Levels of analysis and ACT-R

Cognitive Modelling, Jan. 15, 2010

Sharon Goldwater School of Informatics University of Edinburgh sgwater@inf.ed.ac.uk

Reading: Marr (1982), pp. 19-29; Anderson (1996)

Outline

Two foundational people in cognitive modeling and their ideas:

- David Marr's levels of analysis: a way of thinking about cognitive models.
- John R. Anderson's ACT-R: a unified theory of mind.

David Marr

- Born in Essex, 1945, died of leukemia, age 35.
- Worked in AI Lab and Psychology department at MIT.
- A founder of the field of cognitive neuroscience.
 - First paper (1969) proposed a theory of cerebellar function.
 - "Thirty years later, a significant proportion of researchers working on the cerebellum seem to consider this model as 'generally correct'" (Edelman, 2009).
- Developed influential computational theory of vision.
 - Collected in book, Vision: A Computational Investigation into the Human Representation and Processing of Visual Information, published posthumously in 1982.



Marr's Vision

- Proposes a computational theory of vision:
 - Primal sketch, 2.5D sketch, 3D model.
- Describes three levels of analysis for studying information processing systems (dev. w/ T. Poggio):
 - **Computational theory**: what is the goal of the computation and the logical strategy needed to carry it out?
 - Representation and algorithm: how can the computation be implemented, and what input/output representations are needed?
 - Hardware implementation: what is the physical realization of the algorithm?

Levels of analysis: example



- A cash register:
 - Computational level: what does it do, and why?
 - Computes sum of inputs using theory of addition.
 - Total money owed for goods follows rules of addition.
 - Algorithmic level: what is the representation and algorithm?
 - Arabic numerals.
 - Add least significant digits first, carry remainder, etc.
 - (Unary numerals, counting?)
 - Implementation level: what is the physical realization?
 - Mechanical device, silicon chip, etc.
 - (Matchsticks?)

Cognitive example



- A fly's visual system:
 - Computational level: what does it do, and why?



- If objects in visual field "explode", triggers landing behavior.
- If a small patch appears (e.g., another fly), ψ and ψ ' are delivered to motor system, which is designed to intercept.

Cognitive example



- A fly's visual system:
 - Algorithmic level: what is the representation and algorithm?
 - Not discussed.
 - Implementation level: what is the physical realization?
 - Visual receptor cells, neurons, etc.

To consider

- Which level of analysis do other cognitive models address?
 - Chomsky's generative grammar
 - Associative learning models (e.g., Rescorla-Wagner)
 - Braitenburg's Vehicles
 - Good old-fashioned AI (e.g., Shakey the robot)
- Can all information processing systems be analyzed in this way?
- Is it always easy to separate the three levels?

John R. Anderson

- Born 1947, Vancouver.
- Professor of psychology at CMU since 1978.
- Early pioneer of work on intelligent tutoring.
- Influential work on cognitive architectures (ACT, ACT*)
- Introduced framework for rational analysis (Anderson, 1990)
- Now works with ACT-R, a hybrid approach.



ACT-R

- Provides "important new insights into the integration of cognition" (Anderson, 1996).
 - Unified theory of cognition realized as a production system.
 - Designed to predict human behavior by processing information and generating intelligent behavior.
 - Integrates theories of cognition, visual attention, and motor control.
 - Successfully models a variety of high-level phenomena, e.g., working memory, scientific reasoning, skill acquisition, HCI.

A unified theory of mind

- ACT-R intended to be a single model to capture all aspects of cognitive processing.
 - Good for tackling applied problems.
 - Many other models address only isolated research questions.
- Ex: learning mathematics involves
 - Understanding mathematical expressions
 - Reading (visual and language processing)
 - Problem solving
 - Reasoning and skill acquisition

Example application (Salvucci and Macuga, 2001)

What is the effect of mobile phone use on driving?

- Develop two separate ACT-R models for driving and dialing mobile phone.
- Put them together to predict effects of driving on phone use and vice versa.
 - Compared four ways of dialing.
 - Predicted that only full manual dialing has significant impact on steering abilities.
 - Predictions borne out through later experiments.

Other domains

- **Perception and attention**: visual search, eye movements, task switching, driving behavior, situational awareness.
- Learning and memory: list memory, implicit learning, skill acquisition, category learning, arithmetic, learning by exploration and example.
- Problem-solving and decision-making: use and design of artifacts, spatial reasoning, game playing, insight and scientific discovery.
- Language processing: parsing, analogy and metaphor, learning, sentence memory, communication and negotiation.
- **Other**: cognitive development, emotion, individual differences.

Basic theory

Cognition emerges from the interaction between very many small bits of procedural and declarative knowledge.

- Declarative knowledge: facts.
- Procedural knowledge: encodes processes and skills, represented as production rules.

Declarative knowledge

- Units of declarative knowledge are called chunks.
- Chunks encode things remembered or perceived, as well as current goals:
 - 2+2 = 4
 - Edinburgh is the capital of Scotland.
 - There is a car to my right.
 - I'm trying to get to class.

Procedural knowledge

 Encoded as production rules, consisting of conditions and actions.

IF goal is to add two digits d_1 and d_2 in a column and $d_1+d_2=d_3$ THEN set a subgoal to write d_3 in the column.

- Conditions: may depend on declarative knowledge, buffer contents, and/or sensory input.
- Actions: can change declarative knowledge, goals, or buffer contents, or initiate motor actions.

Modular organization



Modular organization

- Modules: store and process long-term information, which is deposited in buffers.
 - Goal buffer: tracks state in solving problems.
 - Retrieval buffer: holds information retrieved from long-term declarative memory.
 - Visual buffer: tracks visual objects and their identities.
 - Manual buffer: control and sensation of hands.
- Central production system: executive control and coordination of modules.
 - Not sensitive to activity in modules, only to buffer contents.

Timing and coordination

- Within modules, processing is in parallel.
 - Ex: visual system processes entire visual field at once.
- Overall timing determined by serial processing in central production system. In one critical cycle:
 - Patterns in buffers are recognized and a production fires.
 - Buffers are updated for the next cycle.
- Assumptions:
 - Each buffer may contain only one chunk.
 - Only a single production fires each cycle.
 - Cycle takes about 50 ms (based on experimental data).

Hybrid architecture

- Behavior determined by interaction between symbolic and sub-symbolic (statistical) systems.
 - Symbolic: production system.
 - Sub-symbolic: massively parallel processes summarized by mathematical equations.
- Each symbol (production/chunk) has sub-symbolic parameters that reflect past use and determine probability of current use.

Ex: declarative memory module

- Purpose: retrieve chunks formed previously.
- Each chunk has a sub-symbolic activation level, the sum of
 - Base level activation, reflecting general usefulness in past.
 - Associative activation, reflecting relevance to current context.
- Total activation determines probability of being retrieved and speed of retrieval.

Ex: procedural memory

- Many production rules may match at once, but only one can fire.
- Each rule has a sub-symbolic utility function combining
 - The probability that the current goal will be achieved if this rule is chosen (based on past experience).
 - The relative cost (time/effort) and benefit of achieving the current goal.
- The rule with the highest utility is executed.

ACT-R summary

- Complex cognition emerges as the result of (procedural) production rules operating on (declarative) chunks.
- Independent modules encapsulate parallel processing functions, place single chunks in buffers.
- Central production system accesses buffers, detects when rule triggers are satisfied, fires one rule at a time.
- Chunks and rules are symbolic, but sub-symbolic activation levels determine which ones get used.
- Learning involves either acquiring new chunks and productions, or tuning their sub-symbolic parameters.

ACT-R features

- Can predict time-sharing between different tasks.
- Bridges short time-scale processes (retrieval, single productions) with long time-scale processes (e.g., learning to solve algebraic equations), with implications for education.
- Some evidence that modular structure corresponds to different brain regions.

Questionnaire

- Name, degree, and course/specialism.
- For MSc: undergraduate course.
- Background with machine learning, programming, probability theory, cognitive psychology.
- Are there any particular areas of interest to you, things you're hoping to learn about, or reasons for taking this course?

References

- Anderson, John R. 1990. *The Adaptive Character of Thought.* Lawrence Erlbaum Associates, Hillsdale, NJ.
- Anderson, John R. 1996. ACT: a simple theory of complex cognition. *American Psychologist* 51(4):355–365.
- Anderson, John R. 2002. Spanning seven orders of magnitude: A challenge for cognitive modelling. *Cognitive Science* 26:85–112.
- Anderson, John R., D. Bothell, M. Byrne, S. Douglass, C. Lebiere, and Y. Qin. 2004. An integrated theory of the mind. *Psychological Review* 111(4):1036–1060.
- Edelman, Shimon. 2009. David Marr. To appear in the International Encyclopedia of Social and Behavioral Sciences. Available at http://kybele.psych.cornell.edu/~edelman/marr/marr.html.
- Marr, David. 1982. *Vision: A Computational Approach*. Freeman & Co., San Francisco.
- Salvucci, D. D. and K. L. Macuga. 2001. Predicting the effects of cellphone daialing on driver performance. In *Proceedings of the 4th International Conference on Cognitive Modeling*, pp. 25–30.