

Basic Game AI

Technical Game Development II

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Definitions?

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

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

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

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 games
 - interactive storytelling

“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)*
 - scripting *(next week)*
 - minimax search *(next week)*
- **Advanced** game AI
 - used in practice, but in more sophisticated games
 - advanced pathfinding *(tomorrow)*
 - autonomous movement, steering *(next week)*
 - behavior trees (in Halo 3) *(after midterm)*



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Future Game AI ?

- Take IMGD 400X in 2011 (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



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Two Fundamental Types of AI Algorithms

- Search vs. Non-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?

- Panel at GDC AI Summit: “Why so wary of 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

First Basic AI Technique:

Decision Trees

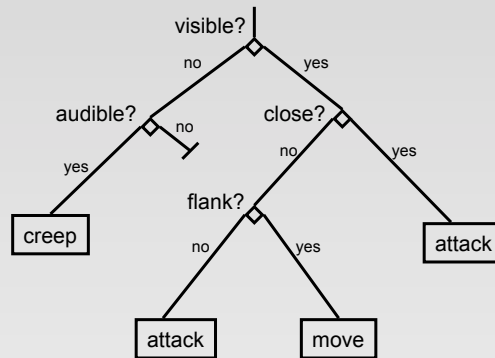
Reference: Millington, Section 5.2

Decision Trees

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

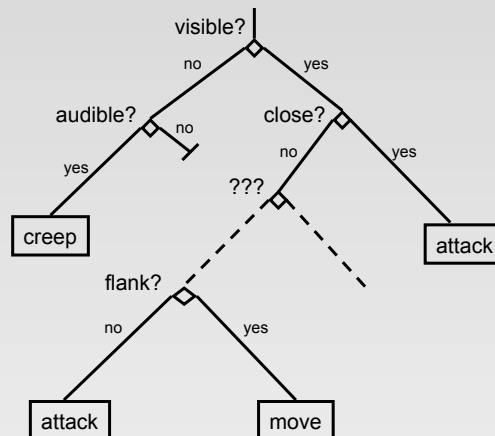
Deciding how to respond to an enemy

```
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)

(see Millington, Section 5.2.3)

```

class Node
  def decide()

class Action : Node
  def decide()
    return this

class Decision : Node

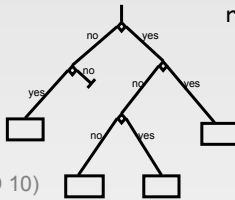
  def getBranch()

  def decide()
    return getBranch().decide()

class Boolean : Decision
  yesNode
  noNode

class MinMax : Boolean
  minValue
  maxValue
  testValue

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



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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 Creep...
  
```

```

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

```

audible.yesNode = creep
  
```

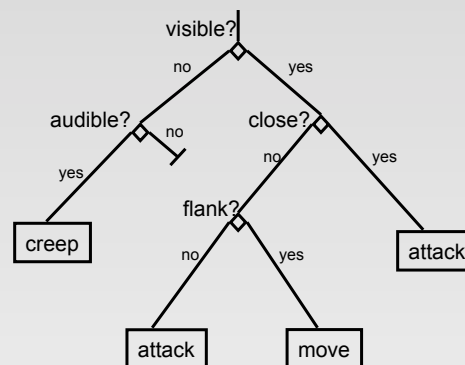
```

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

```

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

...



...or a graphical editor



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Modifying 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 Creep...

```

```

visible.yesNode = close
visible.noNode = audible

```

```

audible.yesNode = creep

```

```

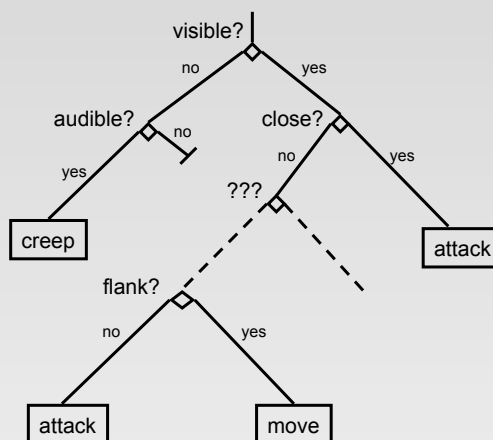
close.yesNode = attack
close.noNode = flank

```

```

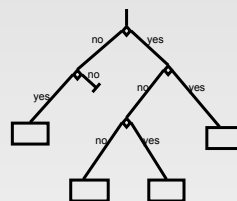
flank.yesNode = move
flank.noNode = attack
...

```



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



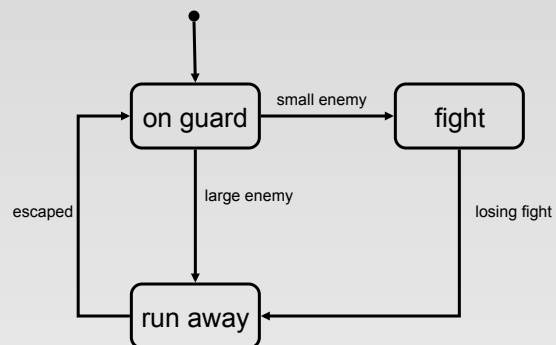
Next Basic AI Technique: (Hierarchical) State Machines

References: *Buckland, Chapter 2*
Millington, Section 5.3

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State Machines



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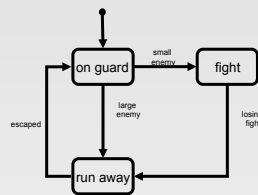
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Hard-Coded Implementation

```
class Soldier
```

```
    enum State  
        GUARD  
        FIGHT  
        RUN_AWAY
```

```
    currentState
```



```
    def update()  
        if currentState = GUARD {  
            if (small enemy)  
                currentState = FIGHT  
                startFighting  
            if (big enemy)  
                currentState = RUN_AWAY  
                startRunningAway  
        } else if currentState = FIGHT {  
            if (losing fight)  
                currentState = RUN_AWAY  
                startRunningAway  
        } else if currentState = RUN_AWAY {  
            if (escaped)  
                currentState = GUARD  
                startGuarding  
        }  
    }
```

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()

class StateMachine (see Millington, Section 5.3.3)
  states
  initialState
  currentState = initialState

  def update()
    triggeredTransition = null

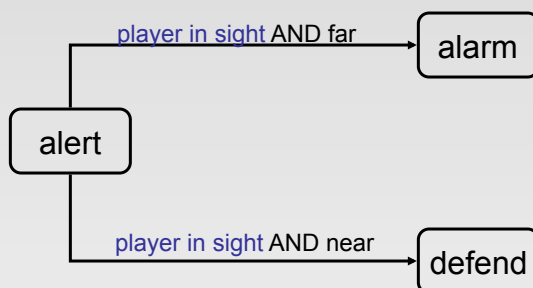
    for transition in currentState.getTransitions()
      if transition.isTriggered()
        triggeredTransition = transition
        break

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

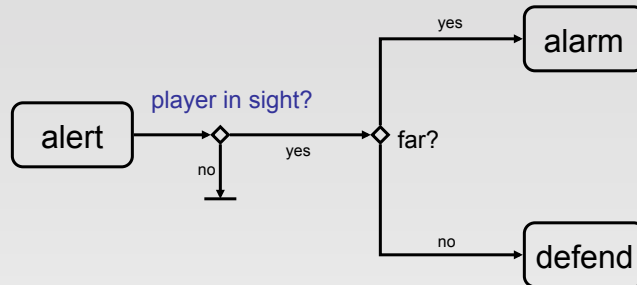
...add tracing

Combining Decision Trees & State Machines

- Why?
 - to avoid duplicating expensive tests

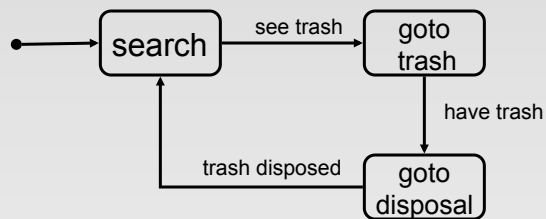


Combining Decision Trees & State Machines

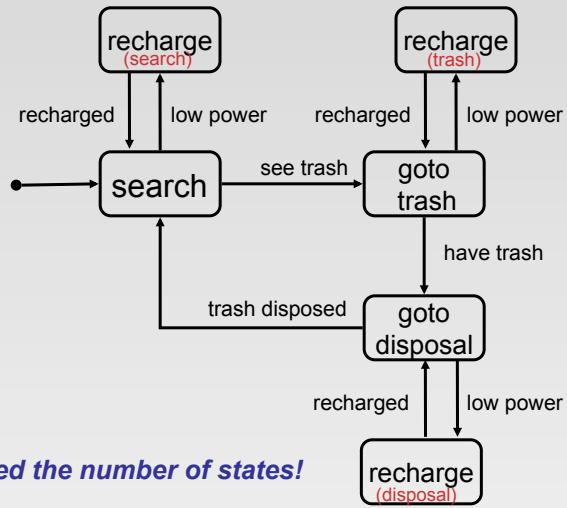


Hierarchical State Machines

- Why?



Interruptions (Alarms)



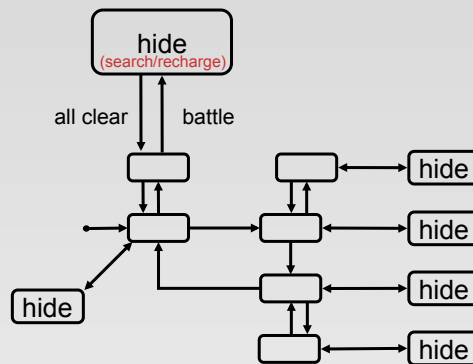
6 - doubled the number of states!



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Add Another Interruption Type



12 - doubled the number of states again!

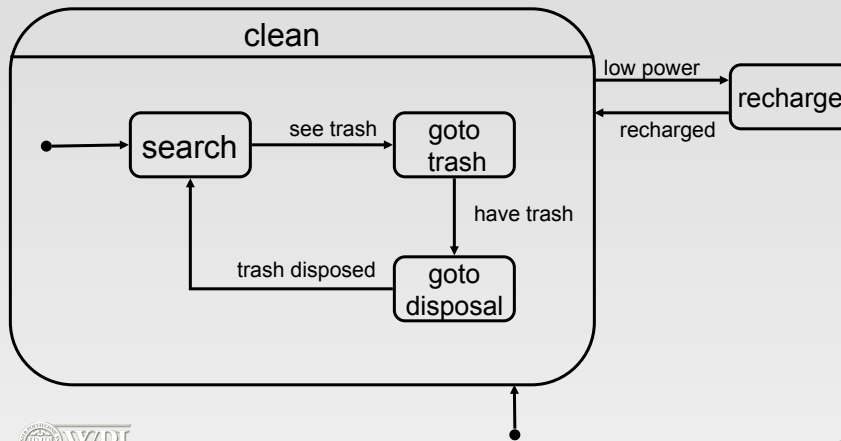


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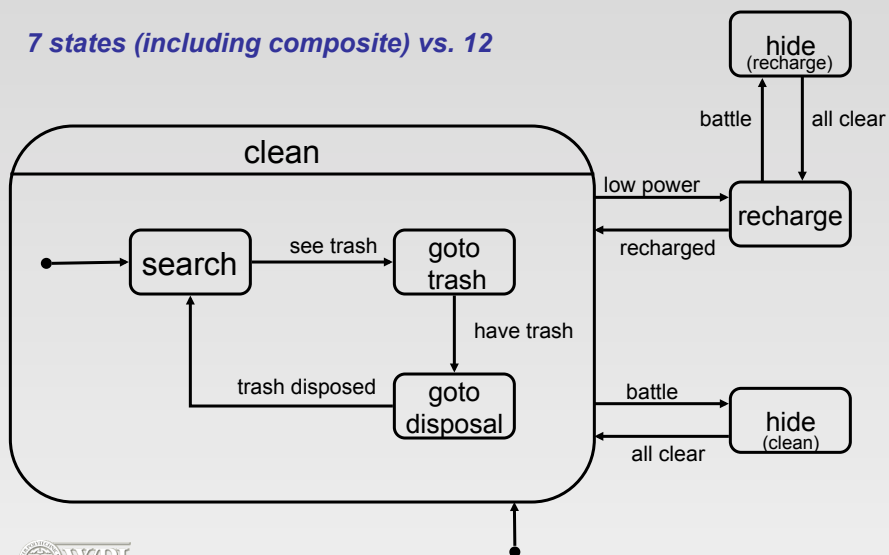
Hierarchical State Machine

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



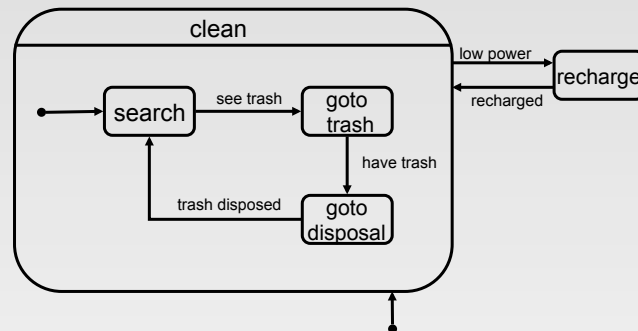
Add Another Interruption Type

7 states (including composite) vs. 12

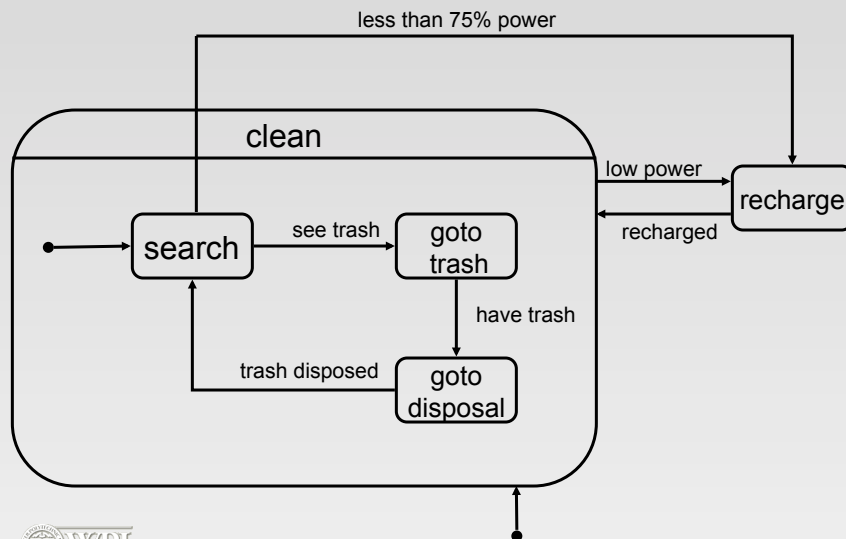


Cross-Hierarchy Transitions

- Why?
 - suppose we want robot to top-off battery if it doesn't see any trash



Cross-Hierarchy Transitions



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 Millington, Section 5.3.9)

