Bob Marinier 27<sup>th</sup> Soar Workshop May 24, 2007

### **SESAME & SOAR: A COMPARISON**



## **SESAME OVERVIEW**

- × SESAME is a theory of human cognition
  - + Stephan Kaplan (University of Michigan)
  - + Modeled at the connectionist level
  - + Mostly theory, not implementation
  - + Basis in perception
  - + Associative (Hebbian) learning used to explain a lot
  - Inspired more by animal and neural studies
     Soar and ACT-R inspired more by human behavior
     Emphasis on cortical areas of brain
     Not basal ganglia, hippocampus, etc.

## MOTIVATION

- SESAME has some striking similarities to Soar, which may provide insight into the basis of those aspects
  - + Neural basis of rules, persistence, etc.
- Different emphasis that should be complementary to Soar's approach
- May provide a useful perspective on lots of things Soar is exploring these days
  - Working memory activation, clustering, sequencing, semantic memory, episodic memory, reinforcement learning, visual imagery

# OUTLINE

- × For each topic:
  - + Describe topic from SESAME's perspective
  - + Compare to Soar
  - + Give possible inspiration/insight/lesson for Soar

### × Topics:

- + Cell Assemblies (Symbols)
- + Memory (LTM and WM)
- + Activation
- + Persistence
- + Learning
- + Sequences
- + Episodic vs Semantic Memory
- + Metacognition
- + "Magic" of Human Cognition
- + Summary

# **CELL ASSEMBLIES**

- How does the brain recognize an object in different situations?
  - + Some (random) neurons fire in response to specific features (e.g. color, size, texture, etc)
  - + Neurons that fire together wire together
  - After many experiences, a group of neurons representing common features for an object become associated as a unit called the cell assembly (CA)

## **CELL ASSEMBLIES**

Cell assemblies are

- + Grounded in perception
- + Categories
- + Concepts
- + Prototypes
- + Symbols (but not in the full Newellian sense)

 Abstraction & Hierarchy: CAs at one level serve as features for the next level of CAs



- × Symbols are CAs
- × CAs are not fully Newellian
- CAs are grounded in perception
- CAs are categories, concepts, prototypes

#### SOAR

- Symbols are abstract basic unit
- × Symbols are fully Newellian
- Symbols can be grounded in perception
- Symbolic structures are categories and concepts, and can be prototypes

Insight: Where symbols come from, properties of symbols

### **COMPARISON: SYMBOLS**

## MEMORY

- CA structures are long-term memories
- × Working memory is the set of active CAs
  - + Activation is in-place (no retrievals or buffers)
- × Limited Working Memory Capacity
  - + Regional Inhibition: When CAs activate, they interfere with other nearby CAs
    - × CAs compete in winner-take-all fashion to become the active representation for object/thought
  - + Limits possible number of active CAs (WM capacity)
    - × Roughly 5±2 for familiar CAs, which tend to be more compact

- × LTM is network of all CAs
- × WM is set of active CAs
  - + Uses existing structure
- × WM is limited

#### SOAR

- LTM includes Production Memory, Semantic Memory, Episodic Memory
- WM is set of elements created or retrieved from LTM
  - + Creates new structure
- × WM is not limited

Insight: Same structure for LTM and WM, WM limitations

### **COMPARISON: MEMORY**

## ACTIVATION

Activity of a CA is dependent on factors including:

- + Connections from other active CAs
  - × Incoming connections may be excitatory or (locally) inhibitory
  - × Required set of active/inactive connections may be complex
- + Reverberation: Positive feedback allows CA to remain active beyond incoming activity
- + Fatigue: As CA remains active, threshold for activation increases
- May be able to describe spread of activation among CAs in rule form:

D

+ If A and B are active and C is inactive, then D activates.

В

- Activation spreads based on rule-like learned connections
- × Activation impacted by incoming connections, reverberation, inhibition, fatigue

### × Spread of activation and CA activation are same thing Lesson: Neurologically-accurate WM activation model

#### SOAR

- × Symbol creation propagates via elaboration rules
- × Activation based on activation of symbols that cause rule match, boost from usage, and decay
- Symbol creation and activation are different

### **COMPARISON: ACTIVATION**

## PERSISTENCE (CONTROL)

- May need to keep a CA around for a while (e.g. to work on a problem)
- × Other "distraction" CAs can interfere
- Inhibitory attention blankets all CAs in (global) inhibition
  - + Highly active CAs are impervious to effect
  - + Weaker distractions are inhibited

- Persistence achieved via inhibitory attention
  - Prevents activation of distractor CAs

#### SOAR

- Persistence achieved via operator selection and application
  - Selection of an operator inhibits selection of other operators (and creation of associated symbols)

Insight: None really – Soar already uses inhibitory mechanism

### **COMPARISON: PERSISTENCE (CONTROL)**

## LEARNING

### × Associative (Hebbian)

 Learns associations between CAs that are often active concurrently (CAs that fire together wire together)

Includes sequentially active CAs, since CAs reverberate

- Learns lack of association between CAs that are not commonly active concurrently
  - × Results in (local) inhibitory connections
- Learning rate is typically slow, but high arousal causes fast learning

- All learning is associative (doesn't really cover RL)
- Learning is typically slow (but modulated by arousal)

#### SOAR

- × Many types of learning
  - + Chunking
  - + Semantic
  - + Episodic
  - + Reinforcement
- Chunking, semantic and episodic are fast, reinforcement is typically slow (but modulated by learning rate)

Insight: Proliferation of learning types in Soar results from proliferation of memory types, role of arousal in learning

### **COMPARISON: LEARNING**

# SEQUENCES

- Sequences are stored in cognitive maps
- Cognitive maps are "landmark"-based maps of problem spaces
  - + Nodes are CAs
  - + Connections represent CAs that have been experienced in sequence
  - + Since experienced sequences overlap, novel sequences are also represented (composability)
- Problem solving involves finding paths through cognitive maps
- Paths may be associated with "affective" codes that help guide the search
  - + Codes learned via reinforcement learning

- Sequences stored in cognitive maps
- Can achieve limited composability
- Problem solving is searching through cognitive map (which represents problem space)
- RL helps improve search

#### SOAR

- Sequences can be stored in operator application rules or in declarative structures
- Can achieve arbitrary composability
- Problem solving is search through problem space
- × RL helps improve search

Insight: Limited composability may be enough

### **COMPARISON: SEQUENCES**

# **EPISODIC VS SEMANTIC**

- CAs are typically derived from multiple overlapping experiences
  - + Thus, tend to be semantic in nature
- A highly-arousing event may be strong enough to form its own CA
  - + Thus, can have episodes
  - Semantic Memory Formation

Episodic Memory Formation



### **EPISODIC VS SEMANTIC**

- In general, there is no clear distinction between semantic and episodic memories
  - + CAs include full spectrum between episodic and semantic
  - + Each time a CA is active, can be modified (allows for episodic memory modification)
- Hippocampus thought to play a role in contextualizing episodic memories, but not in storage

- × No clear distinction
  - + CAs encode both kinds of memories with a smooth transition
- Story on role of hippocampus is not completely worked out
  - Memories are not stored in hippocampus

#### SOAR

- Episodic and semantic memories are learned, stored and retrieved separately
- Episodes are assumed to be initially stored in hippocampus before migrating to cortex

Insight: May not need separate episodic and semantic memories

### **COMPARISON: EPISODIC VS SEMANTIC**

## METACOGNITION

- Brain monitors CA activity to determine current state
  - + Focused, high levels of activation: Clarity
  - + Diffuse, lower levels of activation: Confusion
- Serves as signals about how processing is going
  - + Provides opportunity to change processing
- Clarity/Confusion experienced as pleasure/pain
   + Can influence learning
  - eleanning

- Clarity/Confusion signal how things are going
- Influence learning via pleasure/pain signals
- × Details are sketchy

#### SOAR

- Impasses arise when processing cannot proceed
- Allows for learning via chunking

Lesson: None really – impasses provide same functionality

### **COMPARISON: METACOGNITION**

## **MAGIC OF HUMAN COGNITION**

- × Special mechanisms
  - Human perceptual mechanisms are different than other animals
  - + Leads to different features that CAs learn over
- × Quantitative differences
  - Many animals have CAs and association mechanisms, but the larger quantity in humans may lead to qualitative differences

 In other words: There is no single mechanism that gets us the "magic" -- interaction of all pieces is necessary

x Everything is necessary
x Everything is necessary

SOAR

### Laird's lesson: "There is no magic, just hard work"

### **COMPARISON: MAGIC OF HUMAN COGNITION**

### HIGHLIGHTS: WHAT SOAR CAN LEARN FROM SESAME

- SESAME ideas can provide grounding and inspiration for extensions to Soar
  - + Associative learning can get you:
    - × Non-arbitrary symbols via clustering-type mechanism
    - × Sequences
  - + Working memory
    - × Soar's activation model could account for more features
      - \* Reverberation
      - \* Fatigue
      - Inhibition (local, regional, and global)
    - × Basis for limited capacity
  - + Arbitrary composability may not be necessary
  - + The role of arousal in learning
  - Episodic/Semantic memories may not be as distinct as they are in Soar