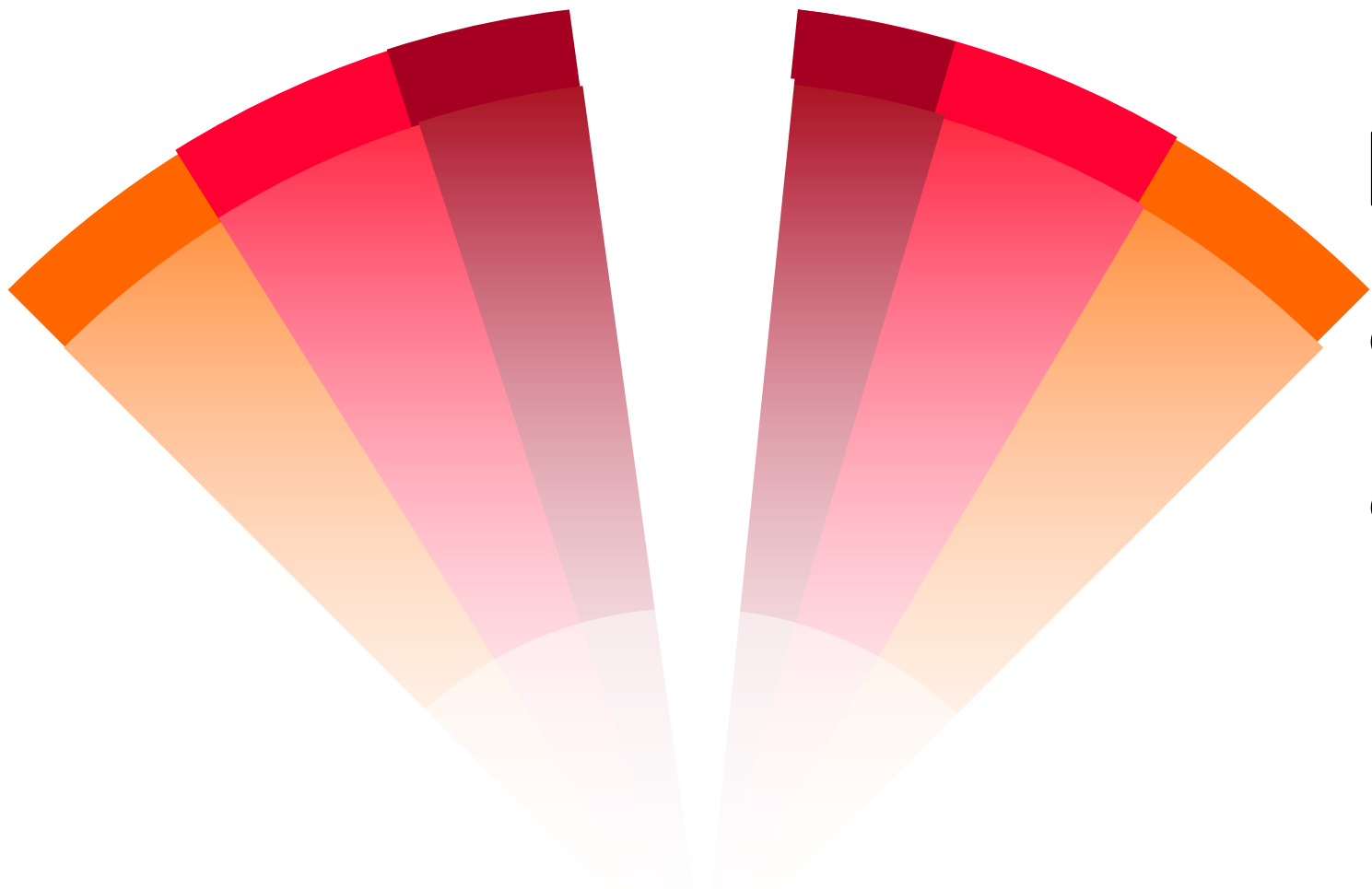


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Vlad Estivill-Castro (2016)

Robots for People

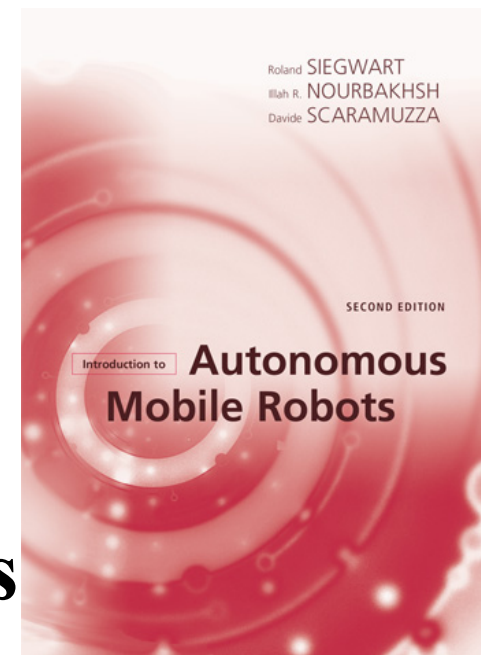
--- A project for intelligent integrated systems

What is the course about?

- textbook
- **Introduction to Autonomous Robots**

- second edition

Roland Siegwart, Illah R. Nourbakhsh, and Davide Scaramuzza



The three key questions in Mobile Robotics

- ▶ **Where am I ?**
- ▶ **Where am I going ?**
- ▶ **How do I get there ?**
 - To answer these questions the robot has
 - to have a model of the environment (given or autonomously built)
 - perceive and analyze the environment
 - find its position/situation within the environment
 - plan and execute the movement

This course will deal with Locomotion and Navigation

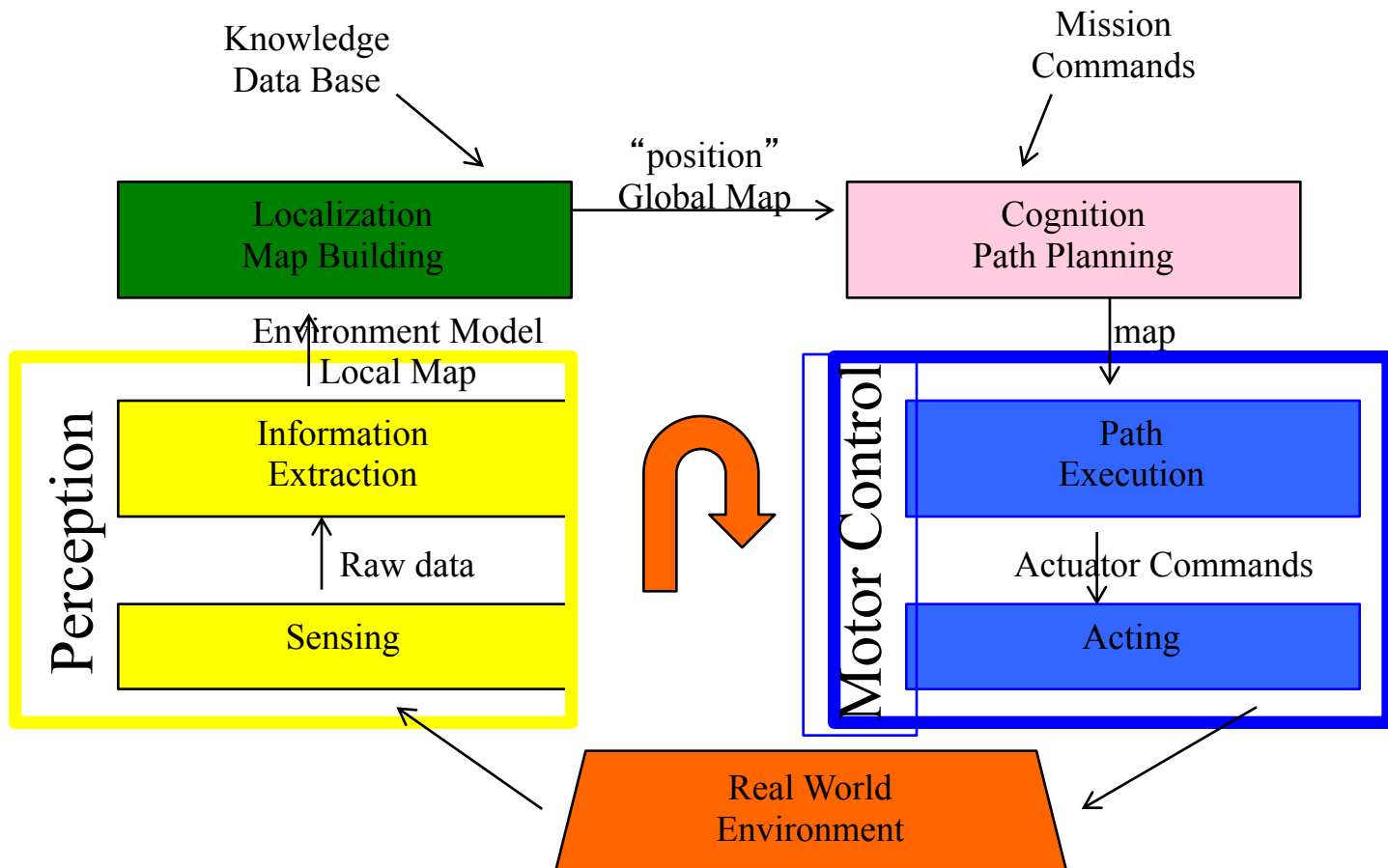
- Perception
- Localization and Mapping
- Planning
- Motion Generation

Course Content

- ▶ *Introduction*
- ▶ *Locomotion (little)*
- ▶ *Mobile Robot Kinematics*
- ▶ *Perception*
- ▶ *Mobile Robot Localization and Mapping*
- ▶ *Planning and Navigation*
- ▶ *- a spin to useful techniques for Humanoid Robots, RoboCup, Nao*

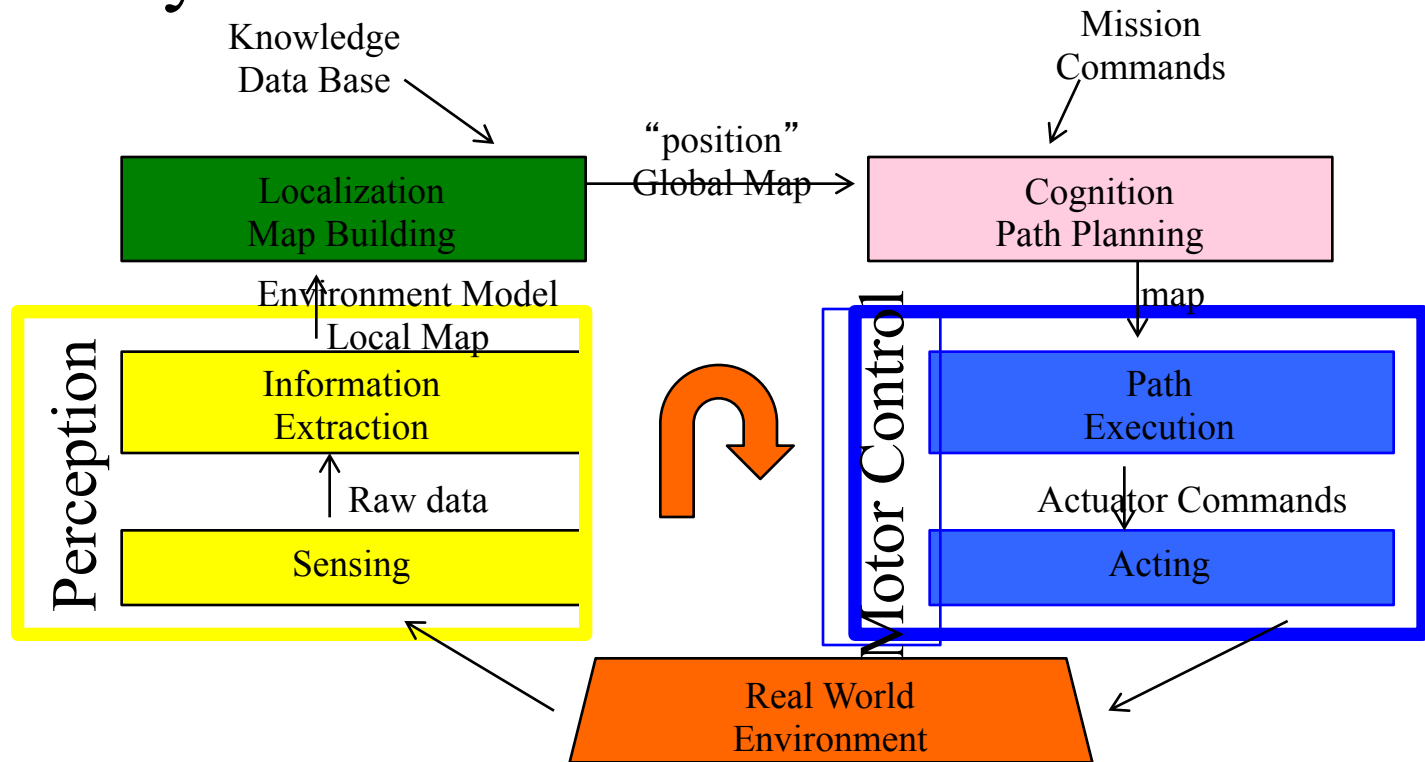
General Control Scheme for Mobile Robot Systems

• (Roland Siegwart, Illah Nourbakhsh, Davide Scaramuzza)



Control loop

- ▀ dynamically changing
- ▀ no compact model available
- ▀ many sources of uncertainties



Two approaches

Classical AI

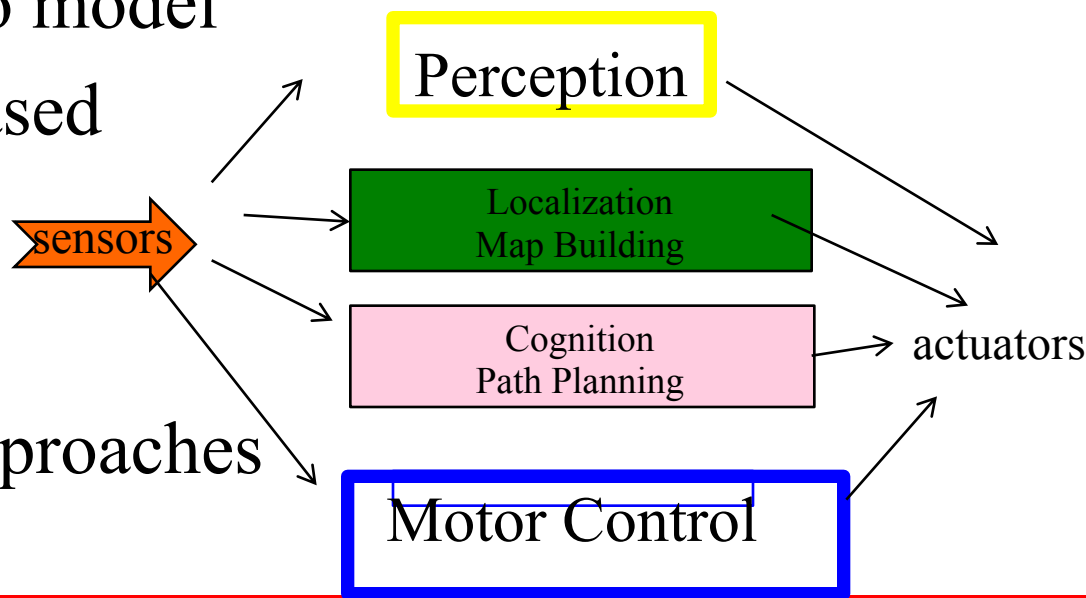
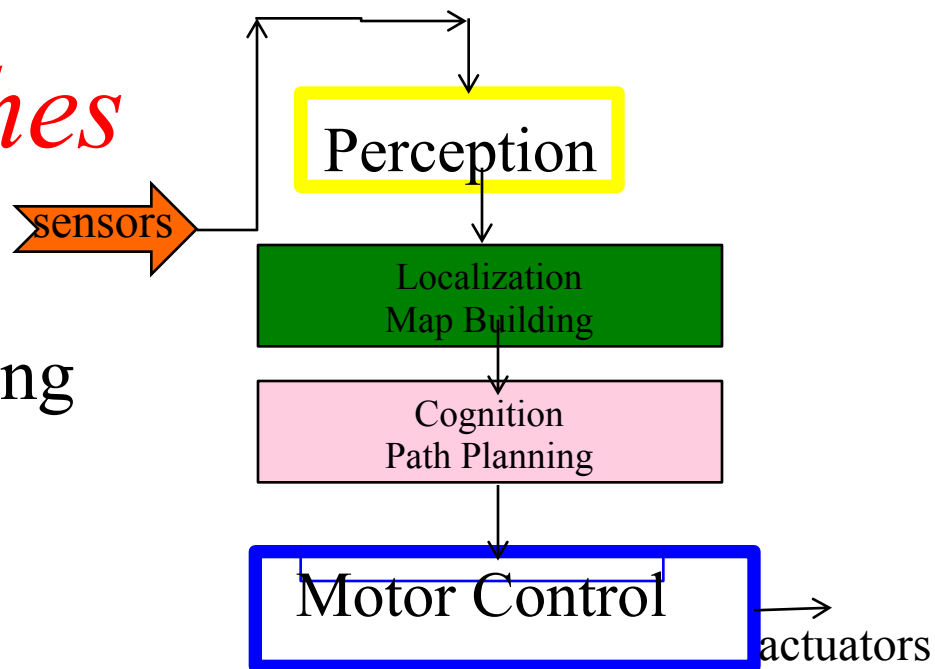
- complete modeling
- function based

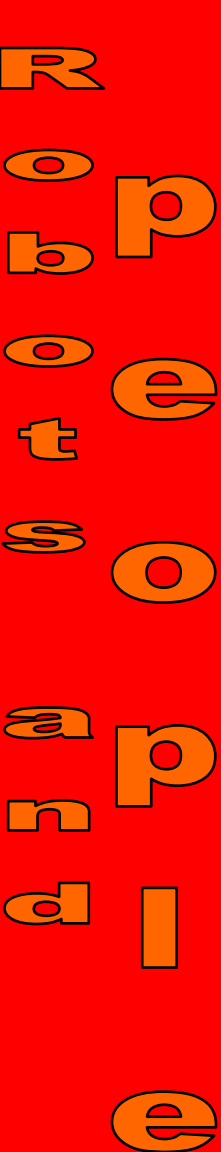
New AI

- sparse or no model
- behavior based

Solution

- combine approaches





The Key for Autonomous Navigation

► Environment Representation

- continuous metric $\rightarrow x, y \Theta$
- Discrete metric \rightarrow metric grid
- Discrete Topological \rightarrow topological grid

► Environment Modeling

- Raw sensor data
 - large volume, low distinctiveness
- Low level features
 - medium volume of data, some distinctiveness
- High level features
 - low volume data high distinctiveness

Environment representation

- ▶ Odometry
 - follow directions/commands like Navigator
 - follow a treasure
- ▶ Modified Environments
 - very specific/encoded information
- ▶ Feature based
 - Landmarks, Affordances of the environment

Human navigation

- Topological and conceptual maps
 - the dimensions of Australia, and what is close, north, south
 - number of buildings vs name of buildings
- Relations at different scales

The Map Categories

1. recognizable landmarks
2. topological maps (graph model)
 - games like snakes and ladders
3. metric topological maps
 - edges have weight equal to distance
4. Full metric maps
 - nodes are geo-positioned

Understanding --- Probabilistic reasoning (Bayesian inference)

- ▶ Reasoning in the presence of **uncertainties** and **incomplete information**
- ▶ Combining **preliminary information** and **models** with **learning from observed data**
- ▶ $P(A|B) = P(B|A) P(A)/P(B)$
 - the probability that the new situation is A once I observed B is related to the chances of noticing B when in situation A and the ratio of observing A with the chances of observing B

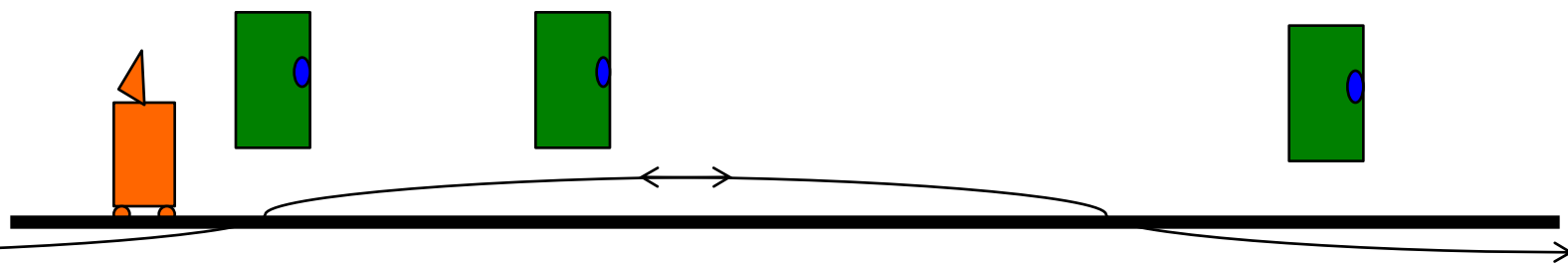
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Reducing Uncertainties

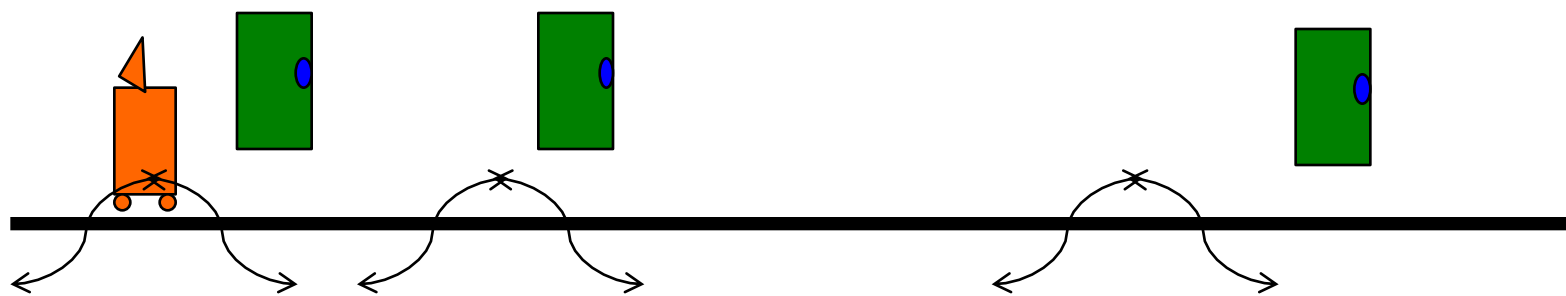
Where can I be?



Anywhere

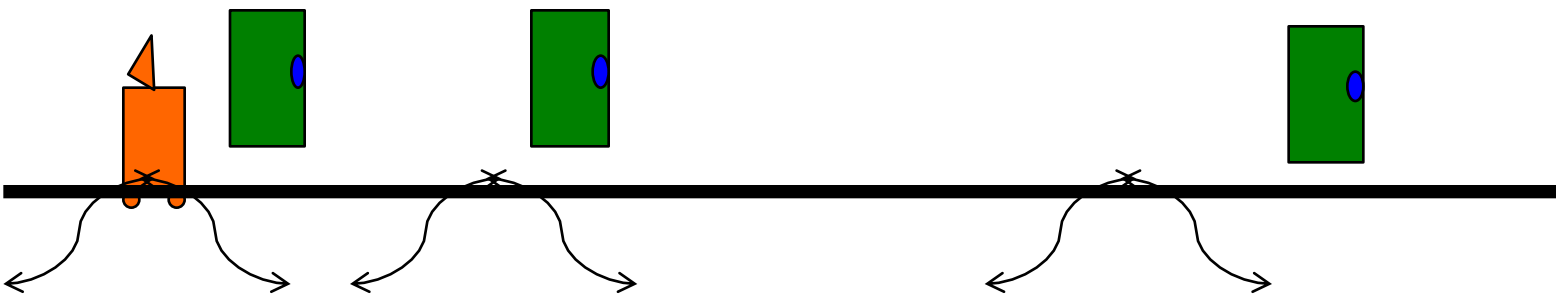


Observe a door

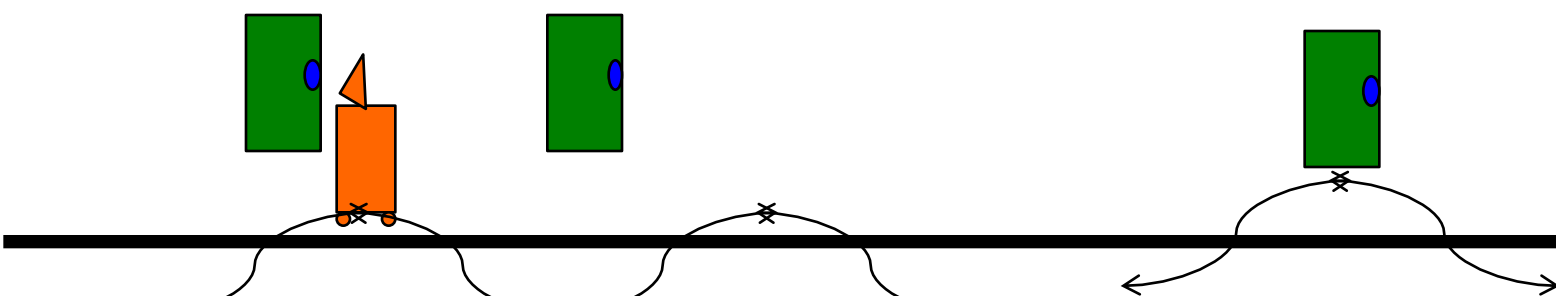


Reducing Uncertainties

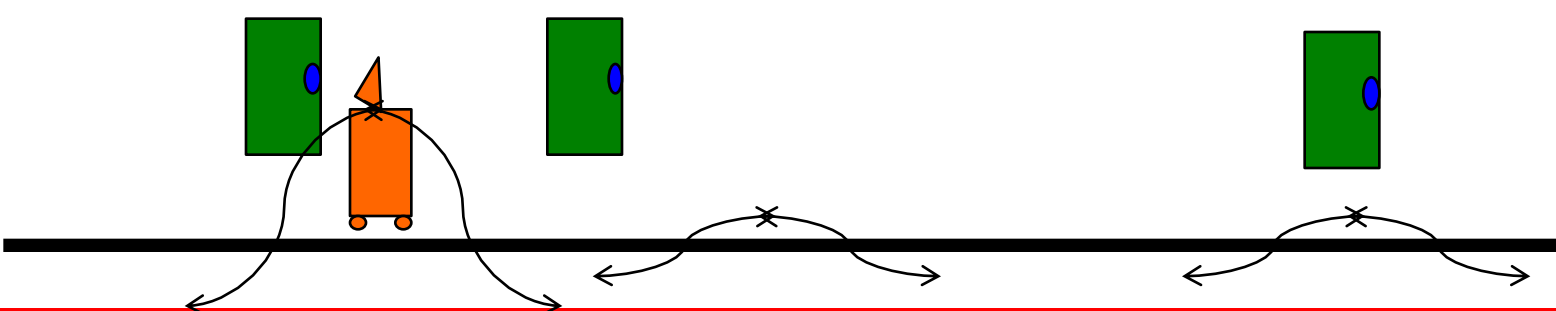
Observe a door



Walk a bit



Observe a door again



Metric Navigation: Probabilistic Position Estimation (Kalman Filter)

- Continuous, recursive and very compact
 - ACT PHASE
 - state prediction via a motion model
 - obtain $x(k+1 | k)$ and $P(k+1 | k)$ given $x(k|k)$ $P(k|k)$
 - SEE PHASE
 - Balance the advice of where I am (result of the motion model) and the advice of where I am (result of the sensor)
 - Depending on how much I trust each

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State prediction: odometry

- Incrementally (dead reckoning)
 - There is a discrepancy into the real trajectory and the belief of where I have been
 - DRIFT
 - cumulative error
- Under no error, current position is accurate

Methods for Localization

: Quantitative Metric Approach

1. A priori, know your environment
 - in graph, map
2. Extract features (line segments, landmarks)
3. Match them with location in map
4. Estimate your position
 - Kalman filter, Markov filters (particle filters)
 - Balance observations with your belief
 - issues:
 - estimation of uncertainties
 - weights to prior statistics

Belief Representation

- ▶ Well known probability models
 - mixture of multi-variate normal distribution

- ▶ Descriptive statistics
 - Histograms

Probabilistic SLAM

- **Simultaneous localization and mapping (SLAM)**
 - **first characteristic question**, *What does the world look like?*
 - **second characteristic question**, *Where am I?*
- **The robotics community claims to have solved the SLAM problem**
 - *lots of variations on the assumptions / sensors/ motions*
 - *issues of complexity*
 - *quality*

Summary

- Mobile robotics
 - the most challenging type of robotics
 - robots are anywhere in the world

- Many architectures, some well accepted but found insufficient, new approaches discussed

- Large body of techniques
 - feature extraction
 - sensor fusion
 - mapping, tracking and/or localization

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THANK YOU

