



Universität Hamburg

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MIN Faculty
Department of Informatics



Object Reconstruction in Robotics

Introduction and Recent Approaches



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Technical Aspects of Multimodal Systems

23. May 2019



Outline

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Applications in Robotics

2. Preliminaries

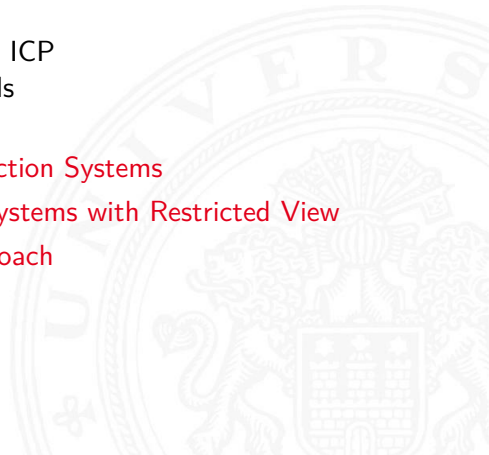
Depth Maps
Point Cloud Matching with ICP
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3. Examples of Active Reconstruction Systems

4. Examples of Reconstruction Systems with Restricted View

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What is Object Reconstruction

- ▶ Capture the shape and dimensions (and appearance, mechanical properties etc.) of real world objects with one or more sensors
- ▶ Construct a model from the obtained data

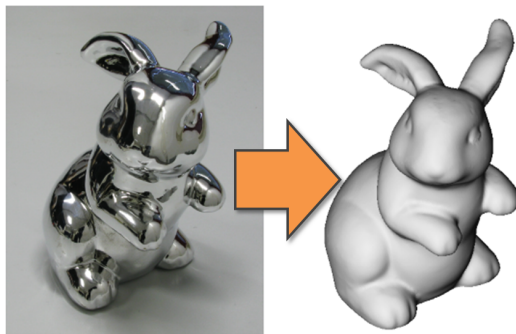
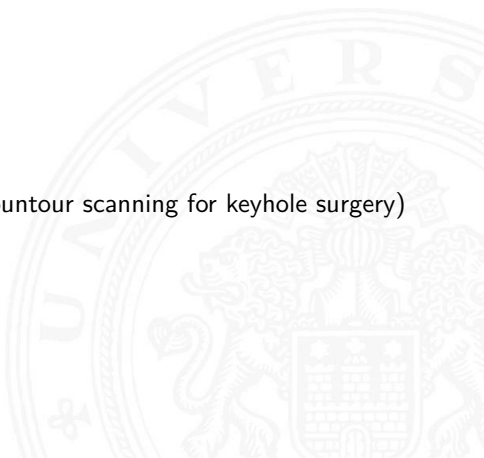


Figure: 1 Bunny figurine and reconstructed model (Weinmann, 2013)



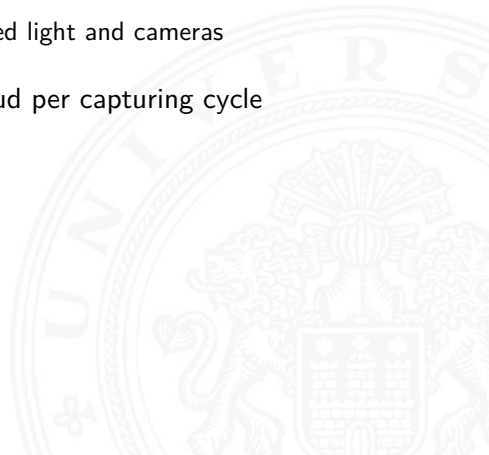
- ▶ Estimation for best grasping position
- ▶ Object recognition
- ▶ Object pose estimation
- ▶ Telerobotics
- ▶ Industry robotics
- ▶ Mapping
- ▶ Also related
 - ▶ Robots used in archaeology
 - ▶ Surgery robots (e.g. inner countour scanning for keyhole surgery)
 - ▶ Tomographic reconstruction
 - ▶ Virtual / augmented reality
 - ▶ ...





Depth Maps

- ▶ Typical sensors and methods for mapping depth
 - ▶ ToF (time-of-flight) cameras
 - ▶ Backscatter LiDAR (Light Detection And Ranging) systems
 - ▶ Triangulation using image cameras
 - ▶ Monocular cues
 - ▶ Triangulation using structured light and cameras
- ▶ The output is a 3D point cloud per capturing cycle



- ▶ Either light pulses or continuous wave
- ▶ Pulses: measure time of flight, continuous wave: measure phase shift

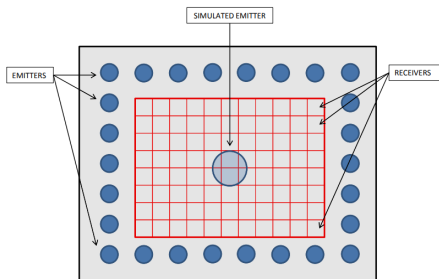


Figure: 2 Scheme of a matricial ToF camera sensor. The CCD/CMOS matrix of lock-in pixels is in red. The emitters (blue) are distributed around the lock-in pixels matrix and mimic a simulated emitter co-positioned with the center of the lock-in pixel matrix (light blue) (Mutto, 2012)

- ▶ Also based on time-of-flight or phase shift measurement

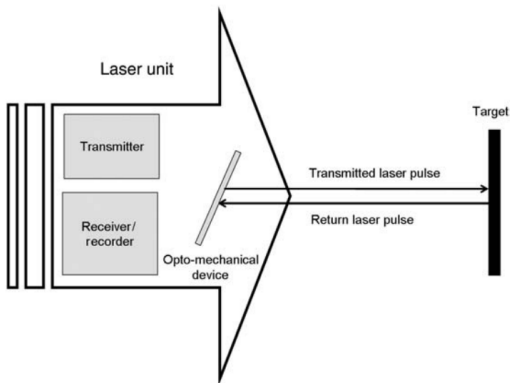


Figure: 4 Basic components of a LiDAR instrument (Heritage, 2009)

Stereo Cameras

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- ▶ 2 (identical) cameras, reference and target camera (Coplanar and aligned, parallel optical axis)
- ▶ Depth by triangulation of matching points in both images
- ▶ Epipolar geometry enhances search for matching points

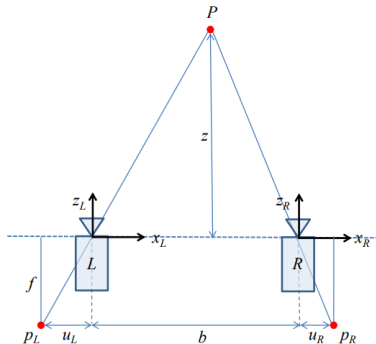


Figure: 7 Triangulation with a pair of aligned and rectified cameras (Mutto, 2012)

- ▶ "Simulate" 2 cameras with mirrors or prisms
- ▶ Track camera position and use images from different positions
 - ▶ Structure from Motion
 - ▶ Using motion sensors, e.g. inertial measurement unit

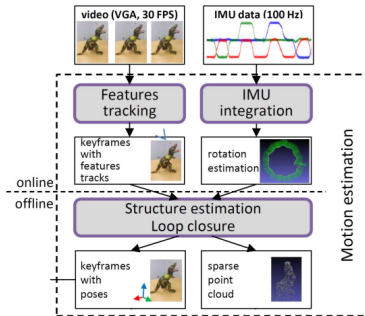
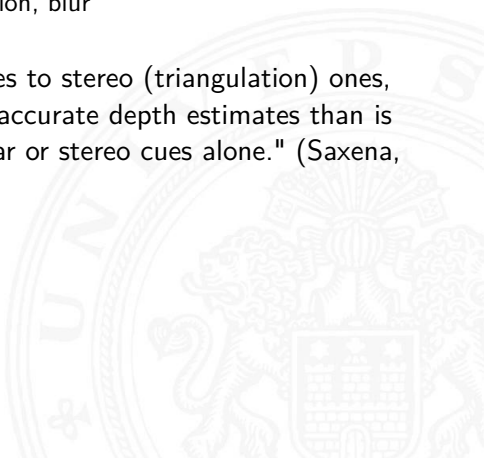


Figure: 8 System overview (Muratov, 2016)



Monocular Cues

- ▶ Infer depth from visual scene properties
 - ▶ Size of objects
 - ▶ Occlusions
 - ▶ Texture gradient and shading
 - ▶ Luminance, contrast, saturation, blur
 - ▶ ...
- ▶ "[...] by adding monocular cues to stereo (triangulation) ones, [...] obtain significantly more accurate depth estimates than is possible using either monocular or stereo cues alone." (Saxena, 2007)



Structured Light

- ▶ Project pattern on scene, various coding schemes to suit environmental conditions
- ▶ Find known pattern features in image to perform active triangulation

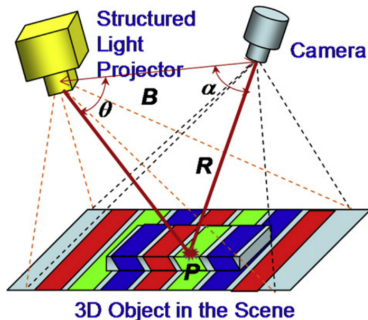


Figure: 10 Illustration of structured light (Geng, 2011)



Point Cloud Matching with ICP

- ▶ Iterative closest point algorithm
- ▶ Combine multiple incomplete (due to e.g. self occlusion) or inaccurate 3D point clouds by matching them
- ▶ Goal
 - ▶ Find corresponding points in both point clouds
 - ▶ Find a rotation R and translation T to match one to the other
- ▶ (Basic) Algorithm (Besl, 1992)
 - ▶ Find closest corresponding point (between target and reference set, for each point)
 - ▶ Compute rotation matrix R and translation vector T
 - ▶ Apply computed transformation and find distances (all corresponding points)
 - ▶ Loop unless distance is below threshold or max. iterations reached

Meshing of 3D Point Clouds

- ▶ Typical approach: Reconstruct surface with triangles (e.g. Delaunay triangulation)
- ▶ Involves steps such as noise reduction, removal of erroneous and redundant triangles, adjusting of spacing, etc.
- ▶ Can be followed by grid construction

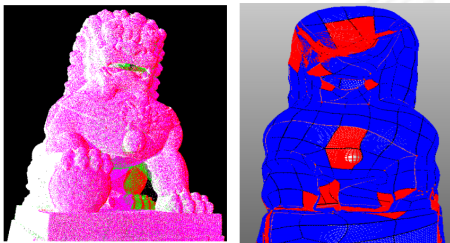


Figure: 11 Five registered point cloud scans, 3D mesh after grid construction (Nguyen, 2012)

Beyond Optical Sensing

- ▶ Many more ways to find shape information
 - ▶ Acoustic
 - ▶ Micro waves
 - ▶ ...
- ▶ Especially popular: tactile sensing
- ▶ Also allows gathering mechanical properties

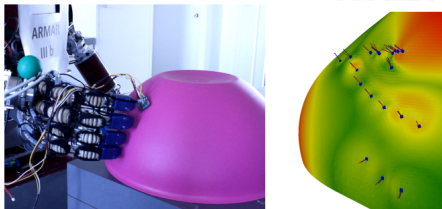


Figure: 12 Exploration of an unknown object [...] reconstructed using the gathered contact positions and normals (Ottenhaus, 2018)

- ▶ System by Jiang et al. (2018)
 - ▶ Workbench with u-shaped gantry above, which moves across
 - ▶ Two 2D laser scanners attached to gantry pointing downwards
 - ▶ Error on 600mm diameter truck wheel:
 - ▶ 0.7mm (53s scan time), 2.6 mm (13s scan time)

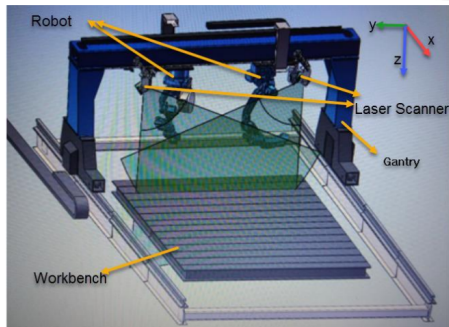


Figure: 13 Large-scale online 3D reconstruction system diagram (Jiang, 2018)

- ▶ Work of Ottenhaus et al. (2018)
 - ▶ Tactile robotic hand with soft fingertips
 - ▶ Robot's fingertips each have IMU and pressure sensor
 - ▶ Object surface is estimated through Gaussian process utilising contact positions and contact normals (local tilt)
 - ▶ Error on bowl: 4.7mm, 17° surface normal
 - ▶ Error on cardboard box: 3.5mm, 24° surface normal

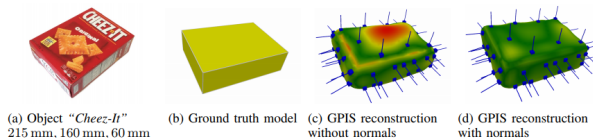


Figure: 14 Comparison of the reconstruction results with included and excluded normal information [...] (Ottenhaus, 2018)

Single View with RGBD Camera

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- ▶ Work of Rodríguez-Jiménez et al. (2013)
 - ▶ Using Kinect (structured light), objects lying on table
 - ▶ RGB-based segmentation on image to fill holes in depth map
 - ▶ Assumes table top normal is extrusion axis to fill unobserved part

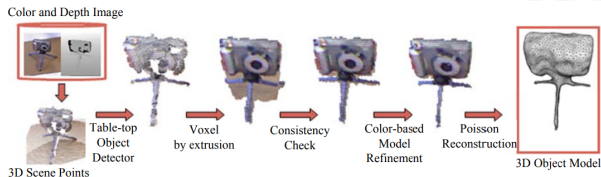
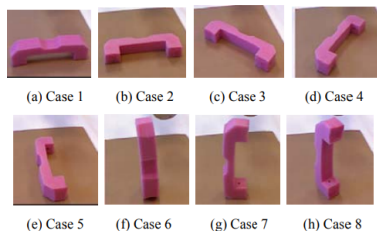


Figure: 15 Overview of [...] model acquisition process (Rodríguez-Jiménez, 2013)

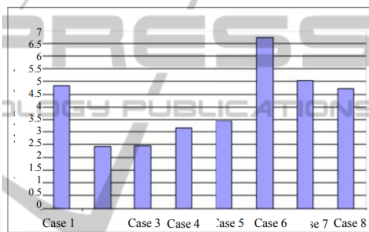
Single View with RGBD Camera

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- ▶ Less than 5mm error on everyday objects (book, can, camera, glue stick, cup, pen, ball, etc.)
- ▶ Higher error on asymmetric or irregular shaped objects, due to incorrect assumption



(a) Fig. 16 [...] object in the 8 evaluated orientations [...] (Rodríguez-Jiménez, 2013)



(b) Fig. 17 Evaluation of the error for 8 orientations (Rodríguez-Jiménez, 2013)

- ▶ Work of Klank et al. (2009)
 - ▶ Stereo camera based object recognition
 - ▶ Internet search and 3D object matching using ICP
 - ▶ Best search for plate on cluttered table: 201.65s (0.98 certainty)

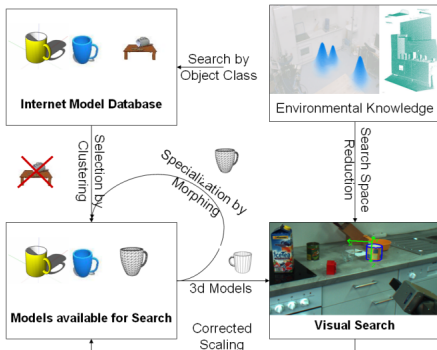
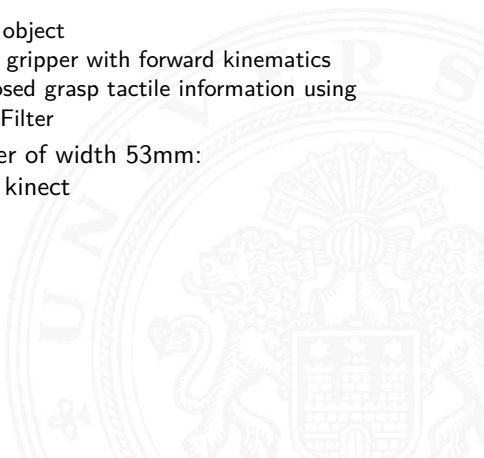


Figure: 18 Overview of the method (Klank, 2009)



- ▶ Work of Ilonen et al. (2013)
 - ▶ Kinect (RGBD) and robotic arm with gripper and tactile sensors
 - ▶ Method:
 1. Construct point cloud model from RGBD data and complete assuming symmetry
 2. Plan and execute grasp on object
 3. Compute contact points of gripper with forward kinematics
 4. Update model based on closed grasp tactile information using Iterated Extended Kalman Filter
 - ▶ Result for cylindrical container of width 53mm:
 - ▶ 45.41mm on avg. using only kinect
 - ▶ 55.43mm after fusion
 - ▶ Error of 2.43mm on avg.

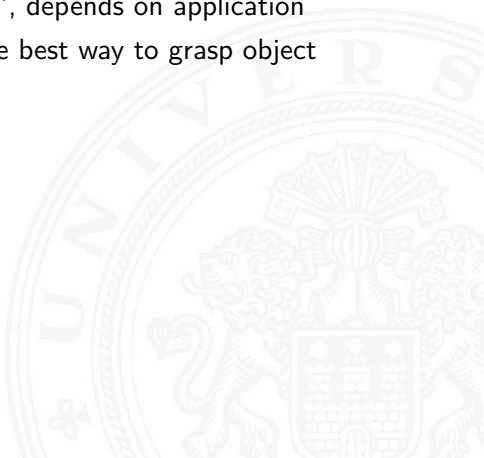


- ▶ Full Laser Scan
 - ▶ Fast and very accurate
 - ▶ Not flexible, expensive hardware
- ▶ Tactile Sensing
 - ▶ Can reconstruct surface accurately through surface normals
 - ▶ Very slow, expensive hardware
- ▶ RGBD Single View
 - ▶ Operates passively, somewhat accurate, comparatively cheap hardware
 - ▶ Makes assumptions about object shape
- ▶ DB Based
 - ▶ Very accurate (if models available), can obtain other information (e.g. weight), comparatively cheap hardware
 - ▶ Classification adds dimension for errors to occur
- ▶ Visual and Tactile Fusion
 - ▶ Somewhat accurate
 - ▶ Object must be accessible, expensive hardware



Conclusion

- ▶ Broad variety of approaches (and counting)
- ▶ Different sensors / modalities to obtain data
- ▶ Often involves tradeoffs (speed, accuracy, flexibility, cost)
- ▶ In general, no "best approach", depends on application
- ▶ Common application: estimate best way to grasp object





References

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- ▶ Signal attenuation, time delay and background illumination account for the signal that is received

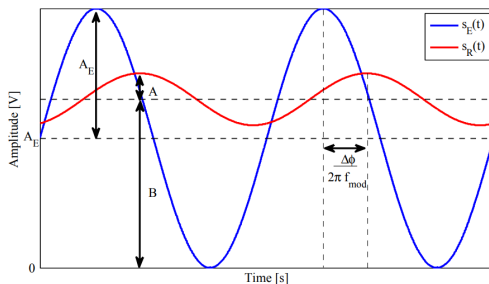


Figure: 3 Example of emitted signal $s_E(t)$ (in blue) and received signal $s_R(t)$ (in red)) (Mutto, 2012)

- Compute position in space from mirror rotation and support position

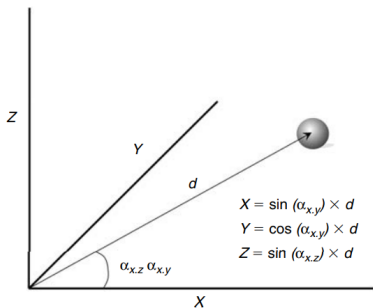


Figure: 5 Illustration of the ability of LiDAR systems to position remote objects in 3D space (Heritage, 2009)