

CS545 Lecture 16

Mobile Robotics



- Mobile Robots
 - Control Paradigms
 - Locomotion
 - Behavior-based robotics
- <http://robotics.usc.edu/~aatrash/cs545>
- Slides based on:
 - *Computational Principles of Mobile Robotics* – Gregory Dudek
 - *Probabilistic Robots* – Sebastian Thrun
 - *Behavior-based Robotics* – Ron Arkin
 - Presentation by Dieter Fox

Mobile Robots



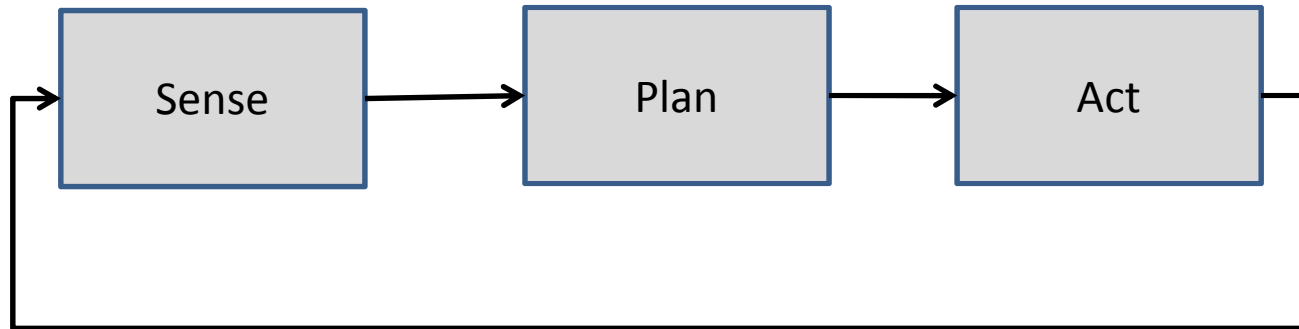
- Robots no longer restricted for factory settings
 - Entertainment, toys
 - Personal services
 - Medical
 - Industrial Automation (mining, harvesting)
 - Hazardous environments (space, underwater)
- Agents need to be mobile

Shakey

- One of first general purpose robots
- Stanford Research Institute
- Use classical planning

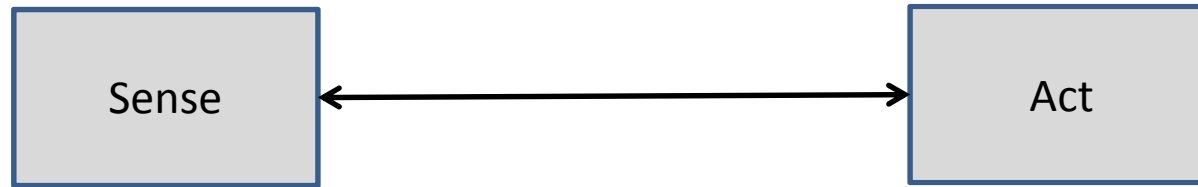


Classical Planning Paradigm



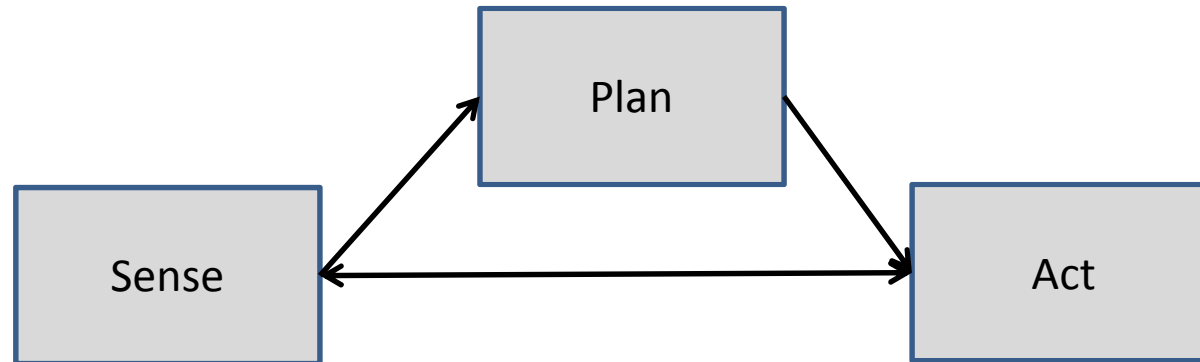
- Direct application of AI tech to robots
- STRIPS (Stanford Research Institute Problem Solver)
- Assumed perfect world. No noise
- SLOW!!!!

Reactive/Behavior-Based



- No models/planning. “World is its own, best model”
- Based on biological systems
- Good for local decisions
- Many limitations

Hybrid



- Combination of other paradigms
 - Use planning components for slower long-term planning
 - Use reactive system for fast, local control

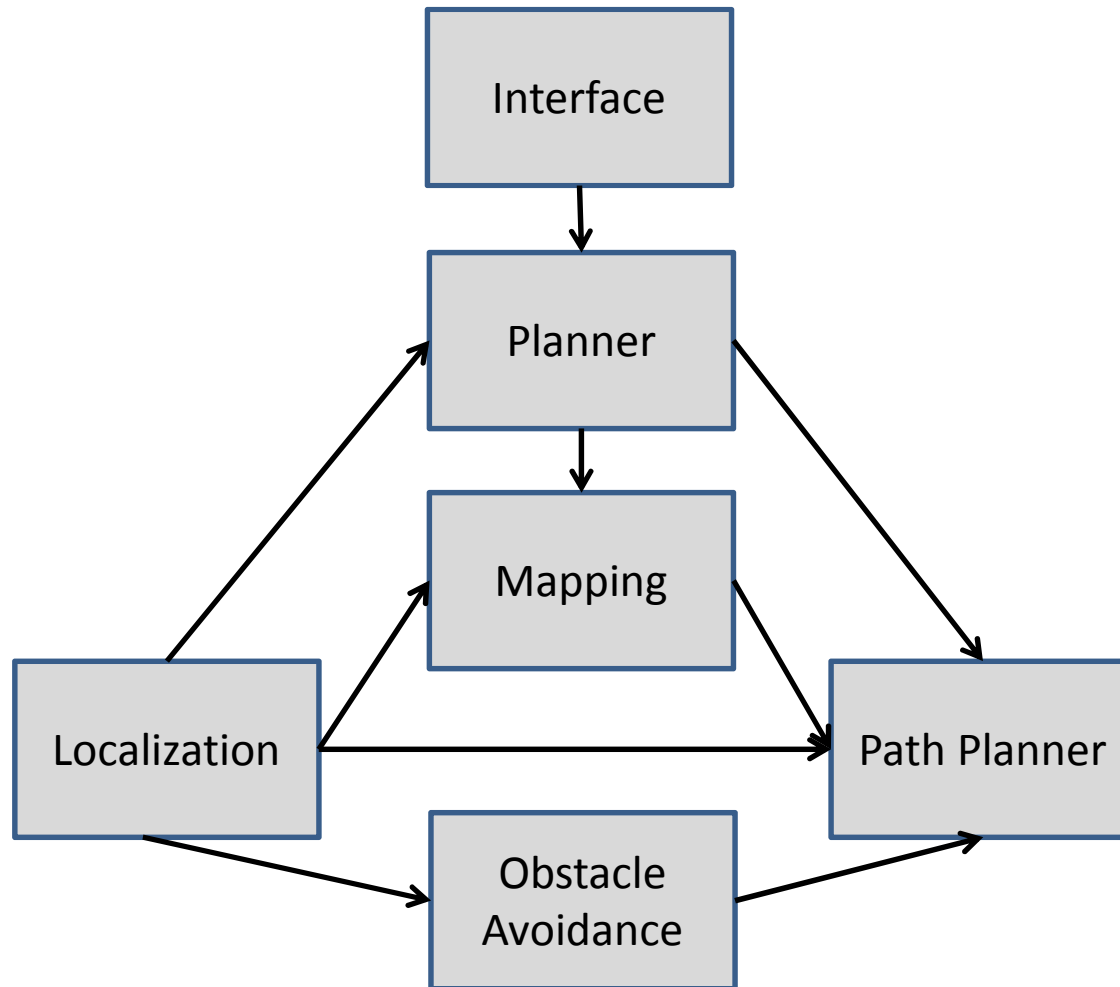
Probabilistic Robots



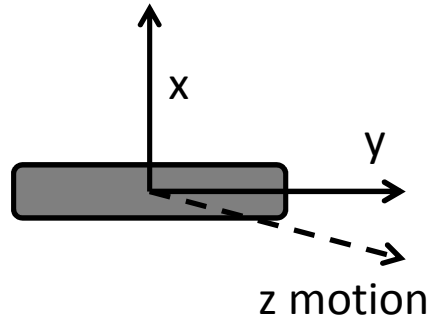
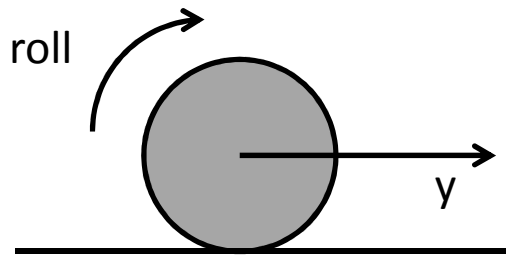
- Assume models and sensors are inaccurate
 - Sensors bad
 - Incomplete information
 - Motors not precise
- Integrate models and sensing
- Bayes Rule:

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

Control Architecture (Example)



Locomotion

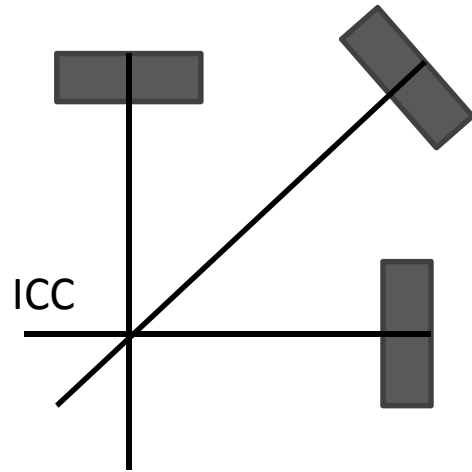


- Common Drives
 - Differential – rotation by speed of wheels
 - Synchronous – can steer wheels
 - Tracked – tanks
 - Car – Ackerman steering
- Holonomic vs. non-Holonomic

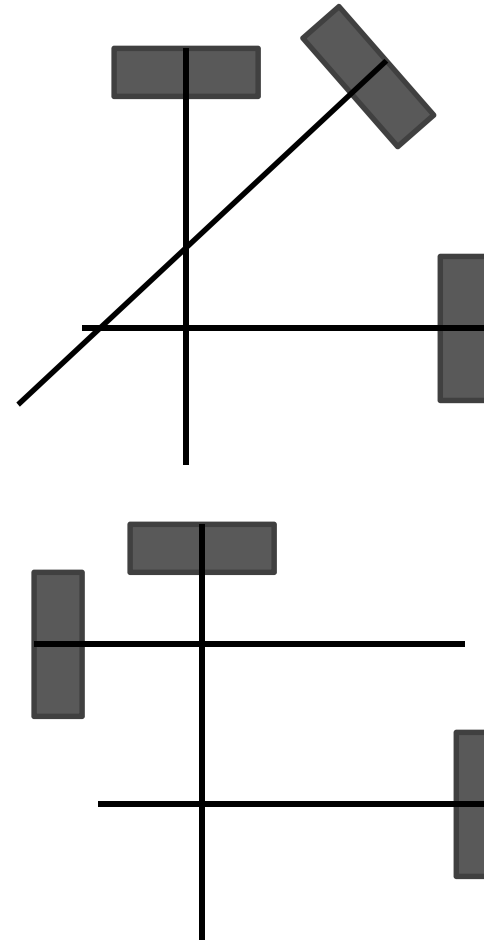
Instantaneous Center of Curvature



- Instantaneous Center of Curvature
 - Intersection of x-axis of wheels



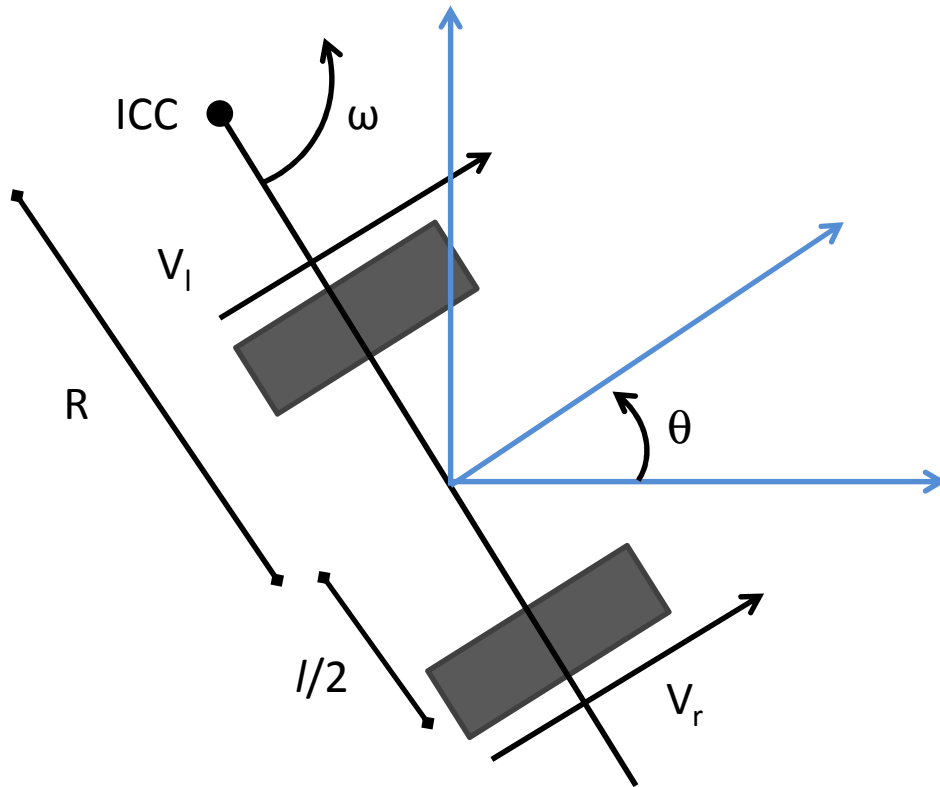
Good!



Bad!!!

Bad!!!

Differential Drives



$$\text{ICC} = [x - R \sin \theta, y - R \cos \theta]$$

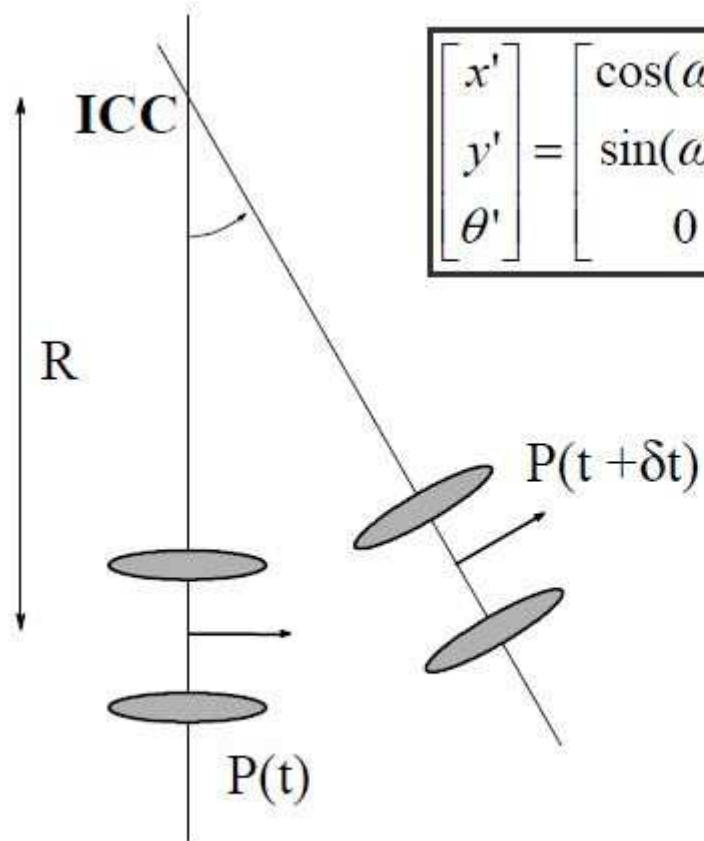
$$\omega(R + l/2) = v_r$$

$$\omega(R - l/2) = v_l$$

$$R = \frac{l (v_l + v_r)}{2 (v_r - v_l)}$$

$$\omega = \frac{v_r - v_l}{l}$$

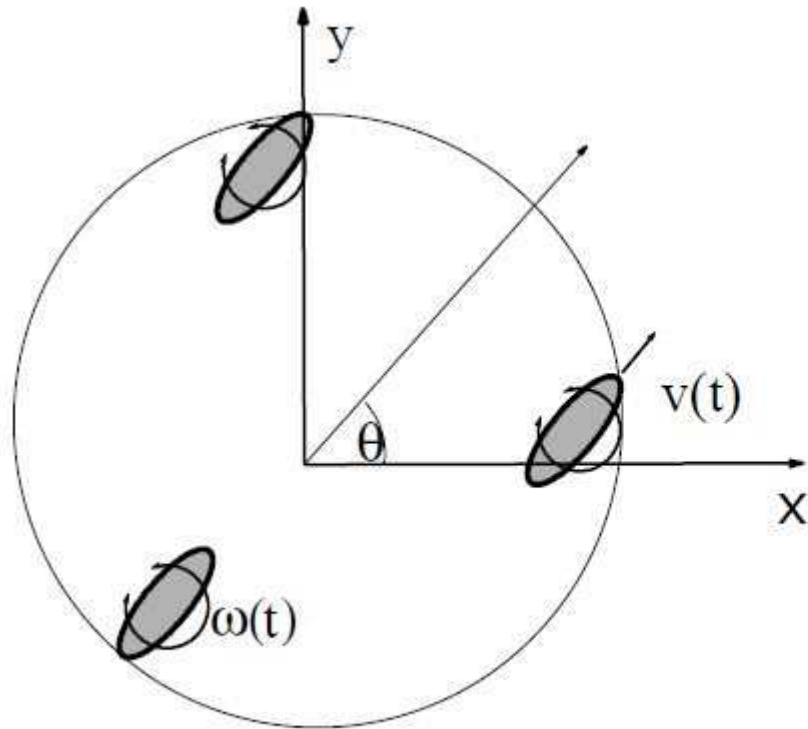
Differential Drive



$$\begin{bmatrix} x' \\ y' \\ \theta' \end{bmatrix} = \begin{bmatrix} \cos(\omega\delta t) & -\sin(\omega\delta t) & 0 \\ \sin(\omega\delta t) & \cos(\omega\delta t) & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x - ICC_x \\ y - ICC_y \\ \theta \end{bmatrix} + \begin{bmatrix} ICC_x \\ ICC_y \\ \omega\delta t \end{bmatrix}$$

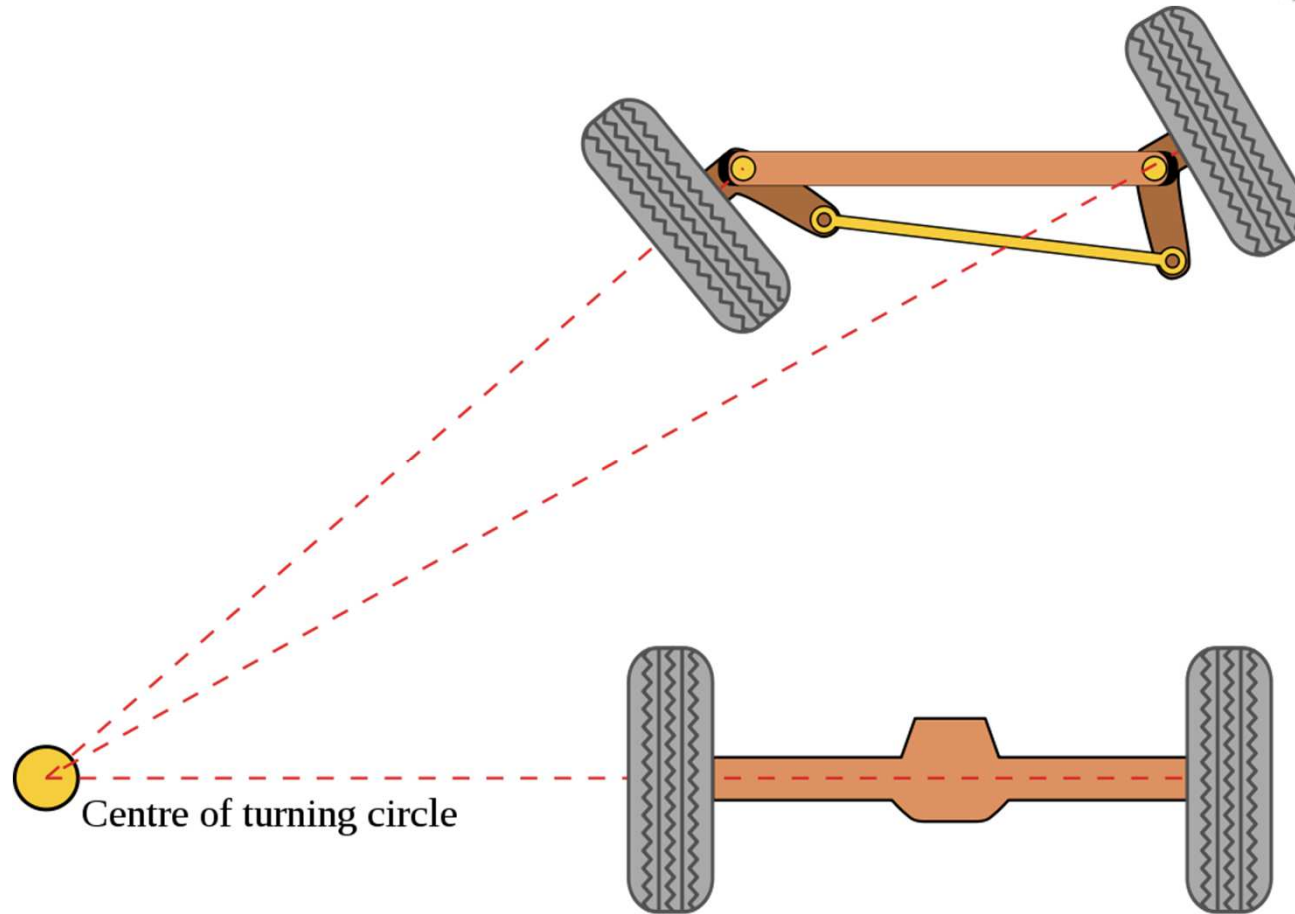
$$\begin{aligned} x(t) &= \frac{1}{2} \int_0^t [v_r(t') + v_l(t')] \cos[\theta(t')] dt' \\ y(t) &= \frac{1}{2} \int_0^t [v_r(t') + v_l(t')] \sin[\theta(t')] dt' \\ \theta(t) &= \frac{1}{l} \int_0^t [v_r(t') - v_l(t')] dt' \end{aligned}$$

Synchronous Drive



$$x(t) = \int_0^t v(t') \cos[\theta(t')] dt'$$
$$y(t) = \int_0^t v(t') \sin[\theta(t')] dt'$$
$$\theta(t) = \int_0^t \omega(t') dt'$$

Ackerman Steering (Kingpin Steering)



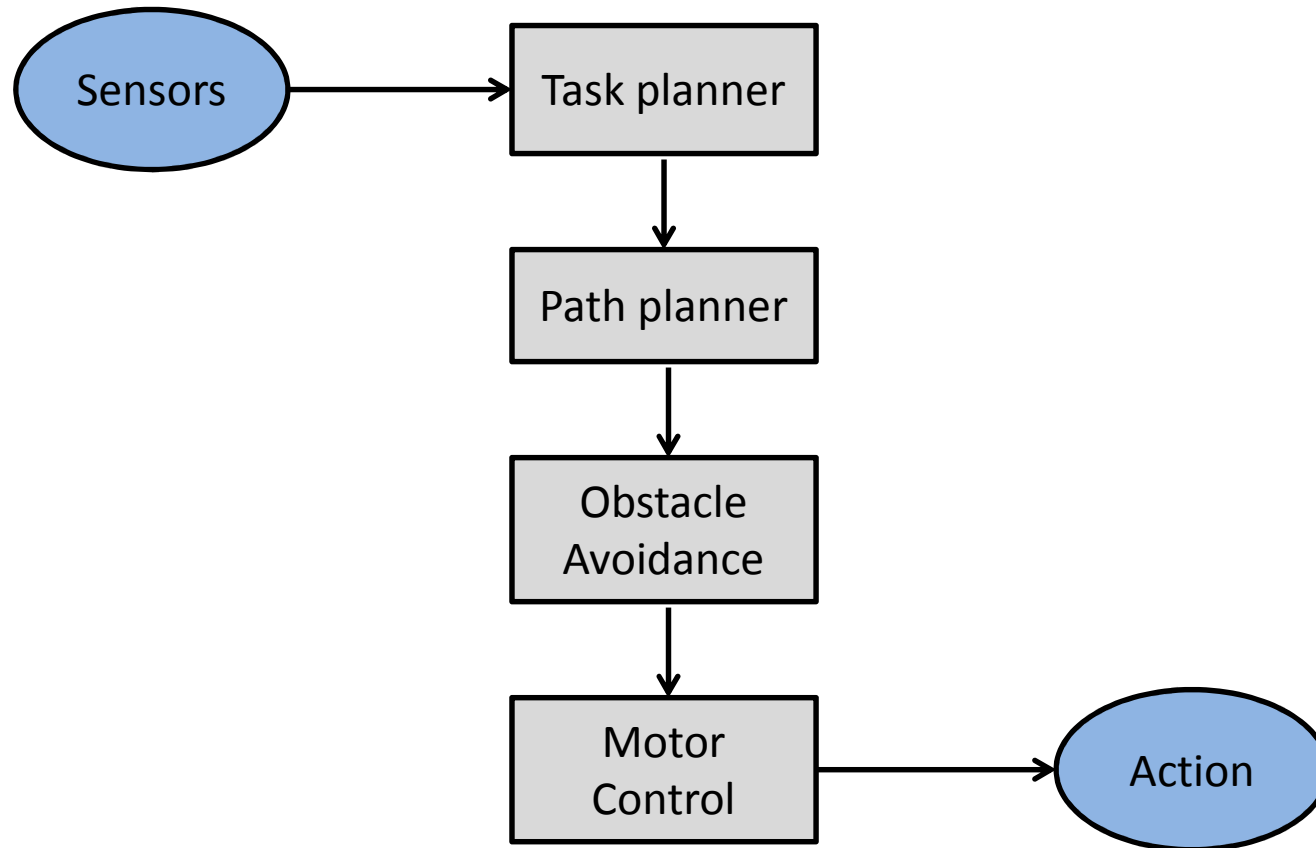
From wikipedia.org

Other examples

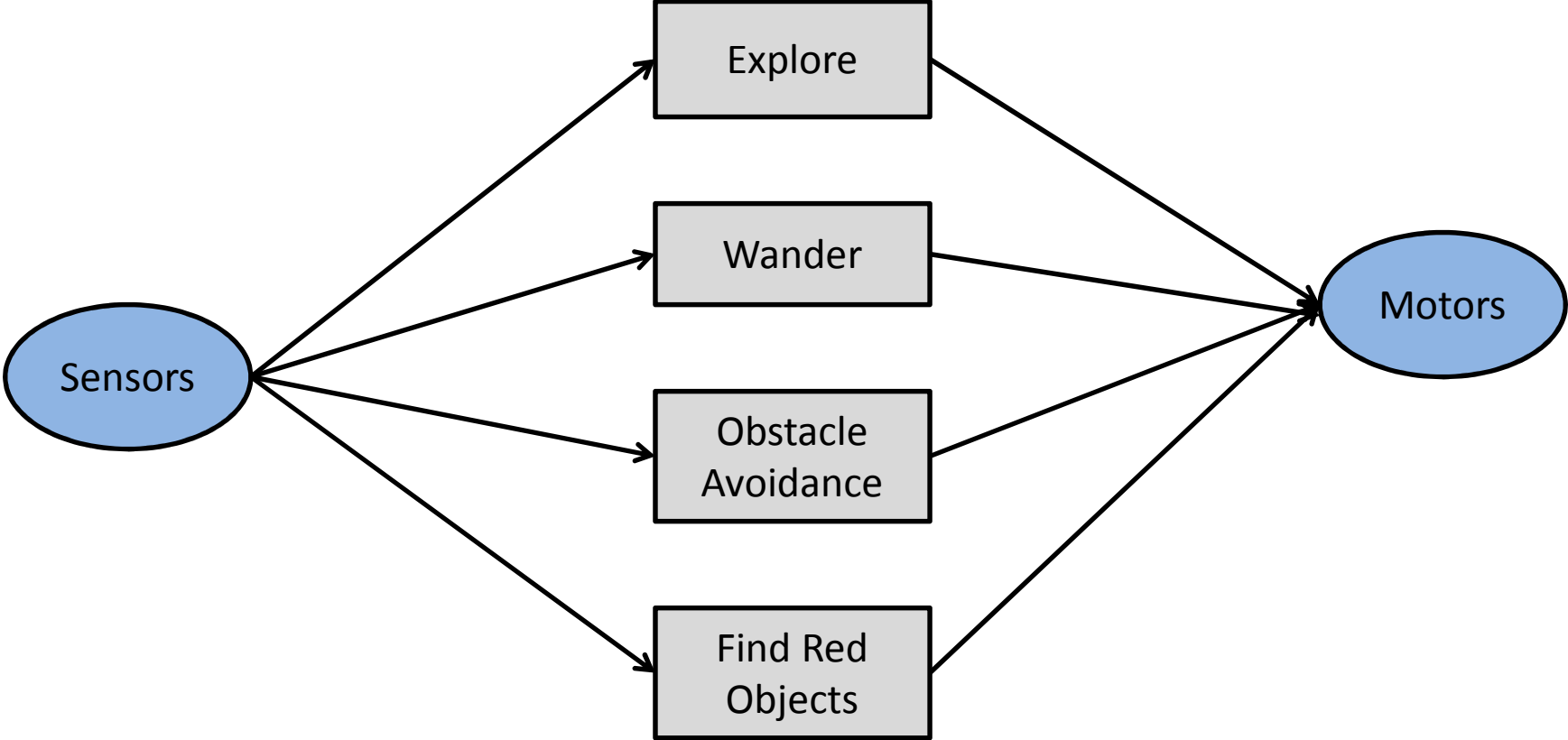
- Bicycle
- Tricycle



Classic Planning



Behavior-based Robotics



Behavior-based Robotics



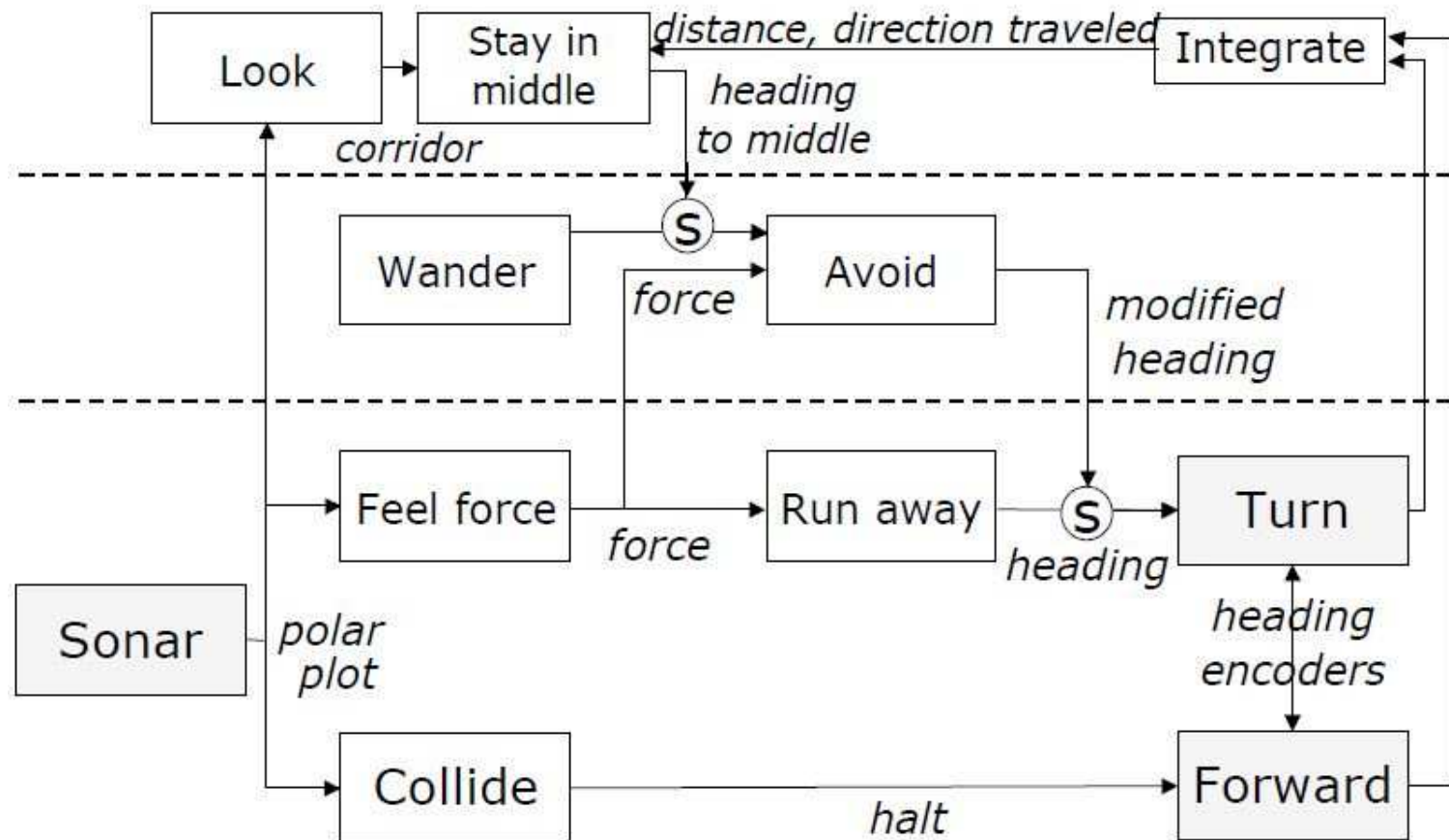
- Reactive systems
- No models. No memory
- Tight coupling between sensors and actuation
- Only local sensing/decision-making
- Based on biology

Behaviors



- Direct mapping from sensing to actuation
- Basic modules
 - Move Forward
 - Wander
 - FollowHallway
 - AvoidObstacles
- Networks of sensing and acting modules (Finite State Automata)
- Subsumption Architecture

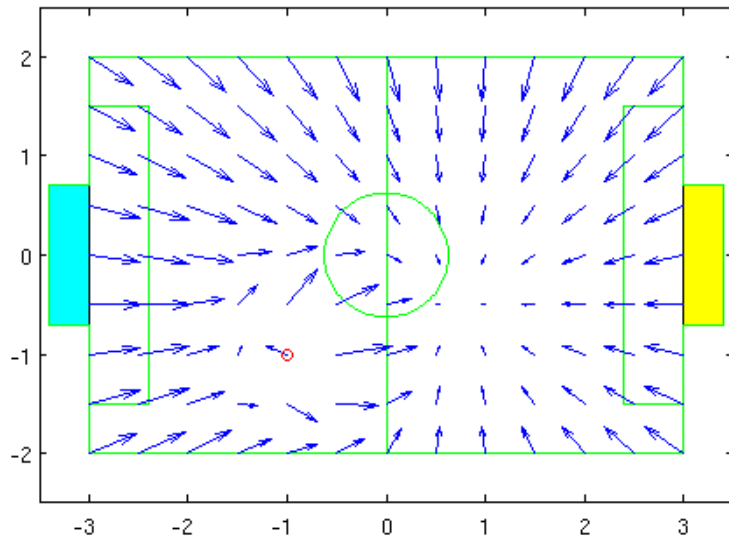
Subsumption Architecture



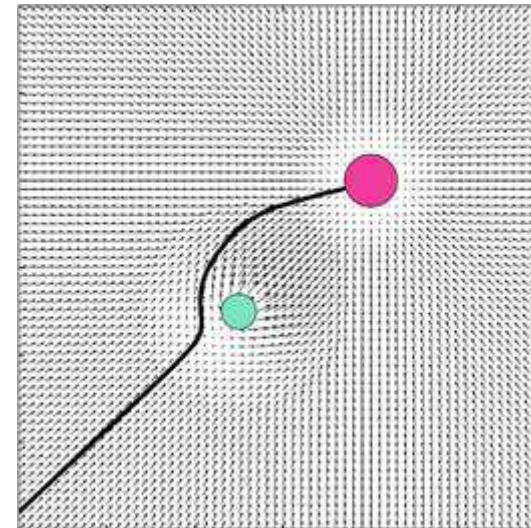
Potential Fields



- Treat robot as particle
- Environment generates potential field vectors
- Direct robot at each point
- Magnitude changes with distance

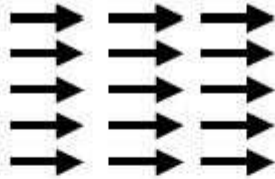


From UPenn

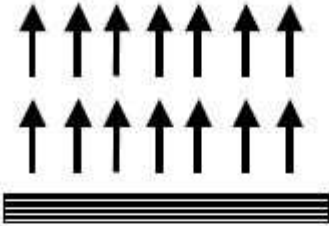


From McGill

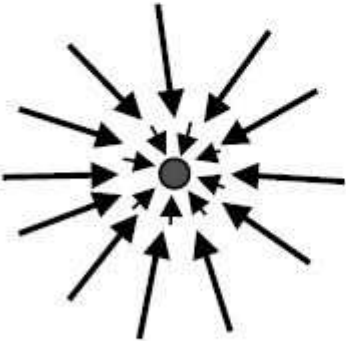
Potential Fields



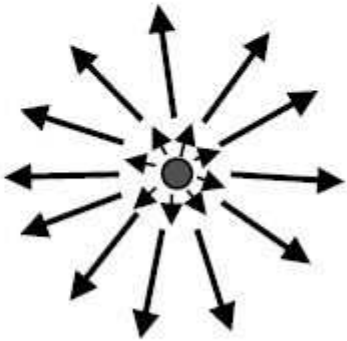
Uniform



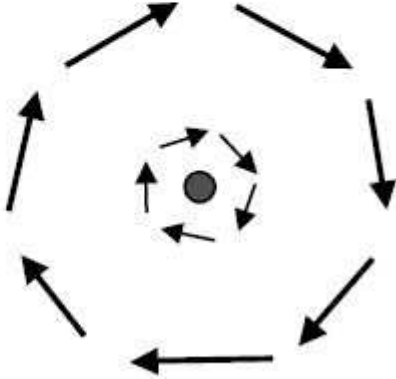
Perpendicular



Attractive



Repulsive



Tangential

Potential Fields



- Obstacle Avoidance - Add repulsive forces around objects
- Wander – Add random field
- Follow Hallway – Two perpendicular forces and uniform force

Characteristics



- Problems with local minima



- Require significant domain knowledge

Upcoming



- Next time: Tracking/Data Fusion
 - Kalman Filters
 - Extended Kalman Filters
 - Bayesian Filters
 - Particle Filters
- Later:
 - Optimal Control (Bellman, MDPs)
 - Adaptive Control
 - Probabilistic Robotics
 - Localization, Mapping