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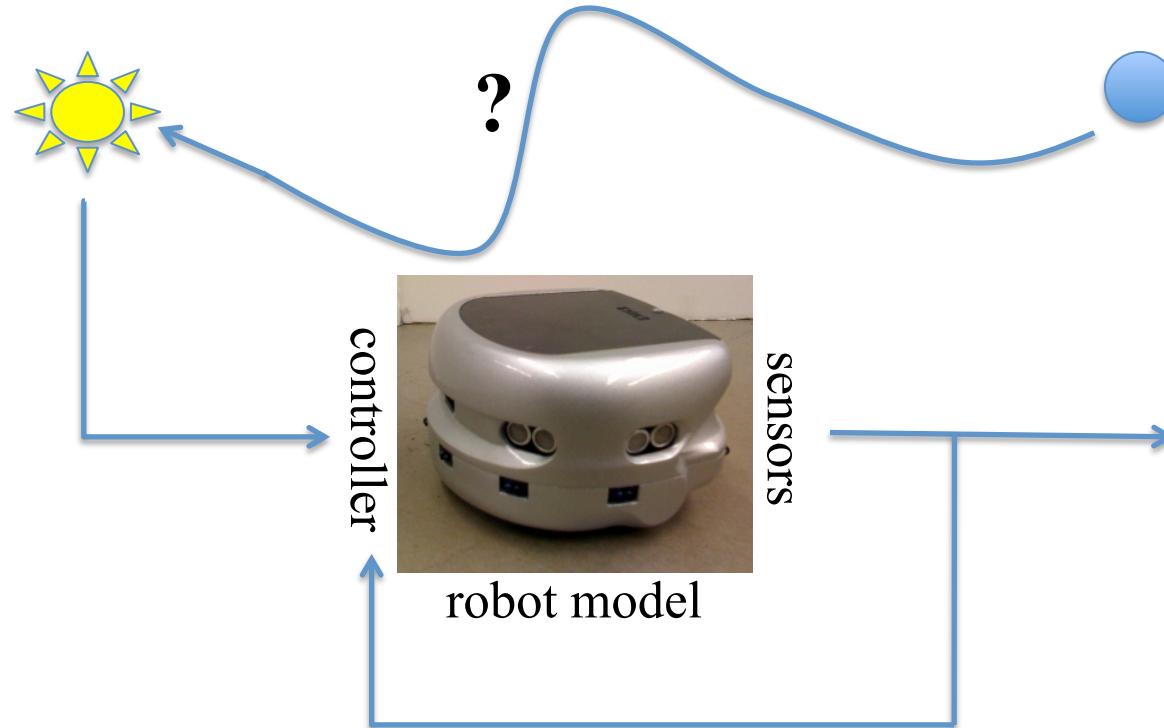
# Control of Mobile Robots

## Module 2 Mobile Robots

*How make mobile robots move in effective, safe, predictable, and collaborative ways using modern control theory?*

# Lecture 2.1 – Driving Robots Around?

- What does it take to drive a robot from point A to point B?



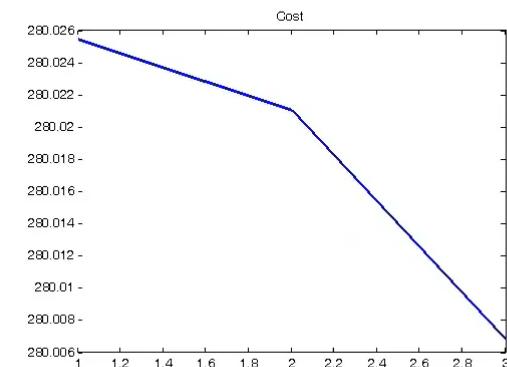
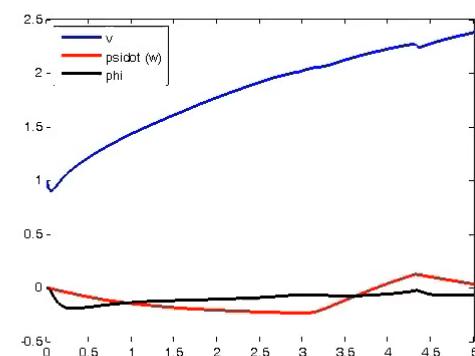
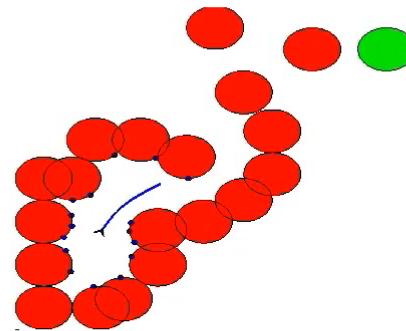
# Divide and Conquer

- The world is dynamic and fundamentally unknown
  - The controller must be able to respond to environmental conditions
  - Instead of building one complicated controller – divide and conquer: Behaviors
    - Go-to-goal
    - Avoid-obstacles
    - Follow-wall
    - Track-target
- ...

# Behaviors

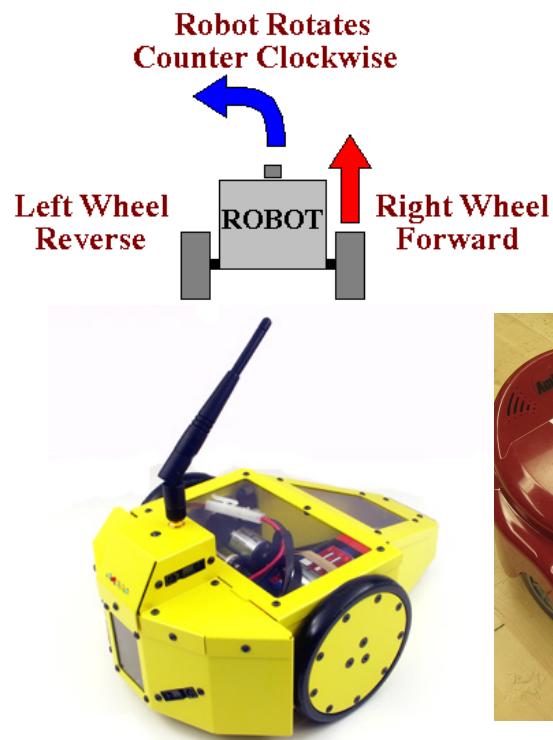


# Behaviors

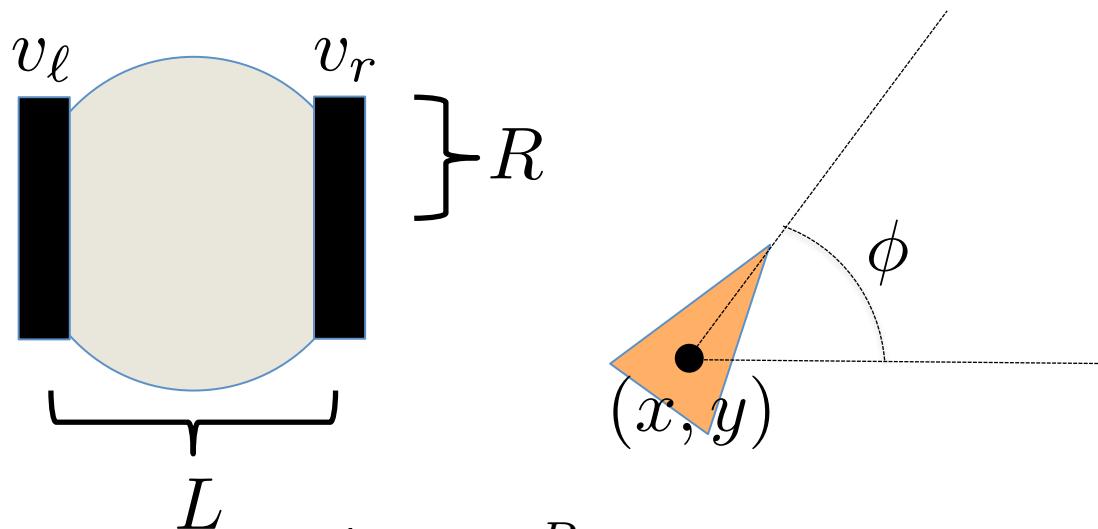


## Lecture 2.2 – Differential Drive Robots

- In order to control mobile robots, we need models
- Differential drive wheeled robots – a very common type



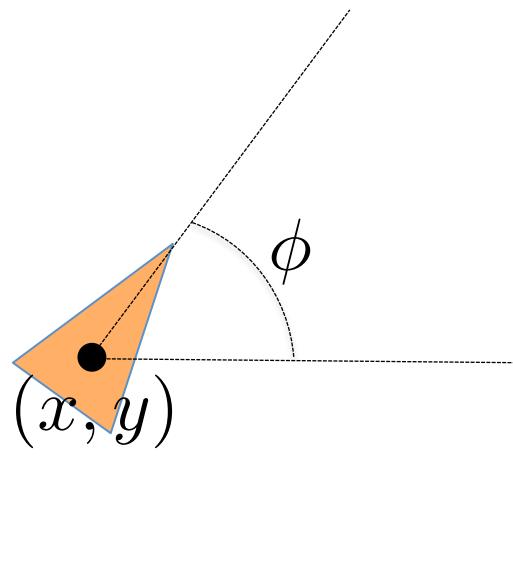
# Model 1.0



$$\left\{ \begin{array}{l} \dot{x} = \frac{R}{2}(v_r + v_\ell) \cos \phi \\ \dot{y} = \frac{R}{2}(v_r + v_\ell) \sin \phi \\ \dot{\phi} = \frac{R}{L}(v_r - v_\ell) \end{array} \right.$$

# The “Unicycle” Model

- But it is not very natural to “think” in terms of wheel velocities
- Go directly for translational and angular velocities



- Inputs:  
 $v$   
 $\omega$
- Dynamics:  
$$\begin{cases} \dot{x} = v \cos \phi \\ \dot{y} = v \sin \phi \\ \dot{\phi} = \omega \end{cases}$$

## Model 2.0

$$\begin{cases} \dot{x} = v \cos \phi \\ \dot{y} = v \sin \phi \\ \dot{\phi} = \omega \end{cases}$$

Design for this model!

$$v = \frac{R}{2}(v_r + v_\ell) \Rightarrow \frac{2v}{R} = v_r + v_\ell$$

$$\omega = \frac{R}{L}(v_r - v_\ell) \Rightarrow \frac{\omega L}{R} = v_r - v_\ell$$

$$\begin{cases} \dot{x} = \frac{R}{2}(v_r + v_\ell) \cos \phi \\ \dot{y} = \frac{R}{2}(v_r + v_\ell) \sin \phi \\ \dot{\phi} = \frac{R}{L}(v_r - v_\ell) \end{cases}$$

Implement this model!

$$v_r = \frac{2v + \omega L}{2R}$$

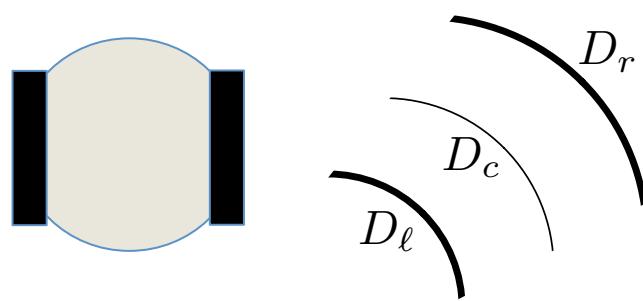
$$v_\ell = \frac{2v - \omega L}{2R}$$

# Lecture 2.3 – Odometry

- The state of the robot is  $(x, y, \phi)$
- *How do we obtain this state information?*
- Two possibilities:
  - External sensors
  - Internal sensors
    - Orientation: Compass, ...
    - Position: Accelerometers, Gyroscopes, ...
    - **Wheel Encoders**

# Wheel Encoders

- Wheel encoders give the distance moved by each wheel
- Assume the wheels are following an arc (short time scale)

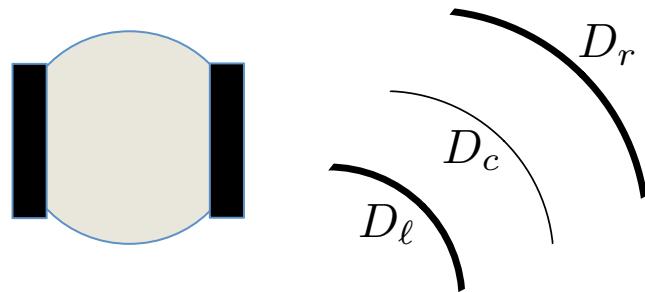


$$D_c = \frac{D_\ell + D_r}{2}$$

$$\begin{cases} x' = x + D_c \cos(\phi) \\ y' = y + D_c \sin(\phi) \\ \phi' = \phi + \frac{D_r - D_\ell}{L} \end{cases}$$

# Wheel Encoders

- But how do we know how far each wheel has moved?



- For both wheels:

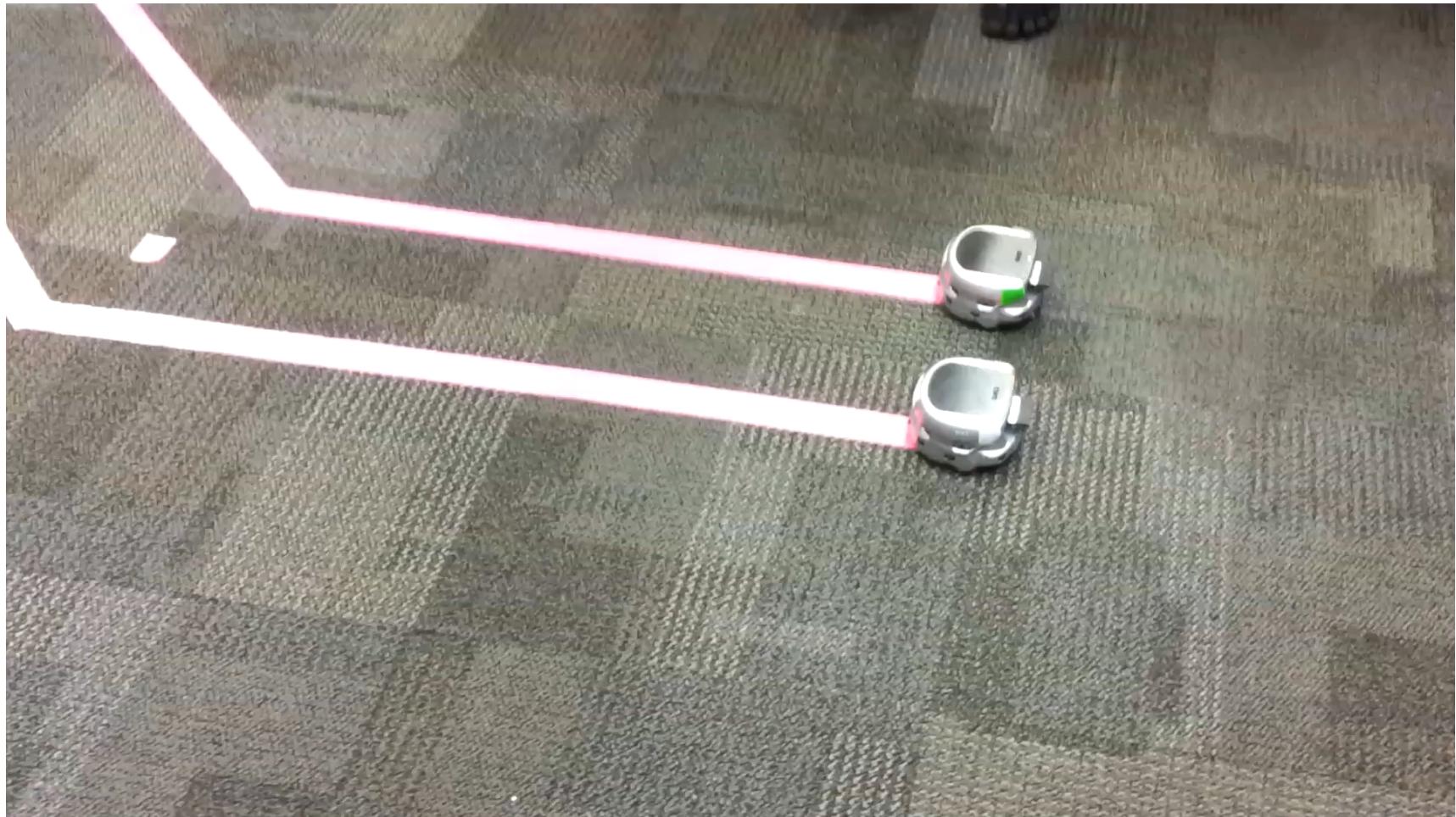
$$\Delta\text{tick} = \text{tick}' - \text{tick}$$

$$D = 2\pi R \frac{\Delta\text{tick}}{N}$$

- Assume each wheel has  $N$  “ticks” per revolution
- Most wheel encoders give the total tick count since the beginning

# A Major Disclaimer

**DRIFT!!!**

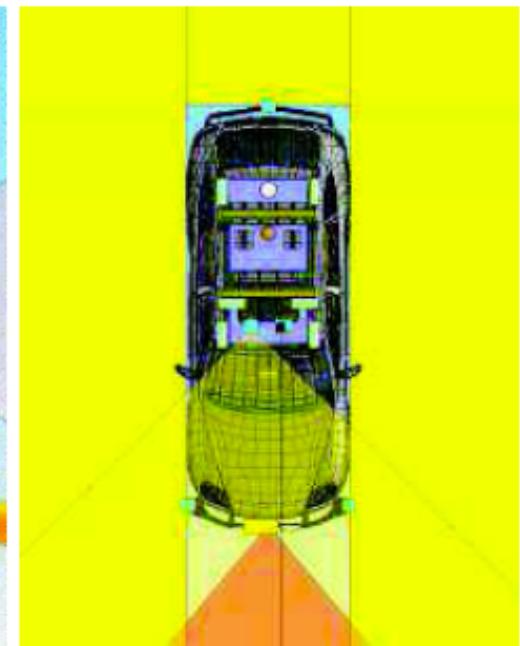
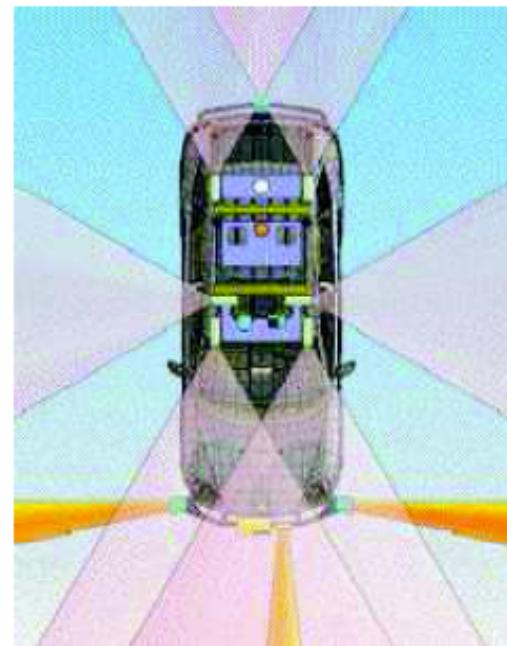
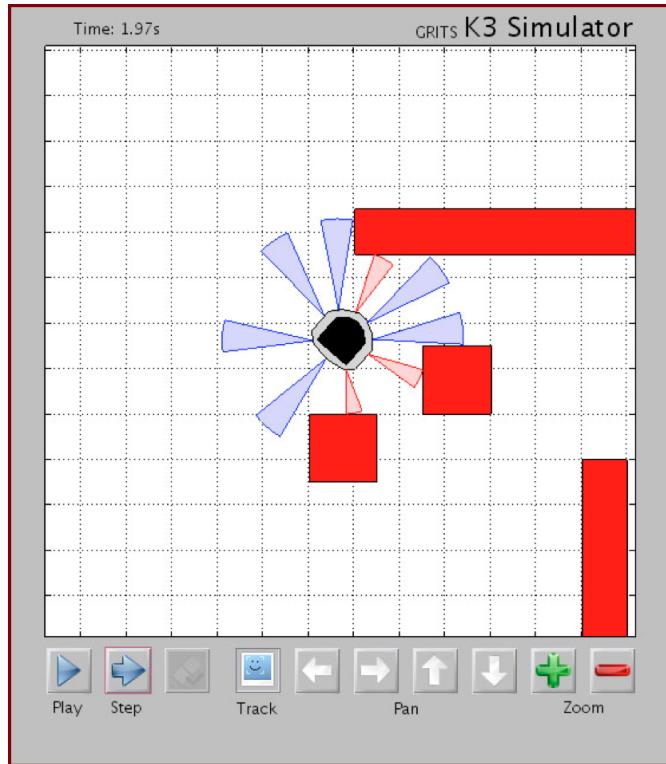


## Lecture 2.4 – Sensors

- Robots need to know what the world around them looks like
- The standard sensor suite includes a “skirt” of range sensors:
  - IR, Ultra-Sound, LIDAR,...
- Other standard external sensors include
  - Vision
  - Tactile
  - “GPS”

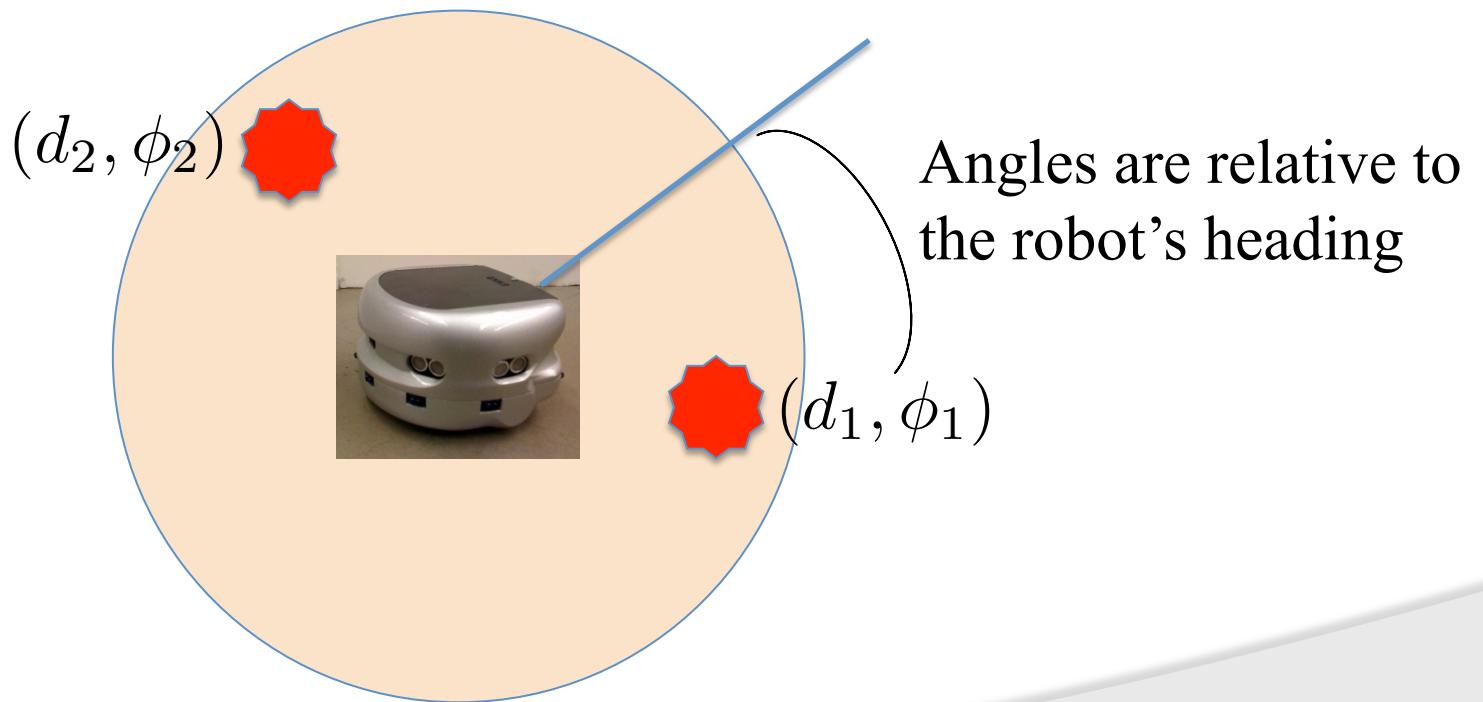
# Range-Sensor Skirts

- We will mainly deal with range-sensors



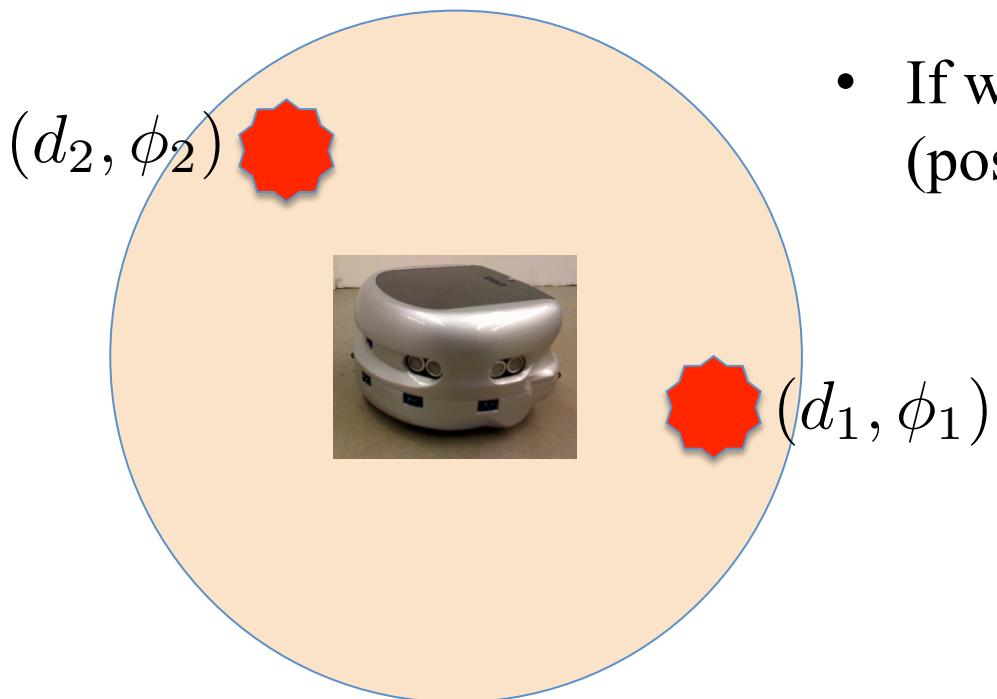
# The Disk Abstraction

- Instead of worrying about the resolution of the sensors, assume we know the distance and direction to all obstacles around us (that are close enough)



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- Instead of worrying about the resolution of the sensors, assume we know the distance and direction to all obstacles around us (that are close enough)

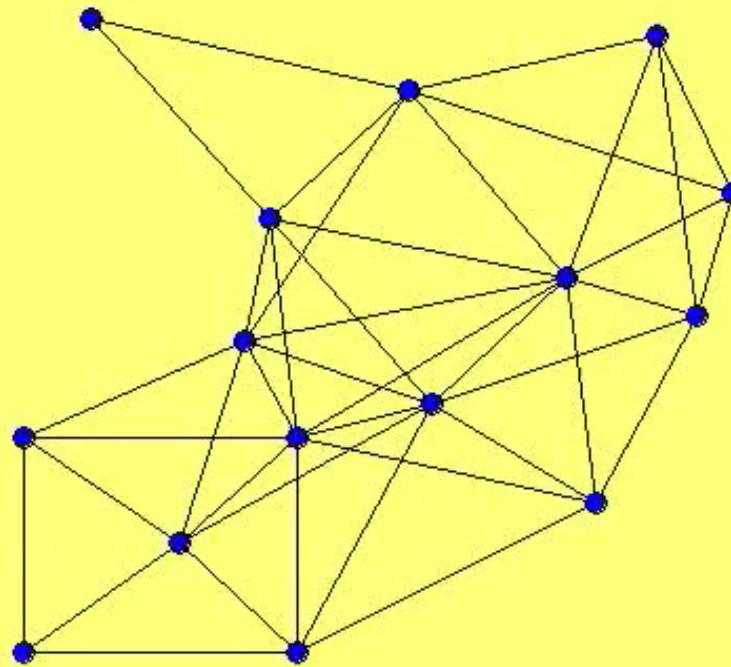


- If we know our own pose (position and orientation):

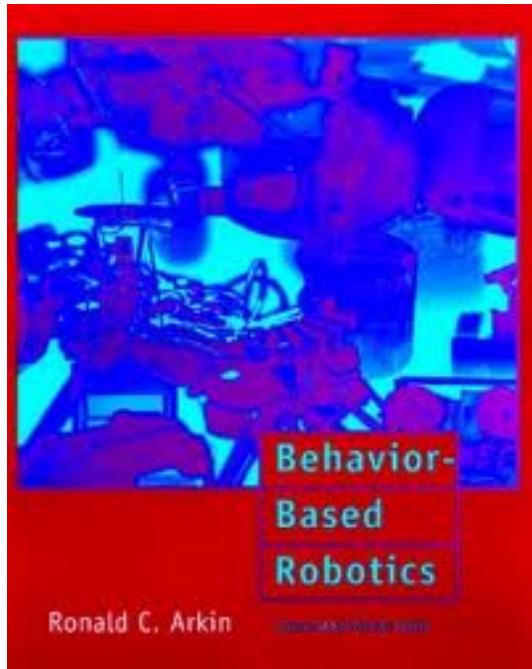
$$x_1 = x + d_1 \cos(\phi_1 + \phi)$$
$$y_1 = y + d_1 \sin(\phi_1 + \phi)$$

# Example: Rendezvous

## Example: Rendezvous



# Lecture 2.5 – Behavior-Based Robotics

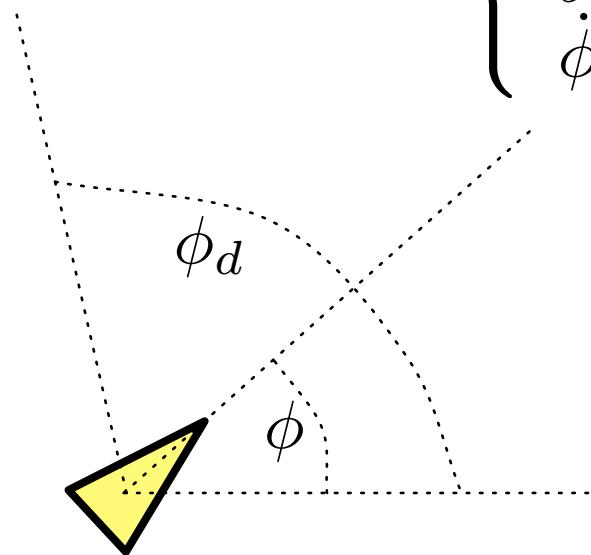


- The world is fundamentally unknown and changing
- Does not make sense to over-plan
- Key idea: Develop a library of useful controllers (=behaviors)
- Switch among controllers in response to environmental changes

# Building a Behavior v.1

- Assume we have a differential-drive, wheeled mobile robot driving at a constant speed

$$\begin{cases} \dot{x} = v_0 \cos \phi \\ \dot{y} = v_0 \sin \phi \\ \dot{\phi} = \omega \end{cases}$$



- Want to drive in a desired heading
- $$\omega = ???$$

# Building a Behavior v.1

- We have a reference, a model, a control input, and a tracking error:

$$r = \phi_d, \quad e = \phi_d - \phi, \quad \dot{\phi} = \omega$$

- We not use PID?

$$\omega = K_P e + K_I \int e d\tau + K_D \dot{e}$$

# Dealing with Angles

- This typically will not work since we are dealing with angles:

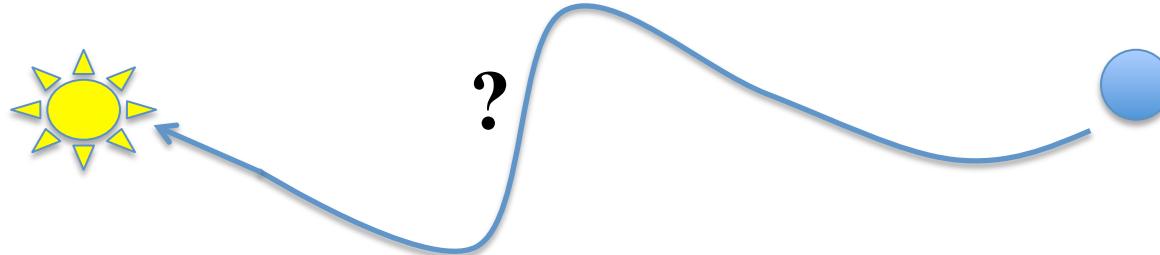
$$\phi_d = 0, \quad \phi = 100\pi \quad \Rightarrow \quad e = -100\pi$$

- Solution: Ensure that  $e \in [-\pi, \pi]$
- Standard trick is to use atan2

$$e' = \text{atan2}(\sin(e), \cos(e)) \in [-\pi, \pi]$$

## Example: Navigation

- Problem: Go to a goal location without bumping in to obstacles:

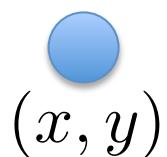
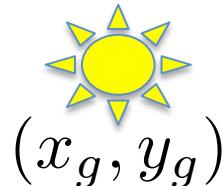


- At a minimum, we need two behaviors: Go-to-goal and Avoid-obstacles

## Lecture 2.6 – Go-To-Goal

- How drive a robot to a goal location?

$$\begin{cases} \dot{x} = v_0 \cos \phi \\ \dot{y} = v_0 \sin \phi \\ \dot{\phi} = \omega \end{cases} \quad e = \phi_d - \phi, \quad \omega = \text{PID}(e)$$



$$\phi_d = \arctan \left( \frac{y_g - y}{x_g - x} \right)$$

# Attempt 1

$$\omega = K(\phi_d - \phi)$$



**ANGLES!!!**

# Attempt 2

$$\omega = K(\phi_d - \phi)$$



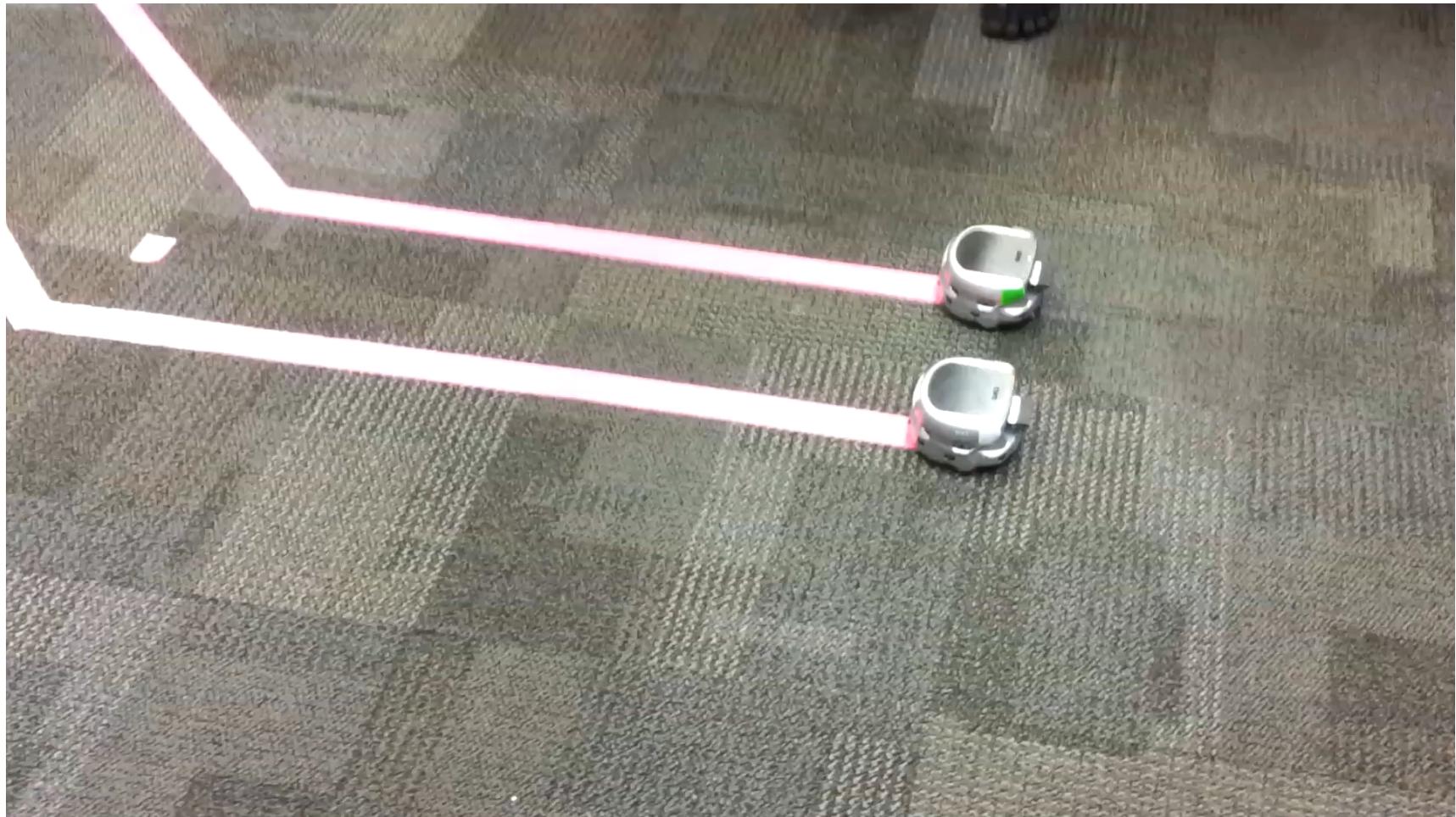
GAIN TOO LOW!!!

# Attempt 3

$$\omega = K_{\text{big}}(\phi_d - \phi)$$

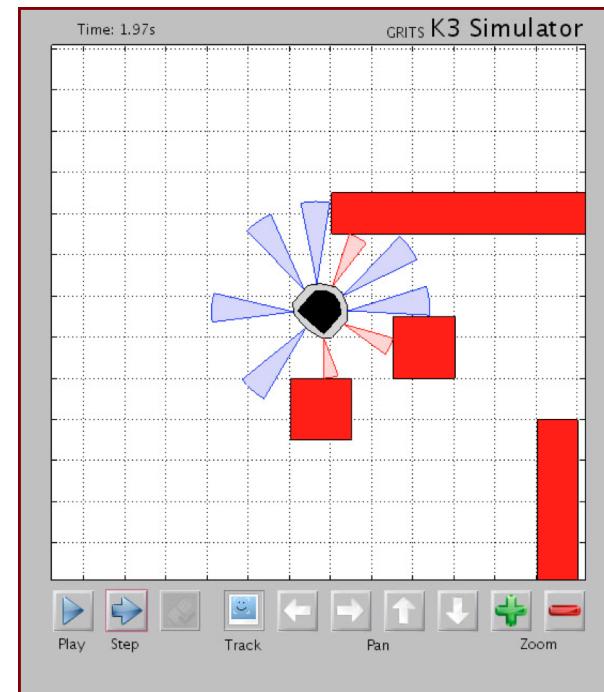


**JUST RIGHT!!!**



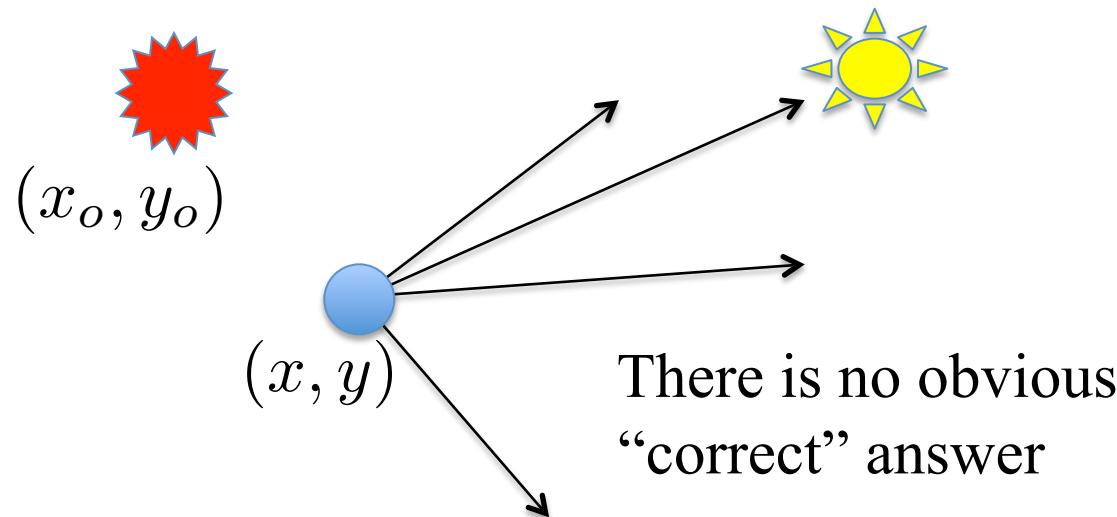
# Lecture 2.7 – The GRITS Robot Simulator

- A MATLAB-based simulator
- Simulates Khepera III differential drive mobile robot(-s) with IR range-sensors and wheel-encoders in cluttered environments
- Available at  
<http://gritslab.gatech.edu/projects/robot-simulator/>  
as stand-alone executable and MATLAB package



## Lecture 2.8 – Obstacle-Avoidance

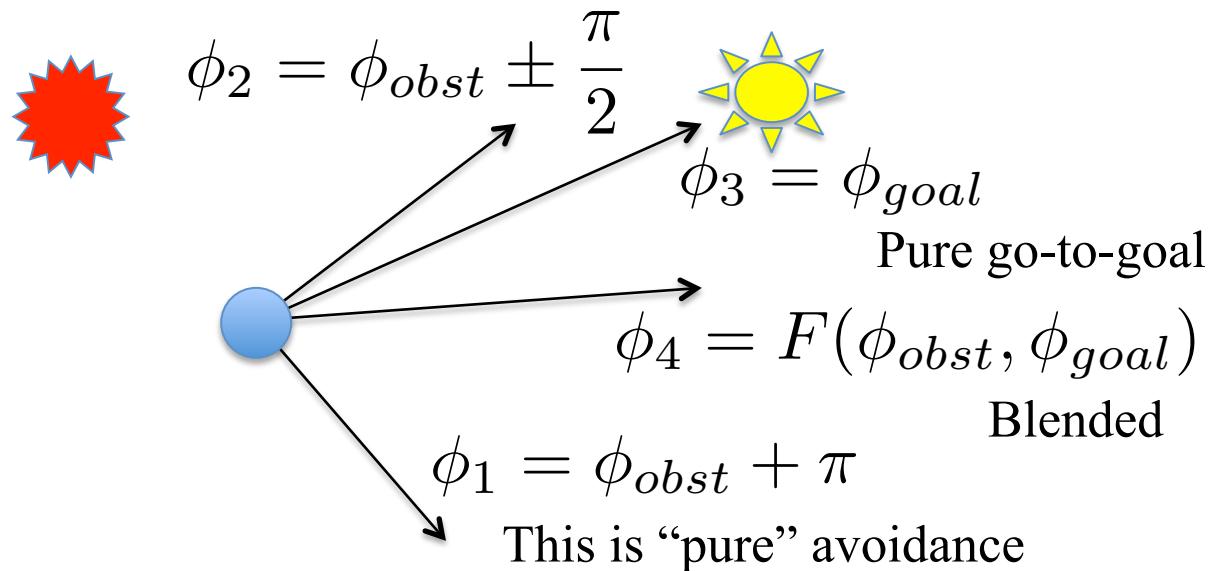
- How avoid driving into obstacles?
- We can use the same idea by defining a desired heading



# Where to Go?

Choice depends on direction to goal.

Not “pure” but “blended”



# Arbitration Mechanisms

- This example illustrate two fundamentally different “arbitration mechanisms”
  - Winner takes all = Hard switches
  - Blending = Combined behaviors
- Both approaches have merit in different situations
  - Performance?
  - Analysis?
- We will see how to design systematic behaviors and arbitration mechanisms

# Module 2: Mobile Robots

- Module 2: Mobile robots
- Module 3: *Enough chit-chat!* Let's start over in a more systematic and formal manner!
  - **Linear Systems**