

# 6.01: Introduction to EECS I

Primitives, Combination, Abstraction, and Patterns

*February 8, 2011*

# PCAP Framework for Managing Complexity

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Python has features that facilitate modular programming.

- **def** combines operations into a procedure and binds a name to it
- **lists** provide flexible and hierarchical structures for data
- **variables** associate names with data
- **classes** associate data (attributes) and procedures (methods)

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|                     | procedures                                  | data                         |
|---------------------|---|------------------------------|
| <b>P</b> rimitives  | <b>+</b> , <b>*</b> , <b>==</b> , <b>!=</b> | numbers, booleans, strings   |
| <b>C</b> ombination | <b>if</b> , <b>while</b> , <b>f(g(x))</b>   | lists, dictionaries, objects |
| <b>A</b> bstraction | <b>def</b>                                  | classes                      |
| <b>P</b> atterns    | higher-order procedures                     | super-classes, sub-classes   |

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## PCAP Framework for Managing Complexity

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We will build on these ideas to manage complexity at higher levels.

- **Programming Styles** for dealing with complexity
- PCAP in Higher-Level Abstractions: **State Machines**

**Reading:** Course notes, chapters 3–4

# Programming Styles for Managing Complexity

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Structure of program has significant effect on its modularity.

## **Imperative** (procedural) programming

- focus on step-by-step instructions to accomplish task
- organize program using structured conditionals and loops

## **Functional** programming

- focus on procedures that mimic mathematical functions, producing outputs from inputs without side effects
- functions are **first-class objects** used in data structures, arguments to procedures, and can be returned by procedures

## **Object-oriented** programming

- focus on collections of related procedures and data
- organize programs as hierarchies of related classes and instances

## Example Program

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Task: Find a sequence of operations (either **increment** or **square**) that transforms the integer  $i$  (initial) to the integer  $g$  (goal).

Example: applying the sequence

**increment increment increment square**

to **1** yields **16**

apply **increment** to **1**  $\rightarrow$  **2**

apply **increment** to **2**  $\rightarrow$  **3**

apply **increment** to **3**  $\rightarrow$  **4**

apply **square** to **4**  $\rightarrow$  **16**

## Check Yourself

---

What is the minimum length sequence of **increment** and **square** operations needed to transform **1** to **100**?

1: <4

2: 4

3: 5

4: 6

5: >6

## Check Yourself

---

What is the minimum length sequence of **increment** and **square** operations needed to transform **1** to **100**?

Try to use as many squares (especially big ones) as possible.

apply **increment** to **1**  $\rightarrow$  **2**

apply **increment** to **2**  $\rightarrow$  **3**

apply **square** to **3**  $\rightarrow$  **9**

apply **increment** to **9**  $\rightarrow$  **10**

apply **square** to **10**  $\rightarrow$  **100**

Five operations.

## Check Yourself

---

What is the minimum length sequence of **increment** and **square** operations needed to transform **1** to **100**? **3: 5**

1: <4

2: 4

**3: 5**

4: 6

5: >6



## Imperative (Procedural) Programming

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Solve the previous problem by writing an imperative program to step through all possible sequences of length 1, 2, 3, ...

```
def increment(n):
    return n+1
def square(n):
    return n**2
def findSequence(initial,goal):
    # construct list of "candidates" of form ('1 increment increment',3)
    candidates = [(str(initial),initial)]
    # loop over sequences of length "i" = 1, 2, 3, ...
    for i in range(1,goal-initial+1):
        newCandidates = []
        # construct each new candidate by adding one operation to prev candidate
        for (action,result) in candidates:
            for (a,r) in [(' increment',increment),(' square',square)]:
                newCandidates.append((action+a,r(result)))
            print i,': ',newCandidates[-1]
            if newCandidates[-1][1] == goal:
                return newCandidates[-1]
        candidates = newCandidates

answer = findSequence(1,100)
print 'answer =',answer
```

# Imperative (Procedural) Programming

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```
1 : ('1 increment', 2)
1 : ('1 square', 1)
2 : ('1 increment increment', 3)
2 : ('1 increment square', 4)
2 : ('1 square increment', 2)
2 : ('1 square square', 1)
3 : ('1 increment increment increment', 4)
3 : ('1 increment increment square', 9)
3 : ('1 increment square increment', 5)
3 : ('1 increment square square', 16)
3 : ('1 square increment increment', 3)
3 : ('1 square increment square', 4)
3 : ('1 square square increment', 2)
3 : ('1 square square square', 1)
4 : ('1 increment increment increment increment', 5)
4 : ('1 increment increment increment square', 16)
4 : ('1 increment increment square increment', 10)
4 : ('1 increment increment square square', 81)
4 : ('1 increment square increment increment', 6)
4 : ('1 increment square increment square', 25)
4 : ('1 increment square square increment', 17)
4 : ('1 increment square square square', 256)
4 : ('1 square increment increment increment', 4)
4 : ('1 square increment increment square', 9)
```

```
4 : ('1 square increment square increment', 5)
4 : ('1 square increment square square', 16)
4 : ('1 square square increment increment', 3)
4 : ('1 square square increment square', 4)
4 : ('1 square square square increment', 2)
4 : ('1 square square square square', 1)
5 : ('1 increment increment increment increment increment', 6)
5 : ('1 increment increment increment increment square', 25)
5 : ('1 increment increment increment square increment', 17)
5 : ('1 increment increment increment square square', 256)
5 : ('1 increment increment square increment increment', 11)
5 : ('1 increment increment square increment square', 100)

answer = ('1 increment increment square increment square', 100)
```

## Imperative (Procedural) Programming

---

This imperative version of the program has three levels of looping.

```
def findSequence(initial,goal):
    # construct list of "candidates" of form ('1 increment increment',3)
    candidates = [(str(initial),initial)]
    # loop over sequences of length "i" = 1, 2, 3, ...
    for i in range(1,goal-initial+1):
        newCandidates = []
        # construct each new candidate by adding one operation to prev candidate
        for (action,result) in candidates:
            for (a,r) in [(' increment',increment),(' square',square)]:
                newCandidates.append((action+a,r(result)))
            print i,': ',newCandidates[-1]
            if newCandidates[-1][1] == goal:
                return newCandidates[-1]
        candidates = newCandidates
```

This approach is straightforward, but nested loops can be confusing.

Challenge is to get the indices right.

# Functional Programming

---

This version focuses on functions as primitives.

```
def apply(opList, arg):
    if len(opList)==0:
        return arg
    else:
        return apply(opList[1:], opList[0](arg))

def addLevel(opList, fctList):
    return [x+[y] for y in fctList for x in opList]

def findSequence(initial, goal):
    opList = [[]]
    for i in range(1, goal-initial+1):
        opList = addLevel(opList, [increment, square])
        for seq in opList:
            if apply(seq, initial)==goal:
                return seq

answer = findSequence(1, 100)
print 'answer =', answer
```

## Functional Programming

---

The procedure **apply** is a “pure function.”

```
def apply(opList, arg):
    if len(opList)==0:
        return arg
    else:
        return apply(opList[1:],opList[0](arg))
```

Its first argument is a list of functions. The procedure applies these functions to the second argument **arg** and returns the result.

```
>>> apply([],7)
7
>>> apply([increment],7)
8
>>> apply([square],7)
49
>>> apply([increment,square],7)
64
```

This list of procedures uses functions as first-class objects.

## Functional Programming

---

The procedure `addLevel` is also a pure function.

```
def addLevel(opList,fctList):  
    return [x+[y] for y in fctList for x in opList]
```

The first input is a list of sequences-of-operations, each of which is a list of functions.

The second input is a list of possible next-functions.

It returns a new list of sequences.

```
>>> addLevel([[increment]], [increment, square])  
[[<function increment at 0xb7480aac>, <function increment at 0xb7480aac>],  
 [<function increment at 0xb7480aac>, <function square at 0xb747b25c>]]
```

# Functional Programming

---

The answer is now a list of functions.

```
def apply(opList, arg):
    if len(opList)==0:
        return arg
    else:
        return apply(opList[1:], opList[0](arg))

def addLevel(opList, fctList):
    return [x+[y] for y in fctList for x in opList]

def findSequence(initial, goal):
    opList = [[]]
    for i in range(1, goal-initial+1):
        opList = addLevel(opList, [increment, square])
        for seq in opList:
            if apply(seq, initial)==goal:
                return seq

answer = findSequence(1, 100)
print 'answer =', answer
```

```
answer = [<function increment at 0xb777ea74>, <function increment at
0xb777ea74>, <function square at 0xb7779224>, <function increment at
0xb777ea74>, <function square at 0xb7779224>]
```



## Functional Programming

---

The functions `apply` and `addLevel` are easy to check.

```
def apply(opList, arg):
    if len(opList)==0:
        return arg
    else:
        return apply(opList[1:], opList[0](arg))
```

```
def addLevel(opList, fctList):
    return [x+[y] for y in fctList for x in opList]
```

```
>>> apply([], 7)
```

```
7
```

```
>>> apply([increment], 7)
```

```
8
```

```
>>> apply([square], 7)
```

```
49
```

```
>>> apply([increment, square], 7)
```

```
64
```

```
>>> addLevel([[increment]], [increment, square])
```

```
[[<function increment at 0xb7480aac>, <function increment at 0xb7480aac>],
 [<function increment at 0xb7480aac>, <function square at 0xb747b25c>]]
```

Greater modularity reduces complexity and simplifies debugging.

## Functional Programming

---

Also notice that the definition of `apply` is **recursive**:  
the definition of `apply` calls `apply`.

```
>>> def apply(opList, arg) :  
...     if len(opList)==0:  
...         return arg  
...     else:  
...         return apply(opList[1:], opList[0](arg))
```

Recursion is

- an alternative way to implement iteration (looping)
- a natural generalization of functional programming
- powerful way to think about PCAP

## Recursion

---

Express solution to problem in terms of simpler version of problem.

Example: raising a number to a non-negative integer power

$$b^n = \begin{cases} 1 & \text{if } n = 0 \\ b \cdot b^{n-1} & \text{if } n > 0 \end{cases}$$

functional notation:

$$f(n) = \begin{cases} 1 & \text{if } n = 0 \\ b f(n-1) & \text{if } n > 0 \end{cases}$$

Python implementation:

```
def exponent(b,n):  
    if n==0:  
        return 1  
    else:  
        return b*exponent(b,n-1)
```

# Recursive Exponentiation

---

Invoking `exponent(2, 6)` generates 6 more invocations of `exponent`.

```
def exponent(b,n):
    if n==0:
        return 1
    else:
        return b*exponent(b,n-1)

exponent(2,6)
    calls exponent(2,5)
        calls exponent(2,4)
            calls exponent(2,3)
                calls exponent(2,2)
                    calls exponent(2,1)
                        calls exponent(2,0)
                            returns 1
                        returns 2
                    returns 4
                returns 8
            returns 16
        returns 32
    returns 64
64
```

Number of invocations increases in proportion to **n** (i.e., linearly).

## Fast Exponentiation

---

There is a straightforward way to speed this process:

If  $n$  is even, then square the result of raising  $b$  to the  $n/2$  power.

$$b^n = \begin{cases} 1 & \text{if } n = 0 \\ b \cdot b^{n-1} & \text{if } n \text{ odd} \\ (b^{n/2})^2 & \text{otherwise} \end{cases}$$

functional notation:

$$f(n) = \begin{cases} 1 & \text{if } n = 0 \\ bf(n-1) & \text{if } n \text{ odd} \\ (f(n/2))^2 & \text{otherwise} \end{cases}$$

# Fast Exponentiation

---

Implement in Python.

```
def fastExponent(b,n):  
    if n==0:  
        return 1  
    elif n%2==1:  
        return b*fastExponent(b,n-1)  
    else:  
        return fastExponent(b,n/2)**2
```

## Check Yourself

---

```
def fastExponent(b,n):  
    if n==0:  
        return 1  
    elif n%2==1:  
        return b*fastExponent(b,n-1)  
    else:  
        return fastExponent(b,n/2)**2
```

How many invocations of **fastExponent** is generated by **fastExponent(2,10)**?

1. 10

2. 8

3. 7

4. 6

5. 5

# Recursive Exponentiation

---

Implement recursion in Python.

```
def fastExponent(b,n):
    if n==0:
        return 1
    elif n%2==1:
        return b*fastExponent(b,n-1)
    else:
        return fastExponent(b,n/2)**2

fastExponent(2,10)
    calls fastExponent(2,5)
        calls fastExponent(2,4)
            calls fastExponent(2,2)
                calls fastExponent(2,1)
                    calls fastExponent(2,0)
                        returns 1
                    returns 2
                returns 4
            returns 16
        returns 32
    returns 1024
1024
```

The number of calls increases in proportion to  $\log n$  (for large  $n$ ).



## Check Yourself

---

```
def fastExponent(b,n):  
    if n==0:  
        return 1  
    elif n%2==1:  
        return b*fastExponent(b,n-1)  
    else:  
        return fastExponent(b,n/2)**2
```

How many invocations of `fastExponent` is generated by `fastExponent(2,10)`? **5**

1. 10

2. 8

3. 7

4. 6

5. 5

# Recursive Exponentiation

---

Functional approach makes this simplification easy to spot.

```
def fastExponent(b,n):  
    if n==0:  
        return 1  
    elif n%2==1:  
        return b*fastExponent(b,n-1)  
    else:  
        return fastExponent(b,n/2)**2
```

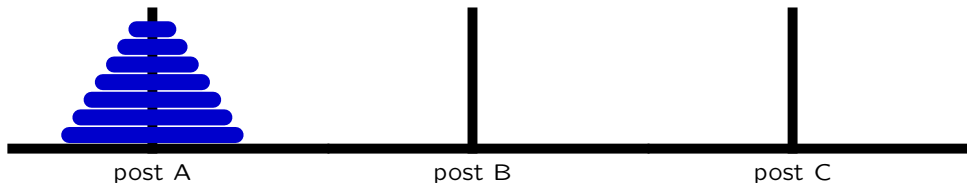
```
fastExponent(2,10)  
    calls fastExponent(2,5)  
        calls fastExponent(2,4)  
            calls fastExponent(2,2)  
                calls fastExponent(2,1)  
                    calls fastExponent(2,0)  
                        returns 1  
                    returns 2  
                returns 4  
            returns 16  
        returns 32  
    returns 1024  
1024
```

Functional approach is “expressive.”

## Towers of Hanoi

---

Transfer a stack of disks from post A to post B by moving the disks one-at-a-time, without placing any disk on a smaller disk.



```
def Hanoi(n,A,B,C):  
    if n==1:  
        print 'move from ' + A + ' to ' + B  
    else:  
        Hanoi(n-1,A,C,B)  
        Hanoi(1,A,B,C)  
        Hanoi(n-1,C,B,A)
```

# Towers of Hanoi

---

Towers of height 3 and 4.

```
> > > Hanoi(3,'a','b','c')
```

```
move from a to b  
move from a to c  
move from b to c  
move from a to b  
move from c to a  
move from c to b  
move from a to b
```

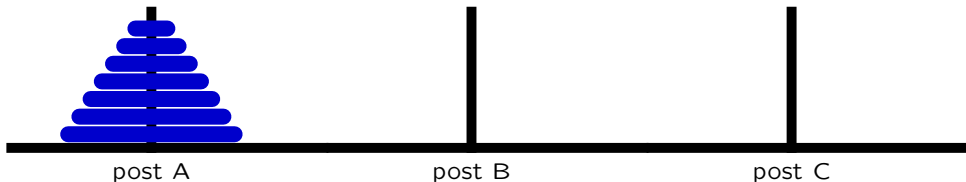
```
> > > Hanoi(4,'a','b','c')
```

```
move from a to c  
move from a to b  
move from c to b  
move from a to c  
move from b to a  
move from b to c  
move from a to c  
move from a to b  
move from c to b  
move from c to a  
move from b to a  
move from c to b  
move from a to c  
move from a to b  
move from c to b
```

## Towers of Hanoi

---

Transfer a stack of disks from post A to post B by moving the disks one-at-a-time, without placing any disk on a smaller disk.



```
def Hanoi(n,A,B,C):  
    if n==1:  
        print 'move from ' + A + ' to ' + B  
    else:  
        Hanoi(n-1,A,C,B)  
        Hanoi(1,A,B,C)  
        Hanoi(n-1,C,B,A)
```

Recursive solution is “expressive” (also simple and elegant).

## Back to the Earlier Example

---

Task: Find a sequence of operations (either **increment** or **square**) that transforms the integer  $i$  (initial) to the integer  $g$  (goal).

**Imperative** (procedural) approach ✓

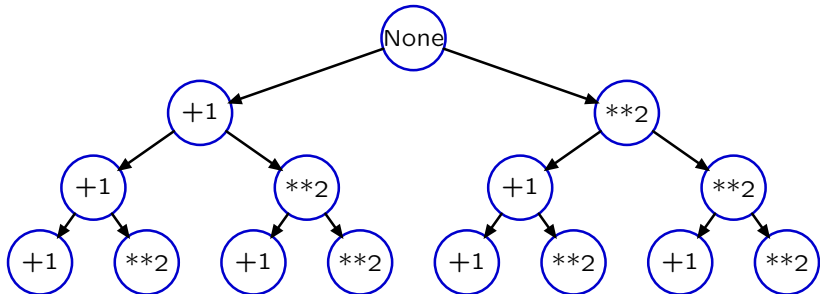
**Functional** approach ✓

**Object-oriented** approach

# OOP

---

Represent all possible sequences in a tree.



Define an object to represent each of these “nodes”:

```
class Node:
```

```
    def __init__(self, parent, action, answer):
```

```
        self.parent = parent
```

```
        self.action = action
```

```
        self.answer = answer
```

```
    def path(self):
```

```
        if self.parent == None:
```

```
            return [(self.action, self.answer)]
```

```
        else:
```

```
            return self.parent.path() + [(self.action, self.answer)]
```

Systematically create and search through all possible **Nodes**

```
def findSequence(initial,goal):
    q = [Node(None,None,1)]
    while q:
        parent = q.pop(0)
        for (a,r) in [('increment',increment),('square',square)]:
            newNode = Node(parent,a,r(parent.answer))
            if newNode.answer==goal:
                return newNode.path()
            else:
                q.append(newNode)
    return None
```

```
answer = findSequence(1,100)
print 'answer =',answer
```

```
answer = [(None, 1), ('increment', 2), ('increment', 3), ('square', 9), ('increment',
10), ('square', 100)]
```

Focus on constructing objects that represent pieces of the solution.

More later, when we focus on effective **search** strategies.



# Programming Styles for Managing Complexity

---

Task: Find a sequence of operations (either **increment** or **square**) that transforms the integer  $i$  (initial) to the integer  $g$  (goal).

**Imperative** (procedural) approach

- structure of search was embedded in loops

**Functional** approach

- structure of search was constructed in lists of functions

**Object-oriented** approach

- structure of search was constructed from objects

Structure of program has significant effect on its modularity.

Now consider abstractions at even higher levels.

## Controlling Processes

---

Programs that control the evolution of processes are different.

Examples:

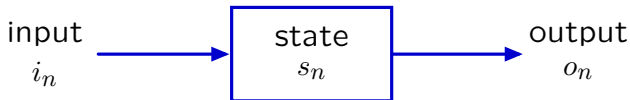
- bank accounts
- graphical user interfaces
- controllers (robotic steering)

We need a different kind of abstraction.

# State Machines

---

Organizing computations that evolve with time.



On the  $n^{\text{th}}$  **step**, the system

- gets **input**  $i_n$
- generates **output**  $o_n$  and
- moves to a new **state**  $s_{n+1}$

Output and next state depend on input and current state

Explicit representation of stepwise nature of required computation.

# State Machines

---

Example: Turnstile

Inputs = {coin, turn, none}

Outputs = {enter, pay}

States = {locked, unlocked}

$$\text{nextState}(s, i) = \begin{cases} \text{unlocked} & \text{if } i = \text{coin} \\ \text{locked} & \text{if } i = \text{turn} \\ s & \text{otherwise} \end{cases}$$

$$\text{output}(s, i) = \begin{cases} \text{enter} & \text{if } \text{nextState}(s, i) = \text{unlocked} \\ \text{pay} & \text{otherwise} \end{cases}$$

$s_0 = \text{locked}$



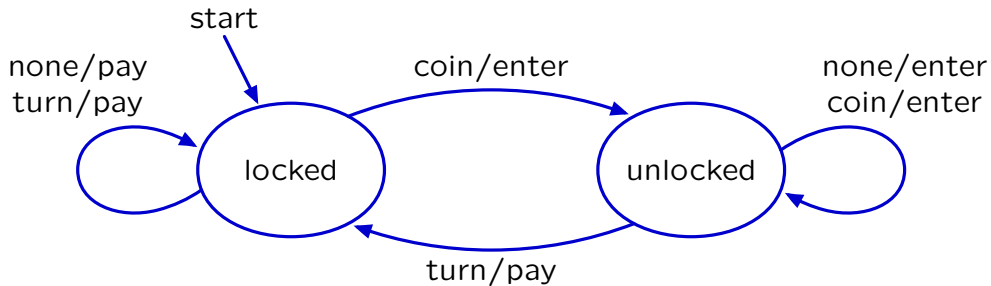
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## State-transition Diagram

---

Graphical representation of process.

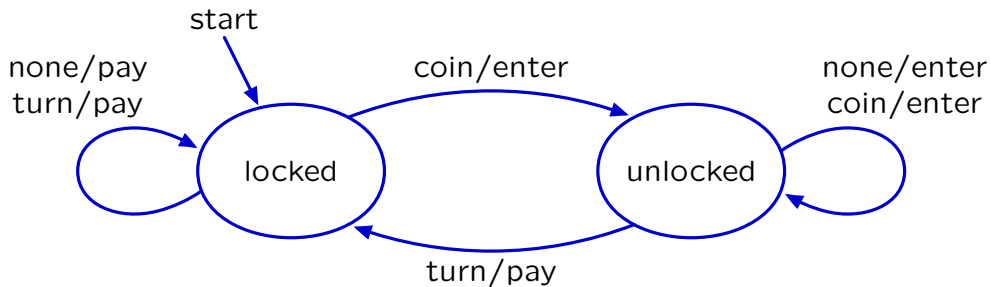
- Nodes represent states
- Arcs represent transitions: label is input / output



## Turn Table

---

Transition table.



---

| time   | 0      | 1      | 2        | 3        | 4      | 5      | 6        |
|--------|--------|--------|----------|----------|--------|--------|----------|
| state  | locked | locked | unlocked | unlocked | locked | locked | unlocked |
| input  | none   | coin   | none     | turn     | turn   | coin   | coin     |
| output | pay    | enter  | enter    | pay      | pay    | enter  | enter    |

---

## State Machines

---

The state machine representation for controlling processes

- is simple and concise
- separates system specification from looping structures over time
- is modular

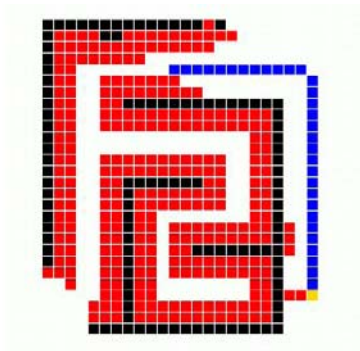
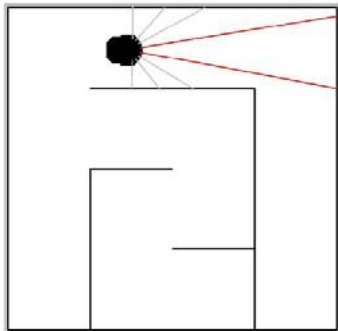
We will use this approach in controlling our robots.

# Modular Design with State Machines

---

Break complicated problems into parts.

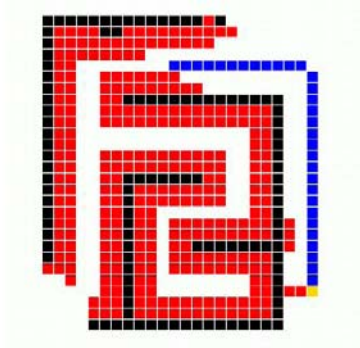
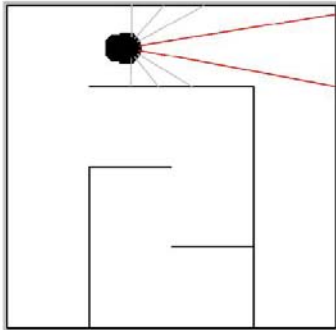
Example: consider exploration with mapping





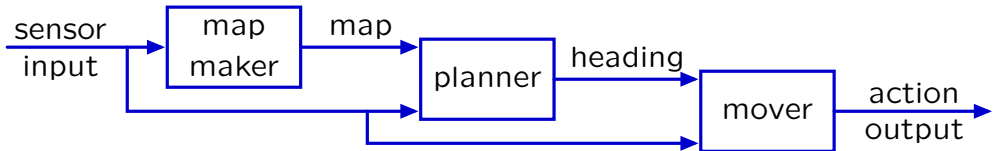
# Modular Design with State Machines

Break complicated problems into parts.



Map: black and red parts.

Plan: blue path, with **heading** determined by first line segment.



## State Machines in Python

---

Represent common features of all state machines in the **SM** class.

Represent kinds of state machines as subclasses of **SM**.

Represent particular state machines as instances.

Example of hierarchical structure

**SM Class:** All state machines share some methods:

- **start(self)** – initialize the instance
- **step(self, input)** – receive and process new input
- **transduce(self, inputs)** – make repeated calls to **step**

**Turnstile Class:** All turnstiles share some methods and attributes:

- **startState** – initial contents of **state**
- **getNextValues(self, state, inp)** – method to process input

**Turnstile Instance:** Attributes of this particular turnstile:

- **state** – current state of this turnstile

## SM Class

---

The generic methods of the **SM** class use **startState** to initialize the instance variable **state**. Then **getNextValues** is used to process inputs, so that **step** can update **state**.

```
class SM:
    def start(self):
        self.state = self.startState
    def step(self, inp):
        (s, o) = self.getNextValues(self.state, inp)
        self.state = s
        return o
    def transduce(self, inputs):
        self.start()
        return [self.step(inp) for inp in inputs]
```

Note that **getNextValues** should not change **state**. The **state** is managed by **start** and **step**.

## Turnstile Class

---

All turnstiles share the same `startState` and `getNextValues`.

```
class Turnstile(SM):
    startState = 'locked'

    def getNextValues(self, state, inp):
        if inp == 'coin':
            return ('unlocked', 'enter')
        elif inp == 'turn':
            return ('locked', 'pay')
        elif state == 'locked':
            return ('locked', 'pay')
        else:
            return ('unlocked', 'enter')
```

## Turn, Turn, Turn

---

A particular turnstyle `ts` is represented by an instance.

```
testInput = [None, 'coin', None, 'turn', 'turn', 'coin', 'coin']
```

```
ts = Turnstile()
```

```
ts.transduce(testInput)
```

```
Start state: locked
```

```
In: None    Out: pay    Next State: locked
```

```
In: coin    Out: enter   Next State: unlocked
```

```
In: None    Out: enter   Next State: unlocked
```

```
In: turn    Out: pay     Next State: locked
```

```
In: turn    Out: pay     Next State: locked
```

```
In: coin    Out: enter   Next State: unlocked
```

```
In: coin    Out: enter   Next State: unlocked
```

```
['pay', 'enter', 'enter', 'pay', 'pay', 'enter', 'enter']
```

# Accumulator

---

```
class Accumulator(SM):  
    startState = 0  
  
    def getNextValues(self, state, inp):  
        return (state + inp, state + inp)
```

## Check Yourself

---

```
>>> a = Accumulator()
>>> a.start()
>>> a.step(7)
>>> b = Accumulator()
>>> b.start()
>>> b.step(10)
>>> a.step(-2)
>>> print a.state,a.getNextValues(8,13),b.getNextValues(8,13)
```

What will be printed?

- 1: 5 (18, 18) (23, 23)
- 2: 5 (21, 21) (21, 21)
- 3: 15 (18, 18) (23, 23)
- 4: 15 (21, 21) (21, 21)
- 5: none of the above

## Classes and Instances for Accumulator

```
a = Accumulator()
```

```
a.start()
```

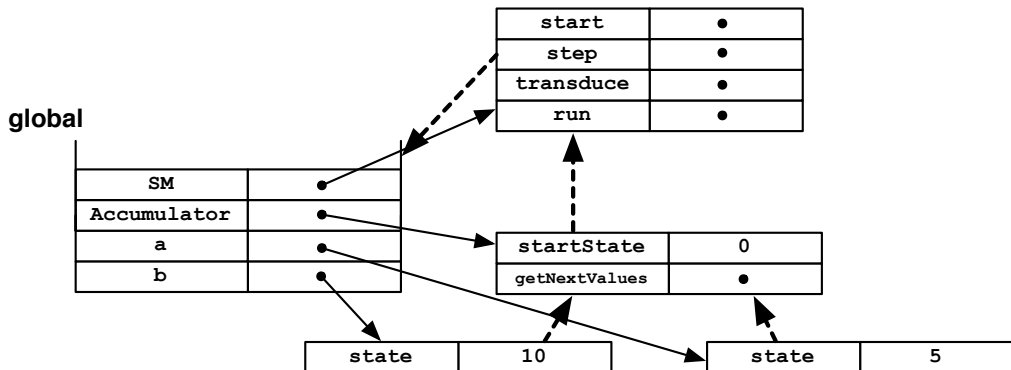
```
a.step(7)
```

```
b = Accumulator()
```

```
b.start()
```

```
b.step(10)
```

```
a.step(-2)
```





## Check Yourself

---

```
>>> a = Accumulator()
>>> a.start()
>>> a.step(7)
>>> b = Accumulator()
>>> b.start()
>>> b.step(10)
>>> a.step(-2)
>>> print a.state,a.getNextValues(8,13),b.getNextValues(8,13)
```

What will be printed? 2

1: 5 (18, 18) (23, 23)

2: 5 (21, 21) (21, 21)

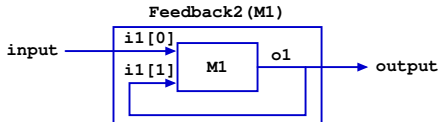
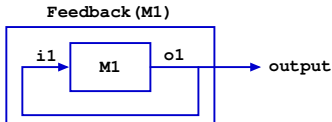
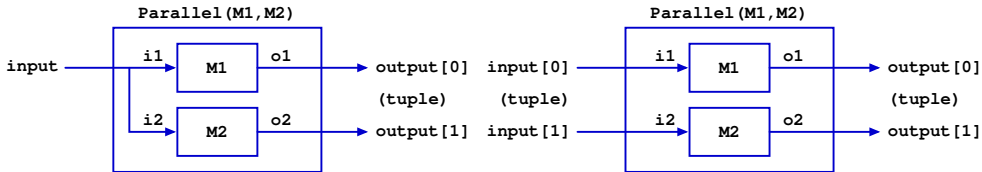
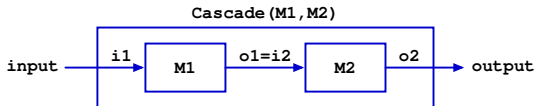
3: 15 (18, 18) (23, 23)

4: 15 (21, 21) (21, 21)

5: none of the above

# State Machine Combinators

State machines can be **combined** for more complicated tasks.



## Check Yourself

---

```
>>> a = Accumulator()
>>> b = Accumulator()
>>> c = Cascade(a,b)
>>> print c.transduce([7,3,4])
```

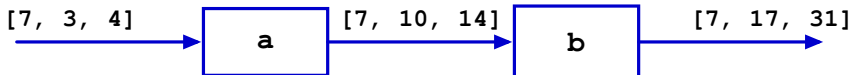
What will be printed?

- 1: [7, 3, 4]
- 2: [7, 10, 14]
- 3: [7, 17, 31]
- 4: [0, 7, 17]
- 5: none of the above

## Check Yourself

---

```
>>> a = Accumulator()
>>> b = Accumulator()
>>> c = Cascade(a,b)
>>> print c.transduce([7,3,4])
```



## Check Yourself

---

```
>>> a = Accumulator()
>>> b = Accumulator()
>>> c = Cascade(a,b)
>>> print c.transduce([7,3,4])
```

What will be printed? **3**

- 1: [7, 3, 4]
- 2: [7, 10, 14]
- 3: [7, 17, 31]**
- 4: [0, 7, 17]
- 5: none of the above

## This Week

---

**Software lab:** Practice with simple state machines

**Design lab:** Controlling robots with state machines

**Homework 1:** Symbolic calculator

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6.01SC Introduction to Electrical Engineering and Computer Science  
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