

Basic Game AI

Technical Game Development II

Professor Charles Rich
Computer Science Department
rich@wpi.edu

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
 - decision trees (Today)
 - (hierarchical) state machines (Today)
 - scripting (Monday)
 - minimax search (Tuesday)
 - basic pathfinding (A*) (IMGD 3000)
- **Advanced** game AI (Later...)
 - used in practice, but in more sophisticated games
 - task/behavior trees in Halo 3
 - autonomous movement, steering
 - advanced pathfinding



Future Game AI ?

- Take IMGD 400X next year (B)
“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

- 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

Scripting?



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7

First Basic AI Technique:

Decision Trees

Reference: Millington, Section 5.2

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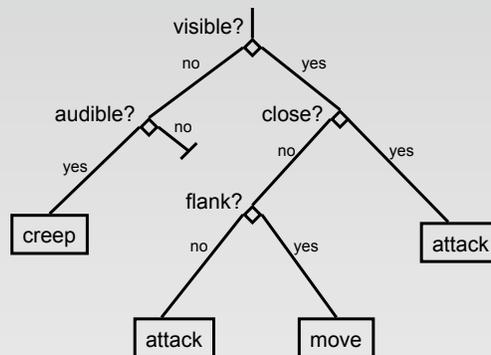
8

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;  
  }  
}
```

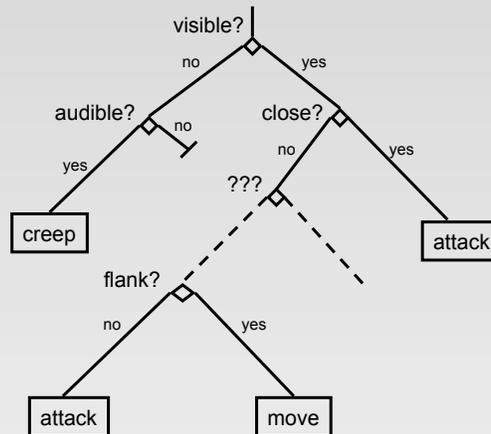


Which would you rather modify?

```

if (visible) {
  if (close) {
    attack;
  } else if (flank) {
    move;
  } else {
    attack;
  }
} else if (audible) {
  creep;
}

```



OO 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

```

```

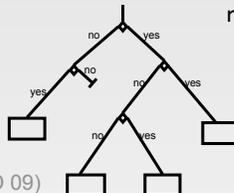
class MinMax : Decision
  minValue
  maxValue
  testValue

```

```

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

```



Building and Maintaining a Decision Tree

```
visible = decision[0] = new Boolean...  
audible = decision[1] = new Boolean...  
close = decision[2] = new MinMax...  
flank = decision[3] = new Boolean...
```

```
attack = action[0] = new Move...  
move = action[1] = new Move...  
creep = action[2] = new Creep...
```

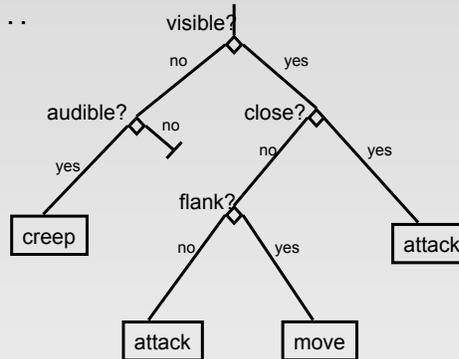
```
visible.yesNode = close  
visible.noNode = audible
```

```
audible.yesNode = creep
```

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

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

```
...
```



...or a graphical editor

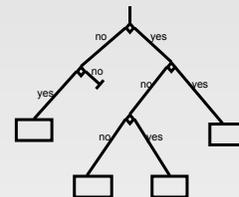


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13

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



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14

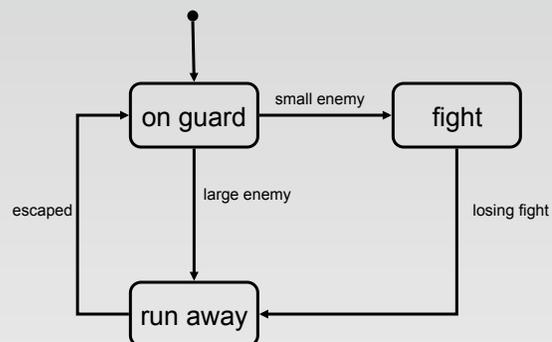
Next Basic AI Technique: (Hierarchical) State Machines

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

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15

State Machines



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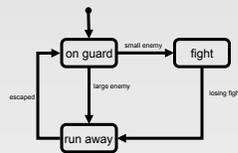
16

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) c  
                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 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()
```

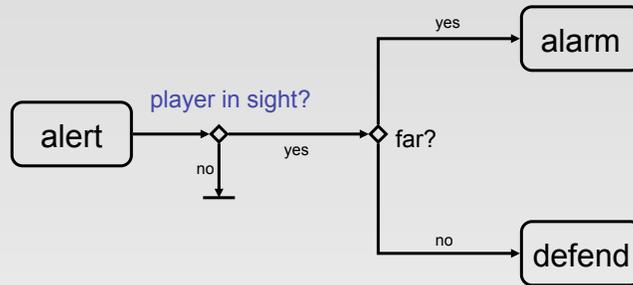
...add tracing

Combining Decision Trees & State Machines

- Why?
 - to avoid duplicating expensive tests

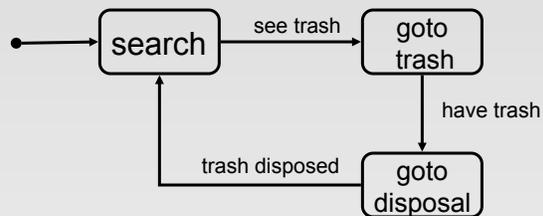


Combining Decision Trees & State Machines

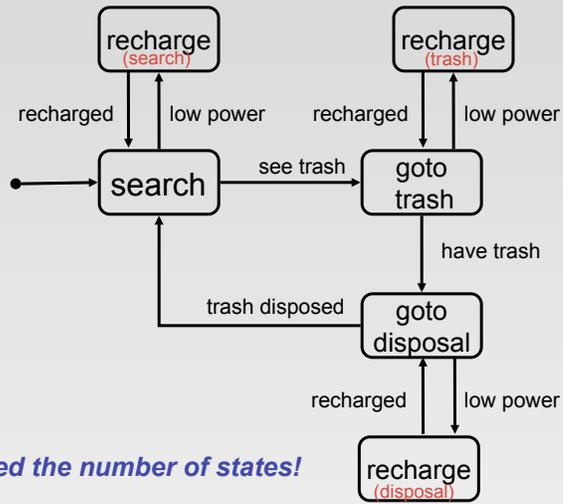


Hierarchical State Machines

- Why?



Interruptions (Alarms)



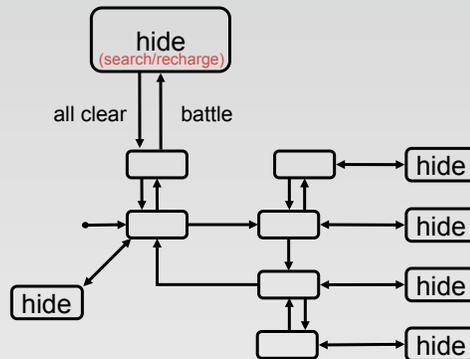
6 - doubled the number of states!



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23

Add Another Interruption Type



12 - doubled the number of states again!

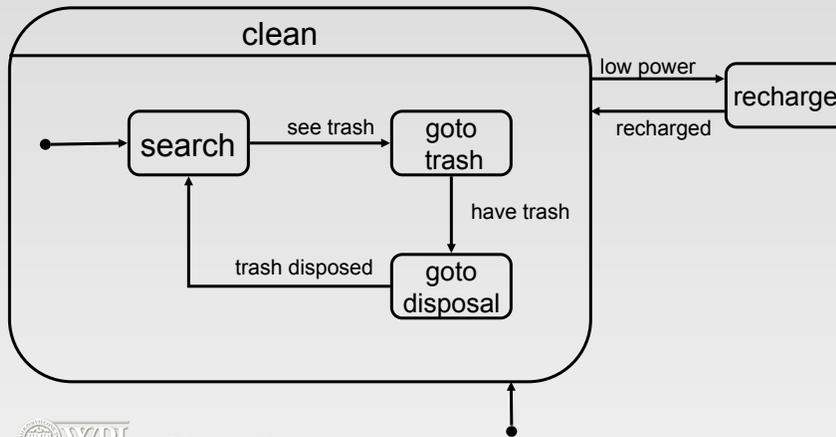


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24

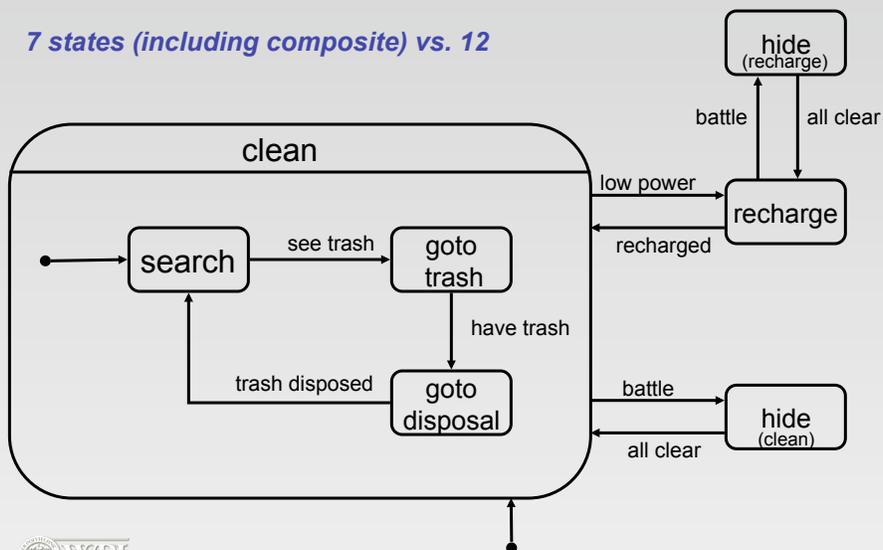
Hierarchical State Machine

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



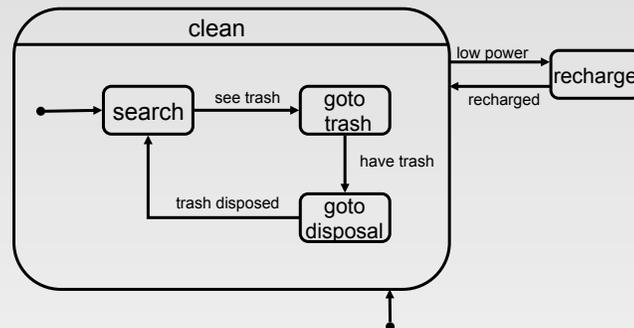
Add Another Interruption Type

7 states (including composite) vs. 12

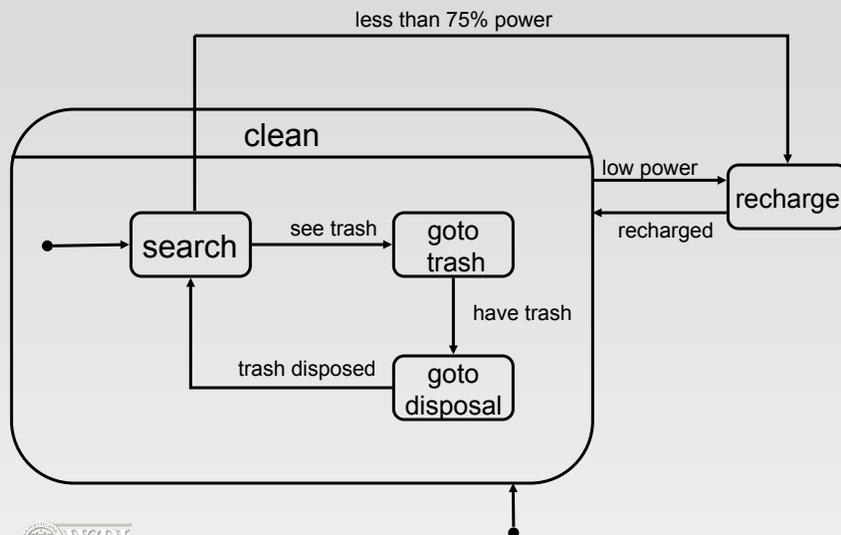


Cross-Hierarchy Transitions

- 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)

