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Pervasive Activation: Applying the mechanism to declarative and procedural memory

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#### SOAR AND SHORT-TERM MEMORY EFFECTS

- Newell (1990) proposed Soar as a candidate UTC.
- UTC constrains mechanisms to those that are *functionally* necessary for producing intelligent behavior.

"[Soar] is entirely functional...No mechanisms...have ever been posited just to produce some empirically known effect..." (pp. 309-310)

• No mechanism for short-term memory effects...

"...the only short-term memory effects...are those rooted in mechanisms that have some functional role..." (Ibid)

• Example: Functional limit on WM capacity in sentence comprehension (Young & Lewis, 1999).

## • Consequences

- *Plausible* and *principled* modeling of some behavior can be difficult or impossible.
- Example: Behavior where performance is influenced by short-term-memory effects.
- With no *architectural* mechanism, the modeler has to create with their own "model" for short-term memory effects.
- Soar contributes little to this important modeling area.
- Solution

"...To exhibit [short-term memory] effects, Soar would need to be augmented with additional architectural assumptions about these mechanisms and their limitations." (Ibid)

- Borrow the activation and decay mechanisms as defined and used in ACT-R 4.0
  - Rudimentary implementation was done in 2000.
  - Significant improvements were recently made.
- Altmann & Schunn (2002) propose a functional role for decay.

"We argue, based on a simple functional analysis, that...distracting information must decay to allow the cognitive system to have any hope of retrieving target information amidst the unavoidable clutter of a well-stocked memory."

• Perhaps this new mechanism is not breaking with the UTC philosophy after all.

# • Basics:

- Based on ACT-R.
- When a WME is created, it is given an initial (*base-level*) activation.
- Activation is a function of the recency and use.
- Activation decays exponentially.
- An element is "forgotten" when its activation falls below the *retrieval threshold*.



- ACT-R
  - All WMEs (working memory elements; chunks) have activation.
- Soar
  - A *partition* of elements in WM have activation.
  - a-memory is the "activated" partition.
  - blip-color is like an ACT-R "chunk-type".
  - items are instances of a type.
  - (bc item bc0) and (bc item bc1) are flagged as having activation.



#### COMPUTATION OF ACTIVATION — SOAR

- Equation 1:  $A_i = B_i + \Sigma W_j S_{ji} + \epsilon_1 + \epsilon_2$ A WME's activation (A<sub>i</sub>) is the sum of its "inherent" activation (B<sub>i</sub>), the contribution of associated WMEs ( $\Sigma W_j S_{ji}$ ) and one noise terms ( $\epsilon_1$ ,  $\epsilon_2$ )
- Equation 2:  $B_i = \beta + \ln(\Sigma t_j^{-d})$

A WME's "inherent" activation ( $B_i$ ) is the sum of its initial (base-level) activation ( $\beta$ ) and a calculation of the recency and frequency of use

• Equation 3:  $\epsilon_{1,2} = ns_{1,2} * log[(1.0 - p) / p]$ p = rand[0.0, 1.0]

Noise terms ( $\epsilon_1$ ,  $\epsilon_2$ ) are sampled from a logistic distribution

#### NUMBER OF PARAMETERS

- ACT-R
  - decay-rate (d)
  - retrieval threshold (rt)
  - base-level constant ( $\beta$ )
  - permanent noise  $(\epsilon_1)$
  - transient noise  $(\varepsilon_2)$
- Soar
  - decay-rate (d)

 $\rightarrow$  permanent noise ( $\epsilon_1$ )

- retrieval threshold (rt)
- base-level constant ( $\beta$ )
- NEW: transient noise (e<sub>2</sub>)

- ACT-R
  - A WME used to fire a production
  - A new WME, created internally or by the environment, is identical to an existing WME; "chunk merging".
- Soar
  - An *activated* WME is used to fire a production (with one exception).
  - NEW: A new activated WME, created internally or by the environment, is identical to an existing WME; "WME merging".
  - NEW: When deciding between a number of competing operators, only the activated WME in the proposal of the selected operator is boosted.

- ACT-R
  - When a WMEs activation falls below threshold, it remains in memory but is not available to match productions.
- Soar
  - Version 0: The sub-retrieval-threshold WME was removed from working memory.
  - This is no longer the case.
  - NEW: The sub-retrieval-threshold WME is removed from the Rete (to prevent it from matching productions) but remains in working memory (to facilitate debugging and WME merging).

#### NEW POSSIBILITIES: ACTIVATION AND THE DECISION CYCLE

### • Activation-based operator selection

- Indifferent preferences direct the decision procedure to randomly pick among candidates.
- Instead of choosing randomly, the decision procedure can be made to choose the proposal that referenced the most highly *activated* WME/s.
- This is similar to activation-based retrieval in ACT-R 4.0; WME activation is one of the criteria used to select which instantiation to fire.

- ACT-R
  - ACT-R uses spreading activation to cause the cue to increase the activation of the target.
- Soar
  - Unimplemented (for the moment).
  - When a WME has been merged, a special recognition WME will be added to WM.
  - This recognition WME has activation and will decay if not used.

#### APPLYING ACTIVATION TO PROCEDURAL MEMORY

- A fundamental feature/commitment of Soar is that learned knowledge cannot be forgotten.
- In general, "Practice makes perfect" is not applicable to Soar models.
- Mechanism only applies to chunks (learned productions).
- Rules written by the modeler are *not* subject to forgetting.
- Frequently used (practiced) chunks have their activation reinforced.
- Infrequently used (unpracticed) chunks would be forgotten.
- Forgotten rules can *usually* be learned again; depends on the context.
- Relearning tends to reduce the likelihood a chunk will be forgotten again.
- Have a basic implementation, but still debugging...

#### NUGGETS

- Combining tested mechanisms from other architectures.
- New Soar modeling opportunities:
  - Used in a model of eye scan patterns and overall performance in a simulated ATC task.
  - Certain errors are emergent.
  - Used in a new Soar category learning model.
  - Models now sensitive to time.
  - Efficiency improvements to the mechanism and explorations in episodic learning and memory—graduate student research @ Michigan.

### COAL

- Runtime costs.
- What's missing?
  - Spreading activation
  - Influence of activation on cycle time
    - activation 🗯 match time 🗰 cycle time
  - An account of interference
- How to "rehearse" chunks?