### 15-494/694: Cognitive Robotics

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#### Lecture 4:

Advanced State Machine Concepts, and Introduction to Particle Filters

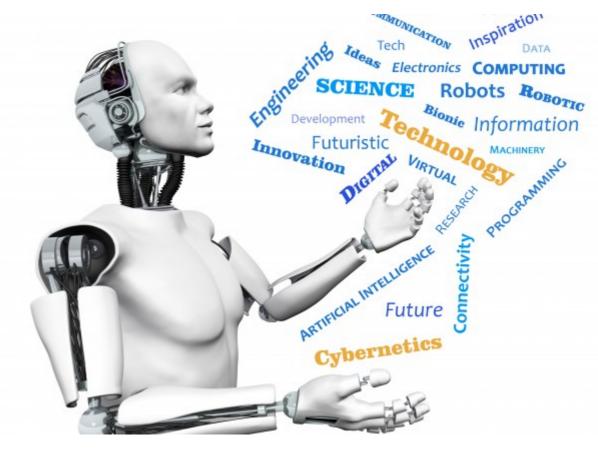


Image from http://www.futuristgerd.com/2015/09/10

### Differences From Classical FSMs

#### 1. Multi-State:

 Multiple states can be active simultaneously (fork), and their completions can be synchronized (join).

#### 2. Hierarchical:

- State machines can nest.

#### 3. Message Passing:

 One state can send a message to another as part of a transition firing.

### More On Hierarchy

- A nested state machine is started automatically when its parent node starts.
- The nested machine can cause its parent to signal completion by:
  - Transitioning to a ParentCompletes node
  - Calling self.parent.post\_completion()
     from inside one of its nodes.
- Similarly for signaling parent success or failure: ParentSucceeds or ParentFails.

#### **Nested State Machines**

```
DingDong has an empty start() method,
but it has a setup() method.
class Doorbell(StateNode):
  $setup {
     ding: Say('ding') =C=>
       dong: Say('dong') =C=>
          ParentCompletes()
                     Doorbell
                      <mark>--</mark>dong<mark>--</mark>l
```

#### **Nested State Machines**

```
class Nested(StateMachineProgram):
  $setup {
     db1: Doorbell() =C=>
        bridge: Say('once again') =C=>
           db2: Doorbell()
                                          Will be triggered
                                           by db1's nested
                                          ParentCompletes
                                              node
   db1: Doorbell
                                db2: Doorbell
                  bridge: Say
    ding → dong → P
```

## Message Passing

 Nodes can signal "data events" that data transitions look for:

```
self.post_data(5)
```

Transitions can match the data item:

```
foo =D(5)=> draw_pentagram
foo =D(6)=> draw_hexagram
```

Transitions can also do wildcard match:

```
foo =D=> draw stuff
```

# Message Passing (cont.)

 When a transition activates a node, the node's start method is passed the event that triggered the transition.

 If this was a DataEvent, the start method can extract the data item and process it.

## Sending Data

```
class Sender(StateNode):
    def start(self, event=None):
        super().start(event)
        value = random.random()
        self.post data(value)
```

### Receiving Data

```
class Receiver(StateNode):
    def start(self, event=None):
        super().start(event)
        if isinstance(event, DataEvent):
        value = event.data
        print('Value received:', value)
```

# Sending and Receiving

```
C> runfsm('SendRecv')
Value received: 0.380313711
```

#### Iteration

Use =CNext=> instead of =Next=> to wait for completion.

### **Default Transitions**

For data events and text message events, value matches take priority over defaults.

```
foo =TM('cat')=> Say('meow')
foo =TM('dog')=> Say('woof')
foo =TM=> Say('wacka-wacka')
```

How does this work? Default (wildcard) transitions have a slight time delay to allow any matching value transition to fire first.

### Tap Events

- The SDK generates tap events when someone taps on a cube.
- We turn these into cozmo\_fsm TapEvents that can be matched by a =Tap=> transition:

```
=Tap(cube2)=>
=Tap=>
```

 We need to check the tap intensity to reject false positives.

#### **Face Events**

- The SDK generates face events whenever a face is detected in the camera image.
- We turn these into cozmo\_fsm
   FaceEvents that can be matched by a
   =Face=> transition:

```
=Face('Dave')=>
=Face=>
```

 Should probably provide separate cases for FaceAppeared and FacePresent.

### The Event Loop

 While the SDK is connected to the robot and simple\_cli is running, the value of asyncio.get\_current\_event\_loop() is available in robot.loop.

 From simple\_cli, in order to run a node we have to schedule it via this event loop.

This is what the now() method does:
 Forward(50).now()

#### Do It "Now"

```
class StateNode(EventListener):
    ...
    def now(self):
        self.robot.loop.
        call soon(self.start)
```

#### EventListener

 Parent class of both StateNode and Transition.

 Includes a polling feature: an instance can request that its poll() method be called every t seconds.

Polling begins when the instance's start()
method is called and ends when stop() is
called.

### Uses of Polling

 DriveForward and DriveTurn use polling to check the robot's progress and decide when to stop.

 TimerTrans uses the polling interval to know when to fire.

 ArucoTrans uses polling to check if a marker has appeared in the camera image.

## **Animation and Trigger Nodes**

 Animation nodes take an animation name as a string argument. There are over 900 to choose from.

```
AnimationNode('anim bored 01')
```

 AnimationTriggerNodes take an \_AnimTrigger object as an argument. AnimationTriggerNode( cozmo.anim.Triggers. CubePouncePounceNormal )

### **Named Transitions**

- A complex state machine may have a lot of CompletionTrans, SuccessTrans, and TimerTrans transitions.
- This makes the trace confusing: what is completiontrans5 doing?
- Solution: assign meaningful names to your transitions.

```
try_grab =grabbed:C=> open_it
try grab =fumbled:F=> reposition
```

### Writing Your Own Transitions

 Rarely necessary, unless you're developing new robot functionality.

#### How to do it:

- · \_\_init\_\_() to store constructor parameters.
- · start() to subscribe to events if needed.
- handle\_event() to examine the events and call self.fire(event) if needed.
- · poll() if polling is needed.

### SeeBoth Transition

```
class SeeBoth(Transition):
 def init (self,thing1,thing2):
    super(). init__()
    self.thing1 = thing1
    self.thing2 = thing2
    self.set polling interval(0.1)
 def poll(self):
    if self.thing1.is visible and
         self.thing2.is visible:
      self.fire()
```

#### See12.fsm

### simple\_cli 'show' commands

- show active
  - Shows the currently active nodes and transitions.
- show viewer
  - Shows the camera viewer
- show worldmap\_viewer
  - Shows the worldmap viewer

### **Particle Filters**

### Intro to Particle Filters

- Odometry is unreliable.
  - Still useful for short trajectories.
  - But error accumulates quickly.
- Solution: use visual landmarks to correct for odometry error.
- But vision is unreliable too!
  - Landmark pose estimation is noisy.
  - Landmarks aren't always available.

### **Probabilistic Robotics**

 Probabilistic robotics is based on the idea that we should embrace the noisiness.

 Instead of discrete values, think in terms of probability distributions.

 Robot's location is not (x,y), but a distribution of possible locations, some more <u>likely</u> than others.

### Modeling Location Distributions

 Particle filters are a way to model distributions.

- Think of each particle as a "guess" (hypothesis) about the robot's location.
- Assume we have a map with landmarks.
- Each guess predicts how the landmarks should look from that location.

### Modeling Location Distributions

- Particles representing good guesses will accurately predict the landmark locations.
  - Good predictions earn a high weight.
- Bad guesses lead to poor predictions.
  - Poor predictions result in a low weight.
- As we accumulate sensor data, we can figure out which particles are the good guesses.

### Particle Filter Demos

 particle\_filter\_demo is linked from the class schedule and can be found in the Class/demos" directory.

 It shows the robot wandering in a maze; walls are landmarks.

- The robot's estimated location (white circle) is the average of all the particles.
- pfdemo.py is another demo you can try.

## Resampling

- Bad guesses are a waste of resources.
- When we've accumulated enough data, we can generate a new set of particles to try to concentrate resources in the region of good guesses.
- Particles with high scores are chosen to spawn new particles.
- Low-scoring particles are unlikely to spawn.

### **Motion Model**

 So far we have a robot that is standing still, receiving sensor data, and trying to figure out its location on the map.

- But the robot needs to move.
  - Stationary robots aren't useful.
  - Motion allows the robot to see more landmarks.

### Motion Model (cont.)

- How can we accommodate motion?
- Solution:
  - As the robot moves, drag the particles along with it.
- But odometry is noisy!
  - Add noise (via a motion model) to the particle locations because we know that motion is unreliable.

#### **SLAM**

What if we don't have a world map?

 SLAM: Simultaneous Localization And Mapping.

 Now each particle represents a slightly different map of the world, <u>plus</u> the robot's estimated location on that map.

We will look at this in the next lecture.