Extending the Reach of Cognitive Architectures

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with support from





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Cognitive Architectures

- Many major successes...
 - several major architectures
 - leaps in scientific understanding of cognition
 - models of hundreds of domains
 - real-world applications (tutoring, gaming, driving)
- ... but still a limited user base.
- How can we extend their reach?

Extending Cognitive Architectures

- Extend functionality with new models
 - General processes (e.g., list memory & recall, analogy)
 - Large-scale knowledge bases
 - Natural language processing rules
 - Fake news filter :)
- Extend functionality with architectural components
 - Production-rule learning (compilation, utility learning, ...)
 - Metacognition
 - Threaded cognition
 - Physiological moderators (fatigue, caffeine, affect, ...)
 - Fatigue modeling ...

Fatigue & Driving

- Experiment (Forsman et al., 2013; Van Dongen et al., 2010, 2011)
 - 41 drivers in a driving simulator
 - Conditions comparing Day- vs. Night-shift driving



Fatigue & Driving

PVT



Driving



Time points within days

(Khosroshahi, Salvucci, Veksler, & Gunzelmann, 2016)

Extending Cognitive Architectures

- Extend computational paradigms for modeling
 - Production systems have been the dominant paradigm
 - primary benefits: flexible for learning, modular for integration
 - But in other places, production systems have largely disappeared, especially from CS curricula
 - and thus, we have a generation of students/programmers who have never learned or seen them
 - Their proposed benefits have largely not been realized
 - flexible for learning? somewhat
 - good for integration? hasn't really happened

Production Systems

- ACT-R code for...
 - storing information about
 Whiskers the cat
 - recalling this information
- Someone new to cognitive modeling will need hours (maybe days) to learn enough to understand this

```
(p rule1
==>
    +imaginal>
         isa cat
         name Whiskers
         owner Jane
    -imaginal>
(p rule2
    •••
    ?retrieval>
         state free
        buffer empty
==>
    +retrieval>
         isa cat
         owner Jane
(p rule3
    =retrieval>
         isa cat
        name =name
==>
```

• Instead, here's a cognitive code version...

Java	<pre>memory.store(new Chunk("cat").set("owner", "Jane").set("name", "Whiskers")); Chunk chunk = memory.recall(new Query("cat").add("owner", "Jane"));</pre>
Python	<pre>memory.store(m.Chunk({'isa': 'cat', 'name': 'Whiskers', 'owner': 'Jane'})) chunk = memory.recall(core.Query('cat').eq('owner', 'Jane'))</pre>

- All college CS students (and many younger students) could read a tutorial and understand this in minutes
- They already know the syntax/semantics of the language, and can focus on understanding the cognition part
- (e.g., what happens if the chunk isn't recalled? hard to guess in ACT-R, easy to guess in Java)

- Core simulation system
 - central Clock maintained across threads
 - threads can delay simulated time as needed:

```
agent.wait(1.0);
memory.store(new Chunk("cat"));
agent.wait(2.0);
Chunk chunk = memory.recall(new Query("cat"));
```

- (as usual, simulation time \neq real time)
- system also includes various helper classes (modules, workers, buffers, items, ...)

- Easy to use just one module
 - e.g., what's the recall probability/time if ... ?

```
Agent agent = new Agent();
                                                Create memory
Memory memory = new Memory(agent, decay);
                                                   and chunk
memory.setActivationNoise(0.5);
Chunk chunk = new Chunk("cat").set("name", "W
memory.store(chunk);
agent.wait(.5);
memory.rehearse(chunk);
                                                Simulate timing
agent.wait(2.5);
                                                  of rehearsal,
memory.rehearse(chunk);
                                                     waiting
agent.wait(3.7);
memory.rehearse(chunk);
agent.wait(6.2);
   memory.computeProbabilityOfRecall(chunk)
...
                                                Calculate results
  memory.computeTimeToRecall(chunk)
```

• Dual-choice task

```
agent.run(() -> {
    Object tone = audition.encode(audition.waitFor(new Query("tone")));
    if (tone.equals("low"))
        speech.say("low");
});
agent.run(() -> {
    Object stimulus = vision.encode(vision.waitFor(new Query("stimulus")));
    if (stimulus.equals("0--"))
        typing.type("1");
});
agent.wait(1.0);
vision.add(new Visual("stimulus", 10, 10, 10, 10, 10), "0--");
audition.add(new Aural("tone"), "low");
agent.waitForAll();
```

• Dual-choice task

0.000	agent	wait for {isa=stimulus}
0.050	agent	<pre>wait for {isa=tone}</pre>
1.000	agent.vision	<pre>found {isa:"stimulus" x:10 y:10 w:10 h:10 seen:false}</pre>
1.000	agent	<pre>encode {isa:"stimulus" x:10 y:10 w:10 h:10 seen:false}</pre>
1.050	agent.audition	<pre>found {isa:"tone" heard:false}</pre>
1.050	agent	encode {isa:"tone" heard:false}
1.185	agent.vision	encoded 0
1.185	agent	type "1"
1.235	agent, hands	typing "1"
1.385		-1F=
1.385	Read	Encode Respond to Respond to
1.435	Procedural stimulus	tone right stim. low tone
1 444		
1 795	Visual	reading
1.705	- <u> </u>	
	encoding	encoding
	Aural delay	low tone
		nressing
		pressing

ring finger

saying

"one"

Manual

Vocal



Game parameters

- Mole size (larger = easier)
- Mole time (longer = easier)





```
agent = new Agent().setLog(OUTPUT);
vision = new Vision(agent);
hands = new Hands(agent);
                                                                                          Setup
mouse = new Mouse(hands, vision).setRandomizeTime(true).addClickListener((visu
      clickMole();
});
•••
agent.run(() -> {
      while (agent.getTime() < RUN_TIME) {</pre>
                                                                                         Model
             Visual visual = vision.waitFor(new Query("mole"));
             mouse.pointAndClickMouse(visual);
      }
});
agent.run(() -> {
      while (agent.getTime() < RUN_TIME + 3.0) {</pre>
             moleVisible = true;
             int x = Utilities.random.nextInt(500);
             int y = Utilities.random.nextInt(500);
             Visual mole = new Visual("mole", x, y, moleSize, moleSize);
                                                                                           Task
             vision.add(mole, "X");
             agent.wait(moleTime);
             moleVisible = false;
             vision.clear();
             agent.wait(1.5 - moleTime);
      }
});
agent.waitForAll();
```

• Score vs. mole size & time in a 30-second game...



Extending the Reach...

- (1) Production rules are a huge limiting factor.
 - Can we maintain 90% of the theory without production rules, and make ACT-R much more accessible?
 - Current "cognitive code" prototypes in Java and Python are heading in this direction
- (2) We need to focus more on out-of-the-box predictions.
 - Most people don't have data and won't do model fitting
 - What can we give them such that, within minutes, they can make predictions about their domain of interest?
 - Give the model a {game board, essay, car interface, ...} the model will give you predicted behaviors + analysis