

DM810

Computer Game Programming II: AI

Lecture 8

Decision Making

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- Hierarchical Pathfinding
- A* variants
- Decision Making
- Decision Trees

1. State Machine

2. Behavior Trees

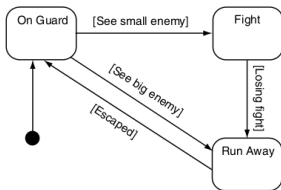
1. State Machine

2. Behavior Trees

Finite State Machines

An FSM is an algorithm used for parsing text, eg, tokenize the input code into symbols that can be interpreted by the compiler.

- **States**: actions or behaviors. Chars are in exactly one of them @ any time.
- **Transitions**: a set of associated conditions, if they are met the char changes state
- **Initial state** for the first frame the state machine is run



In a decision tree, the same set of decisions is always used, and any action can be reached through the tree.

In a state machine, only transitions from the current state are considered, so not every action can be reached.

- set of possible states
- current state
- set of transitions
- at each iteration (normally each frame), the state machine's **update** function is called.
- checks if any transition from the current state is **triggered**
- the first transition that is triggered is scheduled to **fire**
(some actions related to transition are executed)

```
class StateMachine:
    states # list of states for the machine
    initialState
    currentState = initialState
    def update(): # checks and applies
        triggeredTransition = None
        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()
```

```
class MyFSM:
    enum State:
        PATROL
        DEFEND
        SLEEP
    myState # holds current state

    # transition by polling (asking for information explicitly)
    def update():
        if myState == PATROL:
            if canSeePlayer(): myState = DEFEND # access to game state data
            if tired(): myState = SLEEP # access to game state data
        elif myState == DEFEND:
            if not canSeePlayer(): myState = PATROL
        elif myState == SLEEP:
            if not tired(): myState = PATROL

    # transition in an event-based approach (waiting to be told information)
    def notifyNoiseHeard(volume):
        if myState == SLEEP and volume > 10:
            myState = DEFEND

    def getAction():
        if myState == PATROL: return PatrolAction
        elif myState == DEFEND: return DefendAction
        elif myState == SLEEP: return SleepAction
```

State machines implemented like this can often get large and code unclear


```

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

class Transition:
    actions
    def getAction(): return actions
    targetState
    def getTargetState(): return targetState
    condition
    def isTriggered(): return condition.test()

```

Often defined in a data file and read into the game at runtime.

Do not allow to compose questions easily.

Requires condition interface.

```

class Condition:
    def test()

class FloatCondition (Condition):
    minValue
    maxValue
    testValue # ptr to game data
    def test():
        return minValue <= testValue
            <= maxValue

class AndCondition (Condition):
    conditionA
    conditionB
    def test():
        return conditionA.test() and conditionB.test()

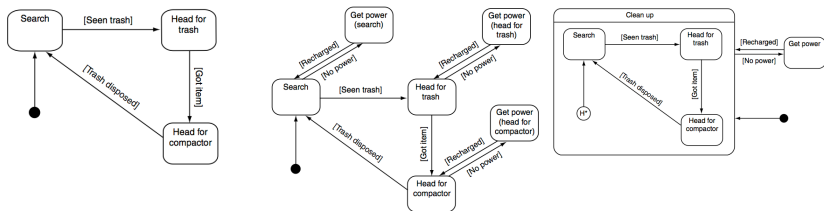
class NotCondition (Condition):
    condition
    def test(): return not condition.test()

class OrCondition (Condition):
    conditionA
    conditionB
    def test():
        return conditionA.test() or conditionB.test()

```

Hierarchical State Machines

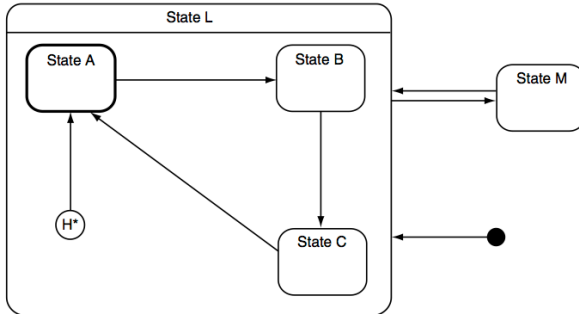
- Alarm mechanism: something that interrupts normal behavior to respond to something important.
- Representing this in a state machine leads to a doubling in the number of states.
- Instead: each alarm mechanism has its own state machine, along with the original behavior.



We can add transitions between layers of machines

Implementation

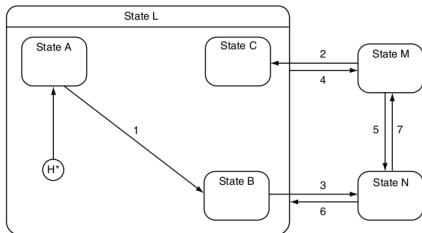
In a hierarchical state machine, each state can be a complete state machine in its own right \rightsquigarrow recursive algorithm



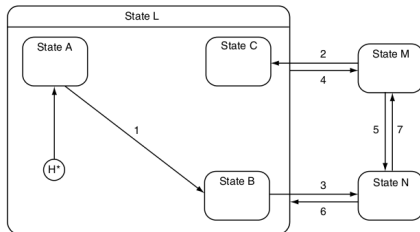
A triggered transition may be: (i) to another state at current level, (ii) to a state higher up, or (iii) to a lower state

Example

- start in State L
- from H* transition to A
update = [L-active, A-entry]
current State [L, A]



- top-level state machine no valid transitions
state machine L: current state [A], triggered transition 1 \rightsquigarrow stay at current level,
transition to B, update = [A-exit, 1-actions, B-entry]
top-level state machine accepts and adds L-active. current State [L, B].
- top level machine: triggered transition 4
transition to State M, update = [L-exit, 4-actions, M-entry].
current State is [M]. (state machine L still keeps State B)
- top level machine: triggered transition 5
transition to State N, update = [M-exit, 5-actions, N-entry]. current State N
- top level machine: triggered transition 6
transitions to State L, update = [N-exit, 6-actions, L-entry].
state machine L has current state still State [L, B] \rightsquigarrow no B-entry action



- top-level state machine no transition; State [L, B] triggered transition 3.
 top-level state machine no triggers state machine L: B, transition has one level up
 update: B-exit
 top-level machine: transition to State N; update += [L-exit, 3-actions, N-entry]
- State N → transition 7 → State M
 ...
- top level machine: triggered transition 2. top-level state machine: transition down ~~~
 updateDown. state machine L: update = C-enter
 top-level state machine changes from State M to State L, update += [M-exit,
 L-entry, 2-actions]

Implementation

```
class HSMBase:
    struct UpdateResult:
        actions
        transition
        level
    def getAction(): return []
    def update():
        UpdateResult result
        result.actions = getAction()
        result.transition = None
        result.level = 0
        return result
    def getStates()

class HierarchicalStateMachine (HSMBase):
    states # List of states at this level
    initialState # when no current state
    currentState = initialState
    def getStates():
        if currentState: return currentState.getStates()
        else: return []
    def update(): ...
    def updateDown(state, level): ...

class State (HSMBase):
    def getStates():
        return [this]
    def getAction()
    def getEntryAction()
    def getExitAction()
    def getTransitions()

class SubMachineState (State, HierarchicStateMachine):
    def getAction(): return State::getAction()
    def update(): return HierarchicalStateMachine::
        update()
    def getStates():
        if currentState:
            return [this] + currentState.getStates()
        else:
            return [this]

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

```
class HierarchicalStateMachine (HSMBBase):
    states # List of states at this level
    initialState # when no current state
    currentState = initialState
    def getStates():
        if currentState: return currentState.getStates()
        else: return []
    def update():
        if not currentState:
            currentState = initialState
            return currentState.getEntryAction()
        triggeredTransition = None
        for transition in currentState.getTransitions():
            if transition.isTriggered():
                triggeredTransition = transition
                break
        if triggeredTransition:
            result = UpdateResult()
            result.actions = []
            result.transition = triggeredTransition
            result.level = triggeredTransition.getLevel()
        else:
            result = currentState.update() # rcrs.
```

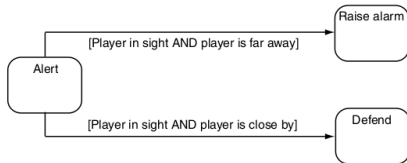
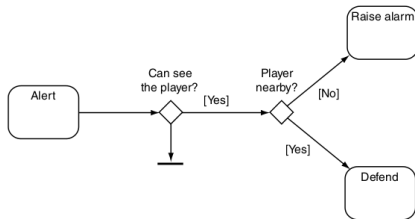
```
if result.transition:
    if result.level == 0: # Its on this level: honor it
        targetState = result.transition.getTargetState()
        result.actions += currentState.getExitAction()
        result.actions += result.transition.getAction()
        result.actions += targetState.getEntryAction()
        currentState = targetState
        result.actions += getAction()
        result.transition = None # so nobody else does it
    else if result.level > 0: # it is for a higher level
        result.actions += currentState.getExitAction()
        currentState = None
        result.level -= 1
    else: # It needs to be passed down
        targetState = result.transition.getTargetState()
        targetMachine = targetState.parent
        result.actions += result.transition.getAction()
        result.actions += targetMachine.updateDown(targetState, -result.level) #
            recursion
        result.transition = None # so nobody else does it
else: # no transition
    result.action += getAction()
return result
```



```
def updateDown(state, level):  
    if level > 0: # continue recursing  
        actions = parent.updateDown(this, level-1)  
    else: actions = []  
    if currentState:  
        actions += currentState.getExitAction()  
    currentState = state # move to the new state  
    actions += state.getEntryAction()  
    return actions
```

Combining DT and SM

Decision trees can be used to implement more complex transitions

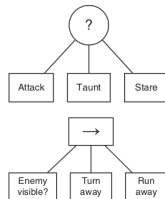


1. State Machine

2. Behavior Trees

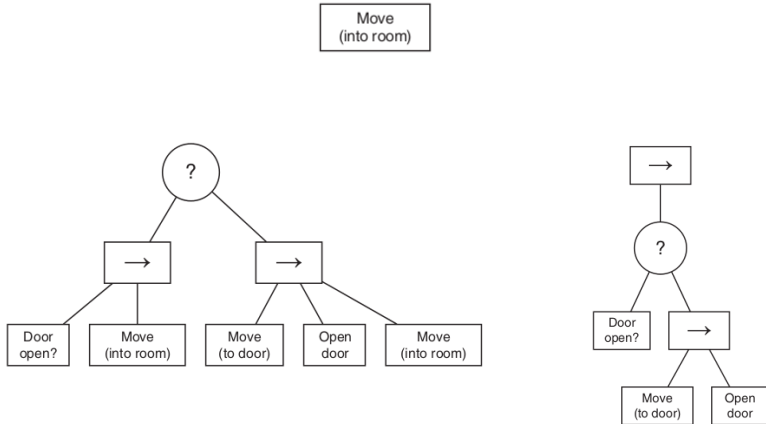
- synthesis of: Hierarchical State Machines, Scheduling, Planning, and Action Execution.
- state: task composed of sub-trees
- tasks are **Conditions**, **Actions**, **Composites**
- tasks return `true`, `false`, `error`, `need more time`
- **Actions**: animation, character movement, change the internal state of the character, play audio samples, engage the player in dialog, pathfinding.
- **Conditions** are logical conditions
- behavior trees are coupled with a graphical user interface (GUI) to edit the trees.

- Both Conditions and Actions sit at the leaf nodes of the tree. Branches are made up of Composite nodes.
- **Composites**: two main types: Selector and Sequence
- Both run each of their child behaviors in turn and decide whether to continue through its children or to stop according to the returned value.
- **Selector** returns immediately with a success when one of its children succeeds. As long as children are failing, it keeps on trying. If no children left, returns failure. (used to choose the first of a set of possible actions that is successful) Eg: a character wanting to reach safety.
- **Sequence** returns immediately with a failure when one of its children fails. As long as children are succeeding, it keeps on trying. If no children left, returns success. (series of tasks that need to be undertaken)

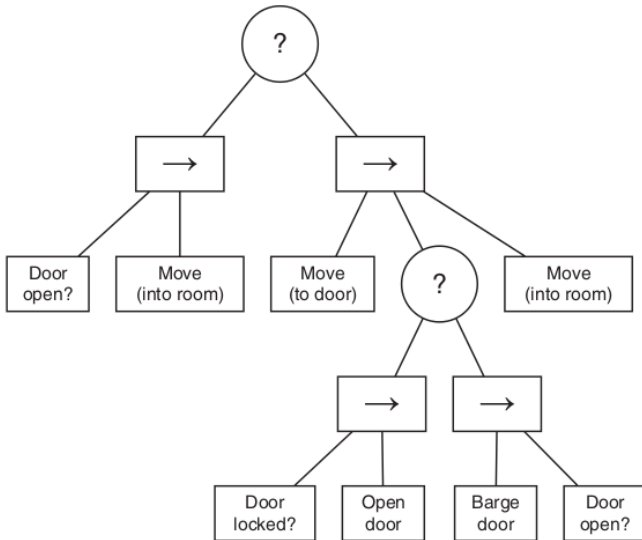


Developing Behaviour Trees

- get something very simple to work initially



Condition task in a Sequence acts like an IF-statement.
If the Sequence is placed within a Selector, then it acts like an IF-ELSE-statement



- behaviour trees implement a sort of **reactive planning**. Selectors allow the character to try things, and fall back to other behaviors if they fail. (look ahead only via actions)
- depth-first search
- could be written as state machines or decision trees but more complicated

Implementation

```
class Task:
    children
    def run() # true/false
```

```
class Selector (Task):
    def run():
        for c in children:
            if c.run():
                return True
        return False
```

```
class Sequence (Task):
    def run():
        for c in children:
            if not c.run():
                return False
        return True
```

```
class EnemyNear (Task):
    def run():
        if distanceToEnemy < 10:
            return True
        return False
```

```
class PlayAnimation (Task):
    animation_id
    speed
    def Attack(animation_id, loop=False,
               speed=1.0):
        this.animation = animation
        this.speed = speed
    def run():
        if animationEngine.ready(): #
            resource checking
            animationEngine.play(animation,
                                   speed)
            return True
        return False
```

- In some cases, always trying the same things in the same order can lead to predictable AIs.
- Selectors: eg, if alternative ways to enter the door, no relevant the order
- Sequences: eg, collect components, no relevant the order “partial-order” constraints in the AI literature. Some
- parts may be strictly ordered, and others can be processed in any order.

```
class NonDeterministicSelector (Task):  
    children  
    def run():  
        shuffled = random.shuffle(children)  
        for child in shuffled:  
            if child.run(): break  
        return result
```

```
class NonDeterministicSequence (Task):  
    children  
    def run():  
        shuffled = random.shuffle(children)  
        for child in shuffled:  
            if not child.run(): break  
        return result
```

by Richard Durstenfeld in 1964 in *Communications of the ACM*, volume 7, issue 7, as "Algorithm 235: Random permutation", and by Donald E. Knuth in volume 2 of his book *The Art of Computer Programming* as "Algorithm P" but originally by Fisher and Yates.

```
def shuffle(original):  
    list = original.copy()  
    n = list.length  
    while n > 1:  
        k = random.integer_less_than(n)  
        n--;  
        tmp = list[k], list[k] = list[n], list[n] = tmp  
    return list
```

