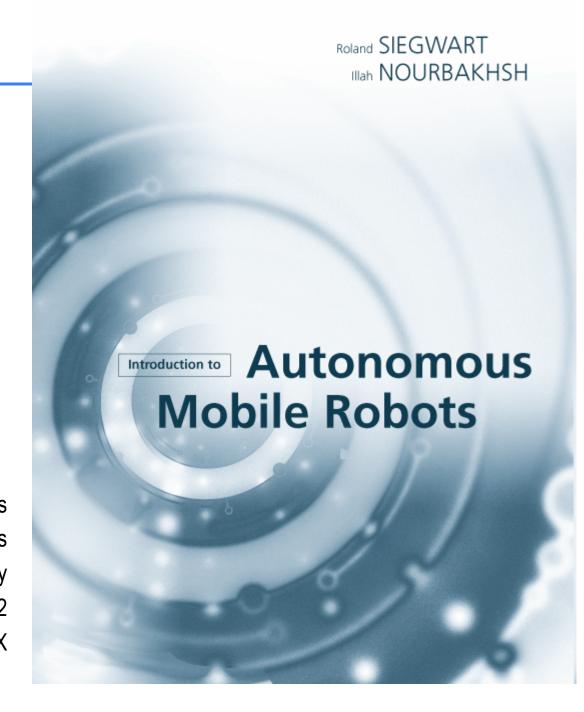
Slides that go with the book

Intelligent Robotics and Autonomous Agents series
The MIT Press
Massachusetts Institute of Technology
Cambridge, Massachusetts 02142
ISBN 0-262-19502-X



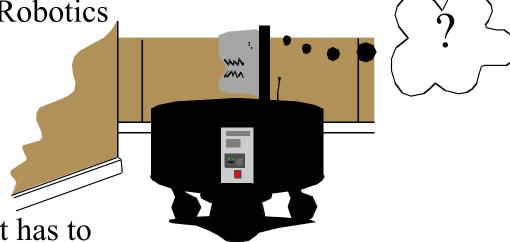
Name this roboticist ...

'70s glasses-

Long legs foretell tall adult

Autonomous Mobile Robots

- 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 within the environment
 - plan and execute the movement
- This course will deal with Locomotion and Navigation (Perception, Localization, Planning and motion generation)

Content of the Course

- 1. Introduction
- 2. Locomotion
- 3. Mobile Robot Kinematics
- 4. Perception
- 5. Mobile Robot Localization
- 6. Planning and Navigation

- Other Aspects of Autonomous Mobile Systems
- > Applications

Autonomous Mobile Robots, Chapter 1

Program

| Date | Room/ Time | Торіс | Responsible |
|-------|-------------------|---|-----------------------------|
| 21.10 | BM 2135, 10 - 12 | Introduction: problem statements, typical applications, video | R. Siegwart |
| 28.10 | BM 2135, 10 - 12 | Locomotion with legs and wheels (2h) | R. Siegwart |
| 28.10 | BM 2127, 13 - 15 | Exercise 1: Introduction to Matlab | Y. Piguet, A. Martinelli |
| 4.11 | BM 2135, 10 - 12 | Mobile Robots Kinematics I: Kinematics model (2h) | R. Siegwart |
| 4.11 | BM 2127, 13 - 15 | Exercise 2: Kinematics model and trajectory calculation of wheeled robots | R. Siegwart |
| 11.11 | BM 2135, 10 - 12 | Mobile Robots Kinematics II: Motion control (1h) Perception I: Sensing and Perception(1h) | R. Siegwart |
| 18.11 | BM 2135, 10 - 12 | Perception II: Sensing and Perception (2h) | R. Siegwart |
| 18.11 | BM 2127, 13 - 152 | Exercise 3: Motion control of a differentially driven robot (Matlab/Khepera) | G. Caprari A. Martinelli |
| 25.11 | BM 2135, 10 - 12 | Perception III: Uncertainty Representation, feature extraction (2h) | R. Siegwart |
| 2.12 | BM 2135, 10 - 12 | Localization I: Introduction, odometry, belief representation (2h) | R. Siegwart |
| 2.12 | BM 2127, 13 - 15 | Exercise 4: Vision and/or laser; take picture, feature extraction; uncertainty representation; belief representation | J. Weingarten B. Jensen |
| 9.12 | BM 2135, 10 - 12 | Localization II: Map representation, introduction to probabilistic map- based localization, Markov localization(2h) | R. Siegwart |
| 16.12 | BM 2135, 10 - 12 | Localization III: Kalman filter localization (2) | R. Siegwart |
| 16.12 | BM 2127, 13 - 15 | Exercise 5: Probabilistic pose estimation with Khepera base on topological map | N. Tomatis A. Tapus |
| 6.1 | BM 2135, 10 - 12 | Localization IV: Other examples of localization systems, map building (2) | R. Siegwart |
| 13.1 | BM 2135, 10 - 12 | Architectures for Navigation I: Introduction, path planning (2) | R. Siegwart |
| 13.1 | BM 2127, 13 - 15 | Exercise 6: Potential Field: Field generation (Matlab), implementation on Khepera | A. Martinelli Y. Piguet |
| 20.1 | BM 2135, 10 - 12 | Architectures for Navigation II: Obstacle avoidance, techniques for decomposition (2) | R. Siegwart |
| 27.1 | BM 2135, 10 - 12 | Architectures for Navigation III: Case Studies: architectures with behaviors (2) | R. Siegwart |
| 27.1 | BM 2127, 13 - 15 | Exercise 7: Obstacle avoidance base on local grid (Matlab), implementation on Khepera | R. Philippsen G. Ramel |
| 3.2 | BM 2135, 10 - 12 | Other aspects of autonomous mobile robots, applications (1h) Research in mobile robotics at ASL - EPFL, summery (1h) | R. Siegwart |

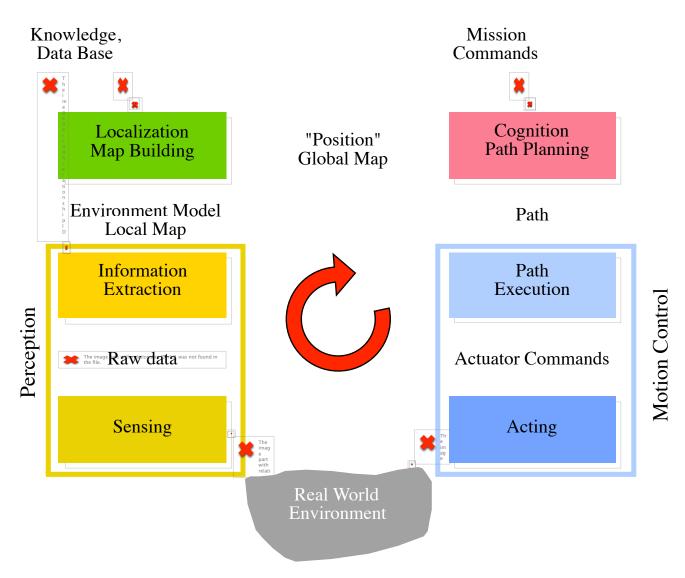
Goal of today's lecture

- Introduce the basic problems of mobile robotics
 - > the basic questions
 - > examples and it's challenges
- Introduce some basic terminology
 - > Environment representation and modeling
- Introduce the key challenges of mobile robot navigation
 - Localization and map-building
- Some examples/videos showing the state-of-the-art

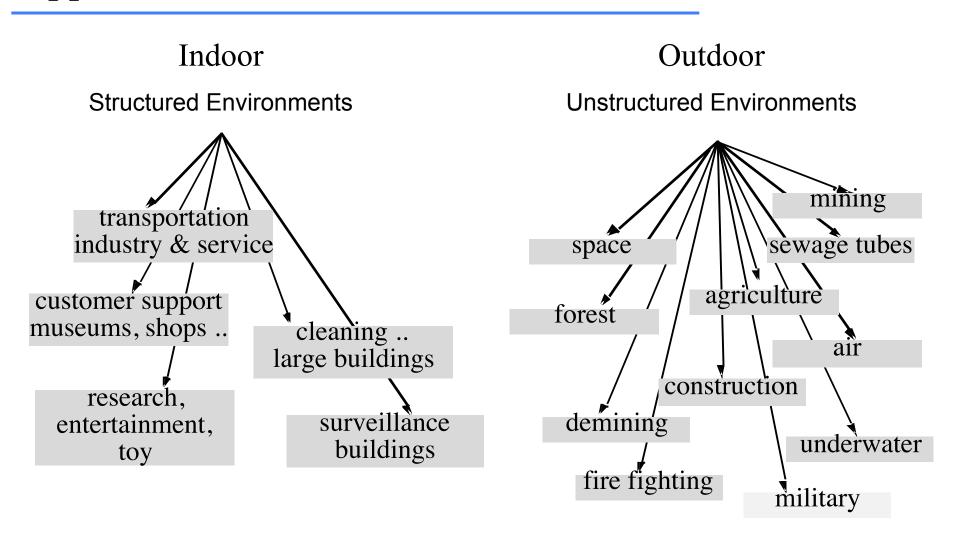
From Manipulators to Mobile Robots



General Control Scheme for Mobile Robot Systems



Applications of Mobile Robots



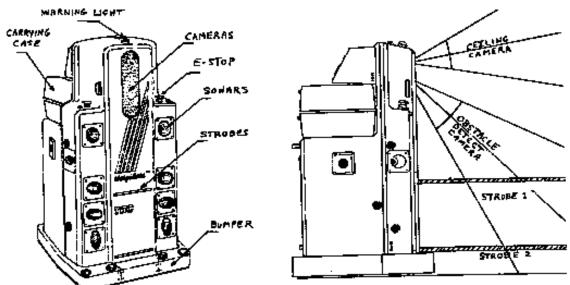
Automatic Guided Vehicles



 Newest generation of Automatic Guided Vehicle of VOLVO used to transport motor blocks from one assembly station to an other. It is guided by an electrical wire installed in the floor but it is also able to leave the wire to avoid obstacles. There are over 4000 AGVs at VOLVO's plants.

Helpmate





 HELPMATE is a mobile robot used in hospitals for transportation tasks. It has various on board sensors for autonomous navigation in the corridors. The main sensor for localization is a camera looking to the ceiling. It can detect the lamps on the ceiling as reference (landmark). http://www.ntplx.net/~helpmate/

The Cye Personal Robot

- Two-wheeled differential drive robot
- Controlled by remote PC (19.2 kb)
- Options:
 - > vacuum cleaner
 - > trailer





The Dyson Vacuum Cleaner Robot

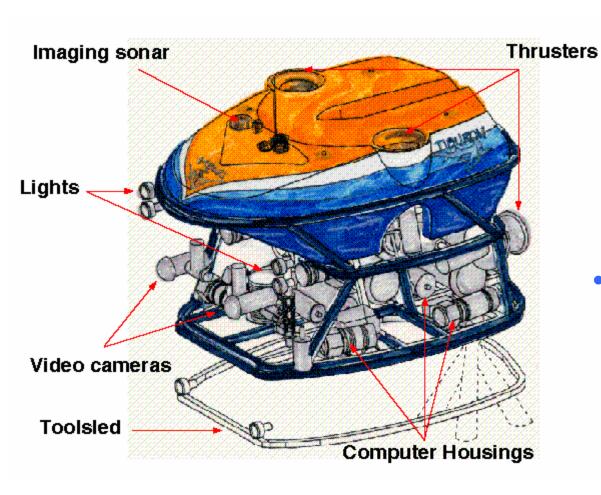


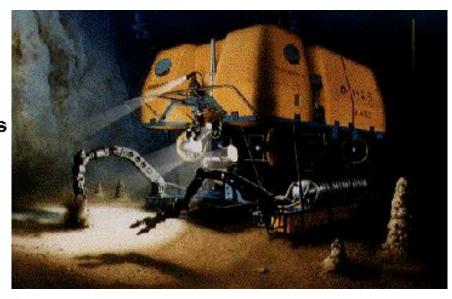
BR700 Cleaning Robot



 BR 700 cleaning robot developed and sold by Kärcher Inc., Germany. Its navigation system is based on a very sophisticated sonar system and a gyro. http://www.kaercher.de

ROV Tiburon Underwater Robot





 Picture of robot ROV Tiburon for underwater archaeology (teleoperated)- used by MBARI for deep-sea research, this UAV provides autonomous hovering capabilities for the human operator.

The Pioneer

 Picture of Pioneer, the teleoperated robot that is supposed to explore the Sarcophagus at Chernobyl



© R. Siegwart, I. Nourbakhsh

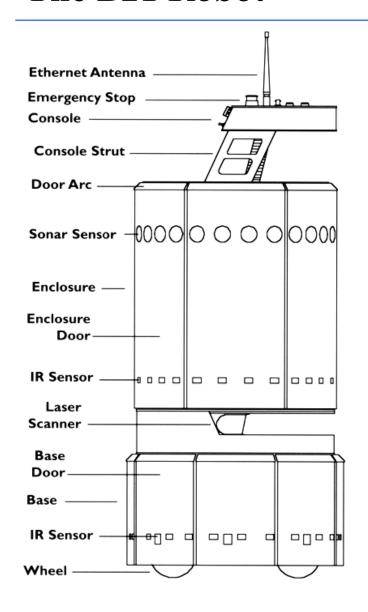
The Pioneer





 PIONEER 1 is a modular mobile robot offering various options like a gripper or an on board camera. It is equipped with a sophisticated navigation library developed at Stanford Research Institute (SRI). http://www.activmedia.com/robots

The B21 Robot

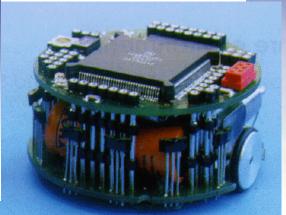




 B21 of Real World Interface is a sophisticated mobile robot with up to three Intel Pentium processors on board. It has all different kinds of on board sensors for high performance navigation tasks. http://www.rwii.com

The Khepera Robot





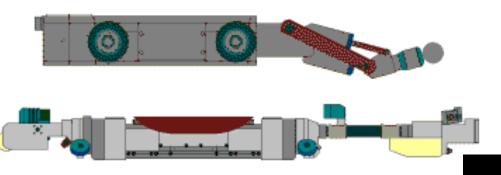
KHEPERA is a small mobile robot for research and education. It sizes only about 60 mm in diameter. Additional modules with cameras, grippers and much more are available. More then 700 units have already been sold (end of 1998). http://diwww.epfl.ch/lami/robots/K-family/ K-Team.html

Forester Robot



 Pulstech developed the first 'industrial like' walking robot. It is designed moving wood out of the forest. The leg coordination is automated, but navigation is still done by the human operator on the robot. http://www.plustech.fi/

Robots for Tube Inspection



 HÄCHER robots for sewage tube inspection and reparation. These systems are still fully teleoperated. http://www.haechler.ch

EPFL / SEDIREP: Ventilation inspection robot

Sojourner, First Robot on Mars



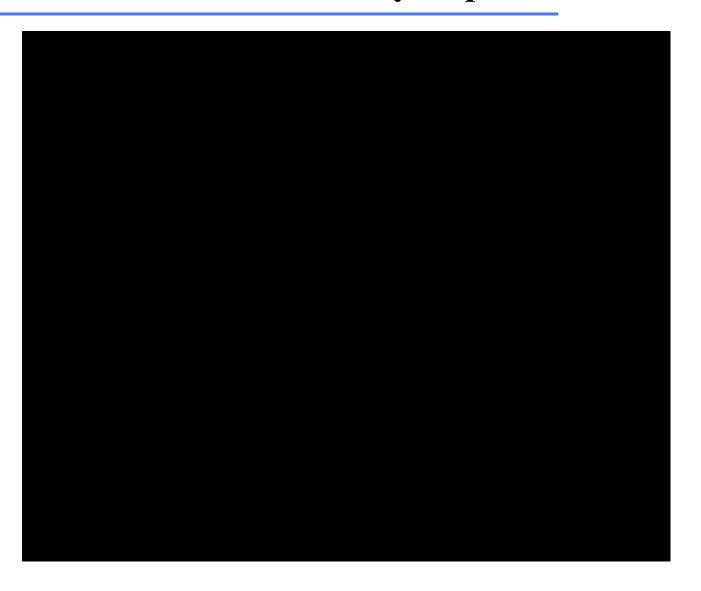
2003 Mars Rover **Press Release Animation**

> Dan Maas dmaas@dcine.com

Sojourner was used during the Pathfinder mission to explore the mars in summer 1997. It was nearly fully teleoperated from earth. However, some on board sensors allowed for obstacle detection. http:// ranier.oact.hq.nasa.g ov/ telerobotics_page/ telerobotics.shtm

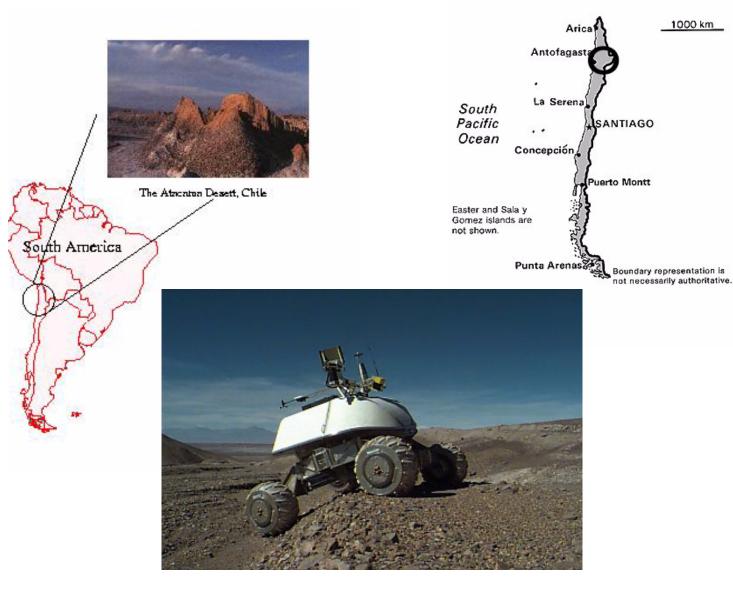
© R. Siegwart, I. Nourbakhsh

Autonomous Robot for Planetary Exploration (ASL-EPFL)



NOMAD, Carnegie Mellon / NASA

http://img.arc.nasa.gov/Nomad/



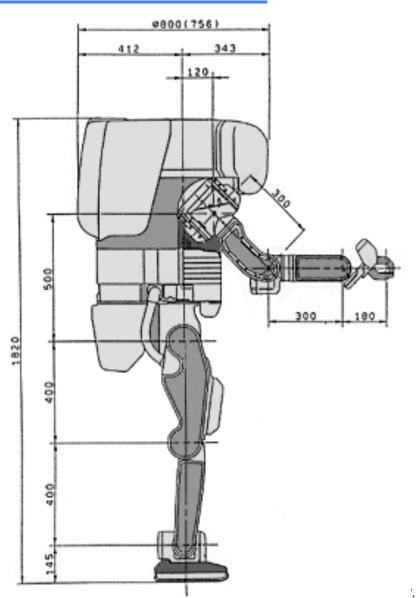


1000 km

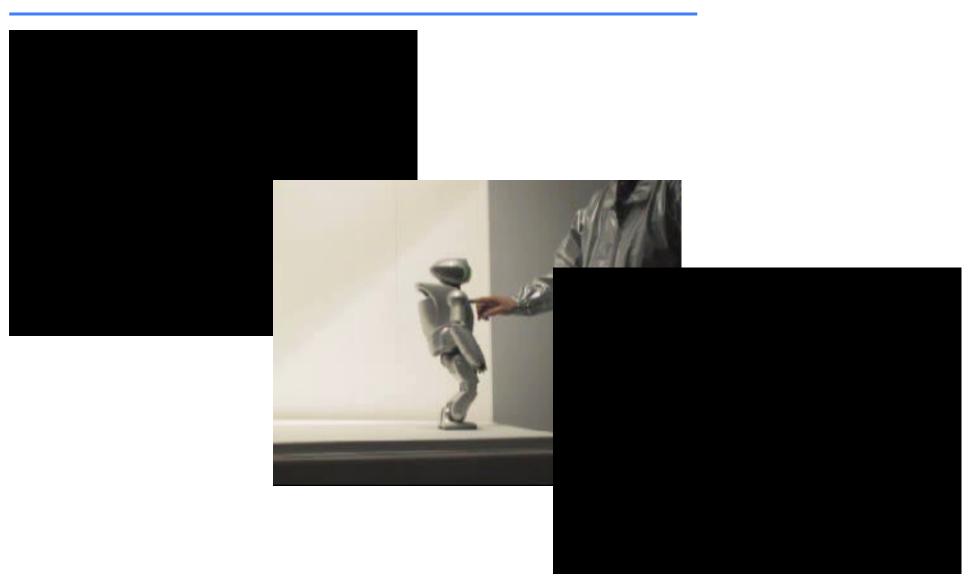


The Honda Walking Robot http://www.honda.co.jp/tech/other/robot.html





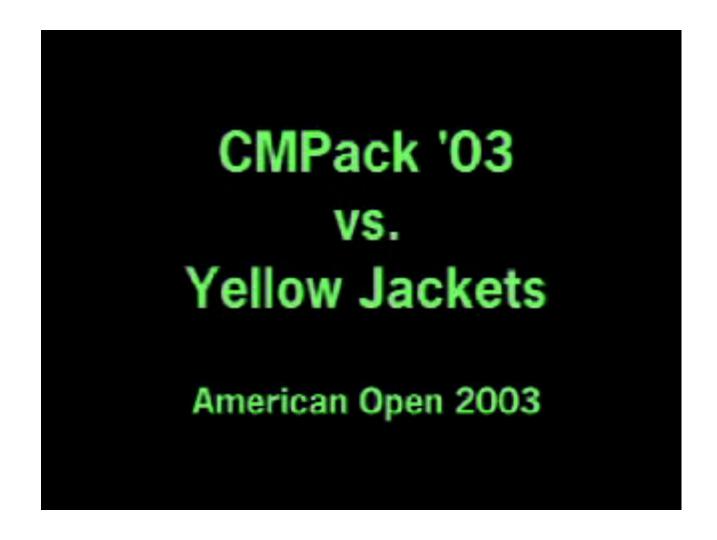
Humanoid Robots (Sony)



Toy Robot Aibo from Sony

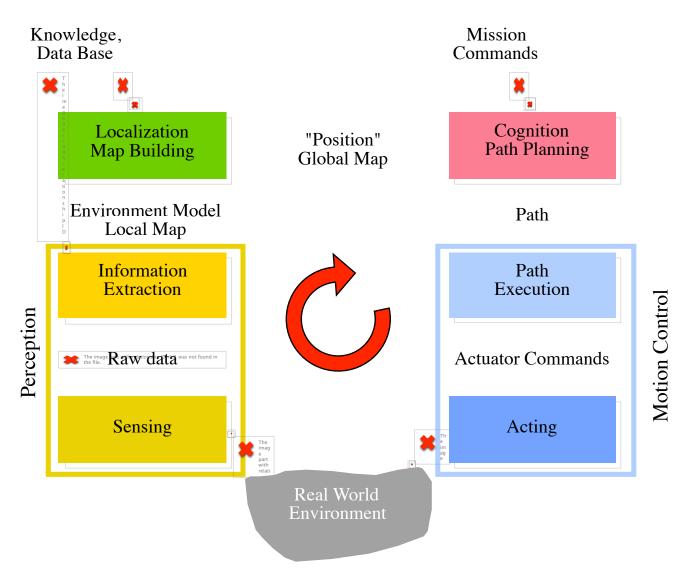
- Size
 - > length about 25 cm
- Sensors
 - > color camera
 - > stereo microphone







General Control Scheme for Mobile Robot Systems

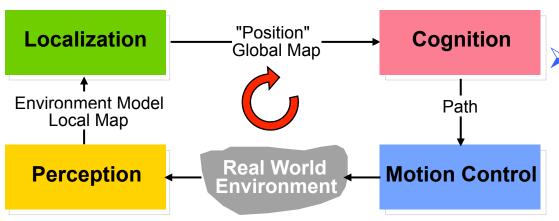


Control Architectures / Strategies

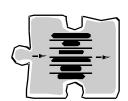
- Control Loop
 - dynamically changing
 - > no compact model available
 - many sources of uncertainty

- Two Approaches
 - Classical AI
 - o complete modeling
 - function based
 - o horizontal decomposition



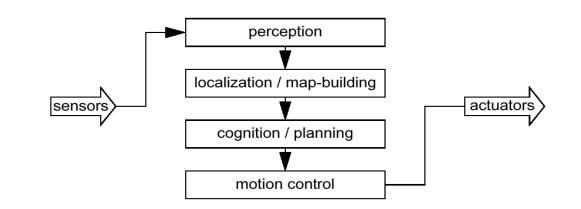


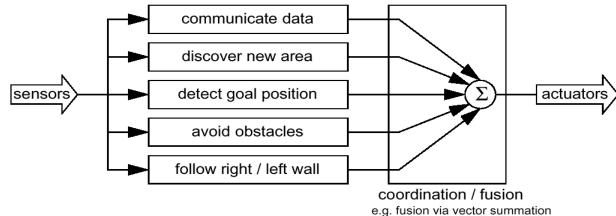
- New AI, AL
 - sparse or no modeling
 - behavior based
 - o vertical decomposition
 - o bottom up

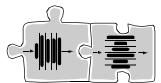


Two Approaches

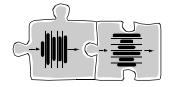
- Classical AI (model based navigation)
 - complete modeling
 - > function based
 - horizontal decomposition
- New AI, AL (behavior based navigation)
 - > sparse or no modeling
 - behavior based
 - vertical decomposition
 - bottom up
- Possible Solution
 - Combine Approaches

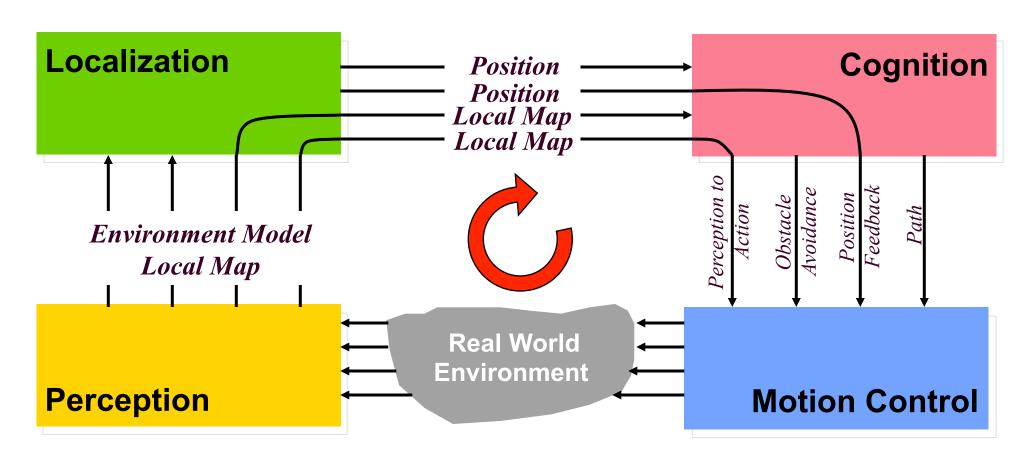






Mixed Approach Depicted into the General Control Scheme





Environment Representation and Modeling:

The Key for Autonomous Navigation

Environment Representation

 \triangleright Continuous Metric $-> x, y, \theta$

Discrete Metric -> metric grid

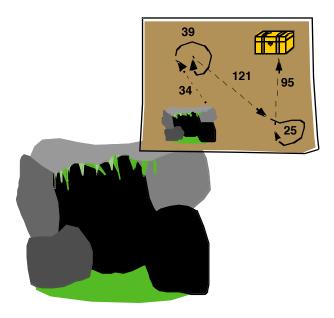
Discrete Topological -> topological grid

Environment Modeling

- Raw sensor data, e.g. laser range data, grayscale images
 - o large volume of data, low distinctiveness
 - o makes use of all acquired information
- Low level features, e.g. line other geometric features
 - o medium volume of data, average distinctiveness
 - o filters out the useful information, still ambiguities
- > High level features, e.g. doors, a car, the Eiffel tower
 - o low volume of data, high distinctiveness
 - o filters out the useful information, few/no ambiguities, not enough information

Environment Representation and Modeling: How we do it!

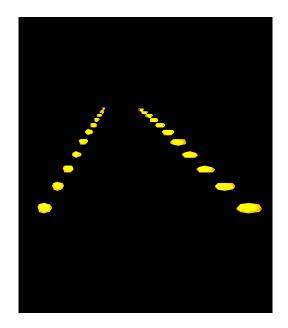
Odometry



How to find a treasure

> not applicable

Modified Environments



Landing at night

expensive, inflexible

Feature-based Navigation



Corridor crossing

Elevator door





Entrance

Eiffel Tower

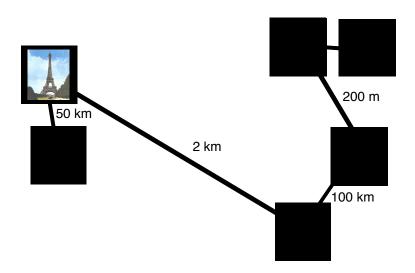
> still a challenge for artificial systems

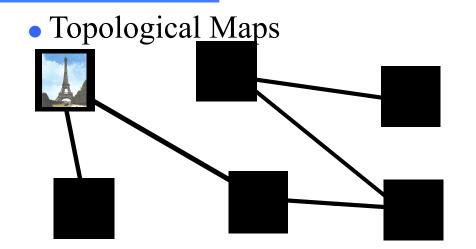
Environment Representation: The Map Categories

Recognizable Locations

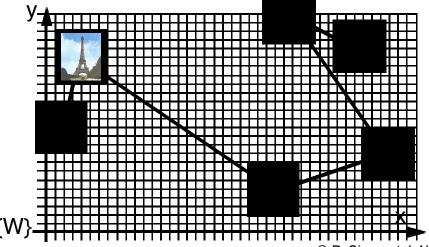


Metric Topological Maps





• Fully Metric Maps (continuous or discrete)



© R. Siegwart, I. Nourbakhsh

Environment Models: Continuous <-> Discrete; Raw data <-> Features

Continuos

 \triangleright position in x,y,θ

Discrete

- > metric grid
- > topological grid

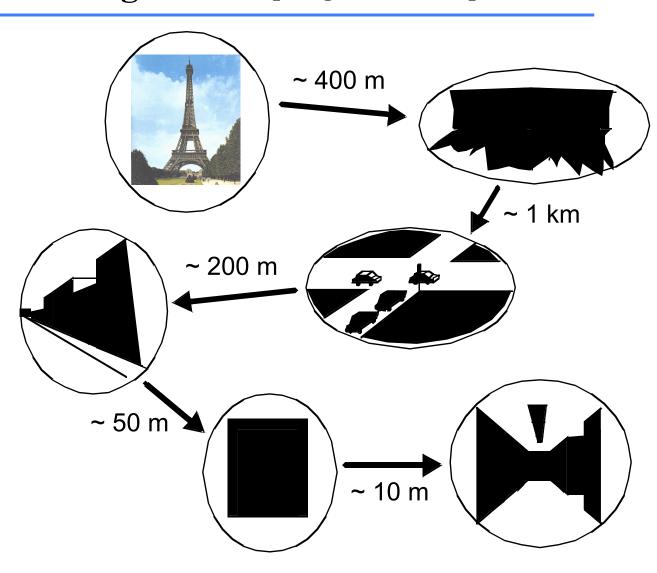
Raw Data

- > as perceived by sensor
- A feature (or natural landmark) is an environmental structure which is static, always perceptible with the current sensor system and locally unique.

Examples

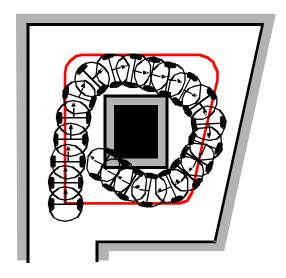
- geometric elements (lines, walls, column ..)
- > a railway station
- a river
- the Eiffel Tower
- a human being
- fixed stars
- skyscraper

Human Navigation: Topological with imprecise metric information



Methods for Navigation: Approaches with Limitations

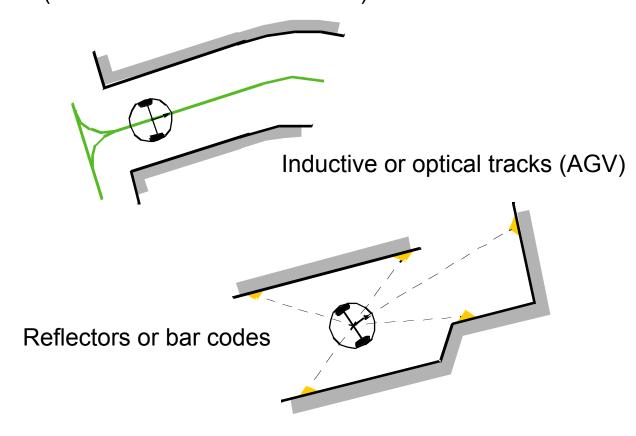
 Incrementally (dead reckoning)



Odometric or initial sensors (gyro)

not applicable

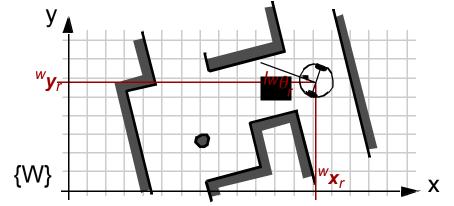
 Modifying the environments (artificial landmarks / beacons)



> expensive, inflexible

Methods for Localization: The Quantitative Metric Approach

1. A priori Map: Graph, metric



2. Feature Extraction (e.g. line segments)

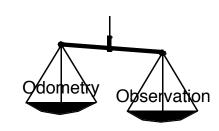
5
4
3
2
E
1
-1
-2
-3
-4
-5
-5 -4 -3 -2 -1 0 1 2 3 4 -5
-5 -4 -3 -2 -1 0 x[m]

3. Matching:Find correspondence

of features

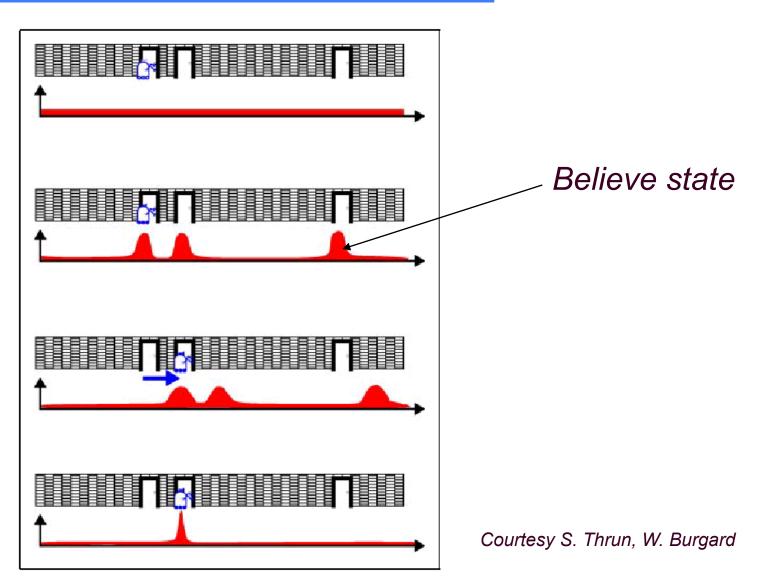
4. Position Estimation:

e.g. Kalman filter, Markov



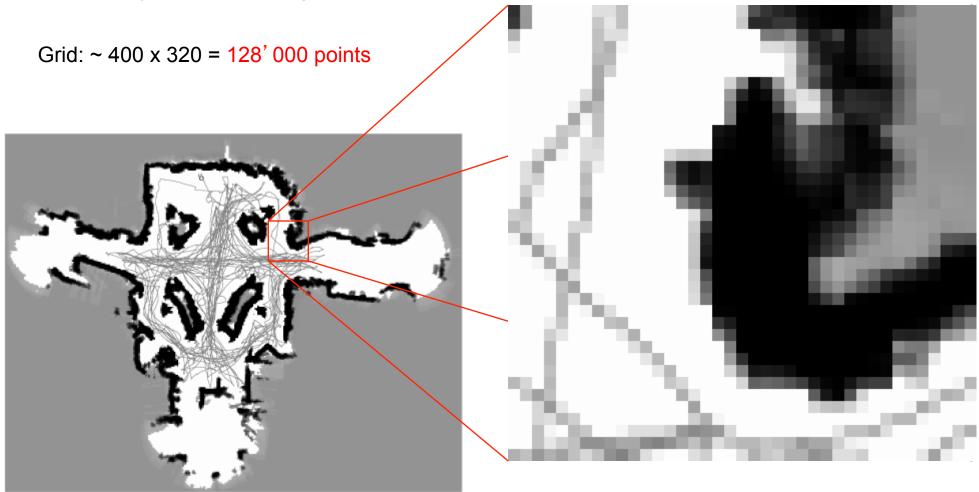
- representation of uncertainties
- optimal weighting acc. to a priori statistics

Gaining Information through motion: (Multi-hypotheses tracking)

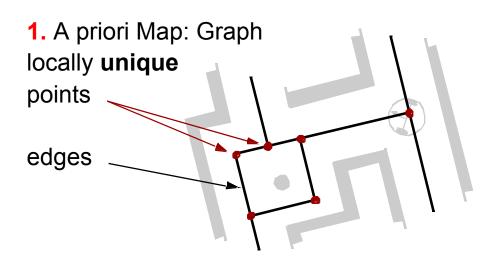


Grid-Based Metric Approach

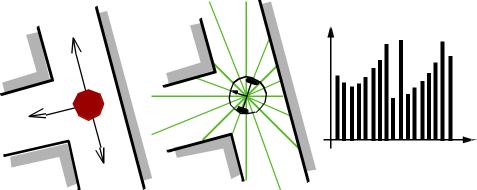
 Grid Map of the Smithsonian's National Museum of American History in Washington DC. (Courtesy of Wolfram Burger et al.)



Methods for Localization: The Quantitative Topological Approach



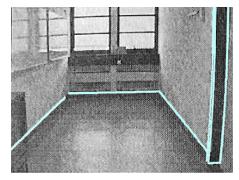
2. Method for determining the local uniqueness

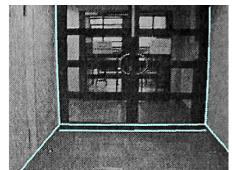


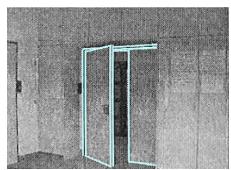
e.g. striking changes on raw data level or highly distinctive features

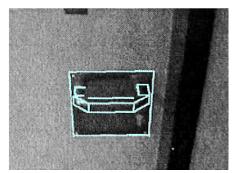
- 3. Library of driving behaviors
- e.g. wall or midline following, blind step, enter door, application specific behaviors

Example: Video-based navigation with natural landmarks







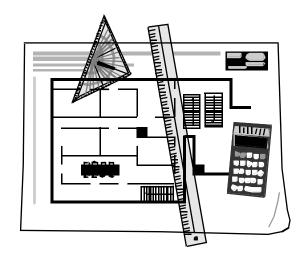


Courtesy of [Lanser et al. 1996]

Courtesy K. Arras

Map Building: How to Establish a Map

1. By Hand



2. Automatically: Map Building

The robot **learns** its environment

Motivation:

- by hand: hard and costly
- dynamically changing environment
- different look due to different perception

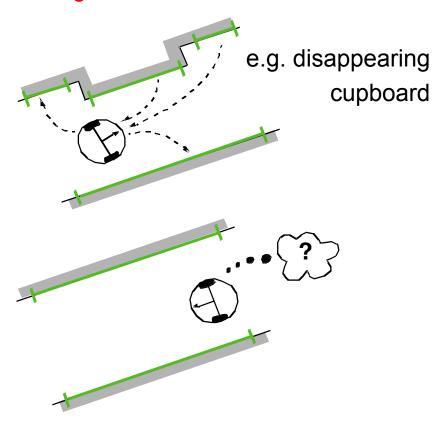
3. Basic Requirements of a Map:

- a way to incorporate newly sensed information into the existing world model
- information and procedures for estimating the robot's position
- information to do path planning and other navigation task (e.g. obstacle avoidance) predictability
- Measure of Quality of a map
 - topological correctness
 - metrical correctness
- But: Most environments are a mixture of predictable and unpredictable features
 - → hybrid approach

model-based vs. behaviour-based

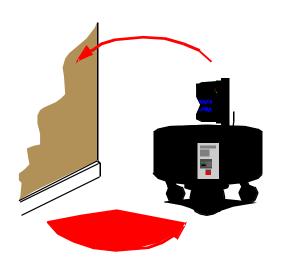
Map Building: The Problems

 Map Maintaining: Keeping track of changes in the environment



 e.g. measure of belief of each environment feature 2. Representation and Reduction of Uncertainty

position of robot -> position of wall

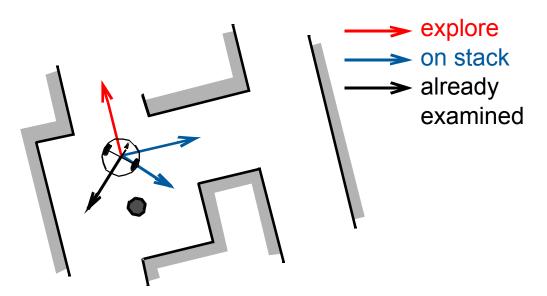


position of wall -> position of robot

- probability densities for feature positions
- additional exploration strategies

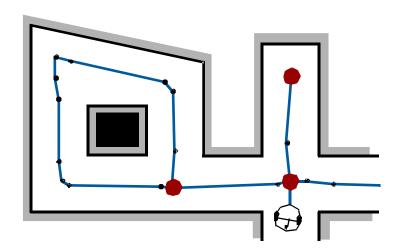
Map Building: Exploration and Graph Construction

1. Exploration



- provides correct topology
- must recognize already visited location
- backtracking for unexplored openings

2. Graph Construction



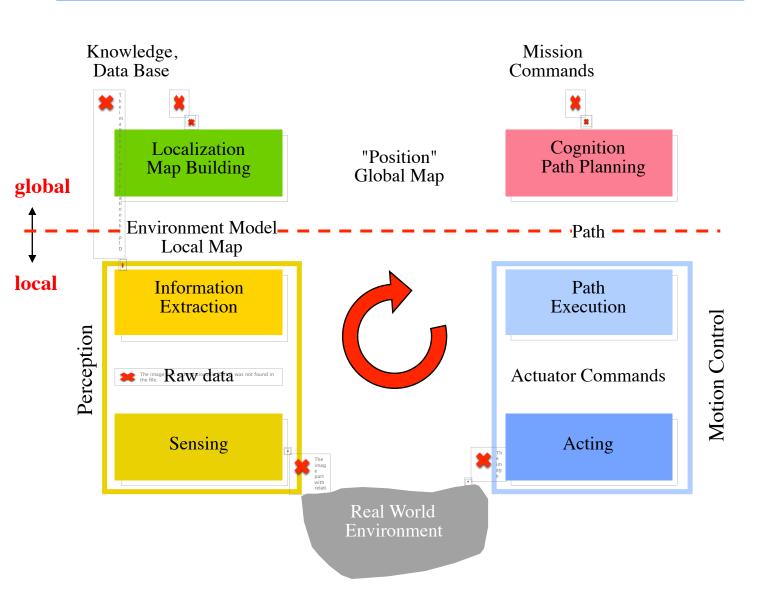
Where to put the nodes?

Topology-based: at distinctive locations



Metric-based: where features disappear or get visible

Control of Mobile Robots

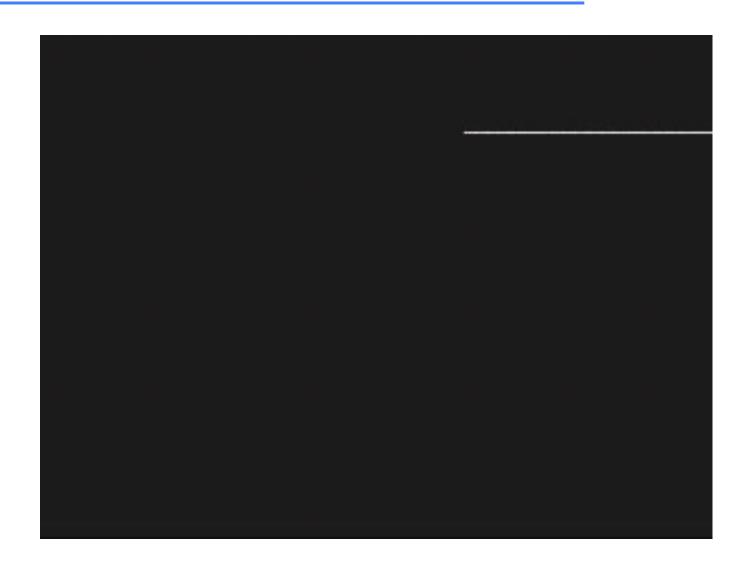


- Most functions for save navigation are 'local' not involving localization nor cognition
- Localization and global path planning
 slower update rate, only when

needed

This approach is pretty similar to what human beings do.

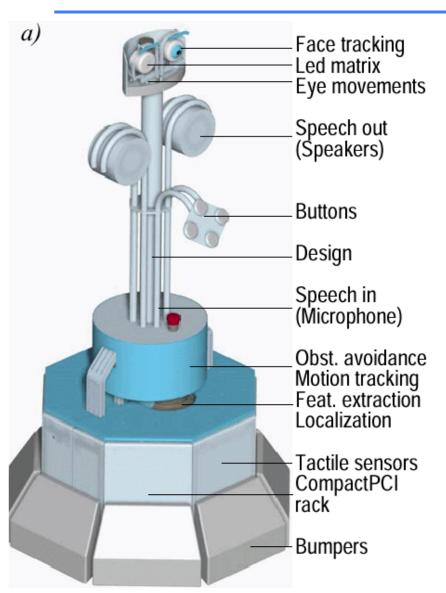
Tour-Guide Robot (Nourbakhsh, CMU)



Autonomous Indoor Navigation (Thrun, CMU)



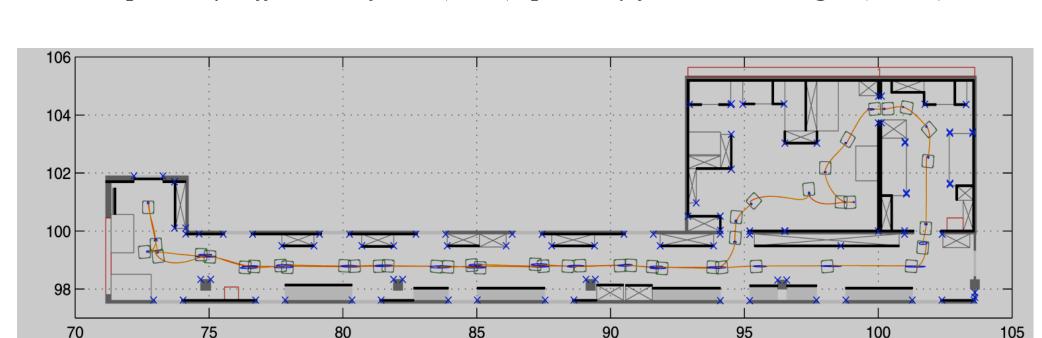
Tour-Guide Robot (EPFL @ expo.02)





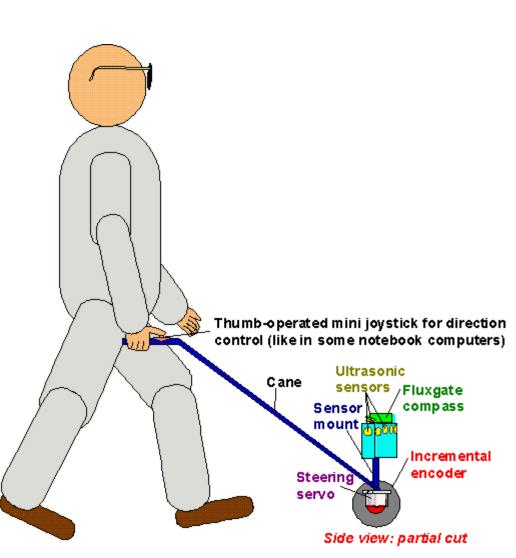
Autonomous Indoor Navigation (Pygmalion EPFL)

- > very robust on-the-fly localization
- > one of the first systems with probabilistic sensor fusion
- > 47 steps, 78 meter length, realistic office environment,
- > conducted 16 times > 1km overall distance
- > partially difficult surfaces (laser), partially few vertical edges (vision)



GuideCane, University of Michigan

http://www.engin.umich.edu/research/mrl/





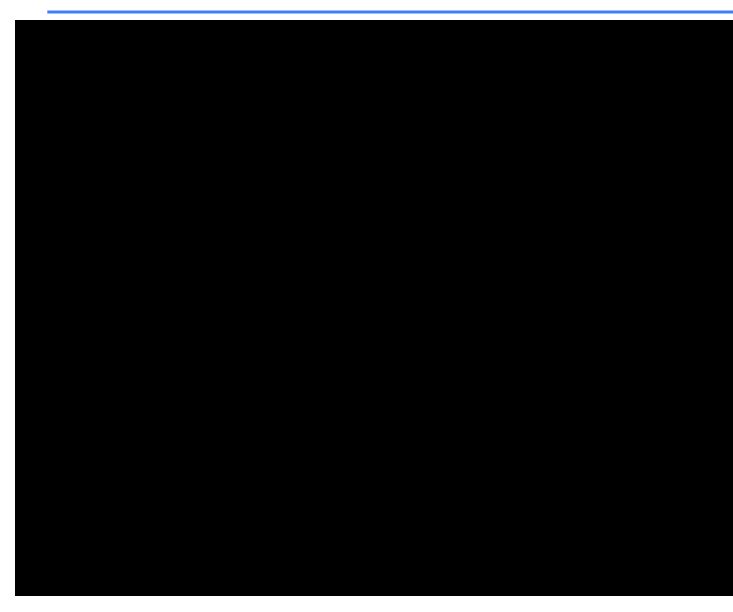
LaserPlans Architectural Tool

(ActivMedia Robotics)





Morpha Project, Germany



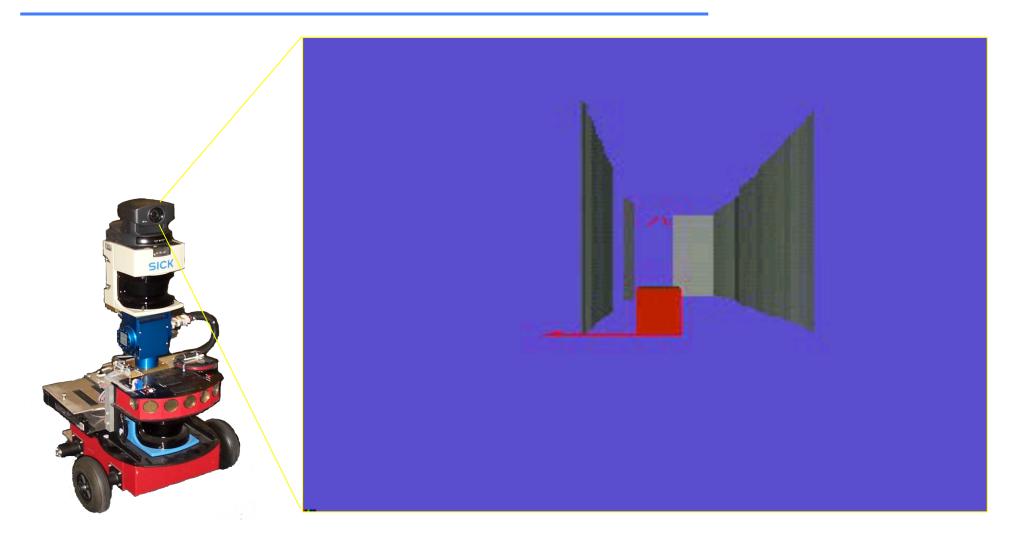
Courtesy of Erwin Prassler

Autonomous Indoor Mapping



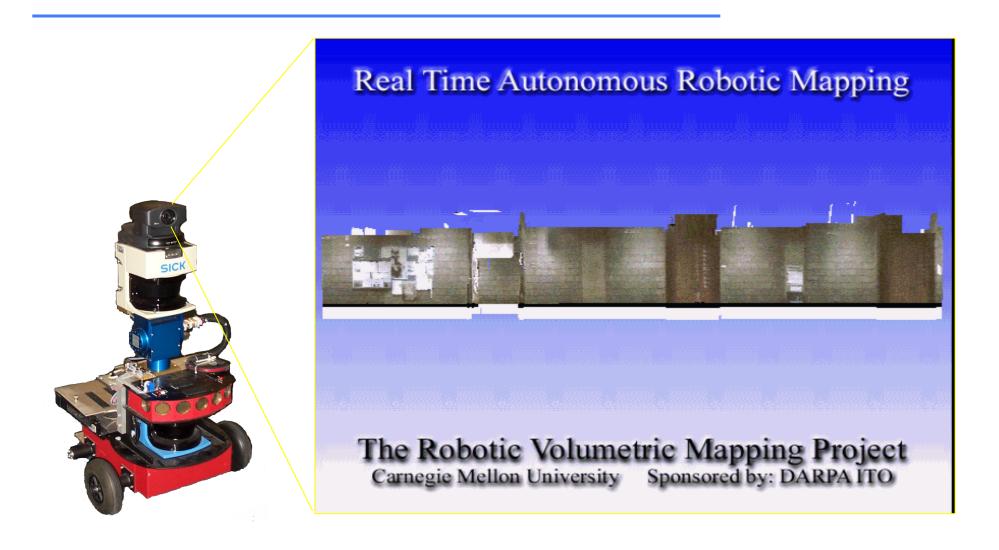
Courtesy of Sebastian Thrun

High-Speed Explotation and Mapping



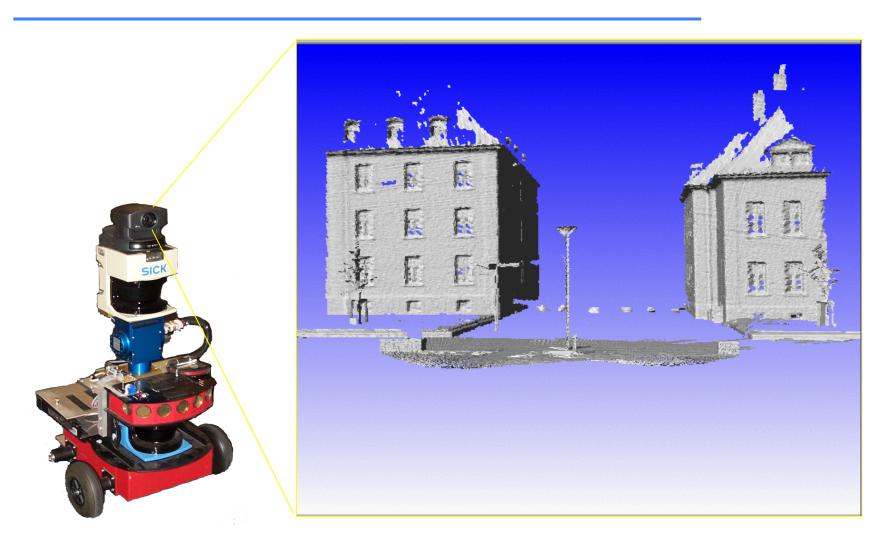
Courtesy of Sebastian Thrun

Turning Real Reality into Virtual Reality



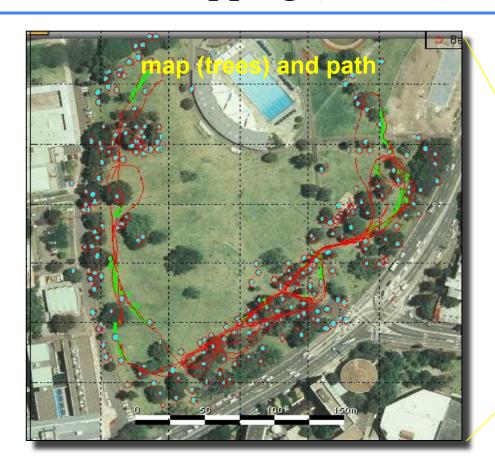
Courtesy of Sebastian Thrun

Urban Reconnaissance



Courtesy of Sebastian Thrun

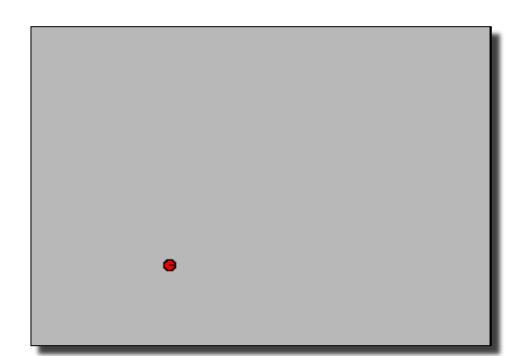
Outdoor Mapping (no GPS)



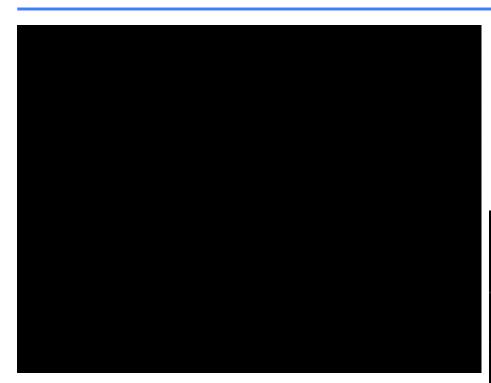


Real-Time Multi Robot Exploration

Multi-Robot Mapping and Exploration Carnegie Mellon October 1999

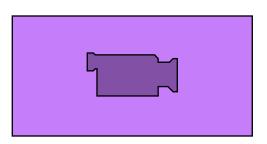


All Terrain Locomotion (Shrimp EPFL)





Human-Robot Interaction (Kismet MIT)



Cye's Navigation Concept

