

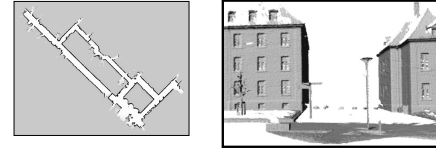
CSE-571 Robotics

Mapping

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Types of SLAM-Problems

Grid maps or scans



Sparse landmarks



RGB / Depth Maps



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Problems in Mapping

- **Sensor interpretation**
 - How do we **extract relevant information** from raw sensor data?
 - How do we represent and **integrate** this information **over time**?
- **Robot locations have to be known**
 - How can we estimate them **during mapping**?

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Occupancy Grid Maps

- Introduced by Moravec and Elfes in 1985
- Represent environment by a grid.
- Estimate the probability that a location is occupied by an obstacle.
- **Key assumptions**
 - Occupancy of individual cells is independent

$$\begin{aligned} Bel(m_t) &= P(m_t | u_1, z_2, \dots, u_{t-1}, z_t) \\ &= \prod_{x,y} Bel(m_t^{[xy]}) \end{aligned}$$

- Robot positions are known!

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Updating Occupancy Grid Maps

- **Idea:** Update each individual cell using a **binary Bayes filter**.

$$Bel(m_t^{[xy]}) = \eta p(z_t | m_t^{[xy]}) \sum_{m_{t-1}^{[xy]}} p(m_t^{[xy]} | m_{t-1}^{[xy]}, u_{t-1}) Bel(m_{t-1}^{[xy]})$$

- **Additional assumption:** Map is static

$$Bel(m_t^{[xy]}) = \eta p(z_t | m_t^{[xy]}) Bel(m_{t-1}^{[xy]})$$

- **Log odds representation:**

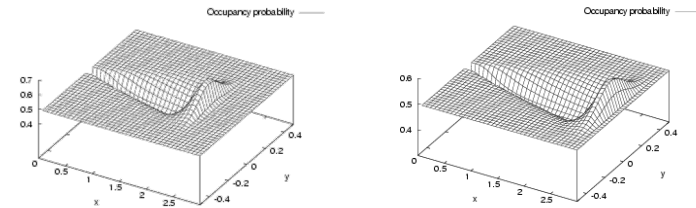
$$l_{t,i} = \log \left(\frac{p(m_i | z_{1:t}, x_{1:t})}{1 - p(m_i | z_{1:t}, x_{1:t})} \right)$$

$$l_{t,i} = l_{t-1,i} + \log \left(\frac{p(m_i | z_t, x_t)}{1 - p(m_i | z_t, x_t)} \right) - \log \left(\frac{p(m_i)}{1 - p(m_i)} \right)$$

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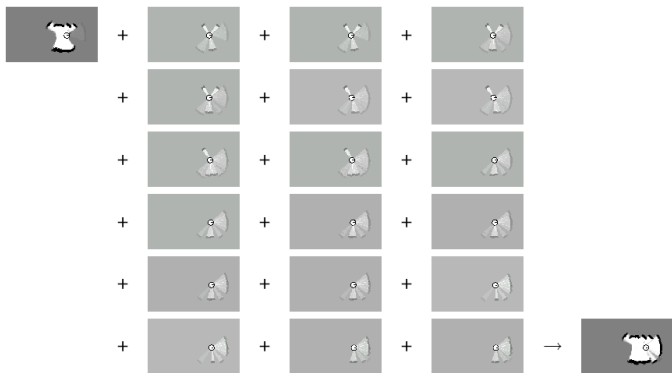
Inverse Sensor Model for Occupancy Grid Maps

Combination of linear function and Gaussian:



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Incremental Updating of Occupancy Grids (Example)



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Alternative: Simple Counting

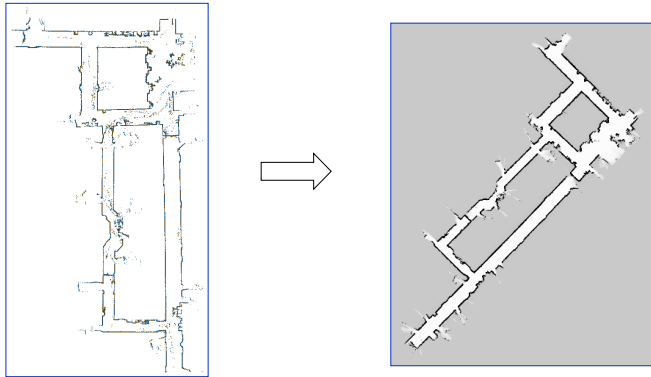
- For every cell count
 - **hits(x,y):** number of cases where a beam ended at $\langle x,y \rangle$
 - **misses(x,y):** number of cases where a beam passed through $\langle x,y \rangle$

$$Bel(m^{[xy]}) = \frac{\text{hits}(x,y)}{\text{hits}(x,y) + \text{misses}(x,y)}$$

- **Assumption:** $P(\text{occupied}(x,y)) = P(\text{reflects}(x,y))$

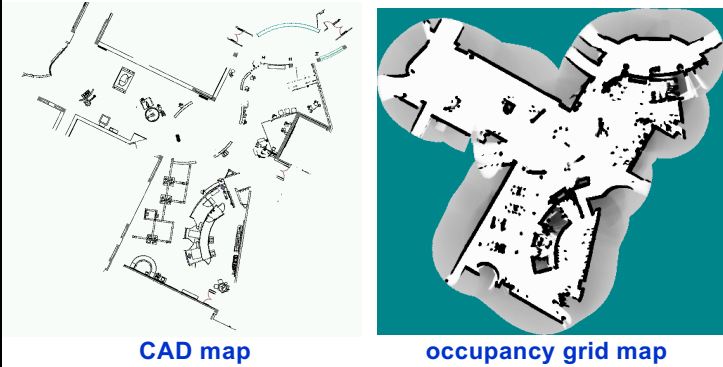
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Occupancy Grids: From scans to maps



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Tech Museum, San Jose



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OctoMap

A Probabilistic, Flexible, and Compact 3D Map Representation for Robotic Systems

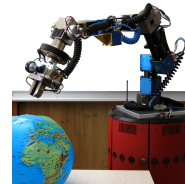
K.M. Wurm, A. Hornung,
M. Bennewitz, C. Stachniss, W. Burgard

University of Freiburg, Germany

<http://octomap.sf.net>

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Robots in 3D Environments



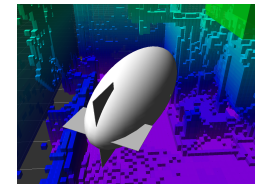
Mobile manipulation



Outdoor navigation



Humanoid robots



Flying robots

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3D Map Requirements

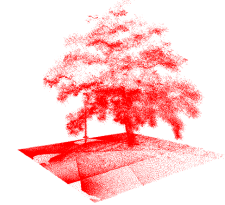
- Full 3D Model
 - Volumetric representation
 - Free-space
 - Unknown areas (e.g. for exploration)
- Can be updated
 - Probabilistic model (sensor noise, changes in the environment)
 - Update of previously recorded maps
- Flexible
 - Map is dynamically expanded
 - Multi-resolution map queries
- Compact
 - Memory efficient
 - Map files for storage and exchange

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Map Representations

Pointclouds

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

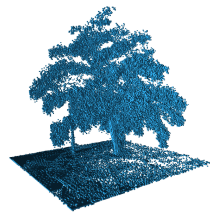


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Map Representations

3D voxel grids

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

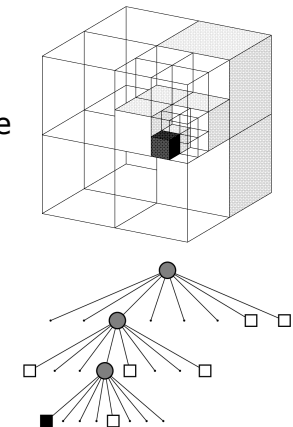


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Map Representations

Octrees

- Tree-based data structure
- Recursive subdivision of space into octants
- Volumes allocated as needed
- Multi-resolution

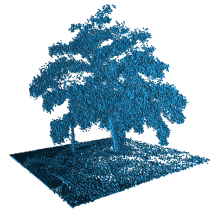


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Map Representations

Octrees

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



▪ Open source implementation as C++ library available at <http://octomap.sf.net>

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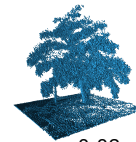
Probabilistic Map Update

- **Clamping policy** ensures updatability [Yguel '07]

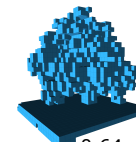
$$L(n) \in [l_{\min}, l_{\max}]$$

- Update of inner nodes enables **multi-resolution queries**

$$L(n) = \max_{i=1..8} L(n_i)$$



0.08 m



0.64 m

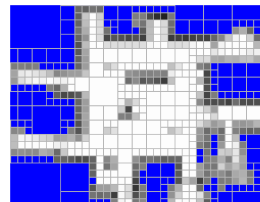
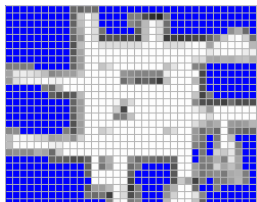


1.28 m

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Lossless Map Compression

- **Lossless pruning** of nodes with identical children
- High compression ratios esp. in free space

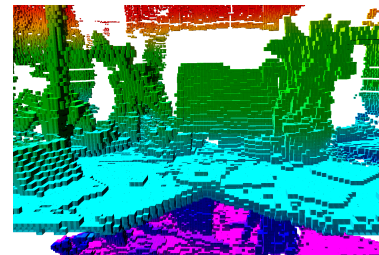


[Kraetzschmar 04]

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Examples

- Cluttered office environment



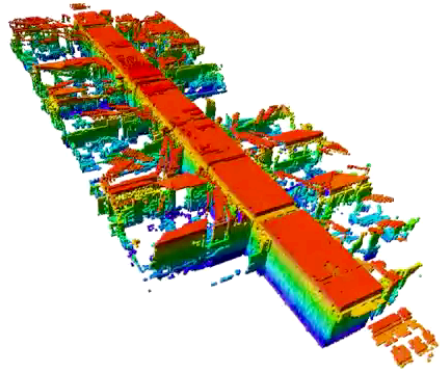
Map resolution: 2 cm



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Examples: Office Building

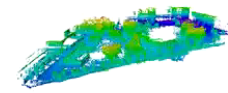
- Freiburg, building 079



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Examples: Large Outdoor Areas

- Freiburg computer science campus
(292 x 167 x 28 m³, 20 cm resolution)



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Examples: Tabletop



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Adding Color

Probabilistic 3D mapping using
OctoMap and RGBDSLAM

Kai M. Wurm, Felix Endres
Autonomous Intelligent Systems Lab
University of Freiburg, Germany



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Memory Usage

Map dataset	Mapped area [m ²]	Resolution [m]	Memory consumption [MB]			File size [MB]	
			Full grid	No compr.	Lossless compr.	All data	Binary
FR-079 corridor	43.8 × 18.2 × 3.3	0.05	80.54	73.64	41.70	15.80	0.67
		0.1	10.42	10.90	7.25	2.71	0.14
Freiburg outdoor	292 × 167 × 28	0.20	654.42	188.09	130.39	49.75	2.00
		0.80	10.96	4.56	4.13	1.53	0.08
New College (Epoch C)	250 × 161 × 33	0.20	637.48	91.43	50.70	18.71	0.99
		0.80	10.21	2.35	1.81	0.64	0.05