

Introduction to Mobile Robotics

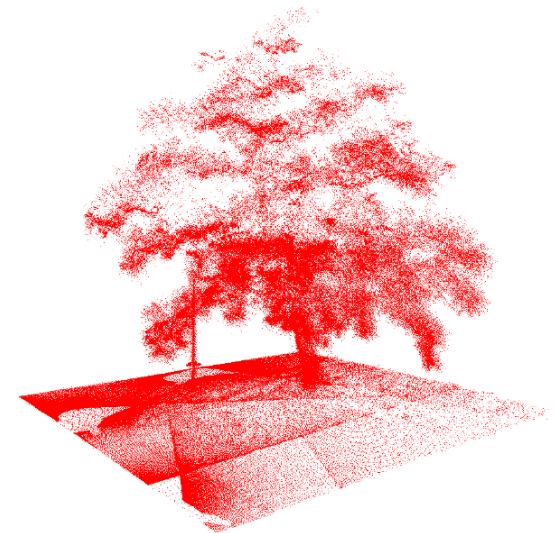
Techniques for 3D Mapping

Wolfram Burgard, Cyrill Stachniss, Maren
Bennewitz, Diego Tipaldi, Luciano Spinello



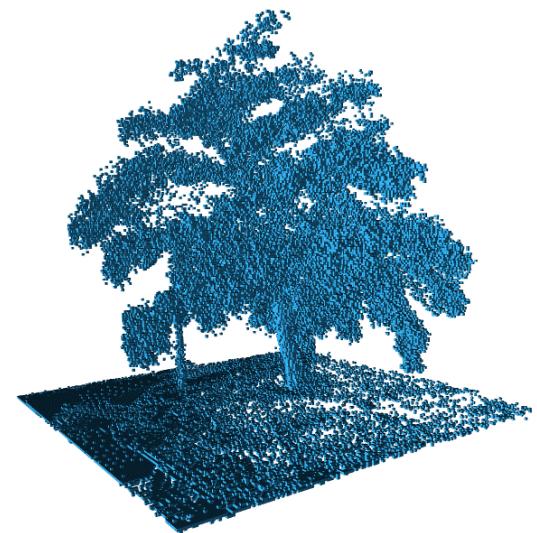
Point Clouds

- **Pro:**
 - No discretization of data
 - Mapped area not limited
- **Contra:**
 - Unbounded memory usage
 - No direct representation of free or unknown space



3D Voxel Grids

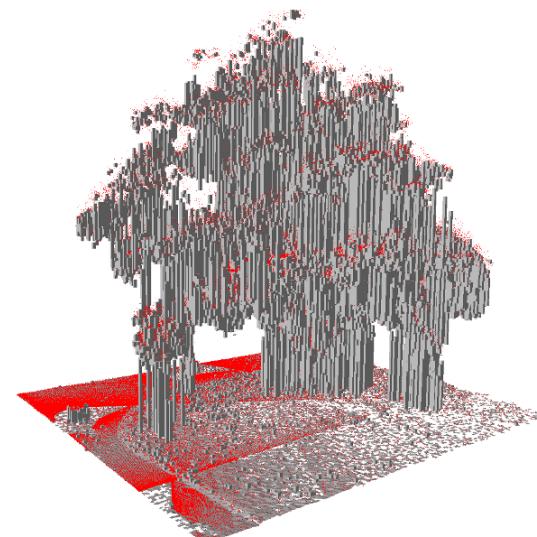
- **Pro:**
 - Volumetric representation
 - Constant access time
 - Probabilistic update
- **Contra:**
 - Memory requirement: Complete map is allocated in memory
 - Extent of the map has to be known/guessed



2.5D Maps: “Height Maps”

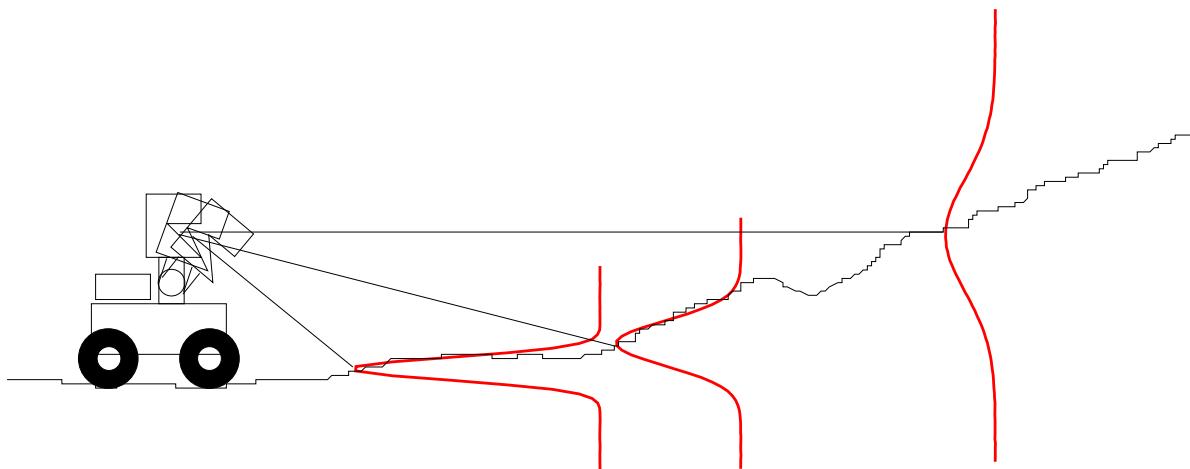
Average over all scan points that fall into a cell

- **Pro:**
 - Memory efficient
 - Constant time access
- **Contra:**
 - Non-probabilistic
 - No distinction between free and unknown space



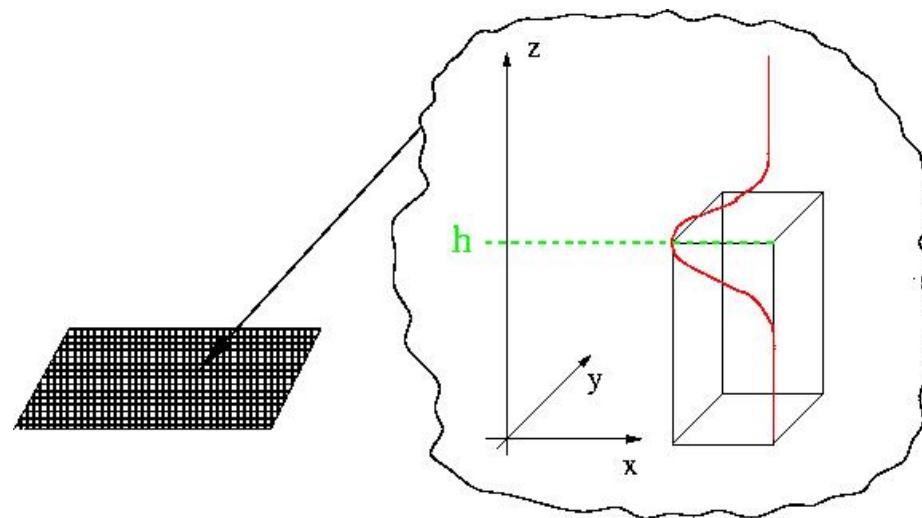
Elevation Maps

- 2D grid that stores an estimated height (elevation) for each cell
- Note: Uncertainty increases with measured distance



Elevation Maps

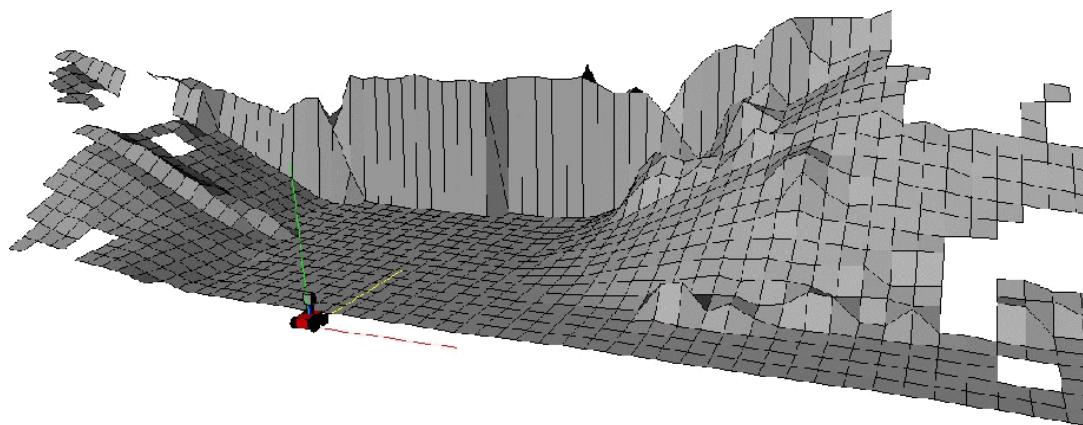
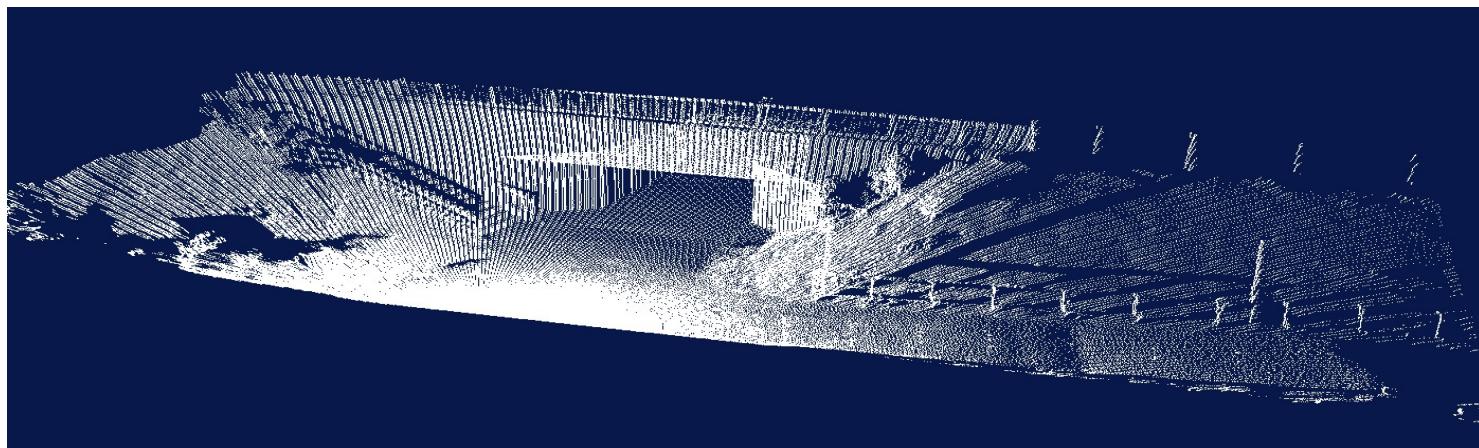
- 2D grid that stores an estimated height (elevation) for each cell
- Note: Uncertainty increases with measured distance
- Kalman update to estimate the elevation



Elevation Maps

- **Pro:**
 - 2.5D representation (vs. full 3D grid)
 - Constant time access
 - Probabilistic estimate about the height
- **Contra:**
 - No vertical objects
 - Only one level is represented

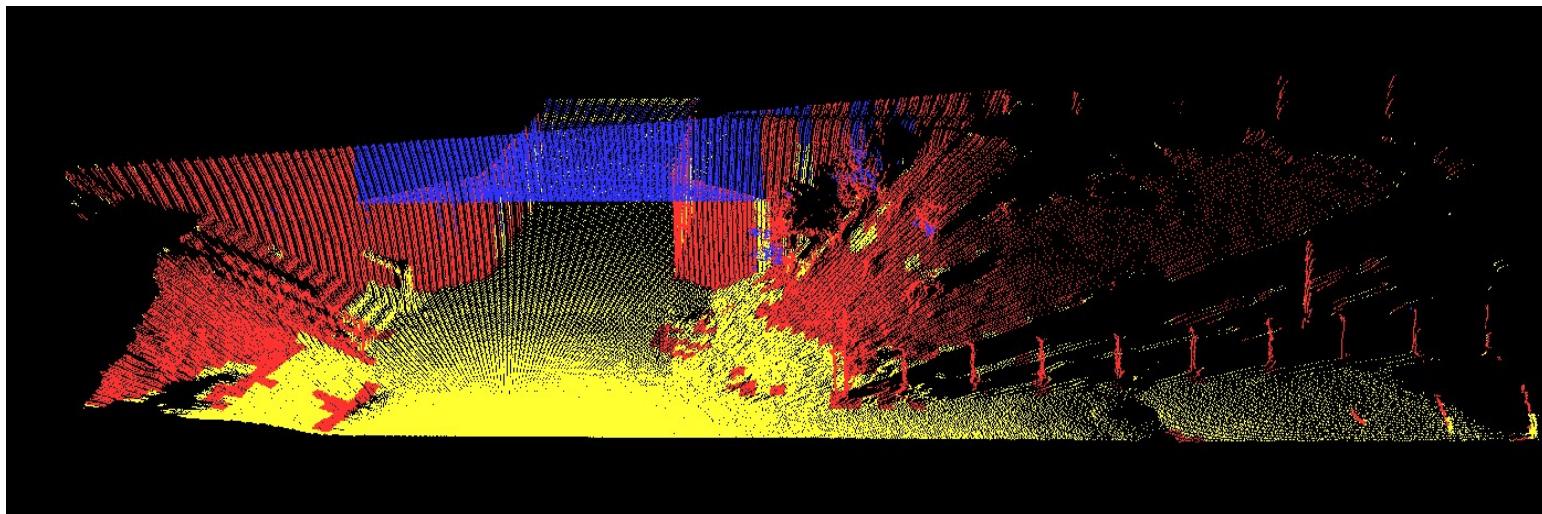
Typical Elevation Map



Extended Elevation Maps

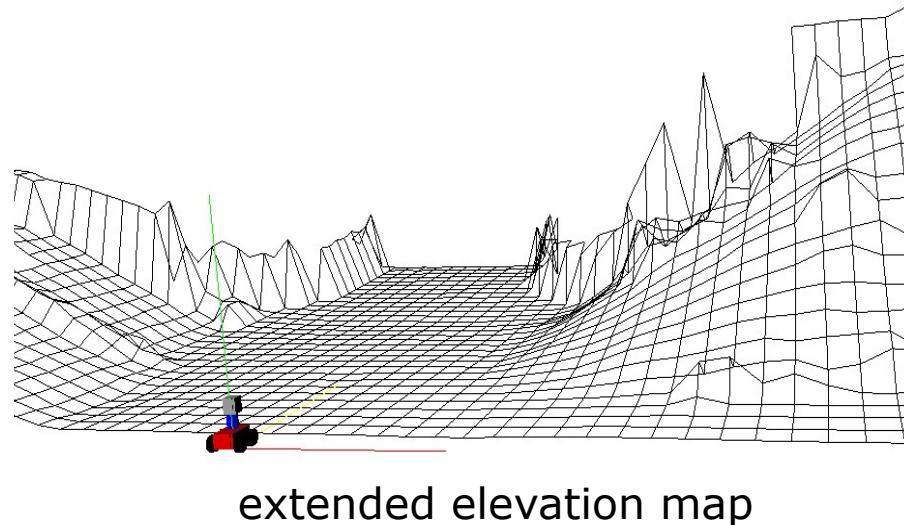
- Identify
 - Cells that correspond to vertical structures
 - Cells that contain gaps
- Check whether the variance of the height of all data points is large for a cell
- If so, check whether the corresponding point set contains a gap exceeding the height of the robot ("gap cell")

Example: Extended Elevation Map



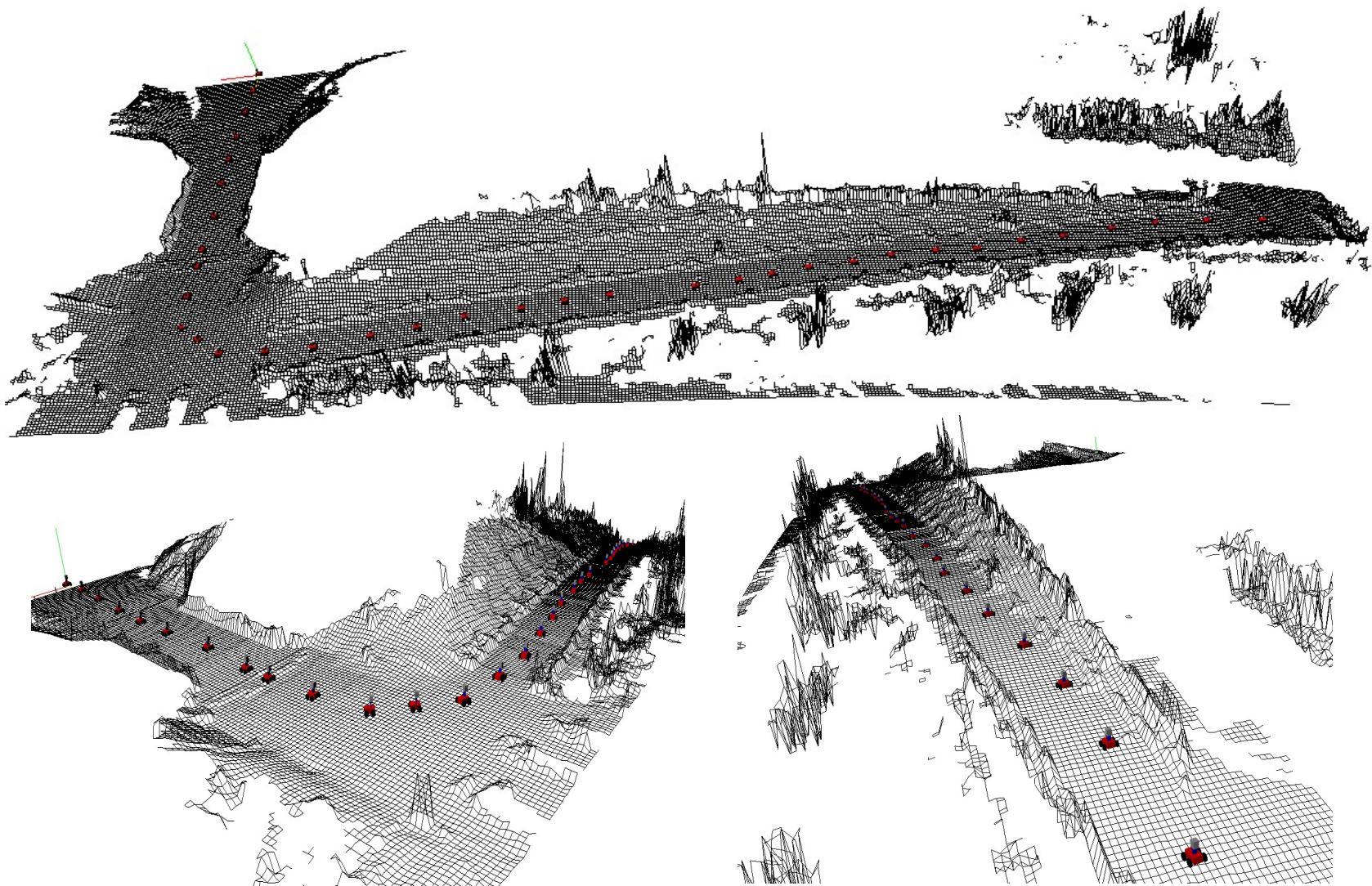
- Cells with vertical objects (**red**)
- Data points above a big vertical gap (**blue**)
- Cells seen from above (**yellow**)

→ use gap cells to determine traversability

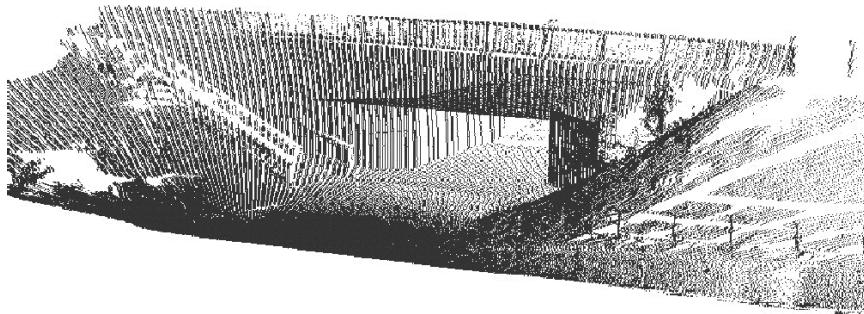


extended elevation map

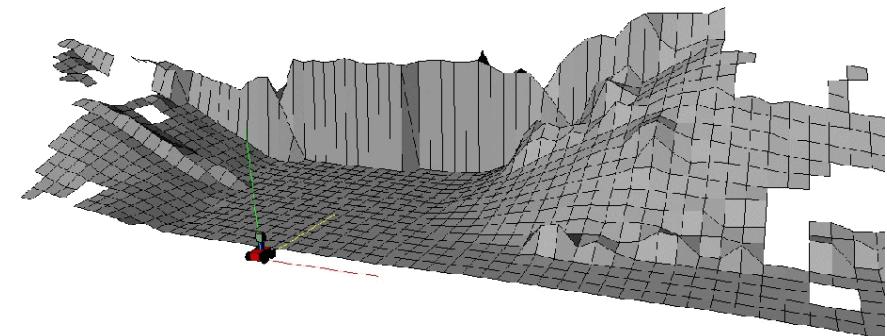
Example Elevation Maps



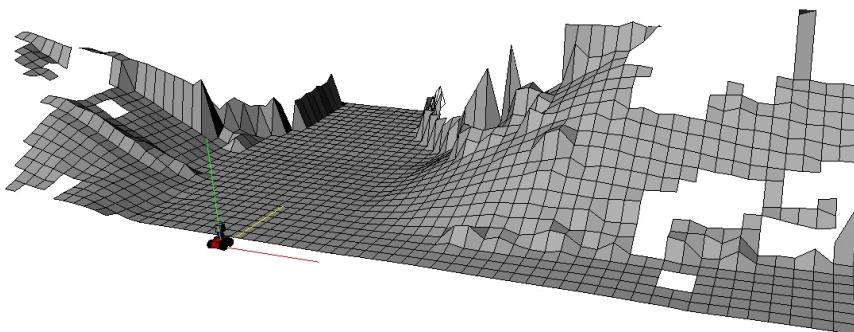
Types of Terrain Maps



Point cloud

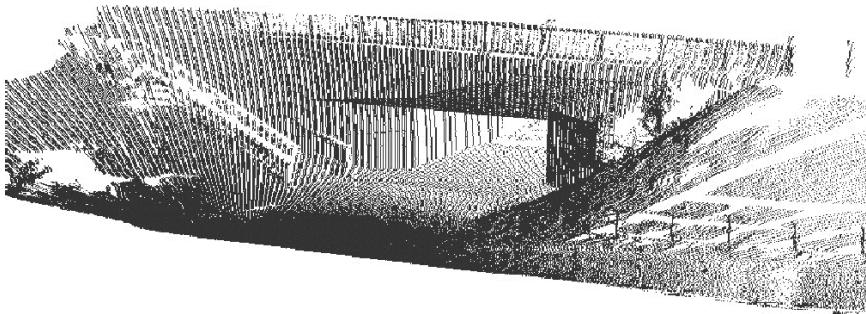


Standard elevation map

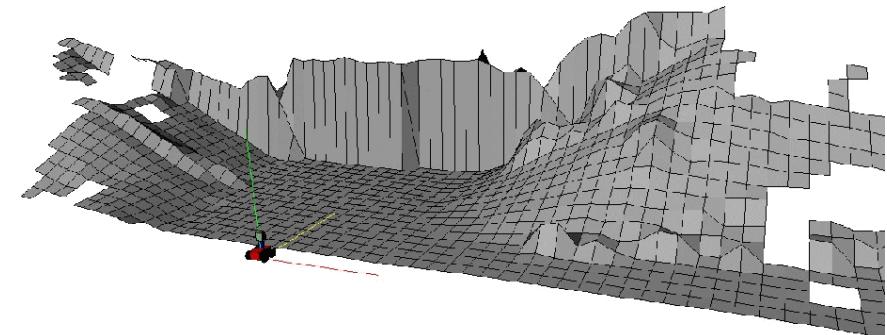


Extended elevation map

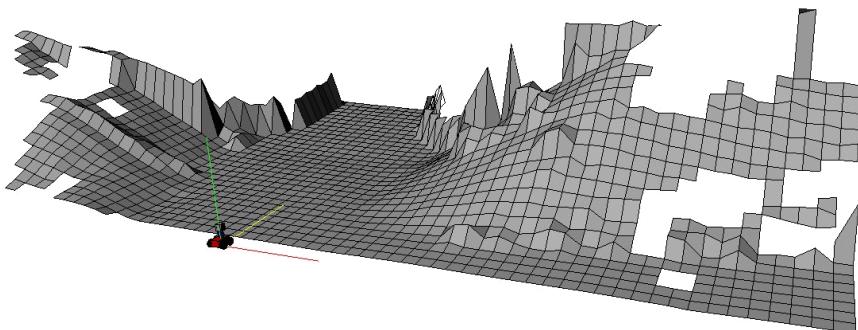
Types of Terrain Maps



Point cloud



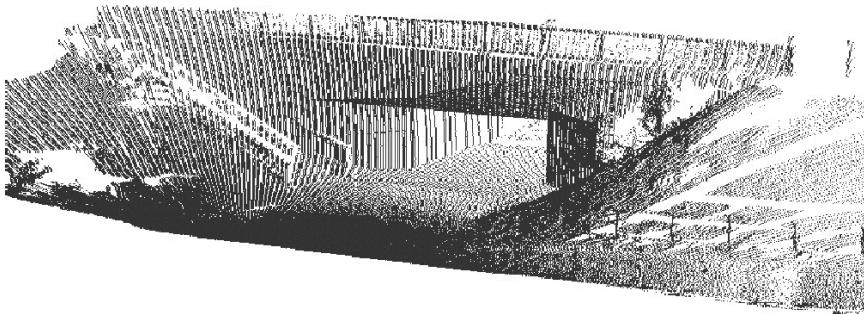
Standard elevation map



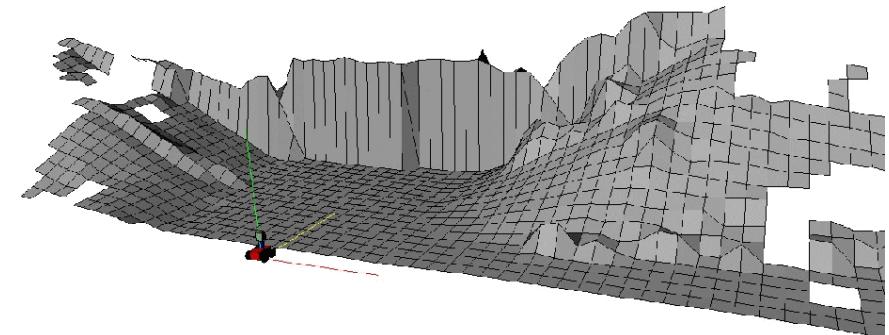
Extended elevation map

- + Planning with underpasses possible
(cells with vertical gaps)
- No paths passing under **and**
crossing over bridges possible
(only one level per grid cell)

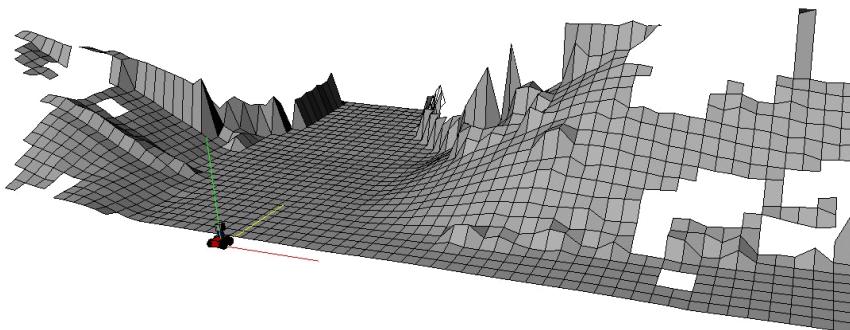
Types of Terrain Maps



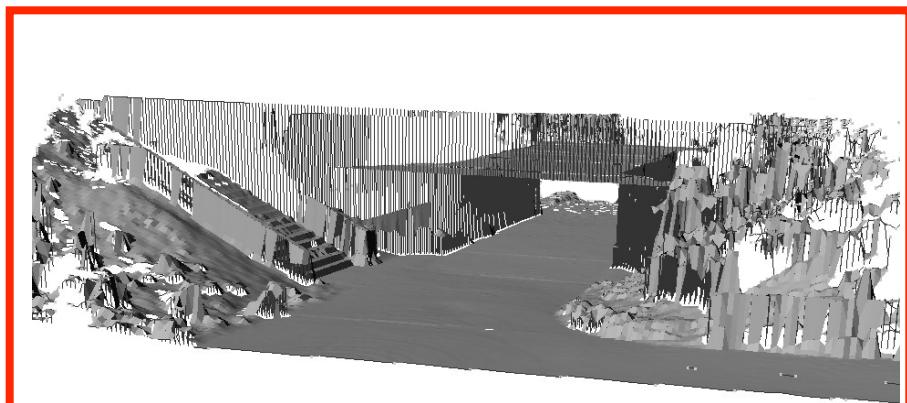
Point cloud



Standard elevation map

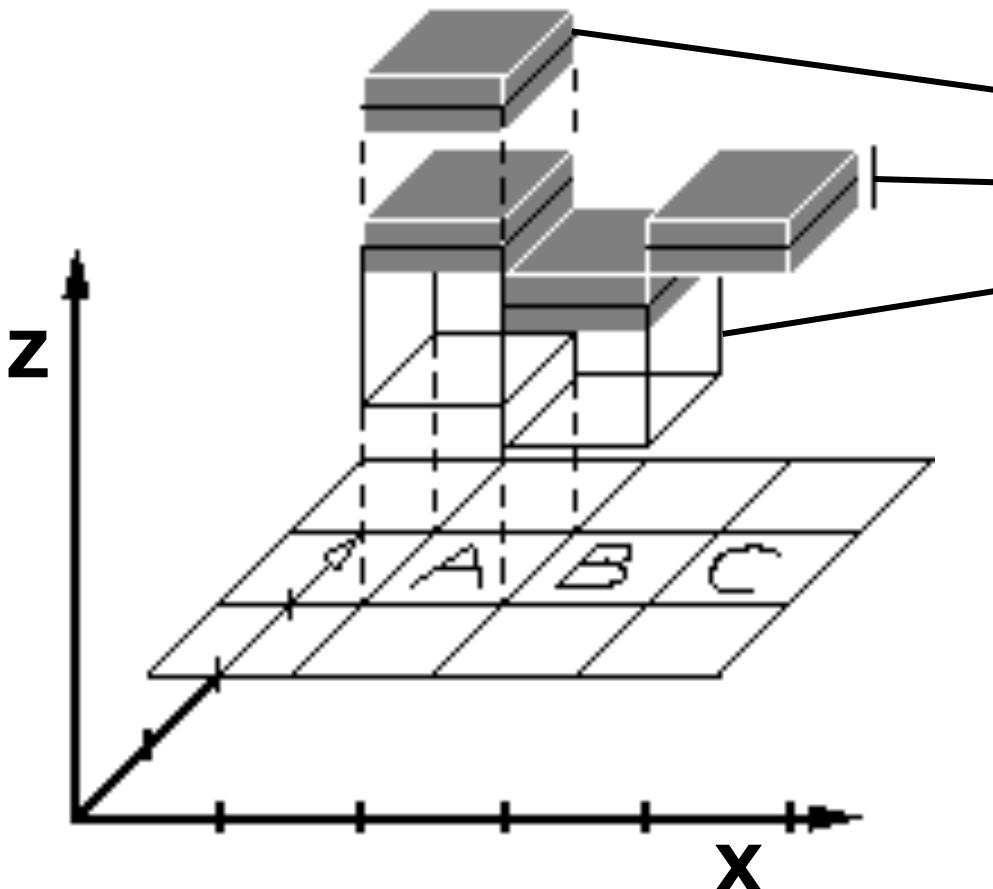


Extended elevation map



Multi-level surface map

MLS Map Representation



Each 2D cell stores various patches consisting of:

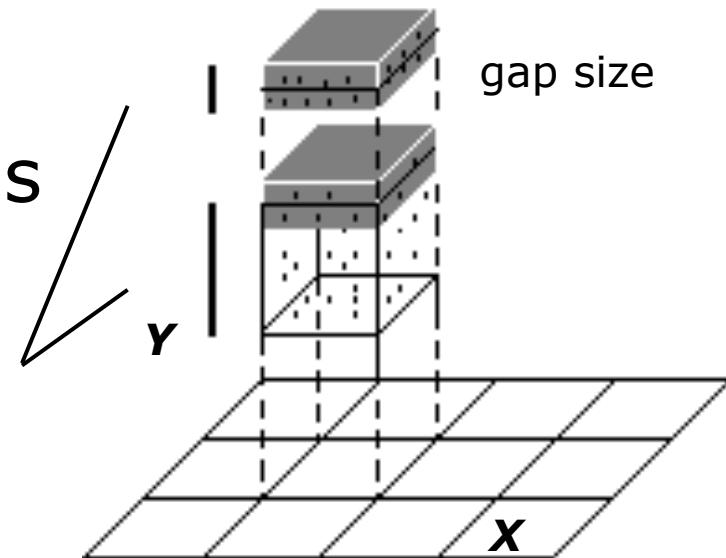
- The height mean μ
- The height variance σ
- The depth value d

Note:

- A patch can have no depth (flat objects, e.g., floor)
- A cell can have one or many patches (vertical gap cells, e.g., bridges)

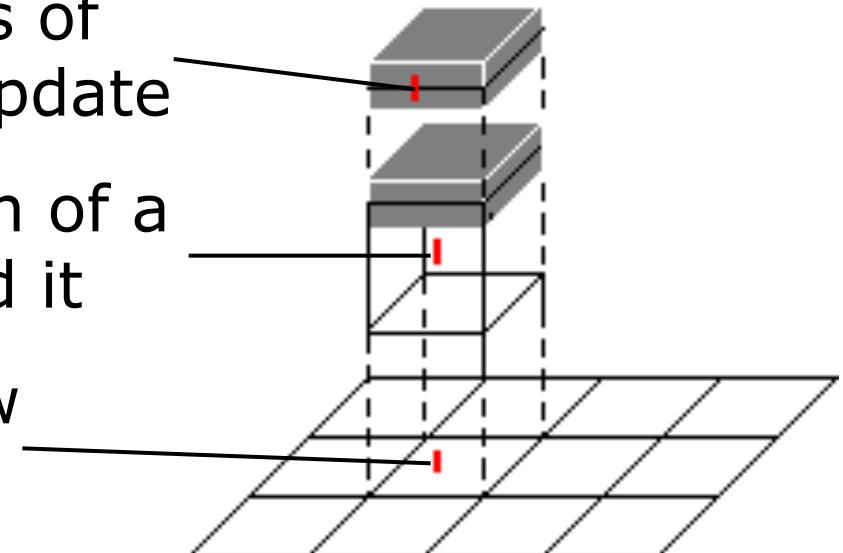
From Point Clouds to MLS Maps

- Determine the cell for each 3D point
- Compute vertical intervals
- Classify into vertical ($>10\text{cm}$) and horizontal intervals
- Apply Kalman update to estimate the height based on all data points for the horizontal intervals
- Take the mean and variance of the highest measurement for the vertical intervals

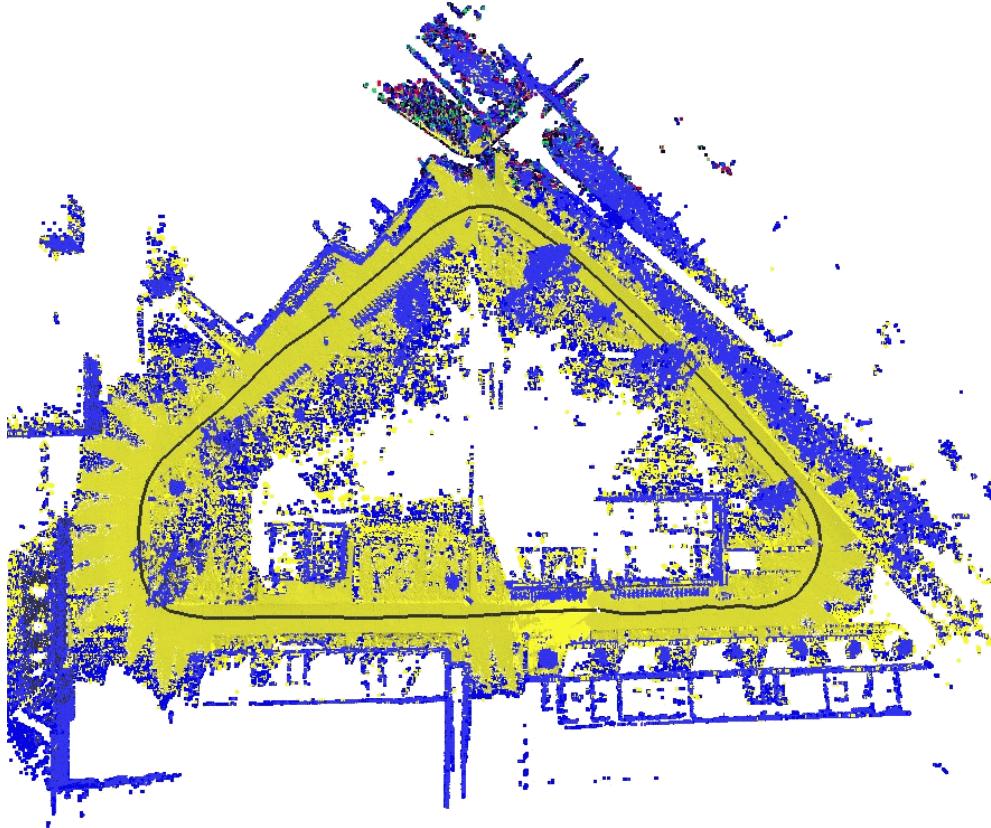
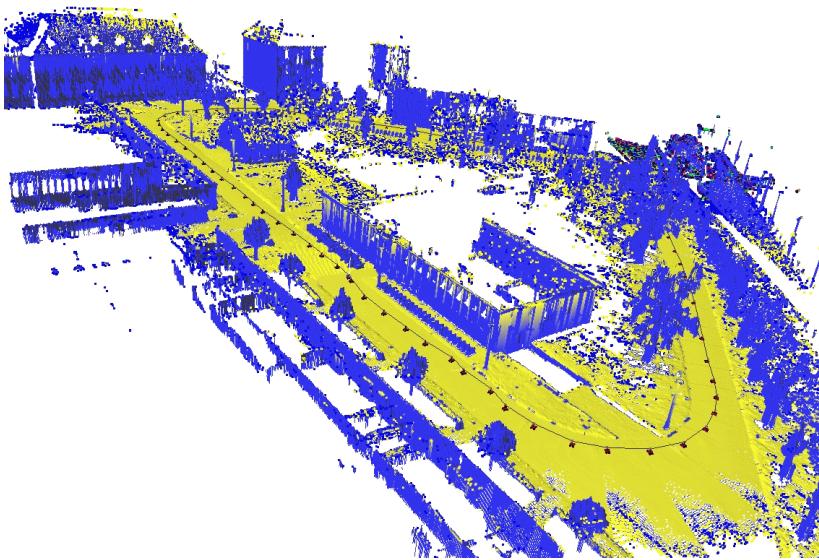


Map Update

- Given: new measurement $z=(\mathbf{p}, \sigma)$ with variance
- Determine the corresponding cell for z
- Find closest surface patch in the cell
- If z is inside 3 variances of the patch, do Kalman update
- If z is in occupied region of a surface patch, disregard it
- Otherwise, create a new surface patch from z

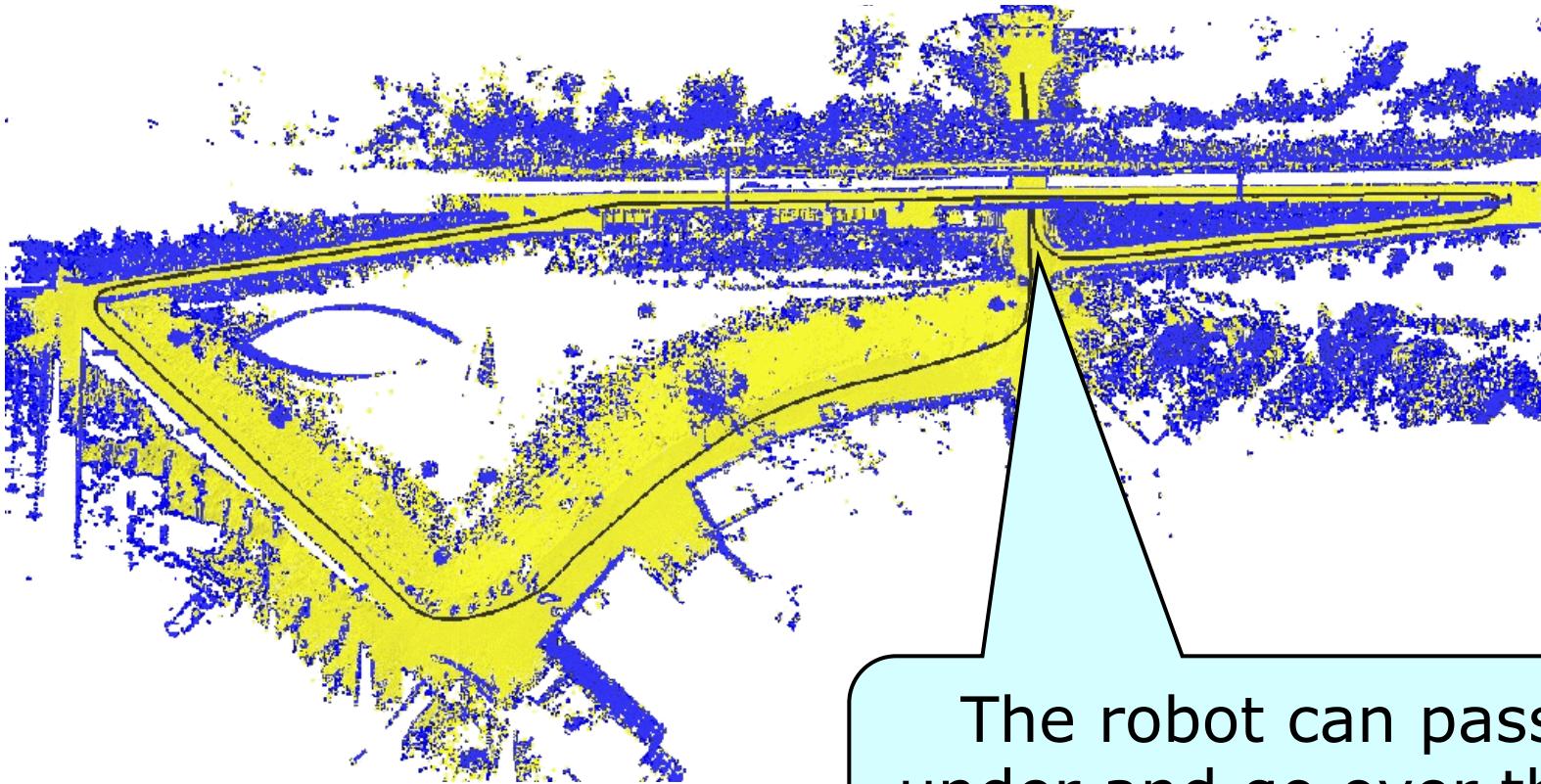


Results



- Map size: 195 by 146 m
- Cell resolution: 10 cm
- Number of data points: 20,207,000

Results



- Map size: 299 by 147 m
- Cell resolution: 10 cm
- Number of data points: 45,000,000

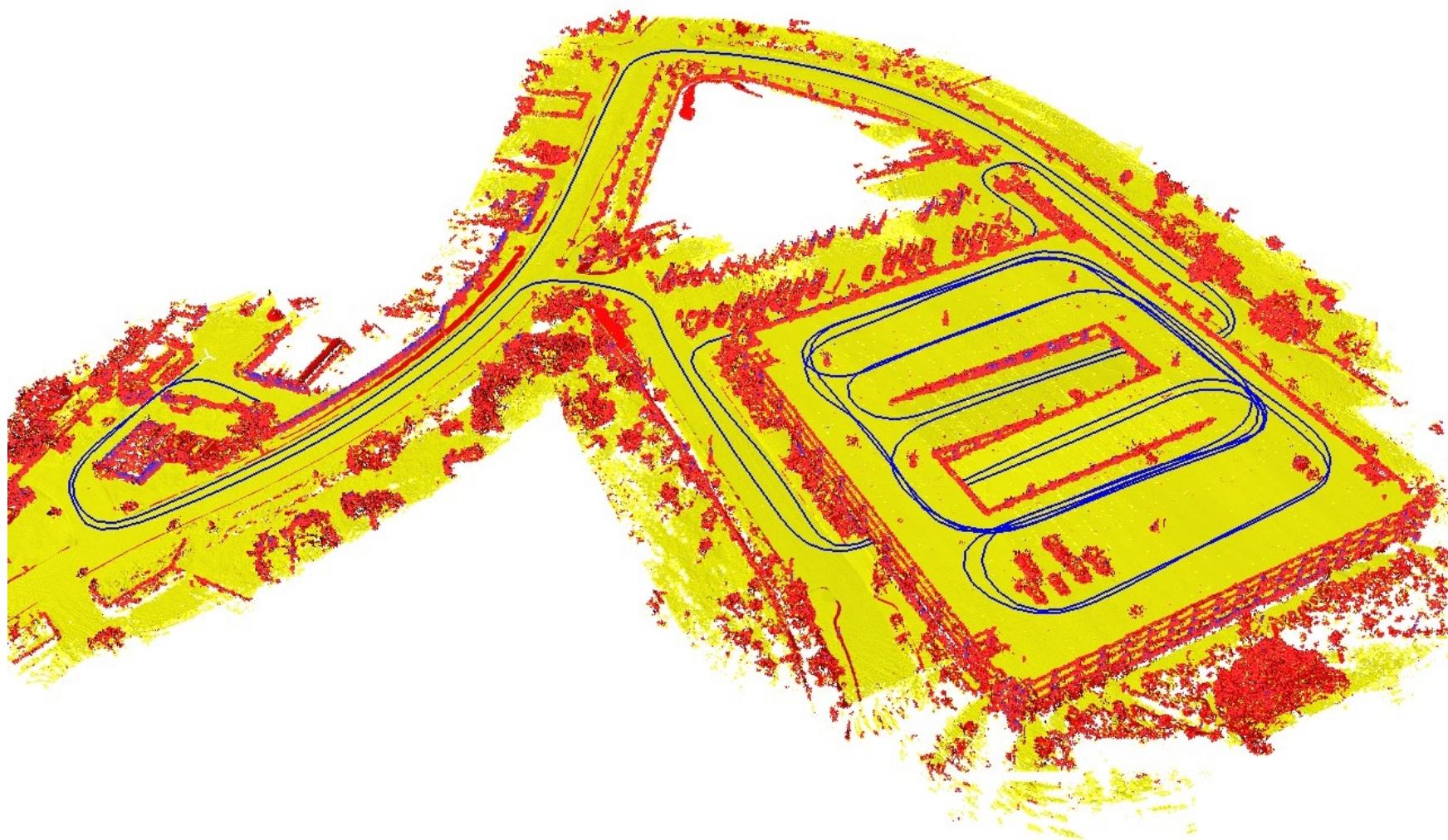
The robot can pass under and go over the bridge

Experiments with a Car

- Task: Reach a parking spot on the upper level



MLS Map of the Parking Garage

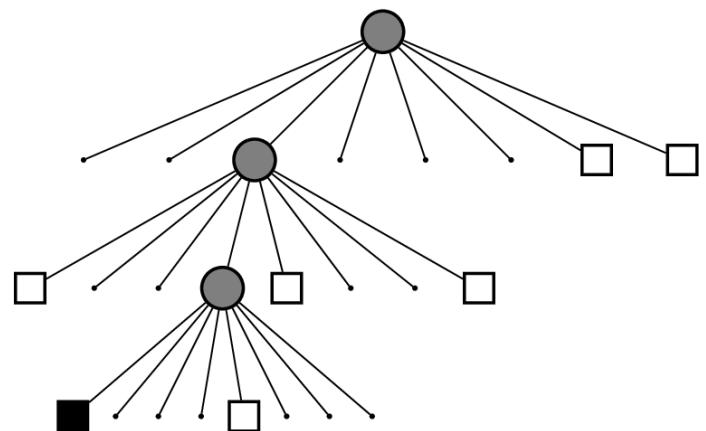
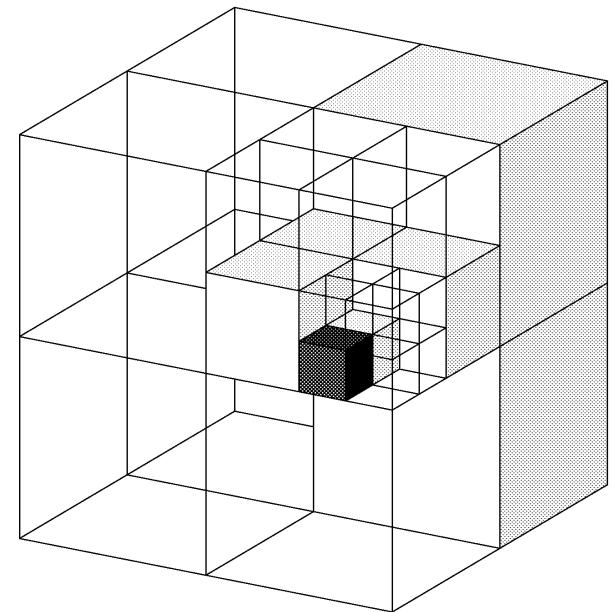


MLS Maps

- Pro:
 - Can represent multiple surfaces per cell
- Contra:
 - No representation of unknown
 - No volumetric representation but a discretization in the vertical dimension
 - Localization in MLS maps is not straightforward

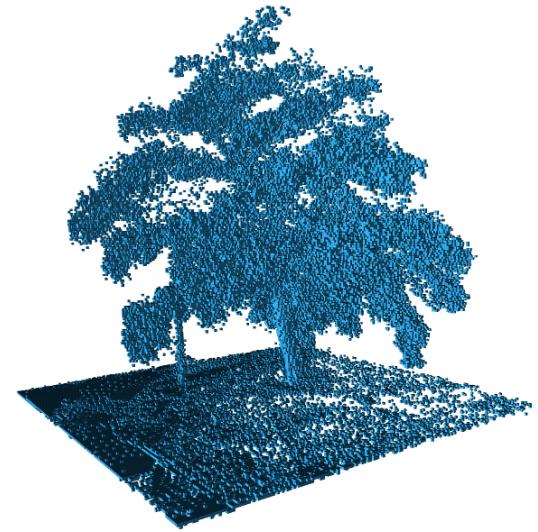
Octree-based Representation

- Tree-based data structure
- Recursive subdivision of the space into octants
- Volumes allocated as needed
- “Smart 3D grid



Octrees

- **Pro:**
 - Full 3D model
 - Probabilistic
 - Inherently multi-resolution
 - Memory efficient
- **Contra:**
 - Implementation can be tricky
(memory, update, map files, ...)



OctoMap Framework

- Based on **octrees**
- **Probabilistic, volumetric** representation of occupancy including unknown
- Supports **multi-resolution** map queries
- **Memory efficient**
- Compact **map files**
- Open source implementation as C++ library available at <http://octomap.sf.net>

Probabilistic Map Update

- Occupancy modeled as recursive
binary Bayes filter [Moravec '85]

$$Bel(m_t^{[xyz]}) = \left[1 + \frac{1 - P(m_t^{[xyz]} | z_t, u_{t-1})}{P(m_t^{[xyz]} | z_t, u_{t-1})} \cdot \frac{P(m_t^{[xyz]})}{1 - P(m_t^{[xyz]})} \frac{1 - Bel(m_{t-1}^{[xyz]})}{Bel(m_t^{[xyz]})} \right]^{-1}$$

- Efficient update using **log-odds** notation

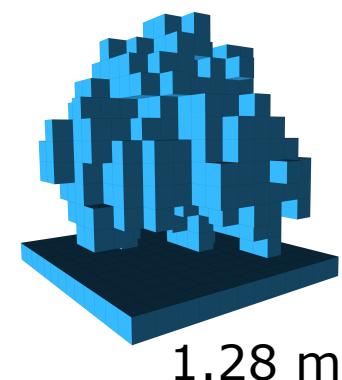
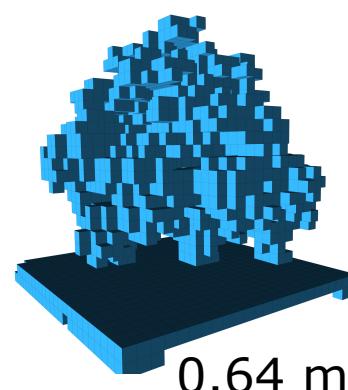
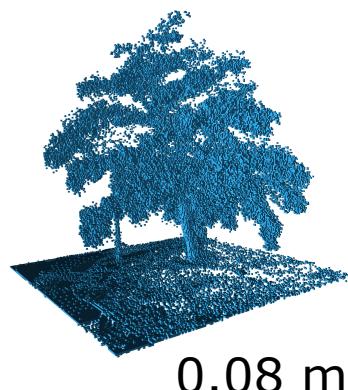
Probabilistic Map Update

- Clamping policy ensures updatability [Yguel '07]

$$Bel(m_t^{[xyz]}) \in [l_{\min}, l_{\max}]$$

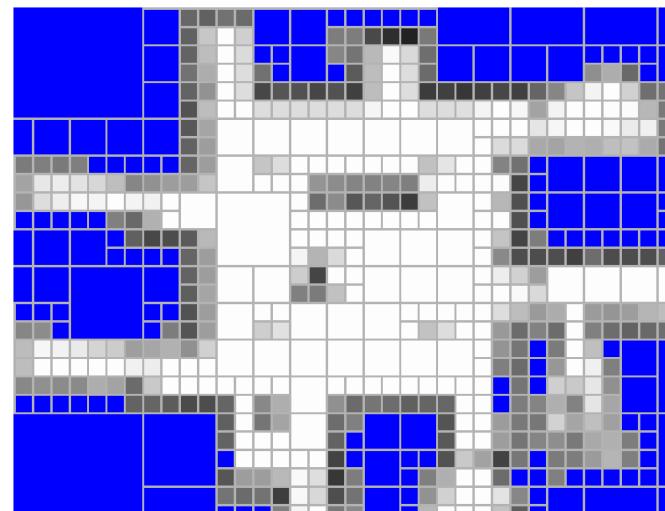
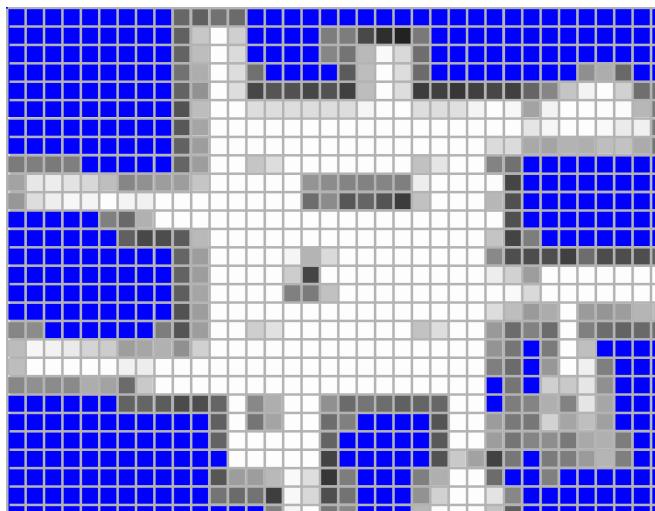
- Multi-resolution queries using

$$Bel(n) = \max_{i=1..8} Bel(n_i), n_i \in \text{children}(n)$$



Lossless Map Compression

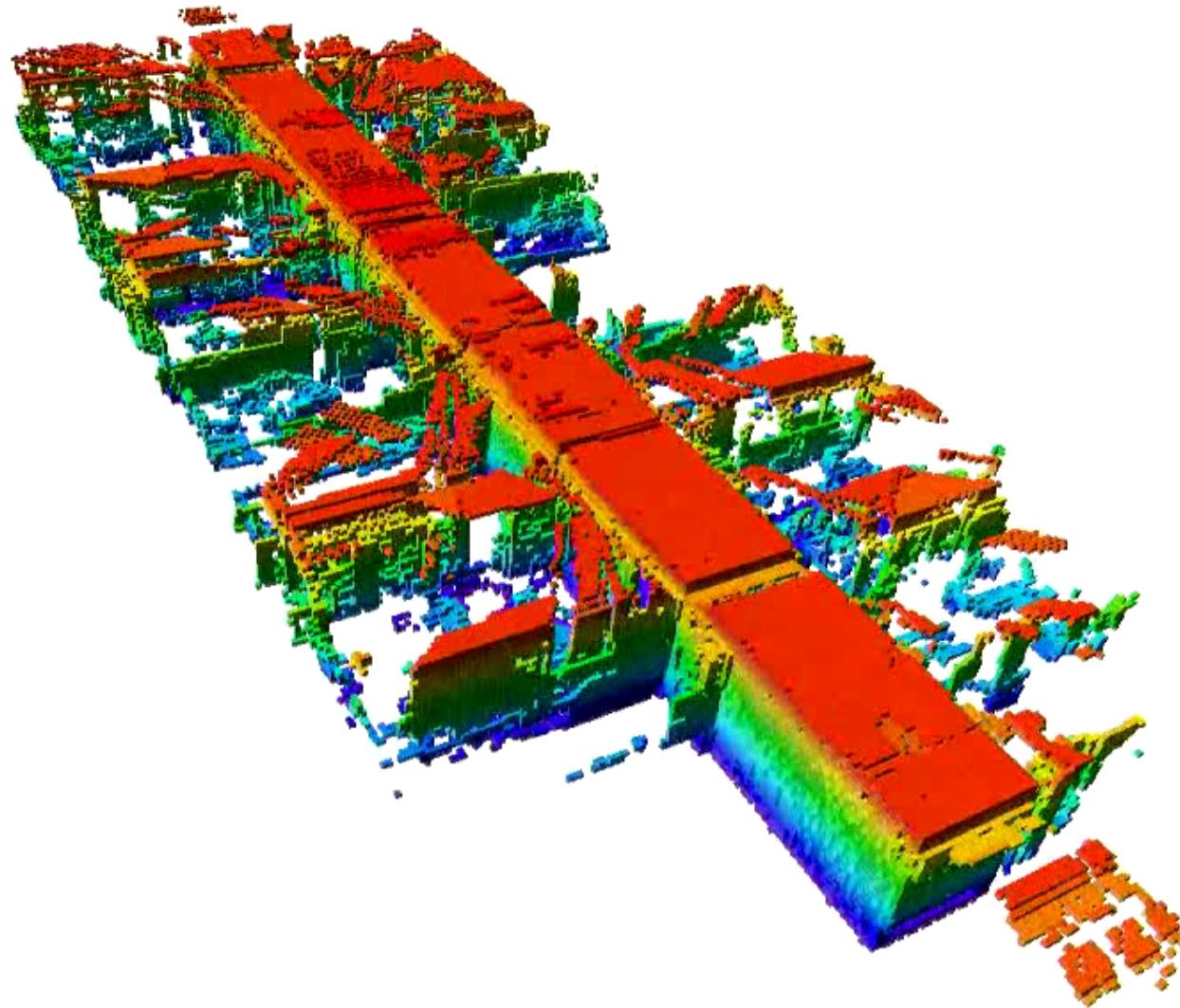
- Lossless pruning of nodes with identical children
- Can lead to high compression ratios



[Kraetzschmar '04]

Video: Office Building

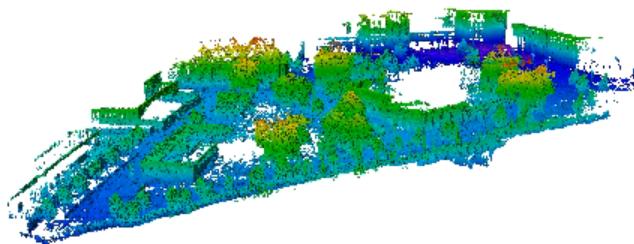
Freiburg, building 079



Video: Large Outdoor Areas

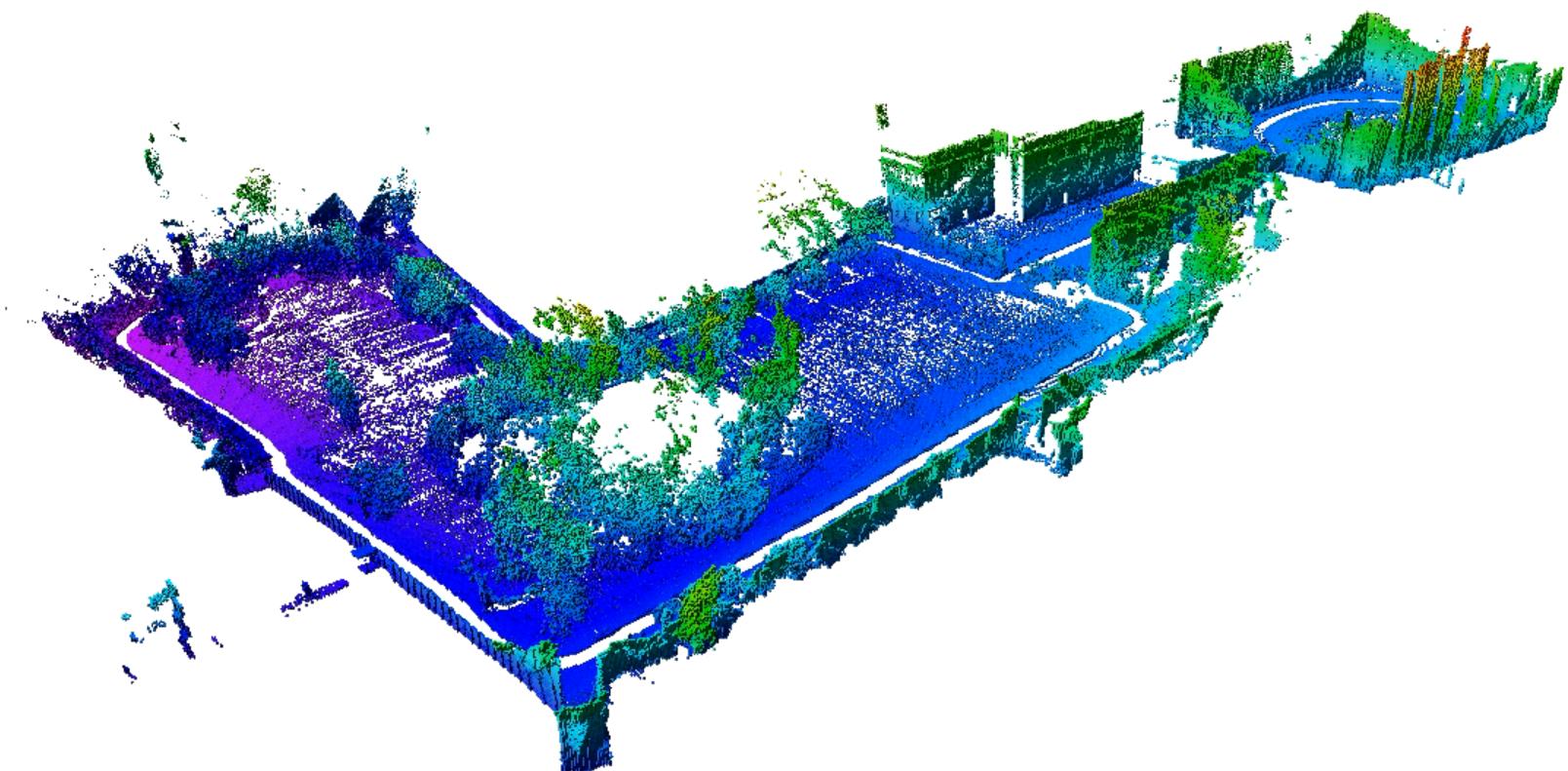
Freiburg computer science campus

($292 \times 167 \times 28 \text{ m}^3$, 20 cm resolution)



Video: Large Outdoor Areas

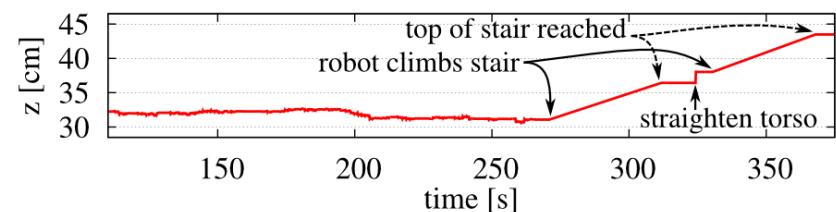
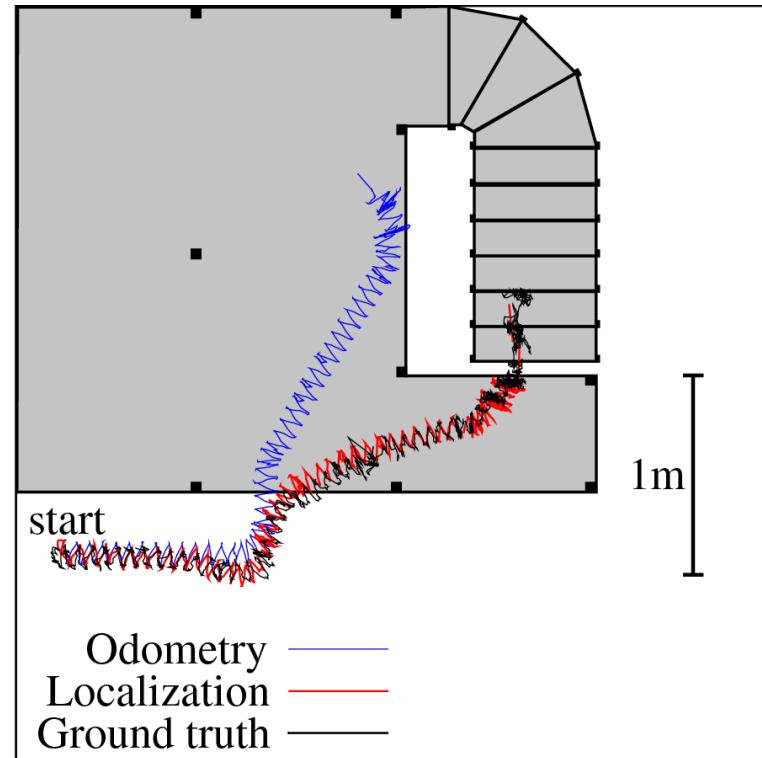
- Oxford New College dataset (Epoch C)
($250 \times 161 \times 33 \text{ m}^3$, 20 cm resolution)



6D Localization with a Humanoid



Goal: Accurate pose tracking while walking and climbing stairs



Video: Humanoid Localization



Summary

- Different 3D map representations exist
- Octomap is currently a popular tool
- Main advantages:
Full 3D model, probabilistic representation,
inherently multi-resolution, memory
efficient