#### Introduction to Mobile Robotics

# **Techniques for 3D Mapping**

Wolfram Burgard



## Why 3D Representations

- Robots live in the 3D world.
- 2D maps have been applied successfully for navigation tasks such as localization.
- Reliable collision avoidance and path planning, however, requires accurate 3D models.
- How to represent the 3D structure of the environment?

#### **Popular Representations**

- Point clouds
- Voxel grids
- Surface maps
- Meshes

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# **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
- Discretization errors

# 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
- Typically, the uncertainty increases with measured distance



#### **Elevation Maps**

- 2D grid that stores an estimated height (elevation) for each cell
- Typically, the 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

# **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



#### **MLS Map Representation**



Each 2D cell stores various patches consisting of:

- $\hfill \hfill \hfill$
- The height variance  $\boldsymbol{\sigma}$
- 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 (>10cm) 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

#### Results



- Cell resolution: 10 cm
- Number of data points: 45,000,000

#### **MLS Map of the Parking Garage**



# MLS Maps

- Pro:
  - Can represent multiple surfaces per cell
- Contra:
  - No representation of unknown areas
  - 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]

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

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

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



## **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 x 167 x 28 m<sup>3</sup>, 20 cm resolution)



#### **6D Localization with a Humanoid**



**Goal:** Accurate pose tracking while walking and climbing stairs





#### **Video: Humanoid Localization**



### Signed Distance Function (SDF)



Negative signed distance (=outside)

Positive signed distance (=inside)

[begin slides courtesy of Jürgen Sturm]

#### Signed Distance Function (SDF)

- Compute SDF from a depth image
- Measure distance of each voxel to the observed surface
- Can be done in parallel for all voxels ( $\rightarrow$  GPU)
- Becomes very efficient by only considering a small interval around the endpoint (truncation)



#### Signed Distance Function (SDF)

- Calculate weighted average over all measurements for every voxel
- Assume known camera poses



#### Visualizing Signed Distance Fields

Common approaches to iso surface extraction:

- Ray casting (GPU, fast) For each camera pixel, shoot a ray and search for zero crossing
- Poligonization (CPU, slow)
  E.g., using the marching cubes algorithm Advantage: outputs triangle mesh

#### Mesh Extraction using Marching Cubes

Find zero-crossings in the signed distance function by interpolation



## Marching Cubes

If we are in 2D: Marching squares

- Evaluate each cell separately
- Check which edges are inside/outside
- Generate triangles according to 16 lookup tables
- Locate vertices using least squares



#### Marching Cubes (3D)



#### **KinectFusion**

- SLAM based on projective ICP (see next section) with point-to-plane metric
- Truncated signed distance function (TSDF)
- Ray Casting



#### **An Application**



[Sturm, Bylow, Kahl, Cremers; GCPR 2013], end courtesy by Jürgen Sturm]

# **Signed Distance Functions**

#### Pro:

- Full 3D model
- Sup-pixel accuracy
- Fast (graphics card) implementation
- Contra:
  - Space consuming voxel grid

#### Summary

- Different 3D map representations exist
- The best model always depends upon the corresponding application
- We discussed surface models and voxel representations
- Surface models support a traversability analysis
- Voxel representations allow for a full 3D representation
- Octrees are a probabilistic representation. They are inherently multi-resolution.
- Signed distance functions also use three-dimensional grids but allow for a sub-pixel accuracy representation of the surface.
- Note: there also is a PointCloud Library for directly dealing with point clouds (see also next chapter).