Hierarchical organization of intelligence Ethology and Al perspectives Joanna J. Bryson Department of Computer Science, University of Bath

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Outline

- History: Ethology and Al
- Parallel-rooted, Ordered, Slip-stack
 Hierarchical (POSH) Action Selection
 - Controlling artificial life
 - Explaining NI task learning
- Hierarchies in brains and other evolved intelligence

Ethology: Controversy

- McDougall, "Hierarchy of Instincts" 1923
- Lorenz, "denied the existence of superimposed mechanisms controlling the elements of groups... particular activity was only dependent on the external stimulation and on the threshold for release of that activity."

(Horst Hendriks-Jansen 1996)

Ethology: Evidence

- Lashley (1951) showed action sequencing happens too fast to wait for sense/act cycle (c.f. Davelaar 2007).
- Dawkins (1975) argued for hierarchy based on ethological parsimony (in a tribute to Tinbergen).
 - Proposed "parsing" structure.

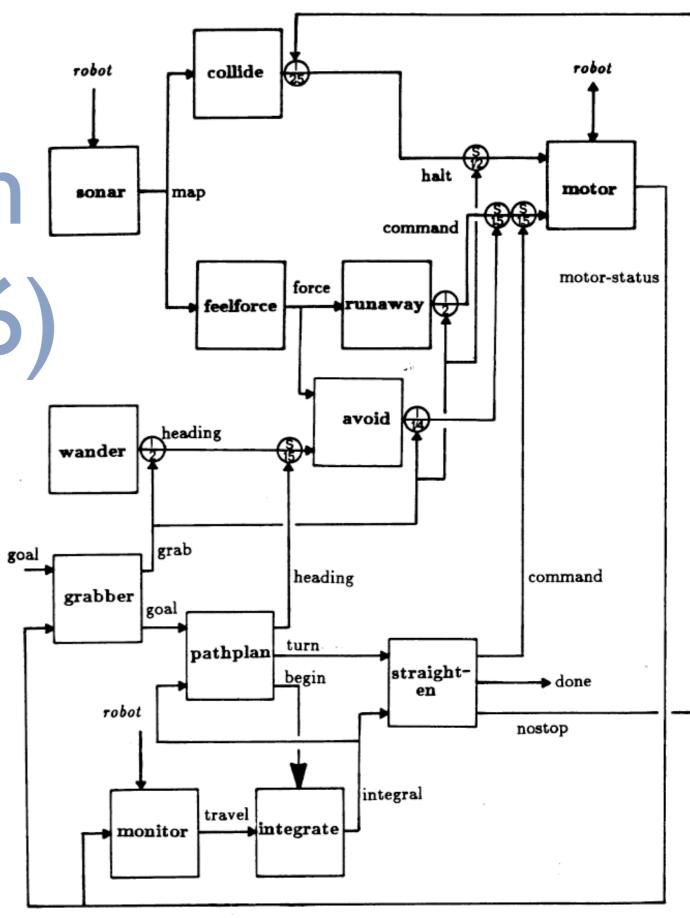
(Bryson 2000)

AI: GOFAI

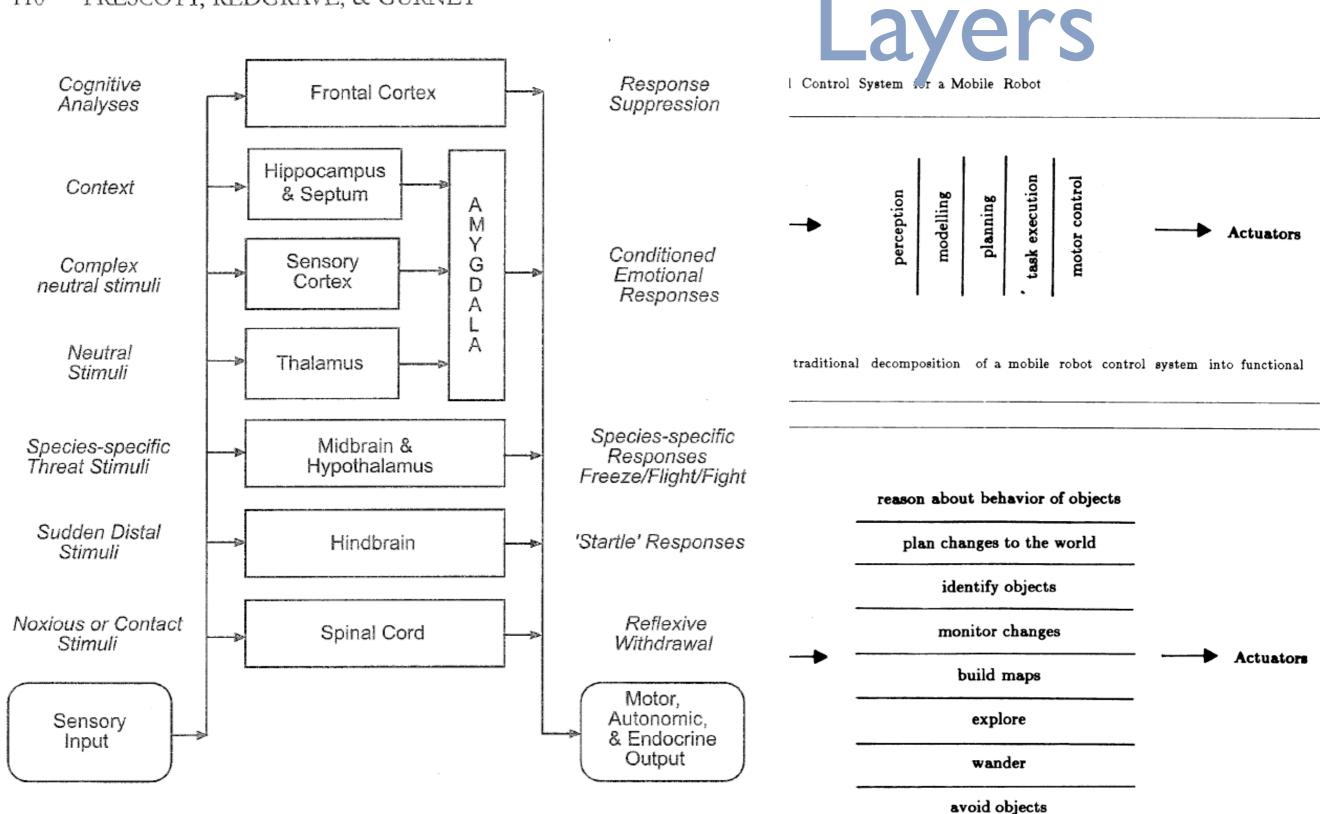
- GPS: You can't get anywhere without problem spaces (combinatorics.)
- SOAR inherited these; ACT-R tried to do without --- briefly!
- Also: notion of subplans.



- Intelligence derives only from constantly concurrent behavior modules.
- And inhibition, suppression, timers...

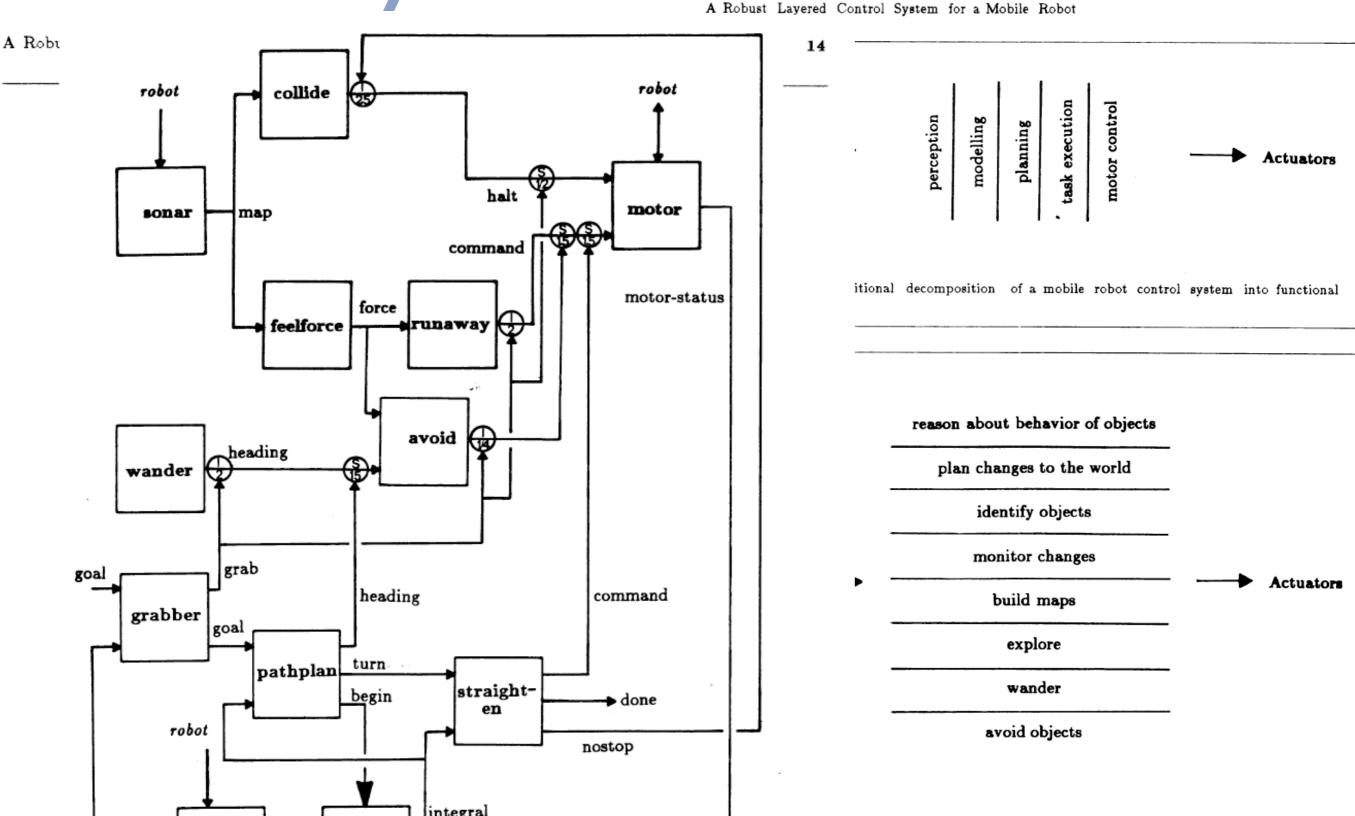


PRESCOTT, REDGRAVE, & GURNEY



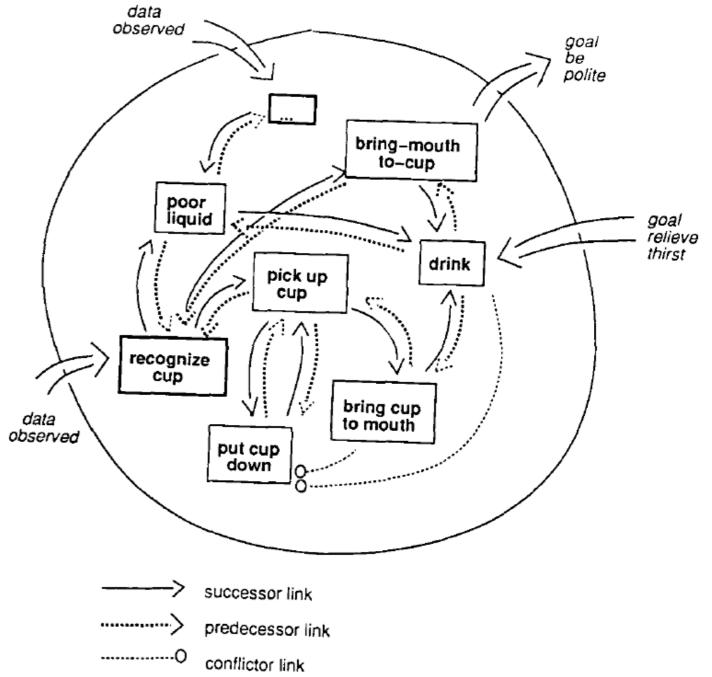
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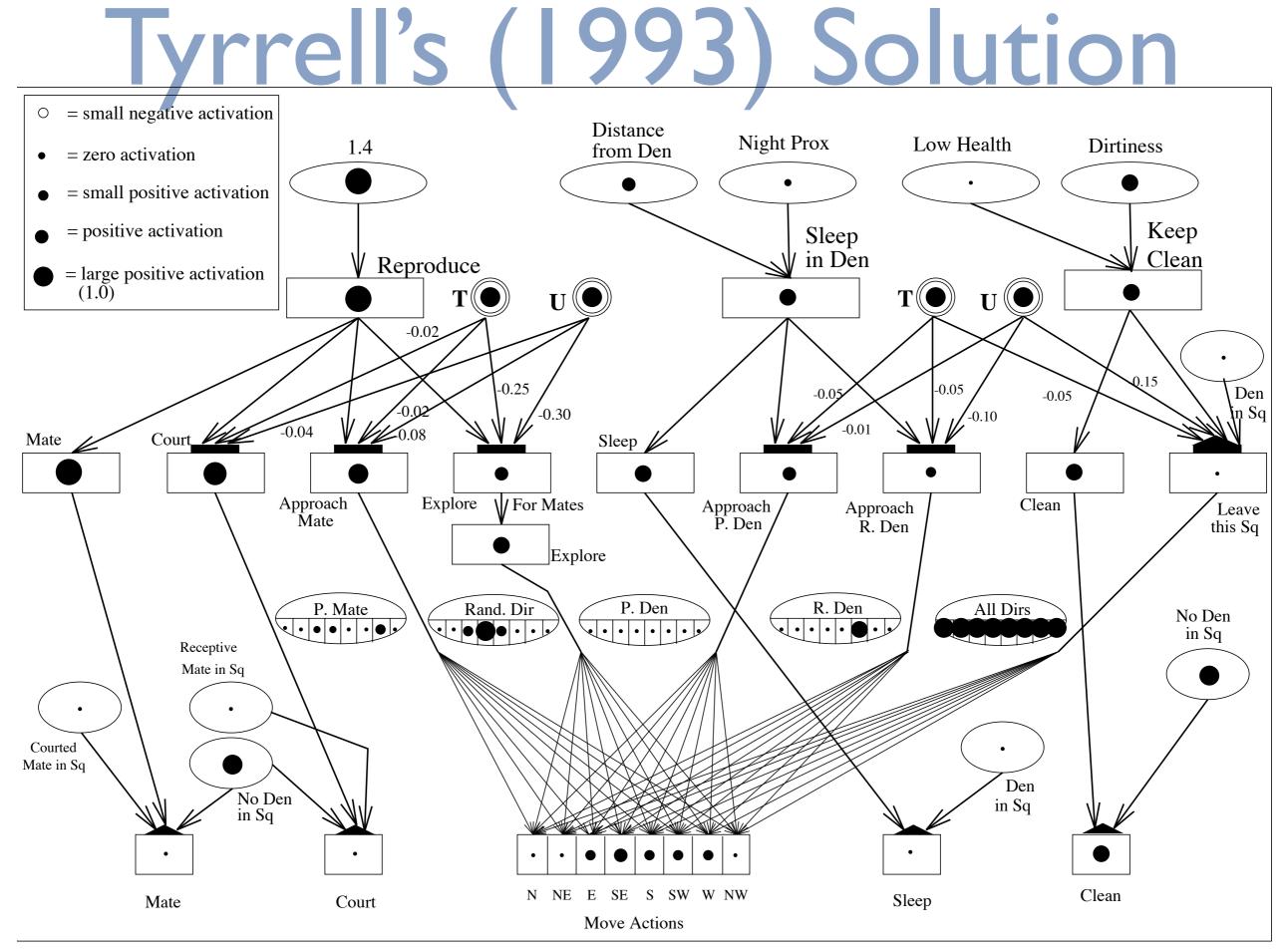
SA: Layers vs. Behaviours



Maes' Spreading Activation Networks

- "Maes Nets" (Adaptive Neural Arch.; Maes 1989)
- "Hierarchies lead to bottlenecks"
- ANA fail to scale (Tyrrell 1994).

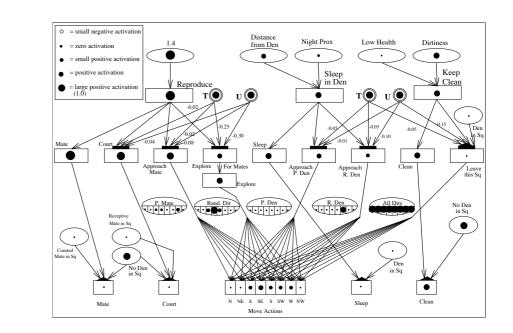




Extended Rosenblatt and Payton Free-Flow Hierarchy

Tyrell's (1993) PhD

Simulated Enviro	nment								
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- Compared Hull, Lorenz, Maes & Freeflow Hierarchy.
- (Hull and Lorenz hard to beat!)

Tests performed in Tyrell's "Simulated Environment"

Stricter hierarchy, more probably correct

O = small negative activation

				= zero activation = small positive activation
				$ \begin{array}{c c} \bullet = \text{positive activation} \\ \hline \bullet = \text{large positive activation} \\ \hline & \bullet \\ (1.6) \end{array} \\ \hline & \bullet \\ \hline \hline & \bullet \\ \hline \hline \hline & \bullet \\ \hline \hline \hline & \bullet \\ \hline \hline \hline \hline & \bullet \\ \hline \hline \hline \hline \hline \hline & \bullet \\ \hline \hline$
		freeze (see_predator t) (covered t) (hawk t)	hold_still	Mate Court 42 0.04 0.05 0.05 0.05 0.05 0.05 0.05 0.05
	flee (C) (sniff_predator t)	run_away (see_predator t)	pick_safe_dir go_fast	P. Mate Receptive Receptive
		look	observe_predator	Counted Marc in Sq Marc in Sq Mar
	mate (C) (sniff_mate t)	inseminate (courted_mate_here t)	copulate	in Sq in Sq Mate Court N NE E SE S SW W NW Move Actions Steep Clean
		court (mate_here t)	strut	
life (D)		pursue	pick_dir_mate go	Edmund vs. ERP in 4 sparse and 4 non-sparse worlds
life (D)	triangulate (getting_lost t)	pick_dir_home go		
	home 1::5 (late t) (at_home \perp)	pick_dir_home go		12- ф
	check 1::5	look_around		10- K
	exploit (C) (day_time t)	use_resource (needed_res_avail t)	exploit_resource	A A A A A A A A A A A A A A A A A A A
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(Bryson SAB 2000)

Parallel-rooted, Ordered Slip-stack Hierarchical (POSH) Action Selection

- Part of Behavior Oriented Design (Bryson 1997, 2000-2007).
- Use hierarchy to describe priorities for behaviour-arbitration in modular, proactive, real-time intelligent systems with multiple conflicting goals.

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What caused cognitivism?

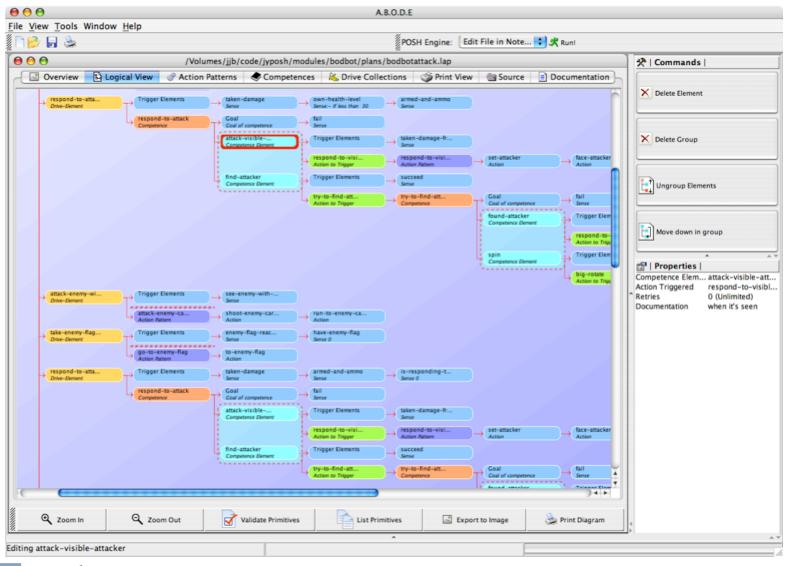
- Latent Learning (Tolman 1930)
- Specialized Learning (Gallistel *et al.* 1991 for review)
 - Pigeons: learn pecking for food, flapping for shock, not vv.
 - Rats: learn smell cues for poison, visual or sound for shock, not vv.



- Modules for perception, action, memory, learning.
 - Each can "run in own thread" / continuous parallel.
- POSH dynamic plans express sequences, competences (subplans), drives (focus of attention.)
 - Action selection necessary where resource contention (Blumberg 1996).

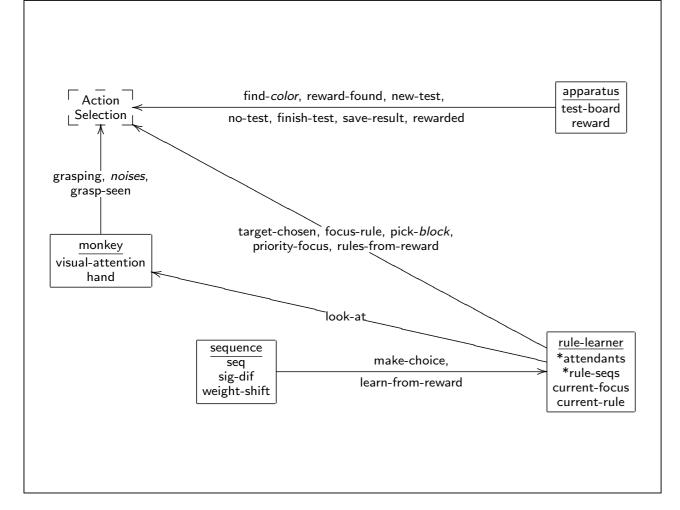
POSH plan in **ABODE** (for UT: Capture the Flag)

(Partington & Bryson 2005)

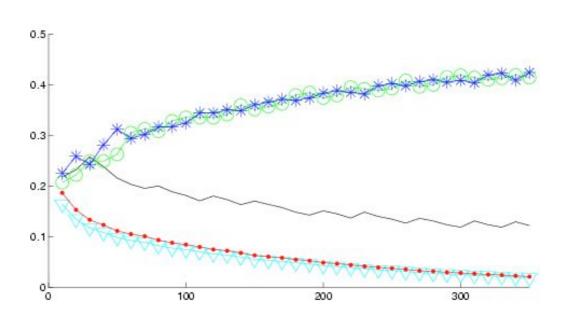


- Advanced BOD Environment.
- Works for realtime or cycle-based systems.
- Games, robots, intelligent environments, alife.

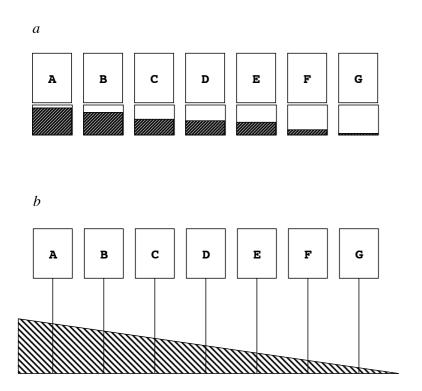
Animals Sacrifice Correctness for Tractability (TI)

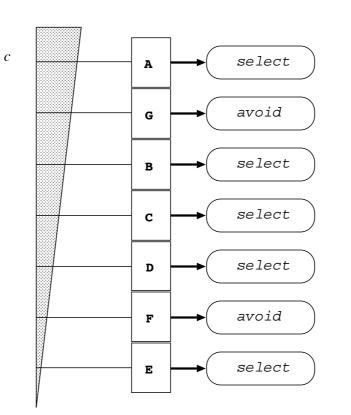






(Bryson & Leong 2007)



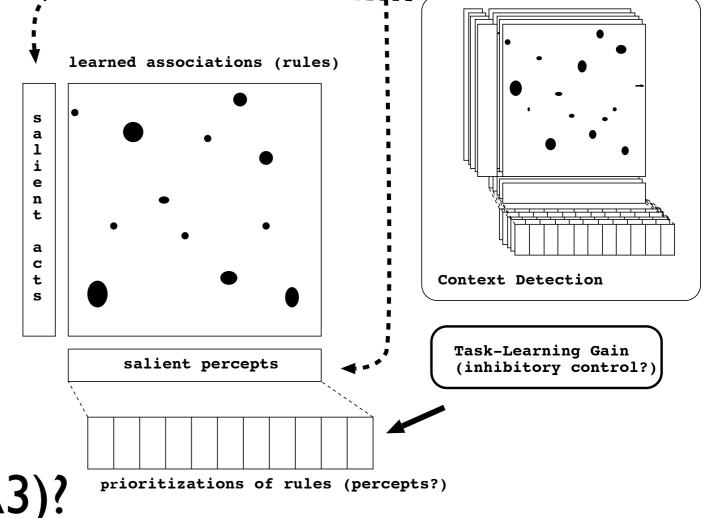


a. Value Transfer

b. Sequence

c. Piroritised productions

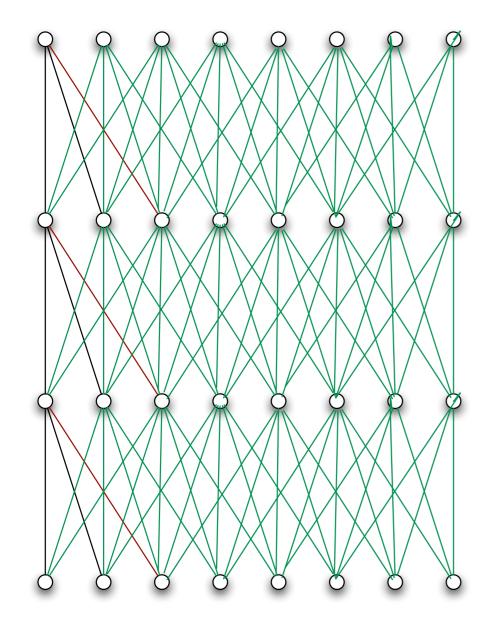
- Action and perception categories learned cortically.
- 2. Associations in parahippocampal gyrus (DG)?
- 3. Priortisations in HP (CA3)?



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Much of cortex is a persistent two-way hierarchy



- Each node informs and is informed by a number of nodes in the next level.
- Two inverted columns of overlapping hierarchies.
- Also has lateral excitation and inhibition.
- Culture is like this?



- Ethology and AI both have "egalitarian" tendencies, but for AS solid evidence that...
 - Ethology: there is more state than just sense/act associations.
 - In AI, that hierarchies can be flexible, efficient, and easier to program.
- Elsewhere, persistent hierarchies are used for recording and developing / evolving information theories.

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POSH Action Selection

- On every cycle, roots / drives are checked to see whether attention should shift.
- For selected drive, execute the next action in sub-plan already attended to (highestpriority production for competence, next item for sequences, primitive...)
- AS must trade off dithering for flexibility.
- Not a strict hierarchy.

Intelligence

- What matters is expressing the right behavior at the right time: action selection.
- Conventional AI planning searches for an action sequence, requires set of primitives.
- Learning searches for the right parameter values, requires primitives and parameters.
 - *parameter*: variable state.
 - Evolution and development are learning.

Combinatorics

• If ...

an agent knows 100 actions (e.g. eat, drink, sleep, step, turn, lift, grasp, poke, flip...), and
it has a goal (e.g. go to Madagascar)

• Then ...

- Finding a one-step plan may take 100 acts.
- -A two-step plan may take 100² (10,000).
- For unknown number of steps, may search forever, missing critical steps or sequence.

Intelligence & Design

- Combinatorics is the problem, search is the only solution.
- The task of intelligence is to focus search.
 - Called bias (learning) or constraint (planning).
 - Most behavior has no or little real-time search.
- For artificial intelligence, most focus comes from design (including physical affordances).

Modularity is not Enough



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Get Fuzzy (Conley 2006)

- I. Initial decomposition \Rightarrow specification.
- 2. Scale the system.
 - i. Code one behavior and/or plan.
 - ii. Test and debug code (test earlier plans).
 - iii. Simplify the design.
- 3. Revise the specification.

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- I. Specify (high-level) what the agent will do.
- Describe activities as sequences of actions.
 competences and action patterns
- 3. Identify sensory and action primitives from these sequences.
- 4. Identify the state necessary to enable the primitives, cluster primitives by shared state. behavior modules
- Identify and prioritize goals / drives. drive collection
- 6. Select a first (next) behavior to implement.

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Simplify the Design

Use the simplest representations.

• Plans:

- primitives, action patterns, competences.
- drives only if need to always check.
- Behavior modules / memory:
 - none, deictic, specialized, general.

(Bryson, AgeS 2003)

Simplify the Design

Trade off representations: plans vs. behaviors

- Use simplest plan structure unless redundancy (split primitives for sequence, add variable state in modules).
- If competences too complicated, introduce primitives or create more hierarchy.
- Split large behaviors, use plans to unify.
- All variable state in modules (deictic).

(Bryson, AgeS 2003)

