Introduction to Mobile Robotics

Robot Control Paradigms

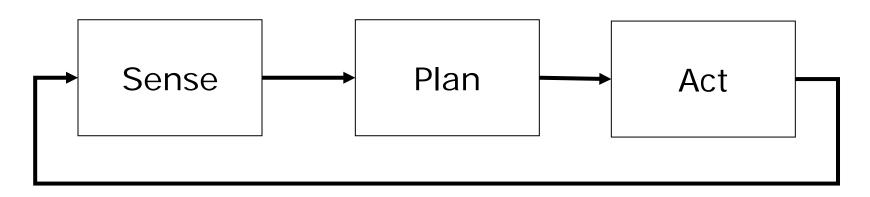
Daniel Büscher



Control Paradigms

- 1. Classical / Hierarchical
- 2. Reactive / Behavior-based
- 3. Potential Field Methods
- 4. Hybrid Deliberative

1. Classical / Hierarchical Paradigm



- 70s
- Focus on automated reasoning and knowledge representation
- STRIPS (Stanford Research Institute Problem Solver): Perfect world model, closed world assumption
- Find boxes and move them to the designated position

Stanford CART 1973



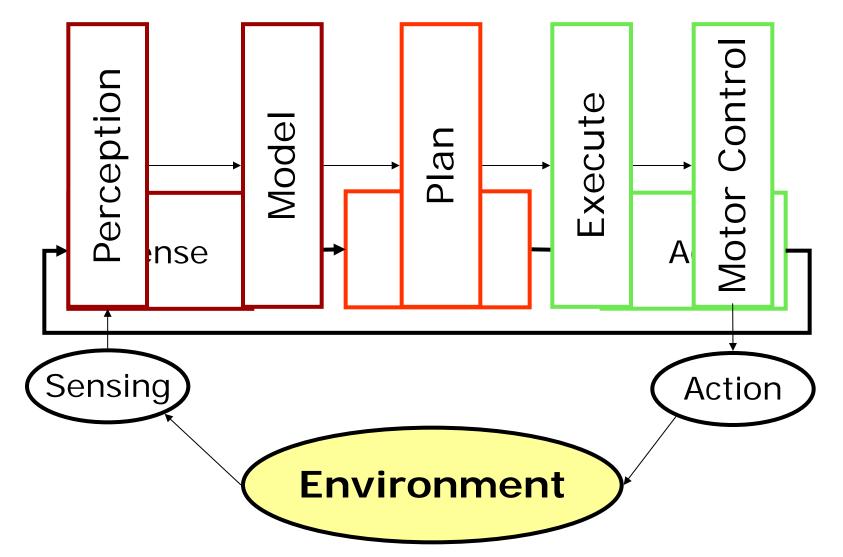
Stanford AI Laboratory / CMU (Moravec)

Classical Paradigm Stanford Cart



- 1. Take nine images of the environment, identify interesting points in one image, and use other images to obtain depth estimates.
- 2. Integrate information into global world model.
- **3**. Correlate images with previous image set to estimate robot motion.
- 4. On basis of desired motion, estimated motion, and current estimate of environment, determine direction in which to move.
- 5. Execute the motion.

Classical Paradigm as Horizontal/Functional Decomposition

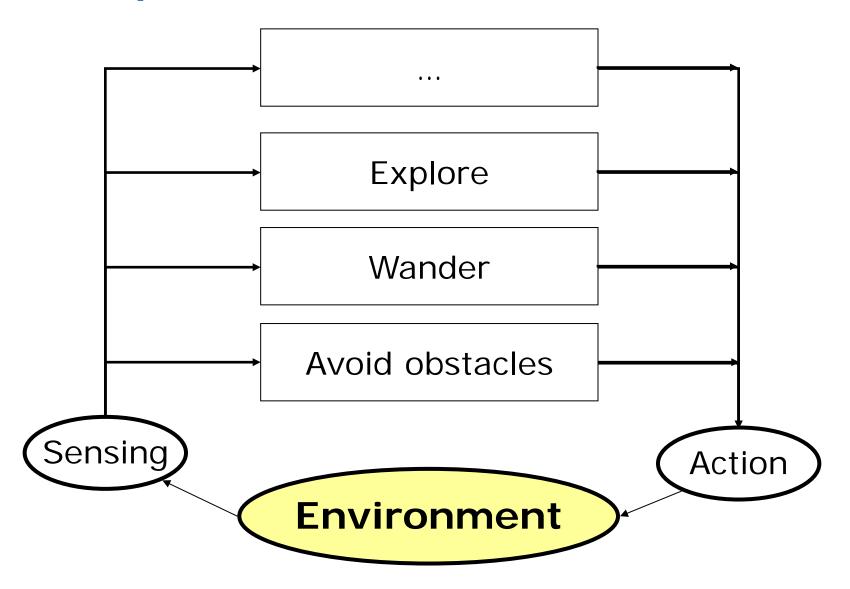


2. Reactive / Behavior-based Paradigm



- 80s
- No models: "The world is its own, best model"
- Early successes, but also limitations
- Investigate biological systems

Reactive Paradigm as Vertical Decomposition



Characteristics of Reactive Paradigm

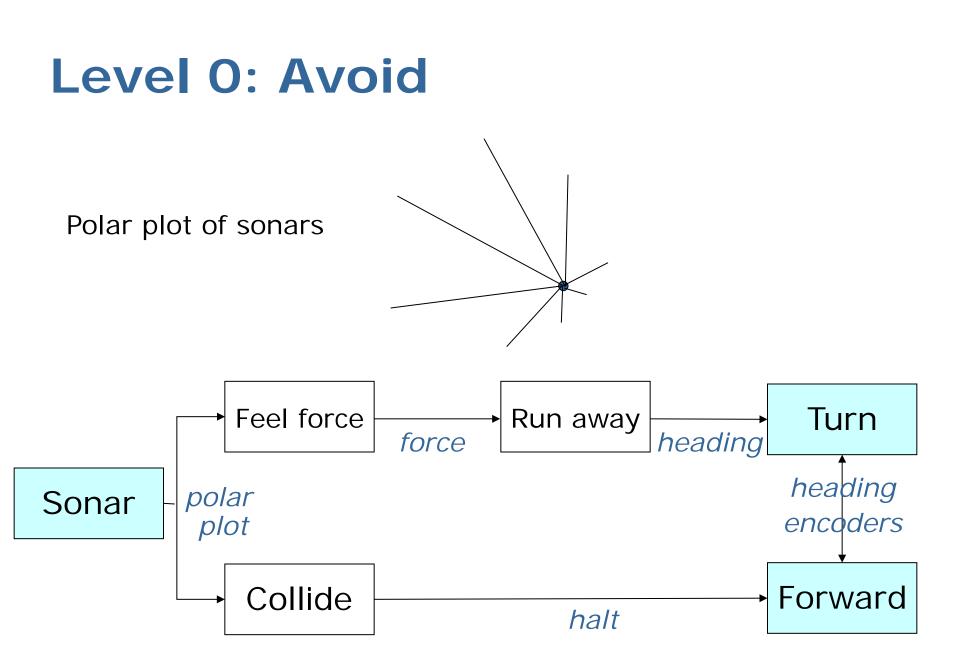
- Situated agent, robot is integral part of its environment.
- No memory, controlled by what is happening in the world.
- Tight coupling between perception and action via behaviors.
- Only local, behavior-specific sensing is permitted (ego-centric representation).

Behaviors

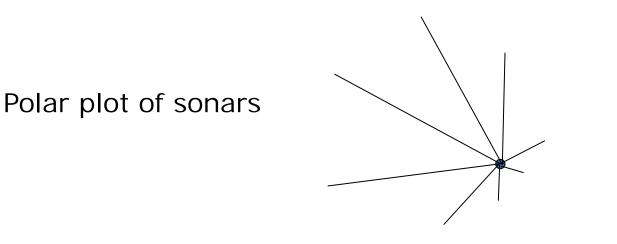
- ... are a direct mapping of sensory inputs to a pattern of motor actions that are then used to achieve a task.
- ... serve as the basic building blocks for robot actions, and the overall behavior of the robot is emergent.
- ... support good software design principles due to modularity.

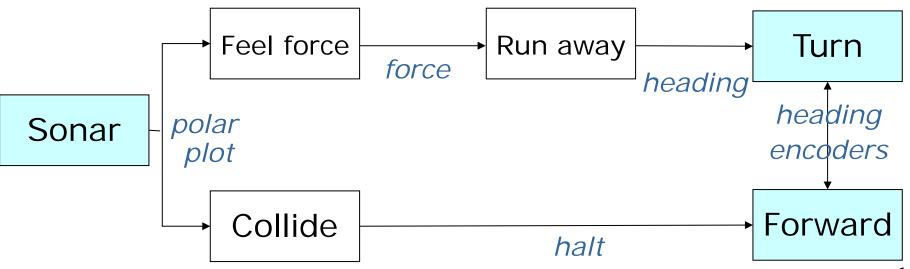
Subsumption Architecture

- Introduced by Rodney Brooks '86.
- Behaviors are networks of sensing and acting modules (augmented finite state machines AFSM).
- Modules are grouped into layers of competence.
- Layers can subsume lower layers.
- No internal state!

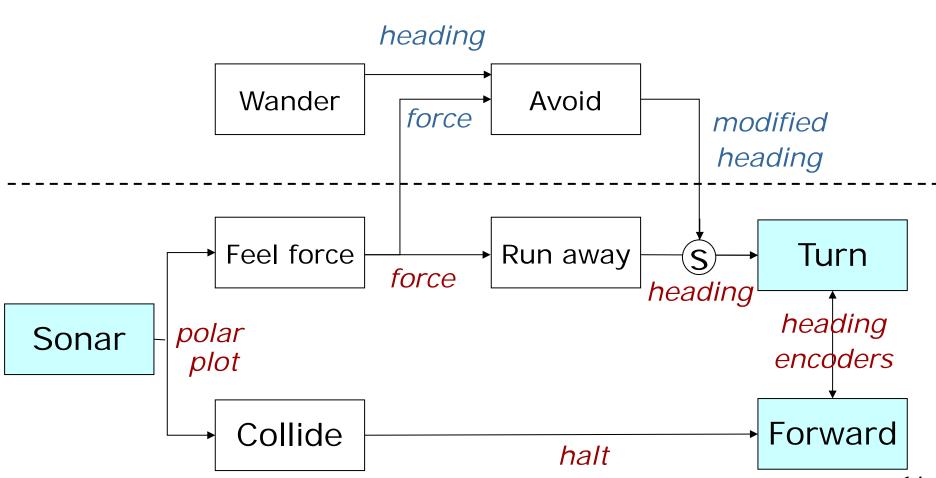


Level O: Avoid

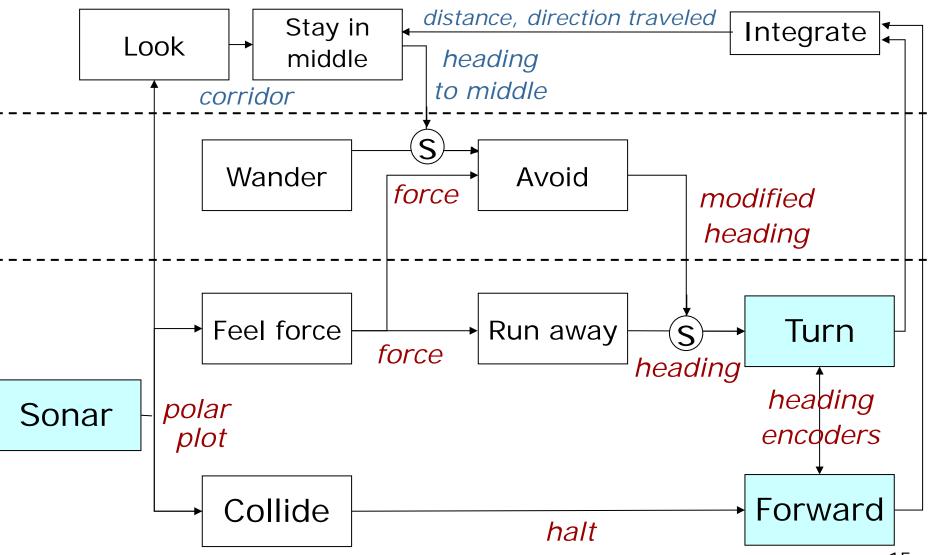




Level 1: Wander



Level 2: Follow Corridor



3. Potential Field Methods

- Treat robot as particle acting under the influence of a potential field
- Robot travels along the derivative of the potential
- Field depends on obstacles, desired travel directions and targets
- Strength of field may change with distance to obstacle/target
- Resulting field (vector) is given by the summation of primitive fields

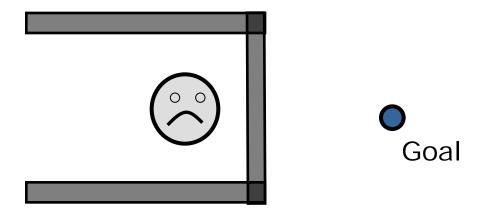
Primitive Potential Fields 1 1 1 1 1 1 1 1 1 1 1 1 1 1 Uniform Perpendicular 0 7 **Attractive** Repulsive **Tangential**

Corridor Following with Potential Fields

- Level O (collision avoidance) is done by the repulsive fields of detected obstacles.
- Level 1 (wander) adds a uniform field.
- Level 2 (corridor following) replaces the wander field by three fields (two perpendicular, one uniform).

Characteristics of Potential Fields

Suffer from local minima

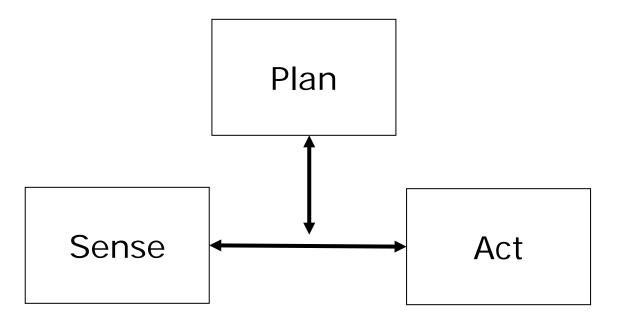


- Backtracking
- Random motion to escape local minimum
- Procedural planner s.a. wall following
- Increase potential of visited regions
- Avoid local minima by harmonic functions

Characteristics of Potential Fields

- No preference among layers
- Pro: Easy to visualize
- Pro: Easy to combine different fields
- Con: High update rates necessary
- Con: Parameter tuning important

4. Hybrid Deliberative Paradigm



Combines advantages of previous paradigms

- Closed loop: reactive fast control
- World model: used for planning

Questions?