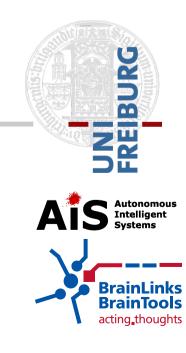
Probabilistic Techniques for Robot Navigation

A Key Technology for Our Future Society

Wolfram Burgard



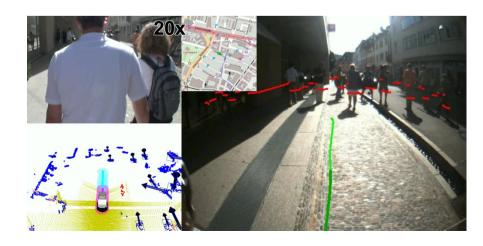
Key Challenges in Navigation

Highly accurate localization Robust mapping Long-term autonomy

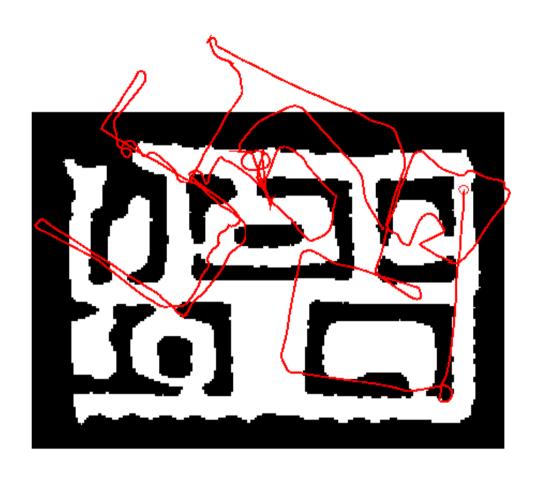


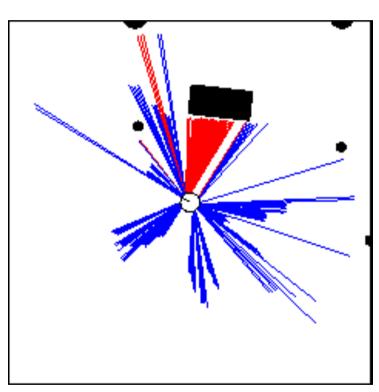






Nature of Data





Odometry Data

Range Data

Probabilistic Robotics

Explicit representation and utilization of uncertainty

Perception = state estimation

• Action = utility optimization

Probabilistic Robotics

Explicit representation and utilization of uncertainty

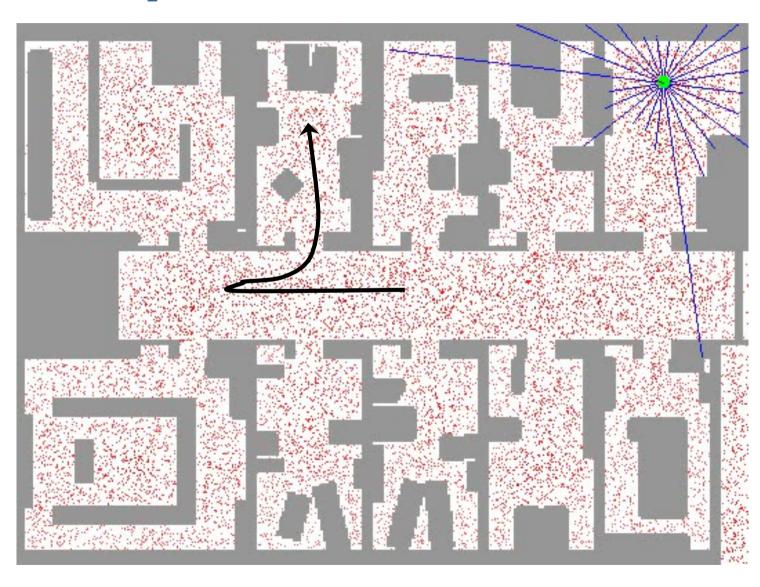
Perception = state estimation

$$Bel(x \mid z, u) = \alpha p(z \mid x) \int_{x'} p(x \mid u, x') Bel(x') dx'$$

Action = utility optimization

$$\pi^*(x) = \underset{u}{\operatorname{argmax}} \sum_{x'} p(x' \mid u, x) V^*(x')$$

MCL: Global Localization (Sonar)



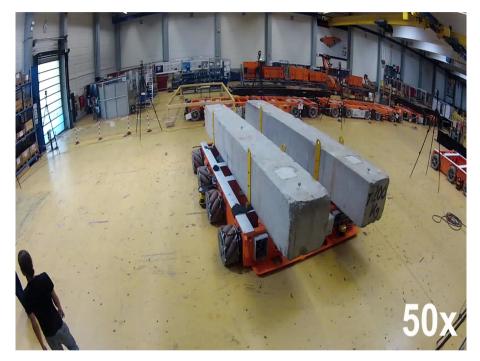
Precise Localization and Positioning for Mobile Robots

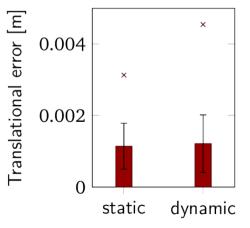


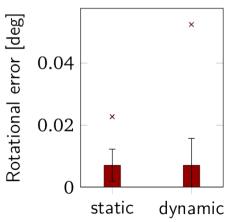


Accurate Localization

- KUKA omniMove (11t)
- Safety scanners
- Error in the area of millimeters
- Even in dynamic environments









26 Units installed at Boeing

- Fuselage assembly
- 20 vehicles to transport industrial robots for drilling and filling of 60,000 fasteners in
- 6 vehicles for logistics of parts, work stands and fuselages





Accurate Indoor RGB-D Localization with a Google Tango Device based on 2D Floor Plans

Wera Winterhalter, Freya Fleckenstein, Bastian Steder, Wolfram Burgard, Luciano Spinello



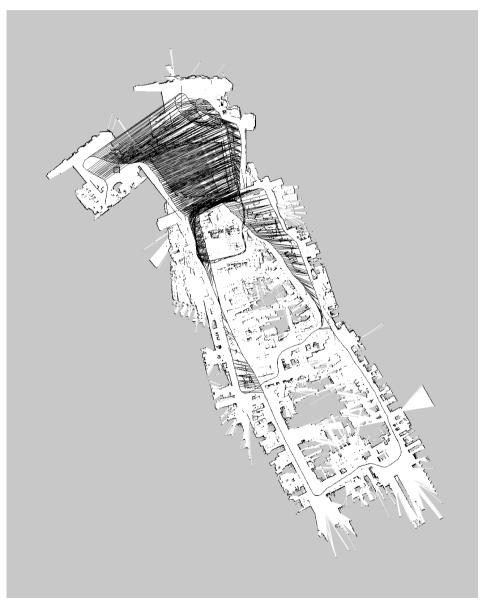




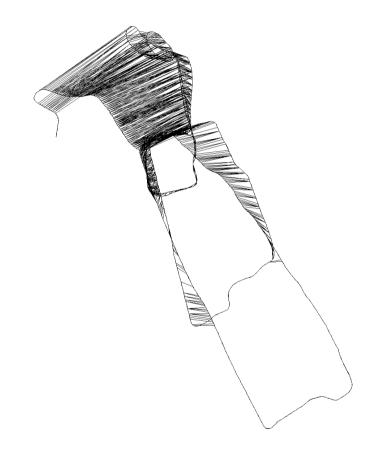
How to Learn a Map: SLAM



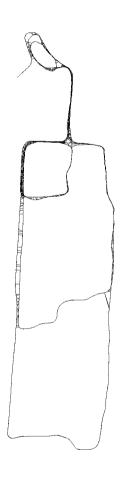
- Problem described as a graph
 - Every node corresponds to a robot position and to a laser measurement
 - An edge between two nodes represents a datadependent spatial constraint between the nodes



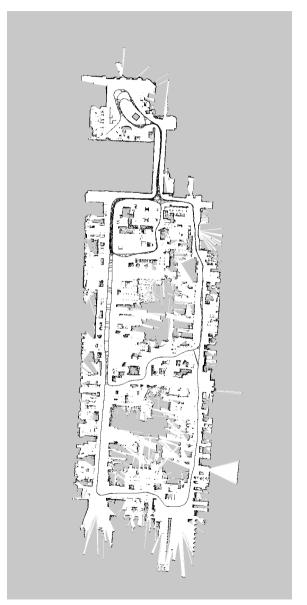
 Once we have the graph, we determine the most likely map by "moving" the nodes



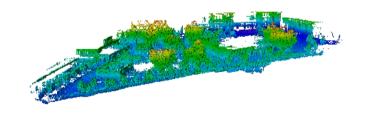
- Once we have the graph, we determine the most likely map by "moving" the nodes
- ... like this.



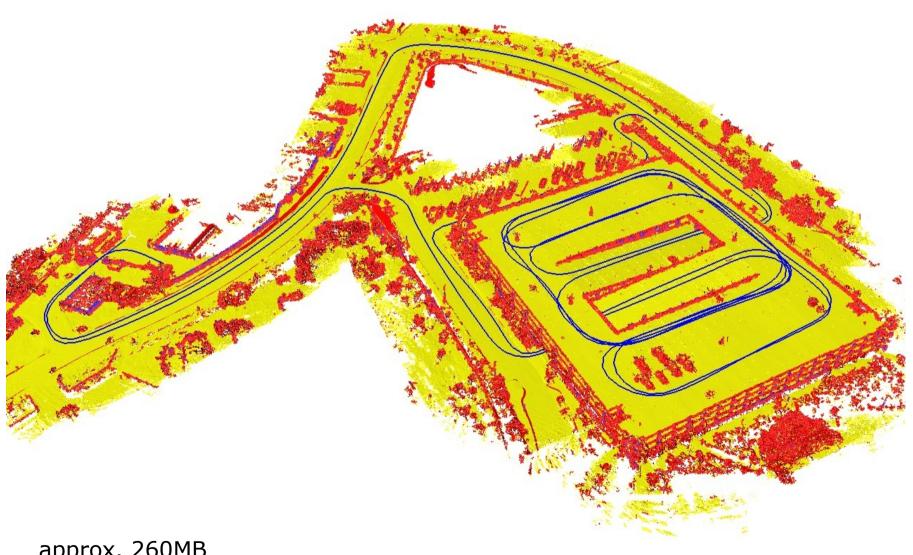
- Once we have the graph, we determine the most likely map by "moving" the nodes
- ... like this.
- Then we render a map based on the known poses



Freiburg Campus Octomap



3D Map of the Stanford Parking Garage



Autonomous Parking



Autonomous Navigation in Urban Areas

Can we build a robot that is able to navigate autonomously through city centers?





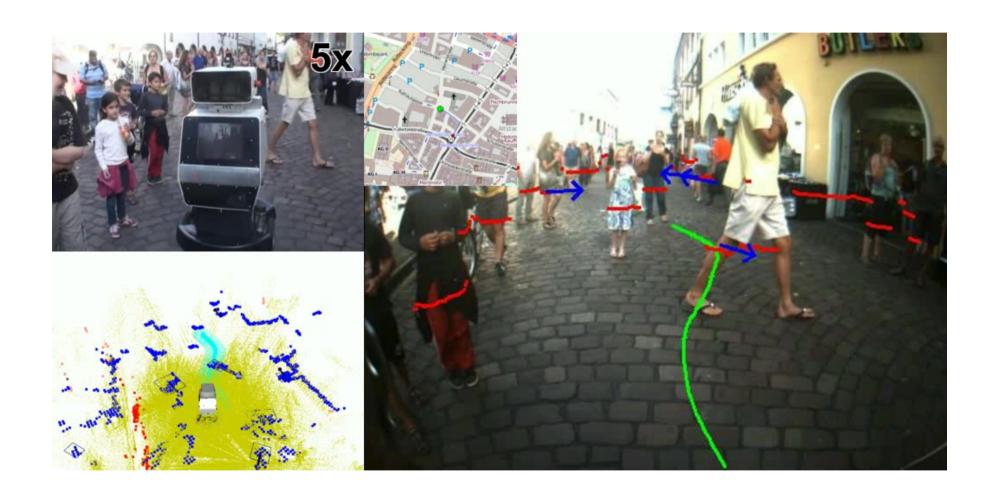




Challenge: Canals



Challenge: Kids



The Tagesthemen-Report



Learning Driving Styles

 Users have different expectations

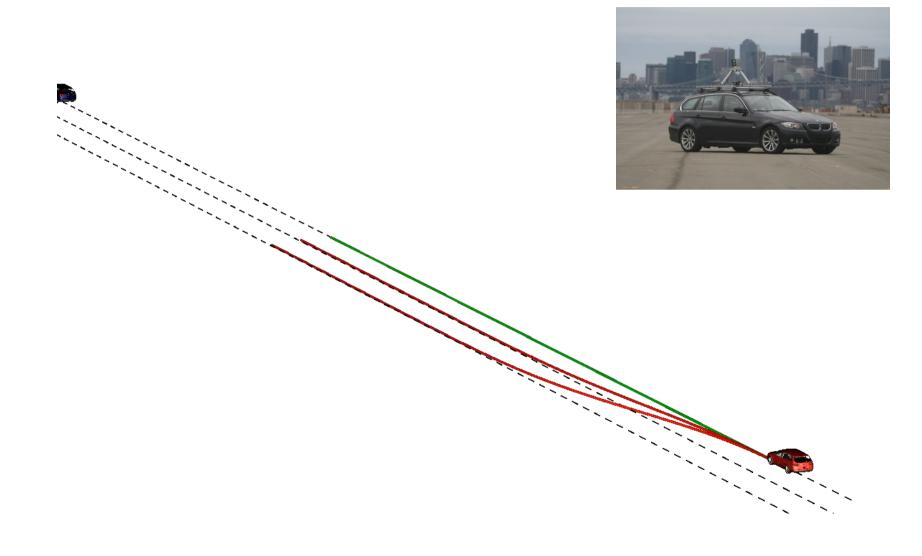


 Many parameters such as accelerations, distances, velocities, etc.

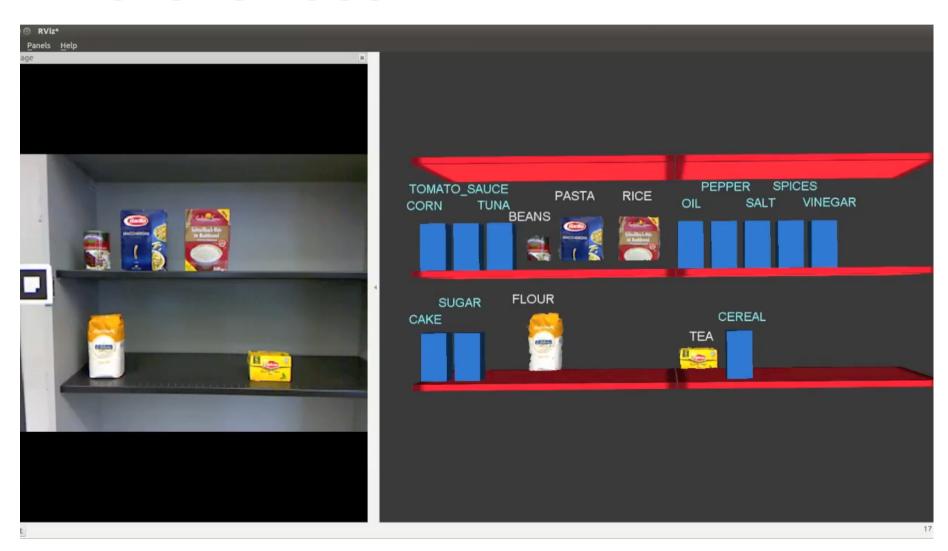


 Difficult to tune parameters for non-technical users

Learning Driving Behavior

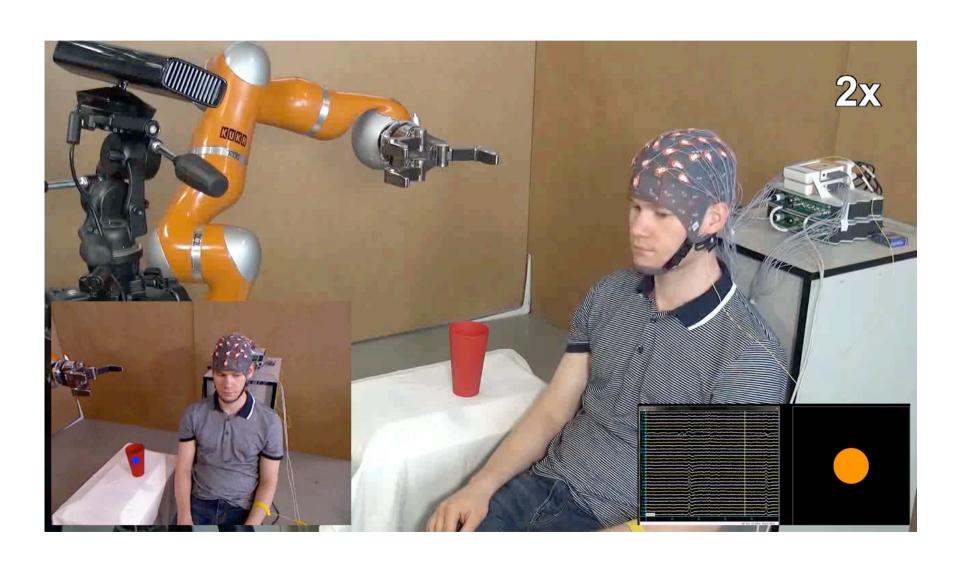


Online Prediction of User Preferences

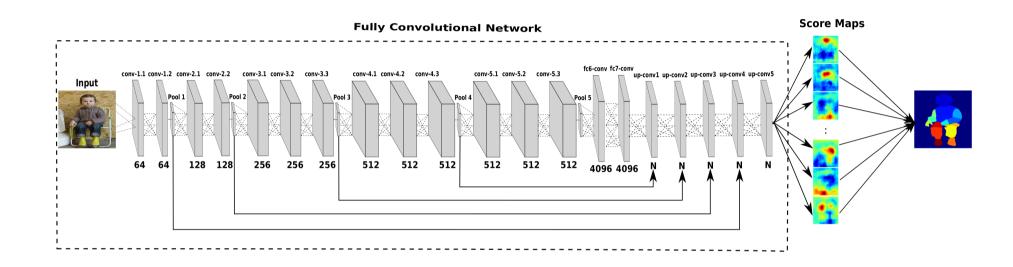


Neurobots





And what's next?



Deep Learning

Applications in Robotics

RGB-D



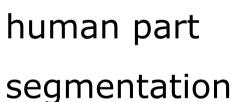
object recognition

Images

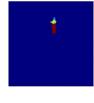






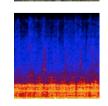




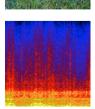


Sound









terrain classification

Terrain Classification using a Late Fusion DCNN Architecture

Snow

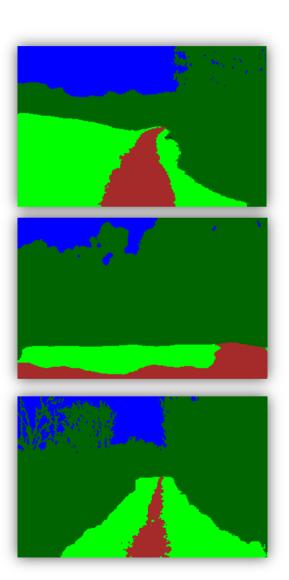


Glare



Low Lighting

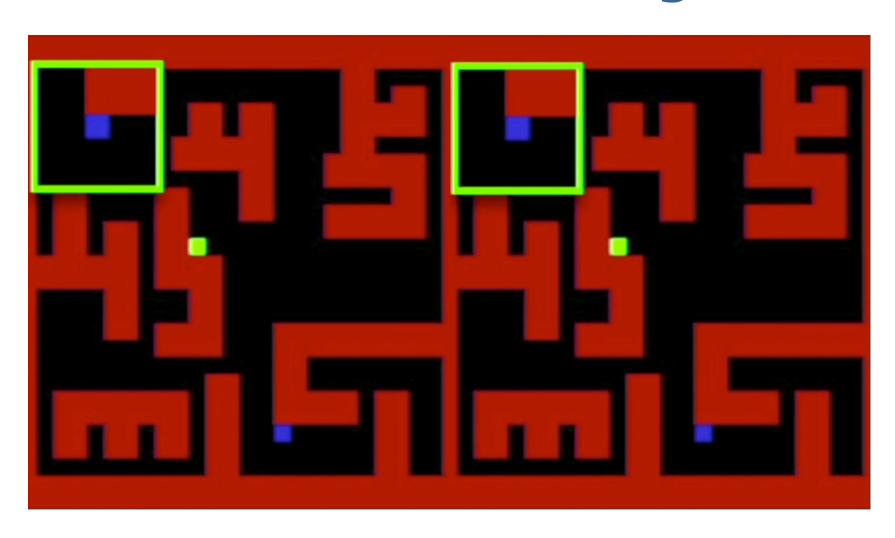




Autonomous Navigation in Outdoor Areas



... and End to End Navigation



Outlook

 With deep learning a new massively parallel and data intensive paradigm has come up that outperforms classical approaches

 Deep learning will play a major role in robotics through the concept of end-to-end learning where no programming is needed.

Summary

 Probabilistic methods are a powerful tool for realizing autonomous systems

 The corresponding state estimation procedures provide the means for robust navigation systems