

mc375: intro to robotics behavior-based systems.

- control
- behavior-based systems
- expressing behaviors
- behavioral encoding
- representations
- example: Toto
- behavior coordination
- emergent behavior

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control: models.

we like to make a distinction between

- classic “model-based” AI
 - symbolic representations
- neo “behavior-based” AI
 - numeric representations

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control: classic models.

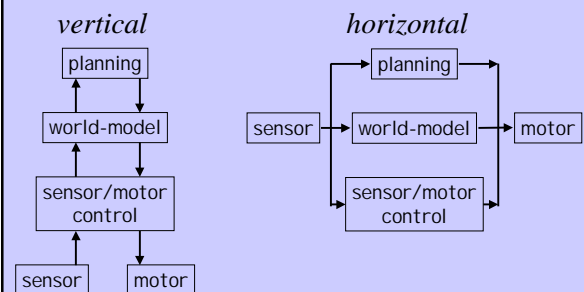
- deliberative... SPA (sense,plan,act)
 - functional decomposition
 - systems consist of sequential modules achieving independent functions
 - sense world
 - generate plan
 - translate plan into actions
- reactive architectures
 - task-oriented decomposition
 - systems consist of concurrently executed modules achieving specific tasks
 - avoid obstacle
 - follow wall

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control: two orthogonal flows.



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control: distinctions.

- key issues that distinguish architectures
 - time scale
 - looking ahead
 - modularity
 - the way in which the architecture decomposes into components

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control: behavior-based systems.

- BBS
- behaviors are the underlying module of the system
- behavioral decomposition
 - systems consist of sequential modules achieving independent functions

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control: robotic behavior.

- generate a motor response from a given perceptual stimulus
- basis in biological studies
 - serves as inspiration for design

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control: behavior vs action.

- | | |
|---|--|
| • behavior | • action |
| – based on dynamic process | – discrete in time |
| • operating in parallel | • well-defined start and end points |
| • lack of a central control | • allows pre- and post-conditions |
| • fast couplings between sensors and motors | – avoidance of side-effects |
| – exploiting emergence | • only one action or few actions at a time |
| • side-effects from combined processes | • conflicts are undesired and avoided |
| • using properties of the environment | – deliberative |
| – reactive | |

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behavior-based systems: ...are reactive systems.

- behaviors serve as building blocks for actions
- abstract representation avoided
- often modeled after animal behaviors
- inherently modular

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behavior-based systems: stimuli.

- presence of stimulus is necessary but not sufficient in behavior-based robot
- stimulus must reach threshold value before response is generated

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behavior-based systems: representation.

- behaviors can be represented/stored in a network, with relationships between them
- strength multiplier, or gain, can turn off behaviors or increase response

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behavior-based systems: properties.

- feedback controllers
- achieve specific tasks/goals
 - avoid others, find friend, go home
- typically execute concurrently
- can store state and be used to construct world models/representations
- can directly connect sensors to effectors
- can take inputs from other behaviors and send outputs to other behaviors (⇒ networks)

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behavior-based systems: properties, 2.

- typically higher-level than actions (go home, not turn left 45°)
- typically closed loop, but extended in time
- when assembled into distributed representations, behaviors can be used to look ahead but at a time-scale comparable with the rest of the system

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behavior-based systems: key properties.

- ability to act in real time
- ability to use representations to generate efficient (not only reactive) behavior
- ability to use a uniform structure and representation throughout the system (so no intermediate layer)

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behavior-based systems: key challenge.

- how can representation be effectively *distributed* over the behavior structure?
 - time scale must be similar to that of real-time components of the system
 - representation must use same underlying behavior structure for all components of the system

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behavior-based systems: challenges, 2.

- some components may be reactive
- not every component is involved with representational computation
- some systems use a simple representation
- as long as the basis is in behaviors and not rules, the system is a BBS

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behavior-based systems: what are behaviors?

- *behavior*: anything observable that the system/robot does
- how do we distinguish internal behaviors (components of a BBS) and externally observable behaviors?
- should we distinguish?

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behavior-based systems: external vs internal.

- reactive robots display desired external behaviors
 - avoiding
 - collecting cans
 - walking
- but controller consists of a collection of rules, possibly in layers
- BBS actually consist and are programmed in the behaviors, which are higher granularity, extended in time, capable of representation

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expressing behaviors.

- behaviors can be expressed with various representations
- when a control system is being designed, the task is broken down into desired external behaviors
- those can be expressed with
 - functional notation
 - stimulus response (SR) diagrams
 - finite state machines/automata (FSA)
 - schema

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expressing behaviors: design paradigms.

- ethological guided/constrained
- situated activity
- experimentally driven

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expressing behaviors: functional notation.

- mathematical model:
 - represented as triples (S,R, β)
S = stimulus
R = range of response
 β = behavioral mapping between S and R
- easily convert to functional languages like LISP

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expressing behaviors: functional example.

```
coordinate-behaviors [
  move-to-classroom (detect-classroom-location),
  avoid-objects (detect-objects),
  dodge-students (detect-students),
  stay-to-right-on-path (detect-path),
  defer-to-elders (detect-elders)
] = motor-response
```

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expressing behaviors: FSA diagrams.

- states and transitions are most easily encoded in finite state automata and drawn as finite state diagrams
- states of the diagram can also be called behaviors
- diagrams show sequences of behavior transitions

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expressing behaviors: formal methods.

- used to verify intentions of designer (not the same as correctness of controller)
- enable automatic generation of code
- use a common language
- support formal analysis
- support high-level programming
- example: situated automata

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expressing behaviors: situated automata.

- formalism for specifying FSA's that are *situated* [Kaelbling & Rosenschein, 1991]
- task described in high-level logic expressions, as a set of goals and a set of operators that achieve (ach) and maintain (mai nt) the goals
- once defined, tasks can be compiled into circuits (using special purpose languages), which are reactive

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expressing behaviors: situated automata example.

```
(defgoal r (ach in-classroom)
  (if (not startUp)
    (maint (and (maint move-to-classroom)
                (maint avoid-objects)
                (maint dodge-students)
                (maint stay-to-right-on-path)
                (maint defer-to-elders))))))
```

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expressing behaviors: subsumption architecture.

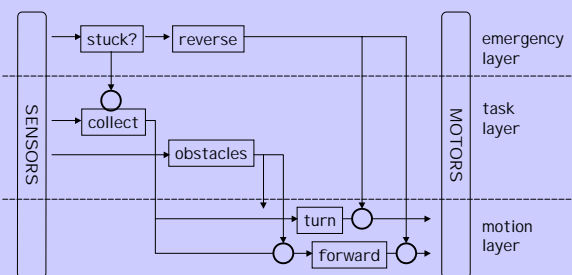
- Rodney Brooks, 1986
- MIT AI lab
- reactive elements
- behavior-based elements
- layered approach based on *levels of competence*

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expressing behaviors: subsumption architecture, 2.

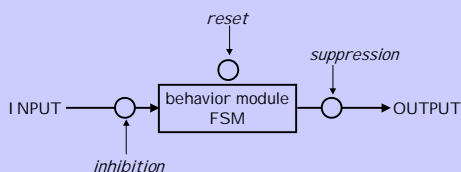


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expressing behaviors: augmented finite state machine (AFSM).



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behavioral encoding.

- behavioral response in physical space has a strength and an orientation
- expressed as (S, R, β)
- S = stimulus, necessary but not sufficient condition to evoke a response (R); internal state can also be used
- β = behavioral mapping categories
 - null
 - discrete
 - continuous

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behavioral encoding: discrete encoding.

- expressed as a finite set of situation-response pairs/mappings
- mappings often include rule-based form IF-THEN
- examples:
 - Gapps [Kaelbling & Rosenschein]
 - subsumption language [Brooks]

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behavioral encoding: continuous encoding.

- instead of discretizing the input and output, a continuous mathematical function describes the input-output mapping
- can be simple, time-varying, harmonic
- examples:
 - potential field
 - schema
- problems with local minima, maxima, oscillatory behavior

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behavioral encoding: motor schemas.

- type of behavior encoding
- based on schema theory
- Ron Arbib, Georgia Tech
- provide large grain modularity
- distributed, concurrent schemas used
- based on neuroscience and cognitive science
- represented as vector fields
- decomposed into "assemblages" by fusion, not competition

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behavioral encoding: assemblages.

- recursively defined aggregations of behaviors or other assemblages
- important abstractions for constructing behavior-based robots

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behavioral encoding: schema representation.

- responses represented in uniform vector format
- combination through cooperative coordination via vector summation
- no predefined schema hierarchy
- arbitration not used -- gain values control behavioral strengths

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behavioral encoding: designing with schemas.

- characterize motor behaviors needed
- decompose to most primitive level, use biological guidelines where appropriate
- develop formulas to express reactions
- conduct simple simulations
- determine perceptual needs to satisfy motor schema inputs
- design specific perpetual algorithms
- integrate/test/evaluate/iterate

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behavioral encoding: strengths and weaknesses.

- strengths
 - support for parallelism
 - run-time flexibility
 - timeliness for development
 - support for modularity
- weaknesses
 - niche targetability
 - hardware retargetability
 - combination pitfalls (local minima, oscillations)

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representation.

- necessary in order to get benefits of BBS
- BBS representations are
 - distributed
 - matched to time-scale of the system
- constructed out of basic components of BBS system: behaviors

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representation: behavior.

- direct feedback loops/control laws: mapping sensors to effectors
- schemas: sensory or motor
- procedures: any combination
 - sensory
 - motor
 - sensory to motor
 - representational

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representation: example task: mapping.

- task: create a robot that can
 - move around safely
 - make a map of its environment
 - use the map to find paths to specific locations
- most common and useful mobile robot task
- many applications!

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representation: representing a map.

- representation needs to be
 - distributed
 - on a short time-scale
 - cannot be a traditional CAD-CAM centralized map

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representation: idea: distribute maps.

- distribute parts of map over different behaviors
- connect parts of the map that are adjacent in the physical world so that they are also adjacent in the map
- result is a network of behaviors representing the map
- map is topological

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example: Toto.

- Maja Mataric, MIT now USC
- behavior-based robot
- first BBS robot to have a distributed representation
- control system consisted of a collection of behaviors
- lowest levels responsible for safe movement of robot (avoiding collisions)
- next levels responsible for keeping robot near walls, boundaries

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Toto: landmarks.

- walls, corridors, messy irregular areas
- robot detected landmarks
- each landmark stored as a behavior
 - landmark type
 - compass heading
 - approximate length/size
- when a new landmark is found, a new behavior is added

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Toto: landmarks, 2.

- adjacent landmarks connected by communication wires
- result is a topological representation of the environment
- also used for path finding

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Toto: active map.

- whenever Toto visited a particular landmark, its associated map behavior would become activated
- if no behavior was activated, then the landmark was new, so a new behavior was created
- if an existing behavior was activated, it inhibited all other behaviors
- localization was based on which behavior was active

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Toto: message passing.

- once Toto had a map, it could find paths from one landmark to another
- the goal behavior/landmark would send messages (send activation) to its neighbors, they would pass it on, etc
- eventually it would reach the current landmark (Toto's current position)

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Toto: continuous map following.

- resulting string of behaviors is a path (or a plan) to the goal
- Toto did not store a string
- messages were passed continuously
- at each behavior in the map, Toto would decide where to go next
- goal reached one behavior at a time

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Toto: path optimization.

- at a junction, how did Toto decide where to go?
- path length computed based on landmark size and number of landmarks from goal to current landmark
- Toto chose shortest path

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behavior coordination.

- BBS consist of collection of behaviors
- execution must be coordinated in a consistent fashion
- coordination can be
 - competitive
 - cooperative
 - combination of the two

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behavior coordination: deciding what to do next.

- action-selection problem
- behavior-arbitration problem

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behavior coordination: behavior arbitration.

- summing up
- average
- winner takes all

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behavior coordination: competitive coordination.

- perform arbitration (selecting one behavior among a set of candidates)
 - priority-based: subsumption
 - state-based: discrete event systems, Bayesian decision theory
 - function-based: spreading of activation action selection

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behavior coordination: cooperative coordination.

- perform command fusion (combine outputs of multiple behaviors)
- voting
- fuzzy (formalized voting)
- superposition (linear combinations)
 - potential fields
 - motor schemas
 - dynamical systems

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emergent behavior.

- important but not well-understood phenomenon
- often found in behavior-based robotics
- robot behaviors “emerge” from
 - interactions of rules
 - interactions of behaviors
 - interactions of either with environment

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emergent behavior: distinction.

- coded behavior
 - in the programming scheme
- observed behavior
 - in the eyes of the observer
 - emergence
- there is no one-to-one mapping between the two!

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emergent behavior: is it magic?

- sum is greater than the parts
- emergent behavior is more than the controller that produces it

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emergent behavior: interaction and emergence.

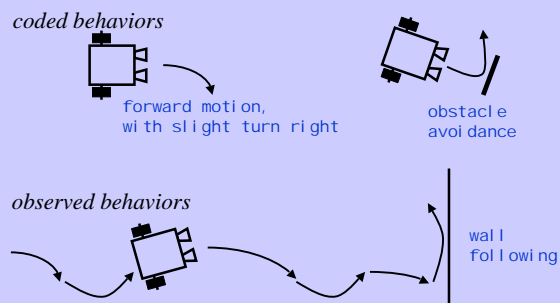
- interactions between rules, behaviors and environment
- source of expressive power for a designer
- i.e., systems can be designed to take advantage of emergent behavior

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emergent behavior: example: wall following.



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emergent behavior: example: wall following, 2.

- can be implemented with these rules:
 - if too far, move closer
 - if too close, move away
 - otherwise, keep on
- over time, in an environment with walls, this will result in wall-following
- is this emergent behavior?

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emergent behavior:
emergent wall following.

- it is argued yes because
 - robot itself is not aware of a wall, it only reacts to distance readings
 - concepts of “wall” and “following” are not stored in the robot’s controller
 - the system is just a collection of rules

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emergent behavior:
emergent flocking.

- program multiple robots:
 - don’t run into any other robot
 - don’t get too far from other robots
 - keep moving if you can
- when run in parallel on many robots, the result is flocking

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emergent behavior:
necessary conditions.

- notion of emergence depends on two aspects:
 - existence of an external observer, to observe and describe the behavior of the system
 - access to the internals of the controller itself, to verify that the behavior is not explicitly specified anywhere in the system

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emergent behavior:
unexpected vs emergent.

- some researchers say the above is not enough for behavior to be emergent, because above is programmed into the system and the “emergence” is a matter of semantics
- so emergence must imply something unexpected, something “surreptitiously discovered” by observing the system

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emergent behavior:
subjective expectations.

- “unexpected” is highly subjective, because it depends on what the observer was expecting
- naive observers are often surprised!
- informed observers are rarely surprised

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emergent behavior:
observation and emergence.

- once a behavior is observed, it is no longer unexpected
- is new behavior then “predictable”?

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emergent behavior: formalization.

- look for behaviors that are not apparent at system level (robot's controller) but are apparent at observer's level

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emergent behavior: execution and emergence.

- so now even simple wall following example given can be called "emergent"
- this means system has to execute in order for behavior to emerge

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emergent behavior: uncertainty and emergence.

- not difficult to achieve -- environment is uncertain, so exact behavior of a system is very hard to predict!
- if behavior contains novel and rich patterns, then it is "emergent"
- if world were completely predictable, then we'd remove "emergent behaviors" by this definition

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emergent behavior: emergence is unavoidable.

- some interesting and unexpected behaviors will always emerge in systems that interact with the physical world
- some may be labeled "emergent"
- some are not desirable!

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emergent behavior: emergent bugs.

- unexpected, emergent behavior that is undesirable
- e.g., oscillations
- try to avoid them, but still want to exploit desirable, unexpected behaviors
- system needs to know how to distinguish between the two

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emergent behavior: sequential vs parallel.

- emergent behaviors can arise from parallel execution of multiple behaviors (e.g., flocking)
- also from sequential interaction with an interesting environment

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emergent behavior:
architectures and emergence.

- different architectures affect the likelihood of generating and using emergent behaviors
- deliberative: always aim to eliminate them
- reactive: aim to exploit them
- hybrid: typically aim to eliminate them
- BBS: aim to exploit them

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emergent behavior:
modularity and emergence.

- modularity directly effects emergence
- reactive and BBS employ parallel rules and behaviors which interact with each other and the environment, thus directly producing and exploiting emergent behavior

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emergent behavior:
avoiding & exploiting.

- hybrid systems follow deliberative model in attempt to produce a coherent, uniform output of the system, minimizing interactions and emergence

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reading.

- *Behavior-Based Robotics*, chapter 3, by Ron Arkin, course pack 2 p203-261
- *Behavior-based Robotics, its scope and its prospects*, by Andreas Birk, course pack 2 p263-268

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