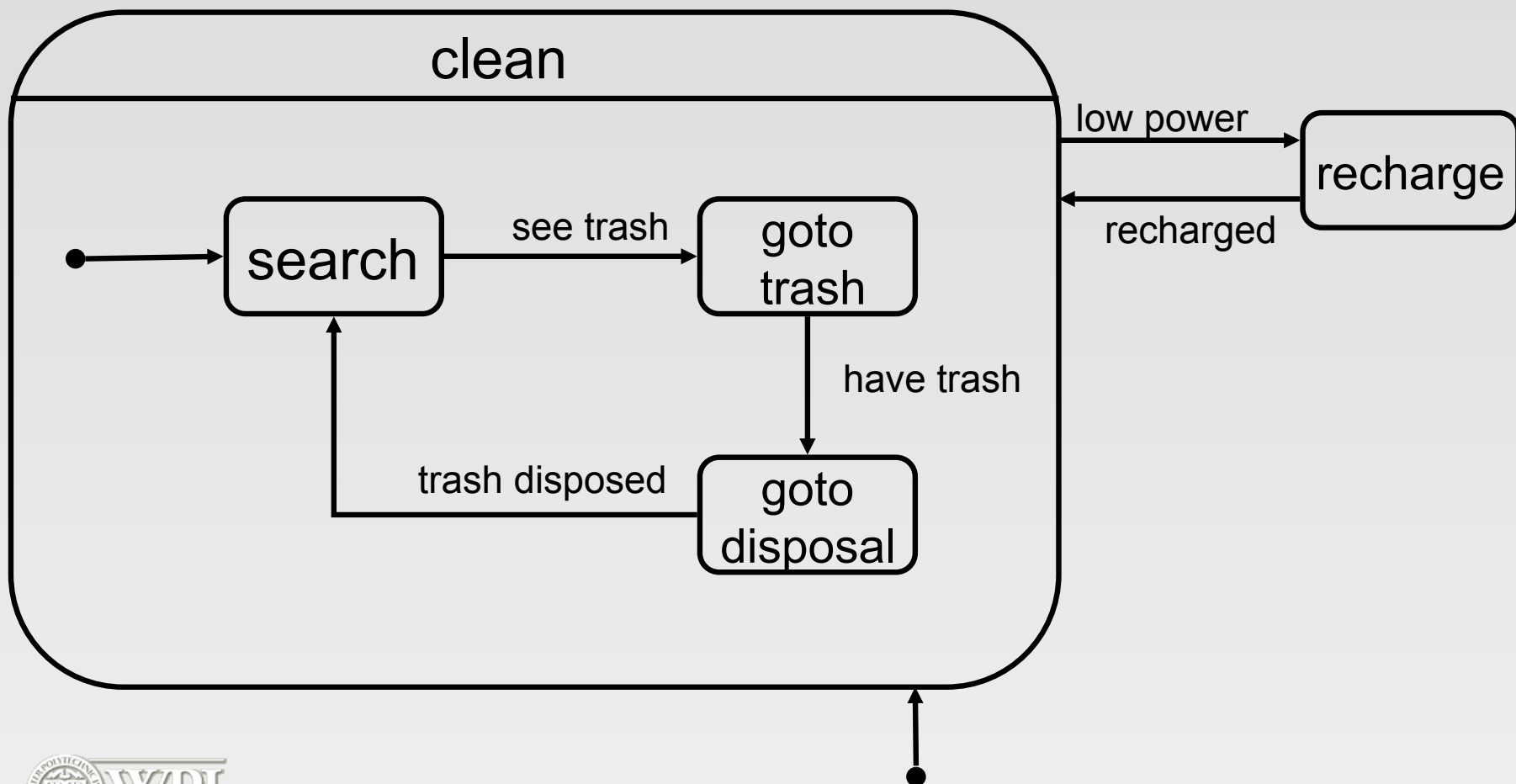
Add Another Interruption Type?



12 - doubled the number of states again!

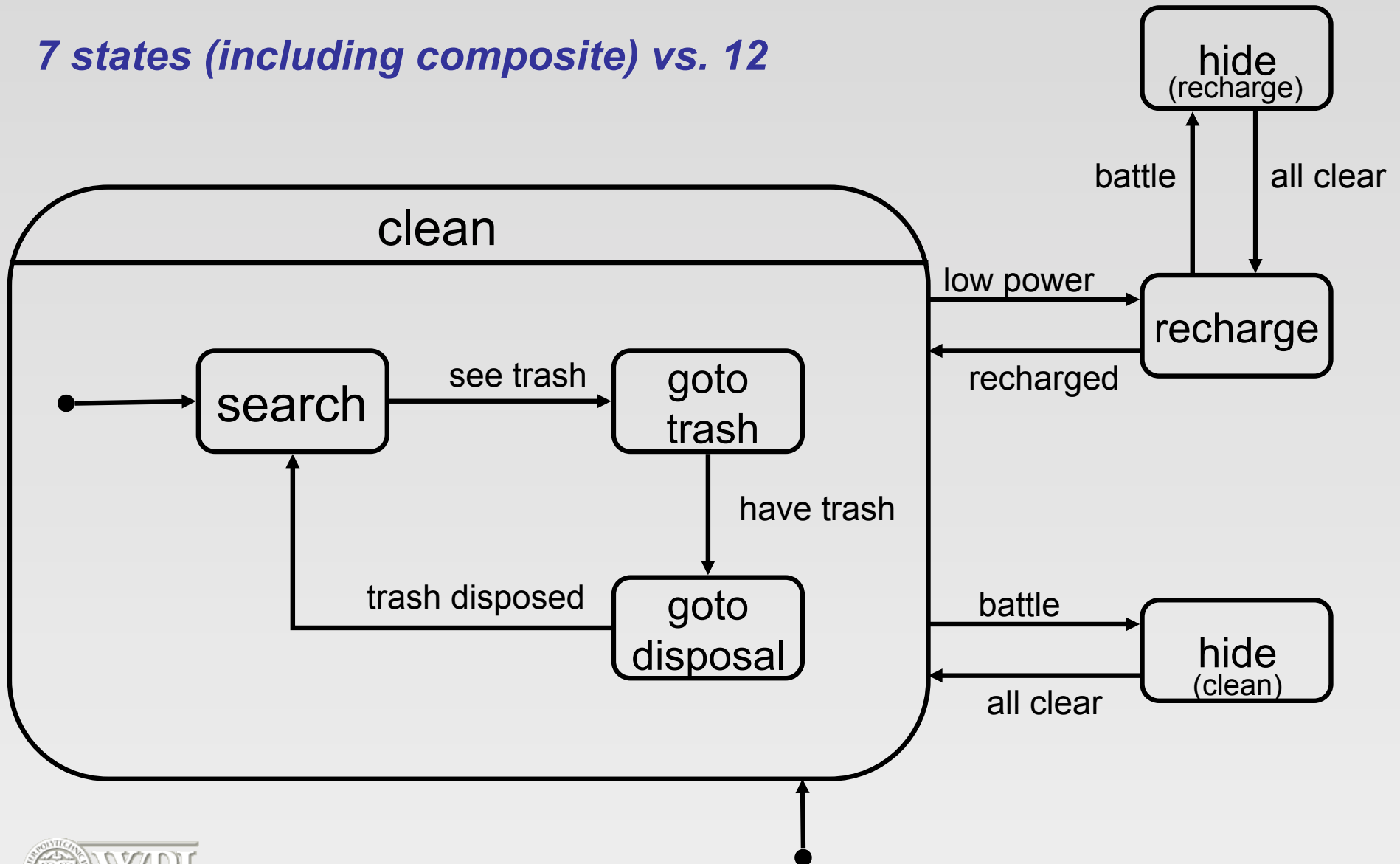
Hierarchical State Machine

- leave any state in (composite) 'clean' state when 'low power'
- 'clean' remembers internal state and continues when returned to from "recharged"



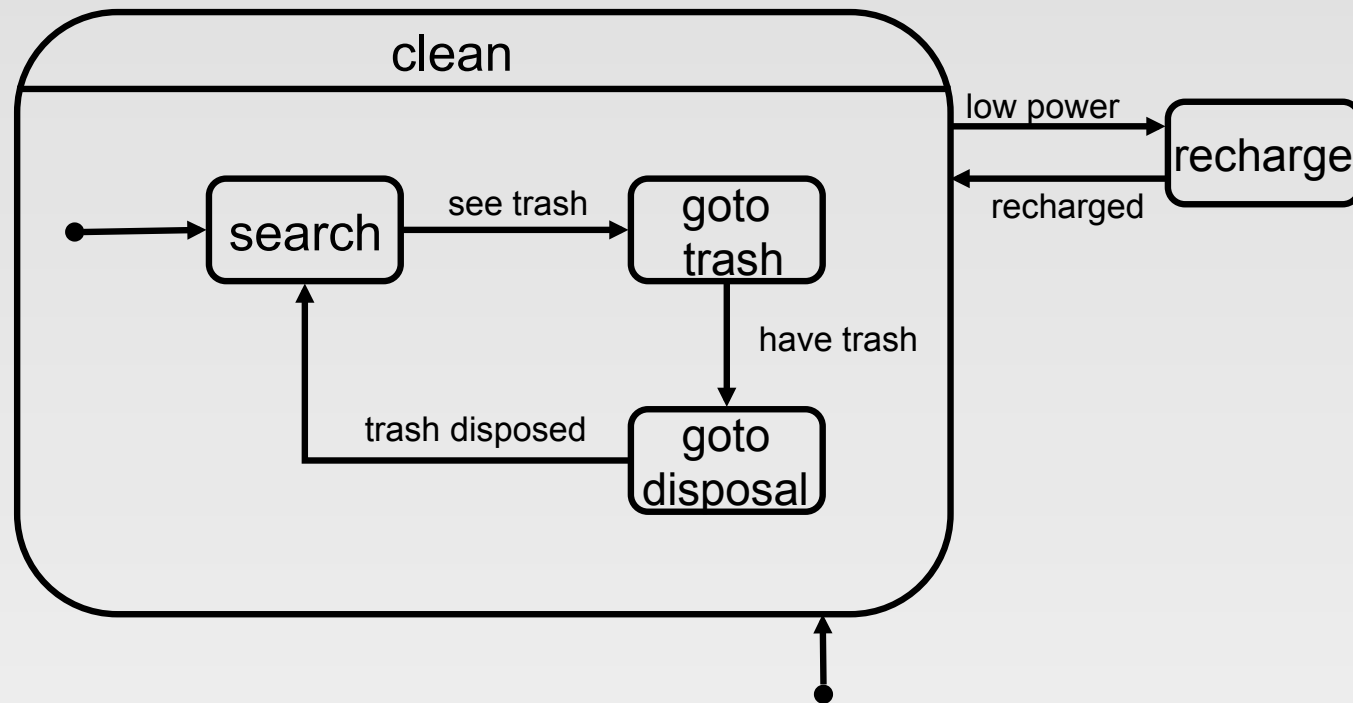
Add Another Interruption Type?

7 states (including composite) vs. 12

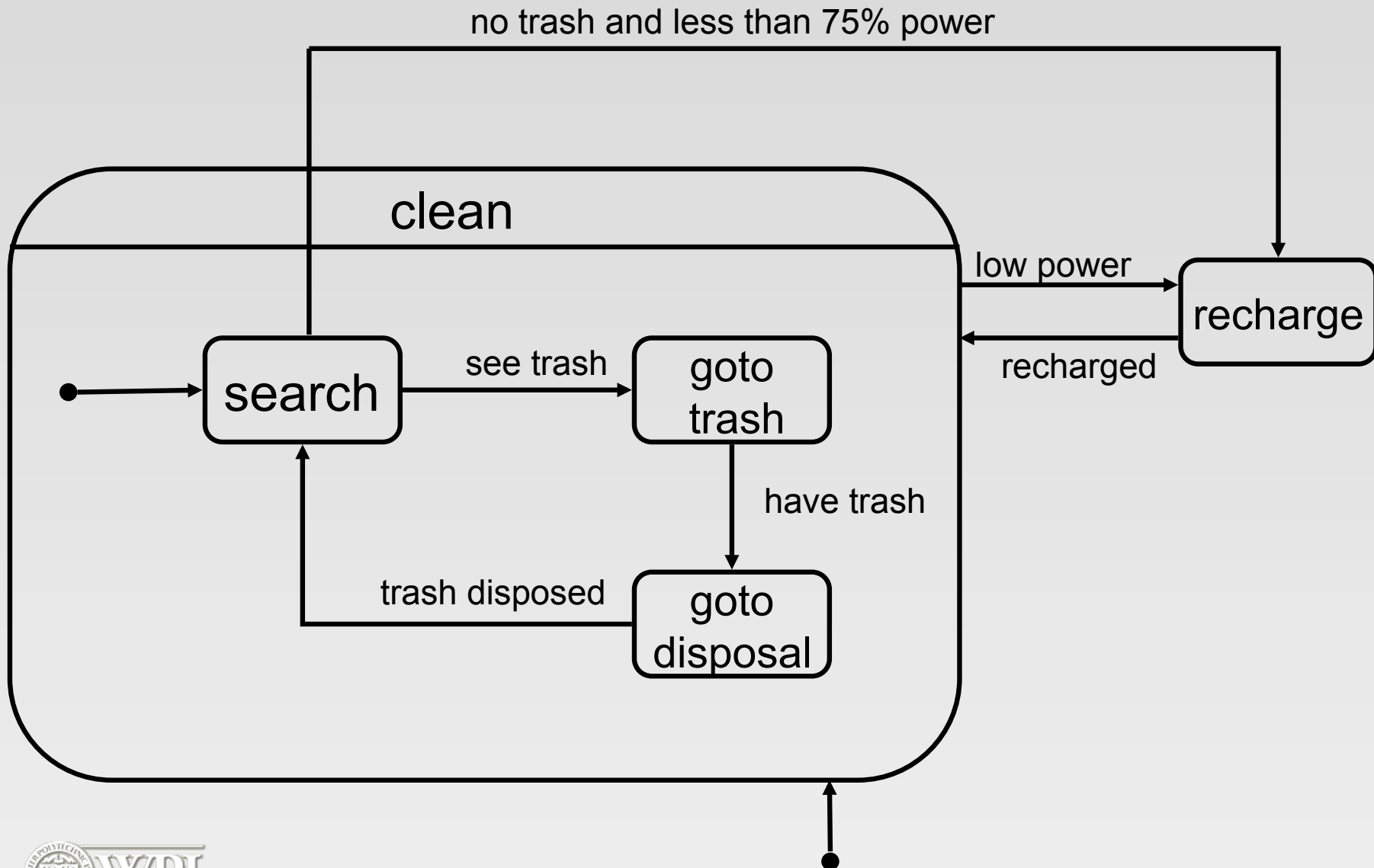


Cross-Hierarchy Transitions

- Why?
 - suppose we want robot to “top off” battery (even if it isn’t low) when it doesn’t see any trash



Cross-Hierarchy Transitions



HFSM Implementation Sketch

```
class State
```

```
// stack of return states
def getStates() return [this]

// recursive update
def update()

// rest same as flat machine
```

```
class Transition
```

```
// how deep this transition is
def getLevel()

// rest same as flat machine
```

```
struct UpdateResult // returned from update
  transition
  level
  actions // same as flat machine
```

```
class HierarchicalStateMachine
```

```
// same state variables as flat machine

// complicated recursive algorithm*
def update ()
```

```
class SubMachine : HierarchicalStateMachine,
                  State
```

```
def getStates()
  push this onto currentState.getStates()
```

* See full pseudo-code at
[http://www.cs.wpi.edu/~rich/courses/
imgd4000-b12/hsm.pdf](http://www.cs.wpi.edu/~rich/courses/imgd4000-b12/hsm.pdf)

Add tracing/debug code!!

