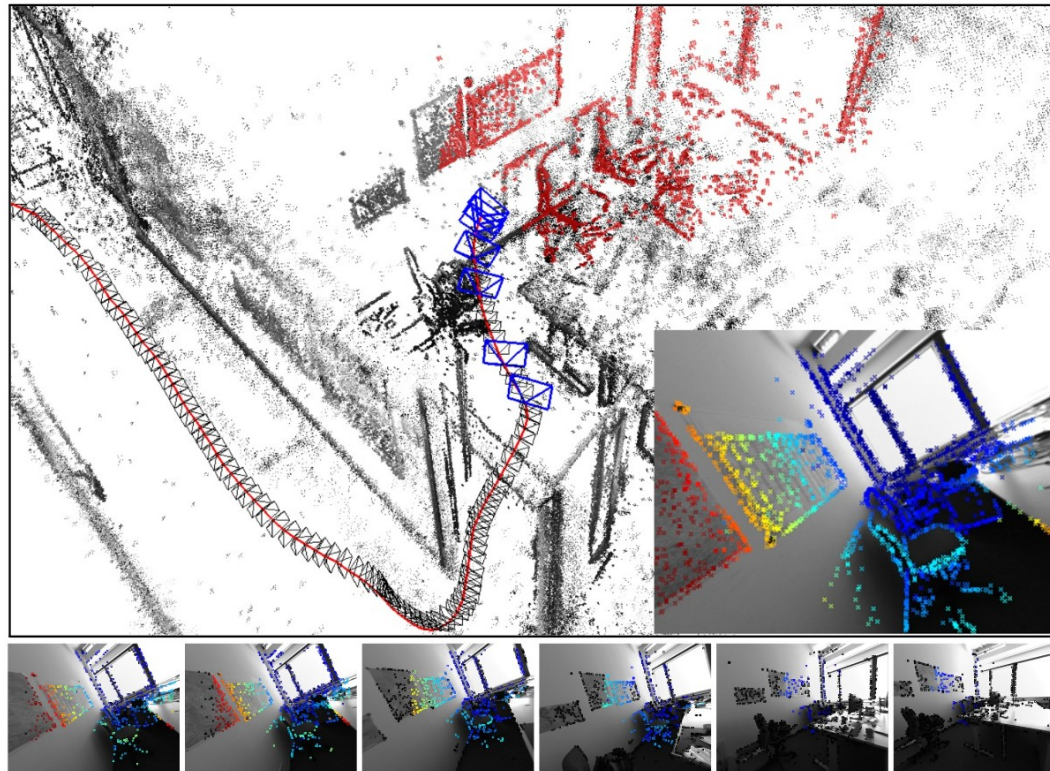


SLAM

Surface Reconstruction

Tobias Sippl
0820552



Overview

- SLAM
- RGBD:
Kinect Fusion
And non-rigid Objects: Dynamic Fusion
- RGB:
How to obtain Depth?
Live Dense Reconstruction with a Single Moving Camera
Monofusion

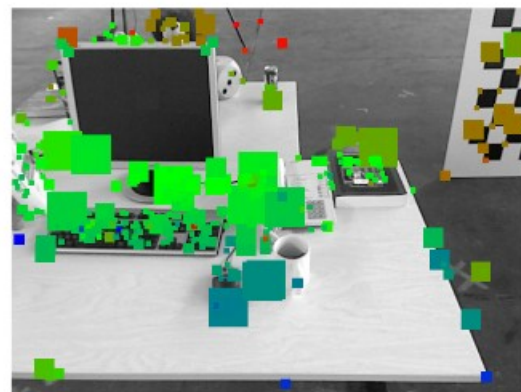
Direct Methods



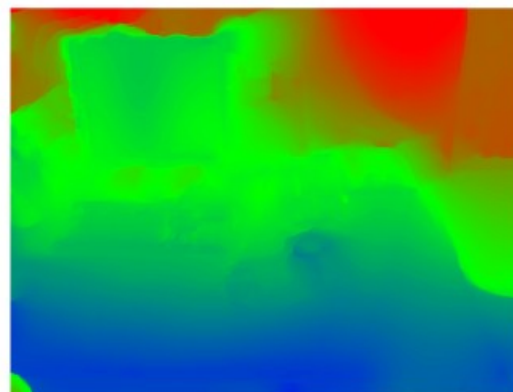
original image



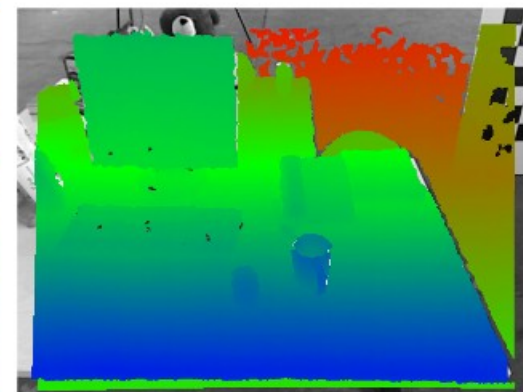
semi-dense depth map



keypoint depth map



dense depth map



RGB-D camera

SLAM

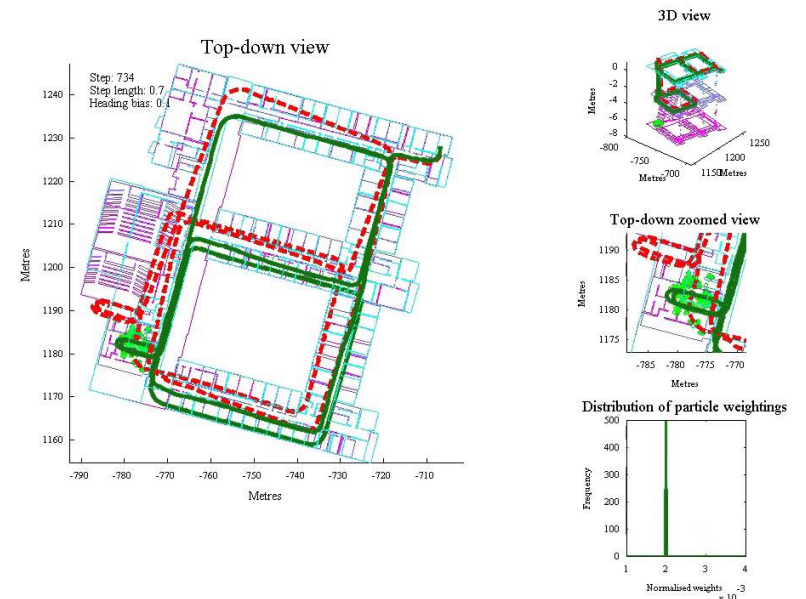
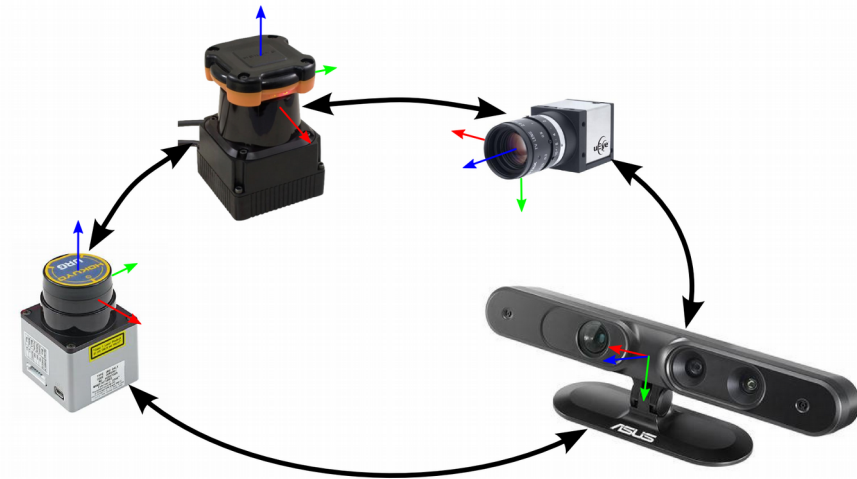
Simultaneous Localization and Mapping

- Inputs:
Cameras, GPS, IMUs, ...

- Output:
Camera parameters,
surrounding map

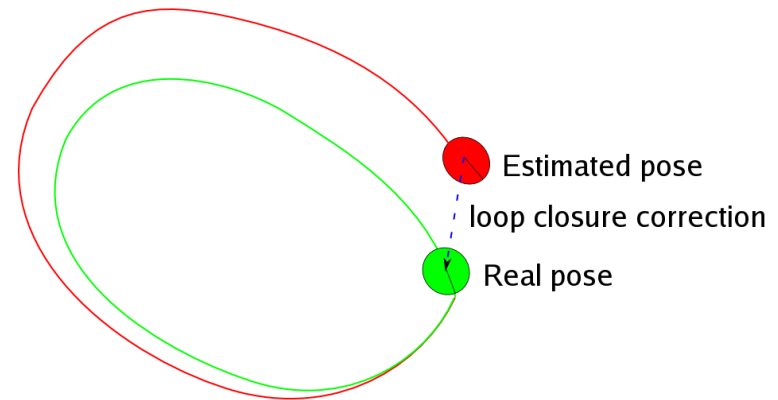
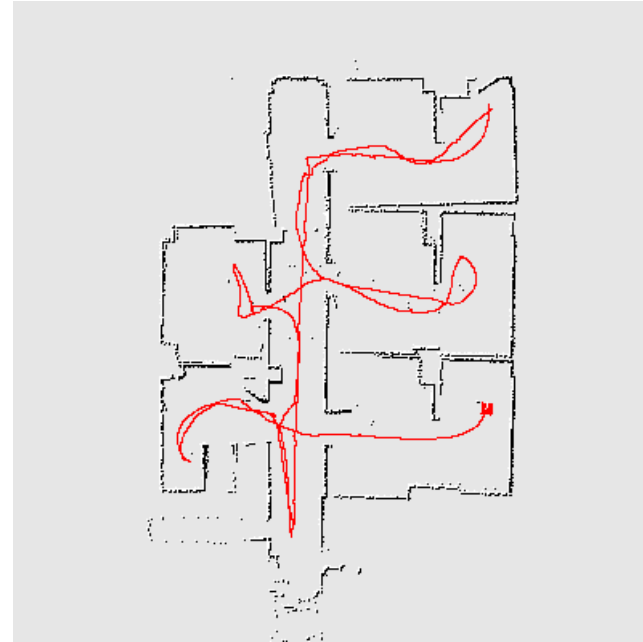
- Applications in:
autonomous robots (cars)
augmented reality

...



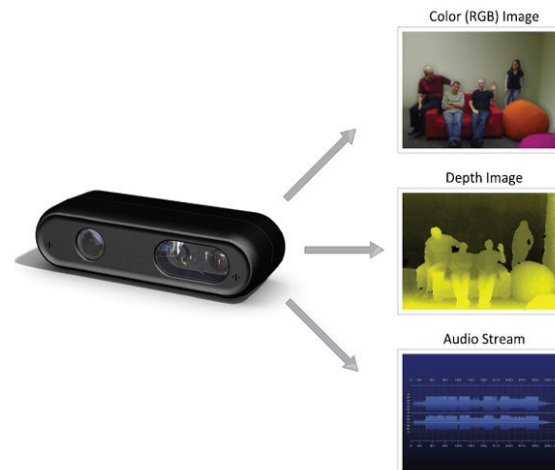
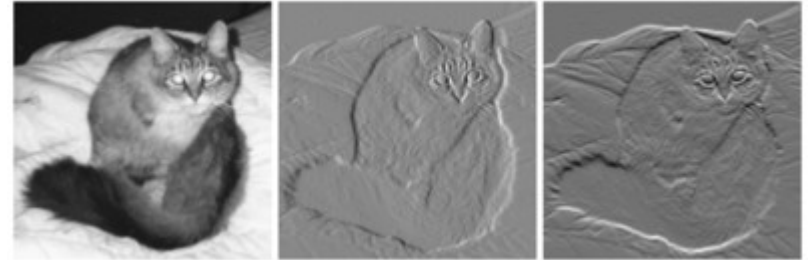
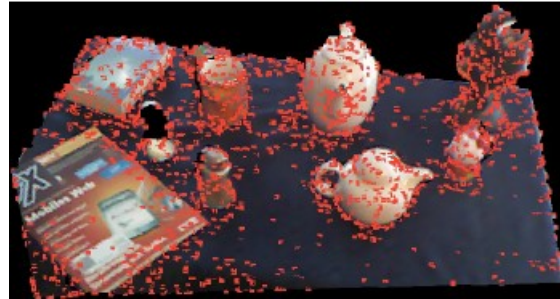
Mapping

- Mapping
- Recognition
- Loop Closure



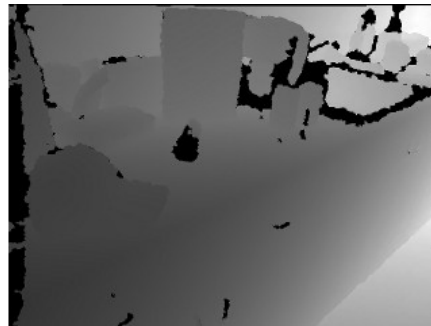
SLAM Terminology

- Indirect
- Direct
- Sparse
- Dense
- RGB
- Depth Sensor



Depth Cameras

- Depth value for each image section
- Noisy image
- Accessible through Kinect

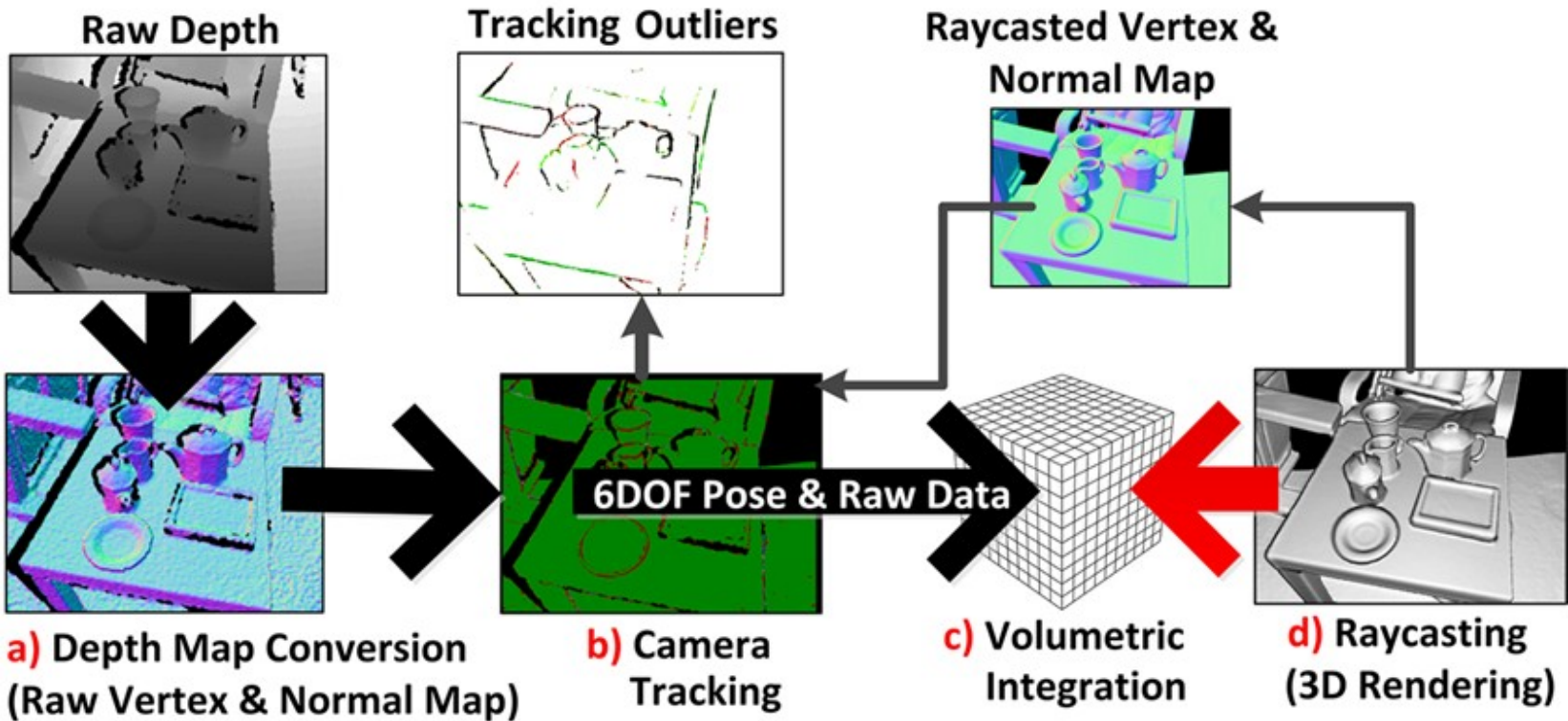


Kinect Fusion

Richard A. Newcombe Shahram Izadi et al. - Cambridge 2011

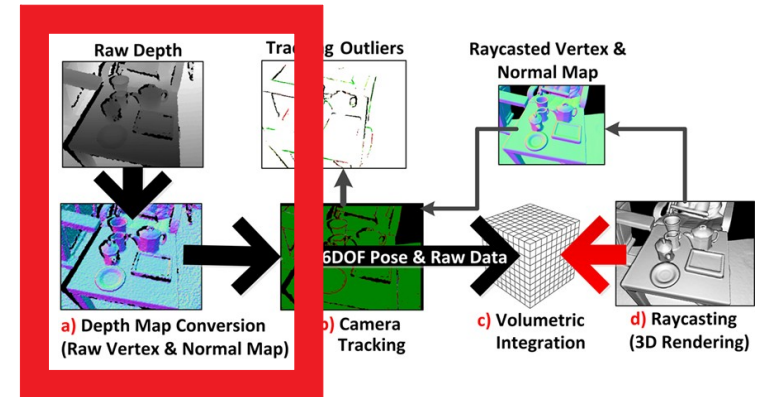


Kinect Fusion



Kinect Fusion

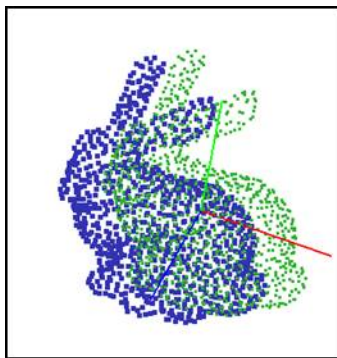
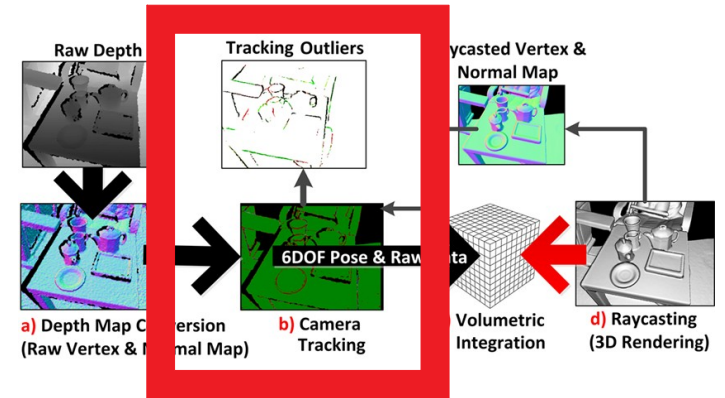
- Convert raw depth map to vertex & normal map



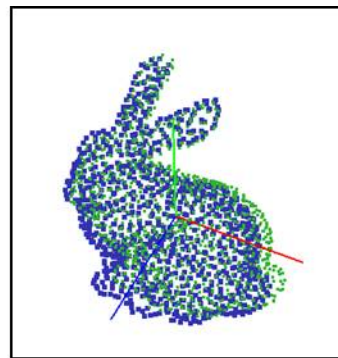
- Vertex Map:
Using Intrinsic calibration
- Normal Map:
neighbouring points

Kinect Fusion

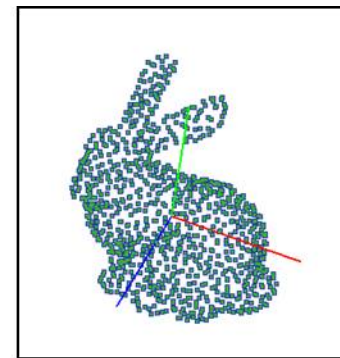
- Camera Tracking:
 - Remove Outliers
 - Iterative Closest Points



Iteration 0



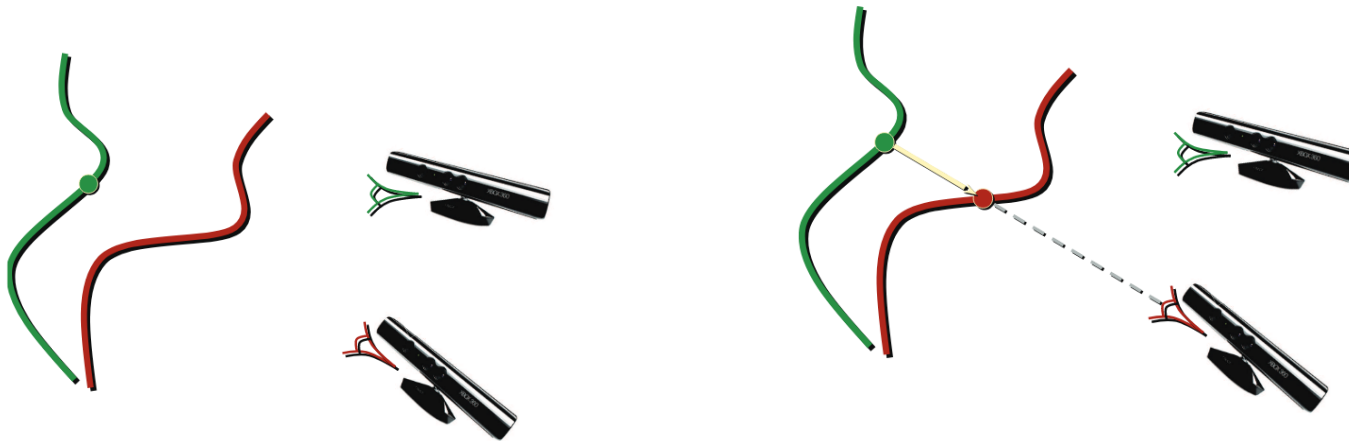
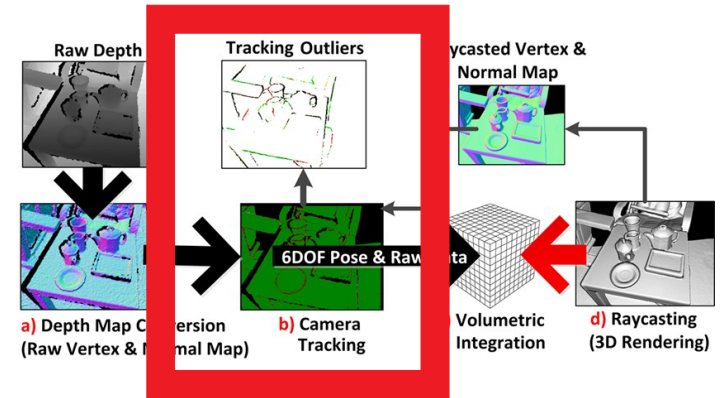
Iteration 1



Iteration 10

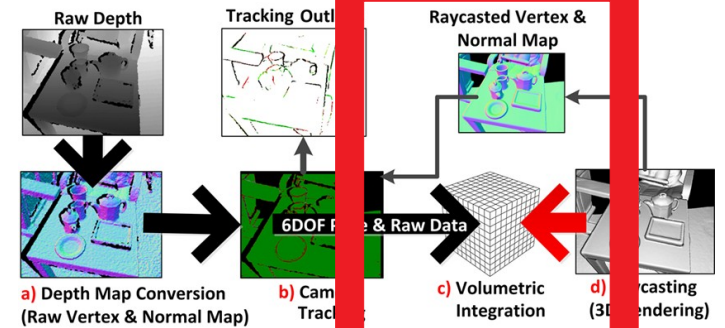
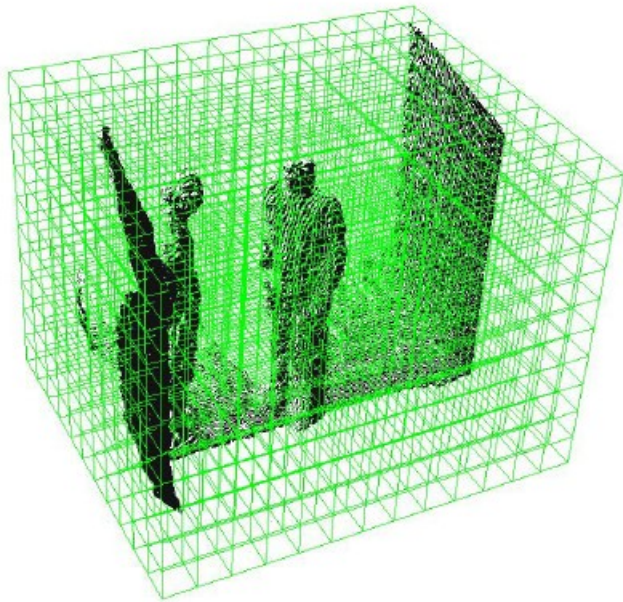
Kinect Fusion

- Find Correspondence
- Improve 6DoF
- Error: Euclidian Distance Normal



Kinect Fusion

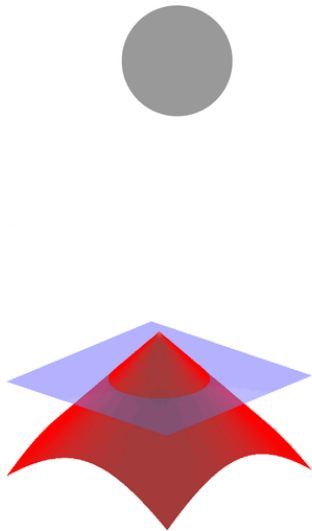
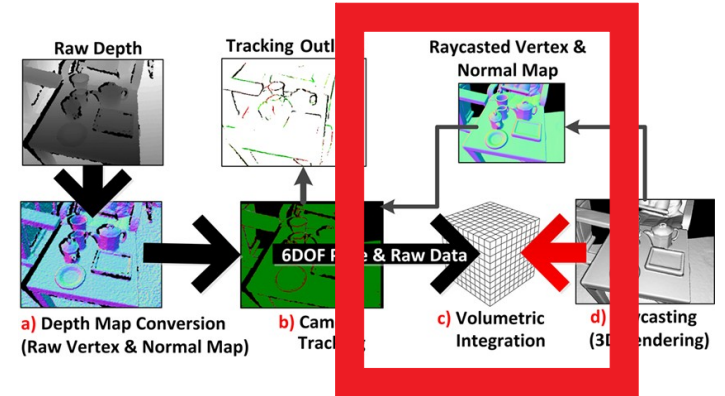
- Voxel Grid



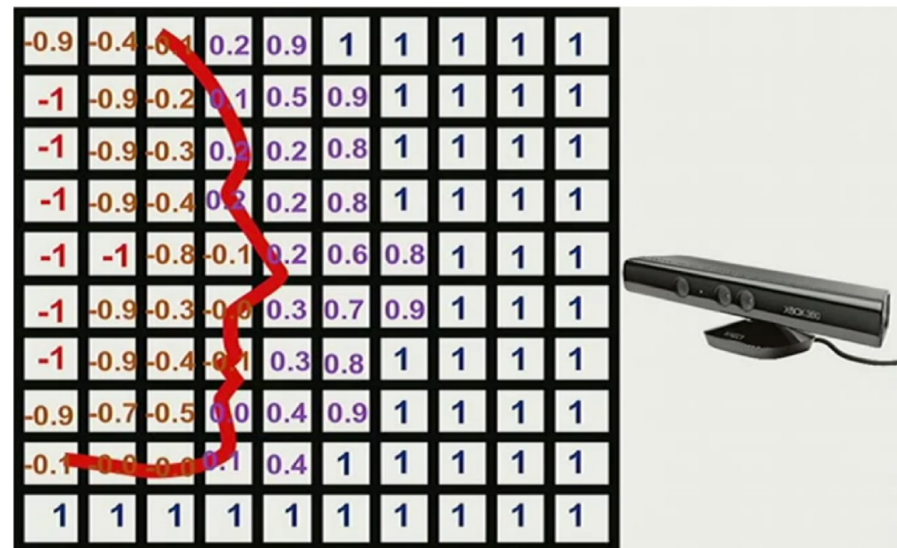
- Volume of 512^3
512 mB memory
2ms per sweep (read write op) on GTX470

Kinect Fusion

- Truncated Signed Distance Field



A Disc and its SDF



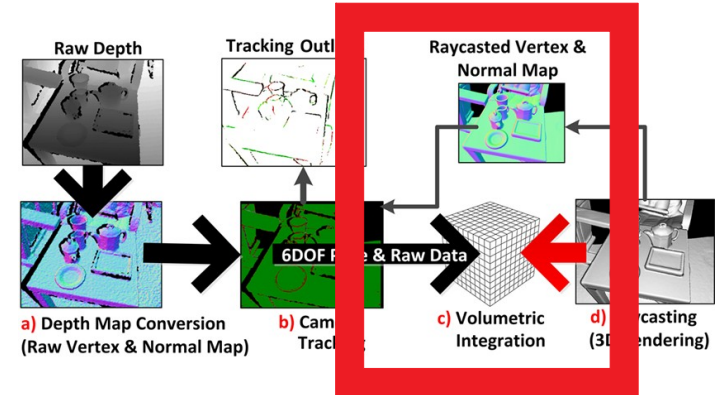
A 2D TSDF

Kinect Fusion

- Volumetric Integration

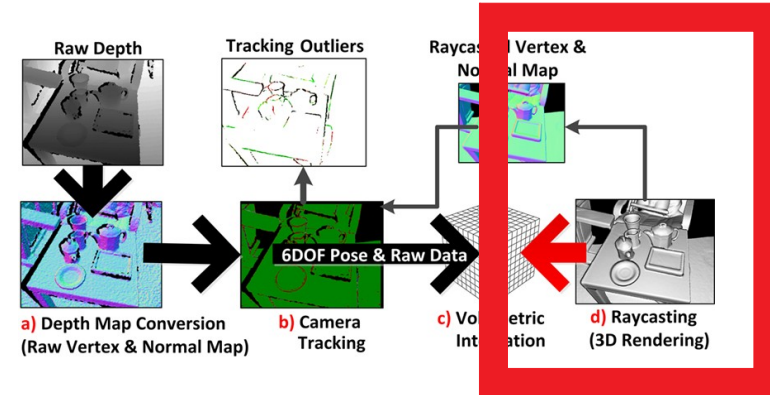
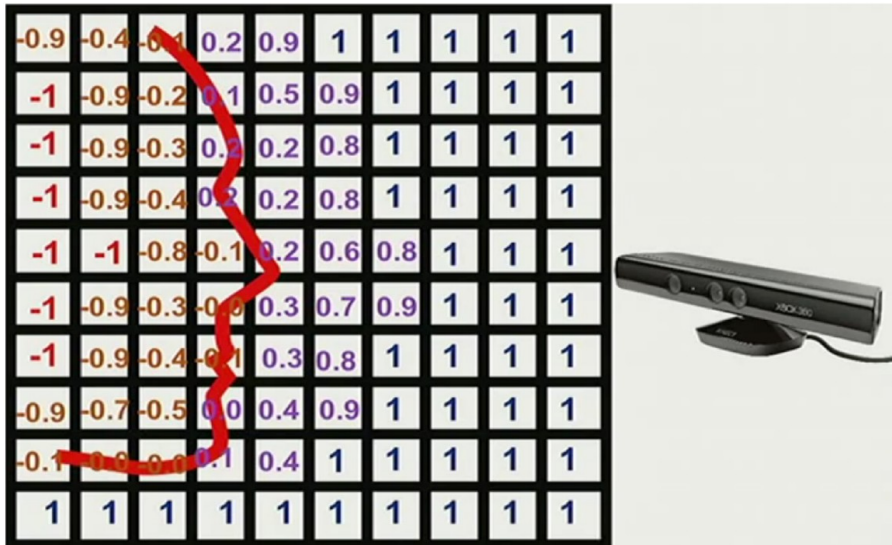
For each Voxel:

- Find Distance from Camera to voxel Position
- Perspective project and compare to Distance in Depth Map
- Apply Thresholds and update Voxel grid



Kinect Fusion

- Raycasting/Rendering
Find the zero-crossing



Kinect Fusion Video

SIGGRAPH Talks 2011

KinectFusion:

**Real-Time Dynamic 3D Surface
Reconstruction and Interaction**

**Shahram Izadi 1, Richard Newcombe 2, David Kim 1,3, Otmar Hilliges 1,
David Molyneaux 1,4, Pushmeet Kohli 1, Jamie Shotton 1,
Steve Hodges 1, Dustin Freeman 5, Andrew Davison 2, Andrew Fitzgibbon 1**

1 Microsoft Research Cambridge 2 Imperial College London

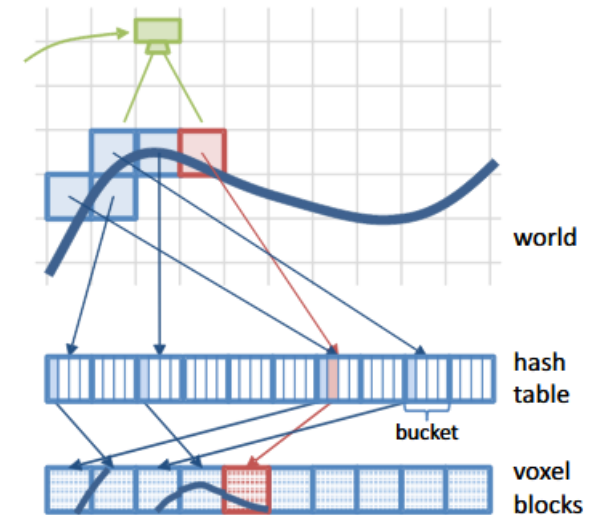
3 Newcastle University

4 Lancaster University

5 University of Toronto

Real-time 3D Reconstruction at Scale using Voxel Hashing

- Empty Voxels not stored
- 8x8x8 Voxelblocks
- Voxels 4mm each
- Efficient GPU algorithms for storage and retrieval



Dynamic Fusion

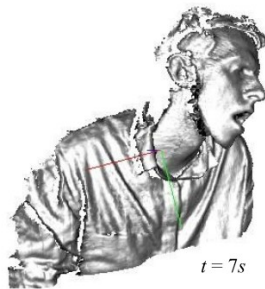
Richard A. Newcombe, Dieter Fox, Steven M. Seitz
University of Washington 2015



$t = 0s$



$t = 2s$



$t = 7s$



$t = 15s$



$t = 16s$



$t = 20s$

Dynamic Fusion

- Extends Kinect Fusion to non-rigid objects
- Store canonical (non-moved-pose) model
- Estimate Warp Field to transform canonical to live frame
- Refine canonical model and warping parameters

Dynamic Fusion

- Estimation of model-to-frame warp:
 - Extract Mesh of canonical model with marching cubes
 - Transform it with current warp field
 - Render it to a buffer (projective)
 - Compare to rendering of live depth map
 - Produces an Error term
- Add regularization Term and Deformations
- Solve with Gauss Newton

Dynamic Fusion



(a) Initial Frame at $t = 0s$



(b) Raw (noisy) depth maps for frames at $t = 1s, 10s, 15s, 20s$



(c) Node Distance



(d) Canonical Model

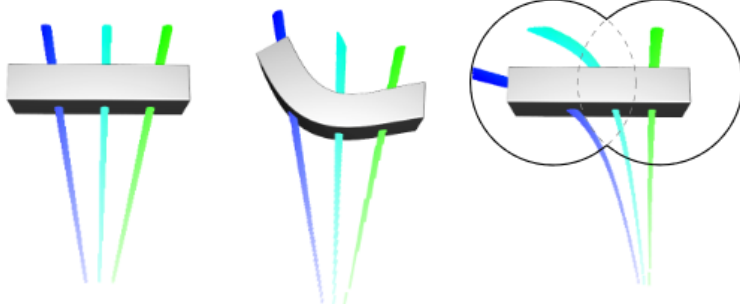


(e) Canonical model warped into its live frame

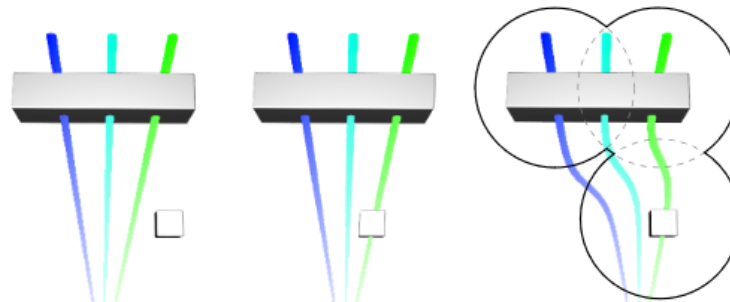


(f) Model Normals

Non-rigid scene deformation



Introducing an occlusion



Dynamic Fusion Video

*Dynamic***Fusion:**

Reconstruction & Tracking of Non-rigid Scenes in *Real-Time*

Richard Newcombe, Dieter Fox, Steve Seitz

Computer Science and Engineering,
University of Washington

Feature Based Methods

(e.g. ORB-SLAM2)

- Finding Depth

Image sequence



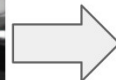
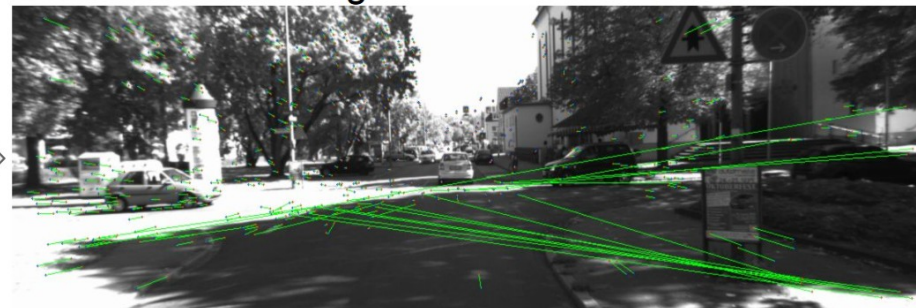
RANSAC, PnP



Feature detection

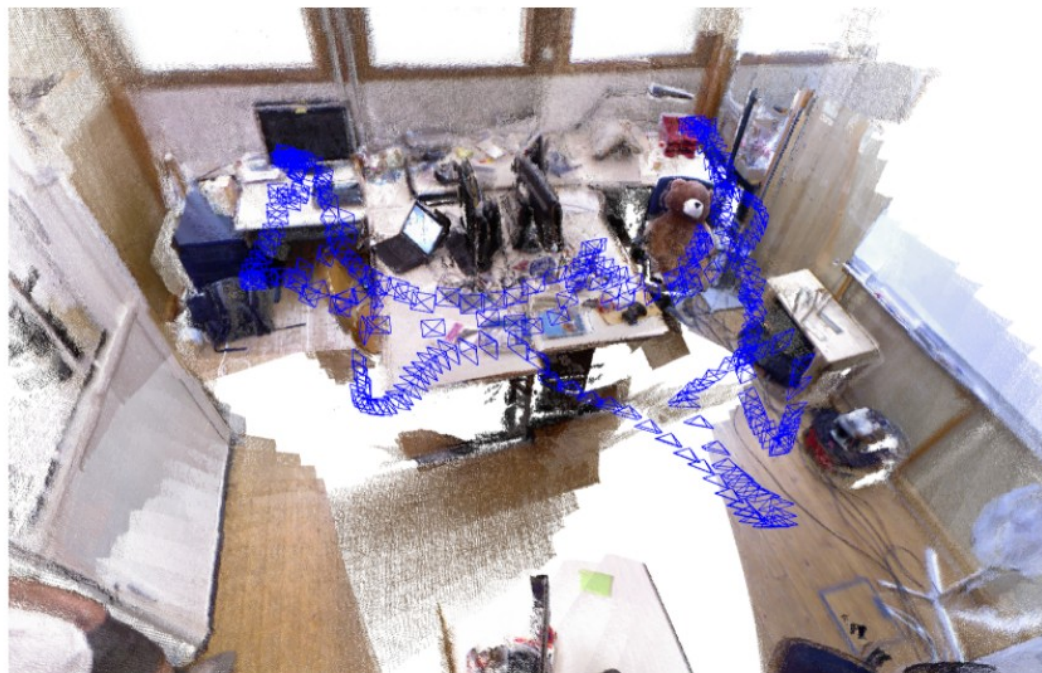


Feature matching



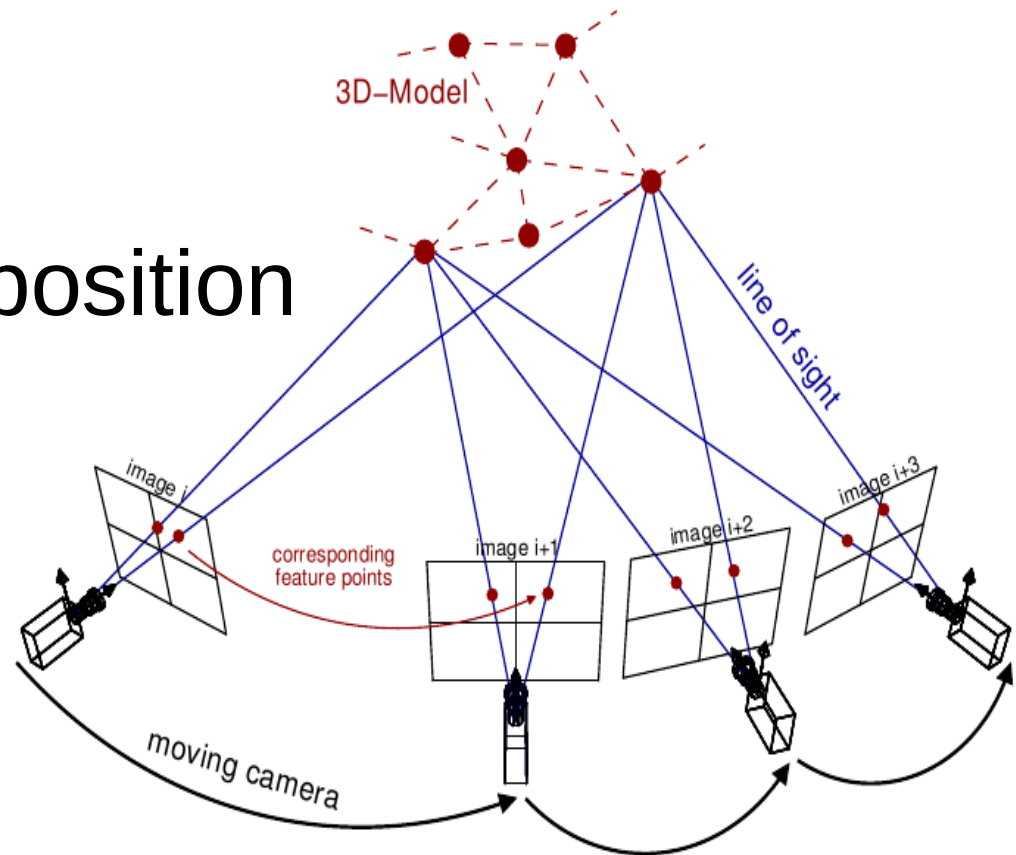
ORB-SLAM 2

Raúl Mur-Artal and Juan D. Tardós



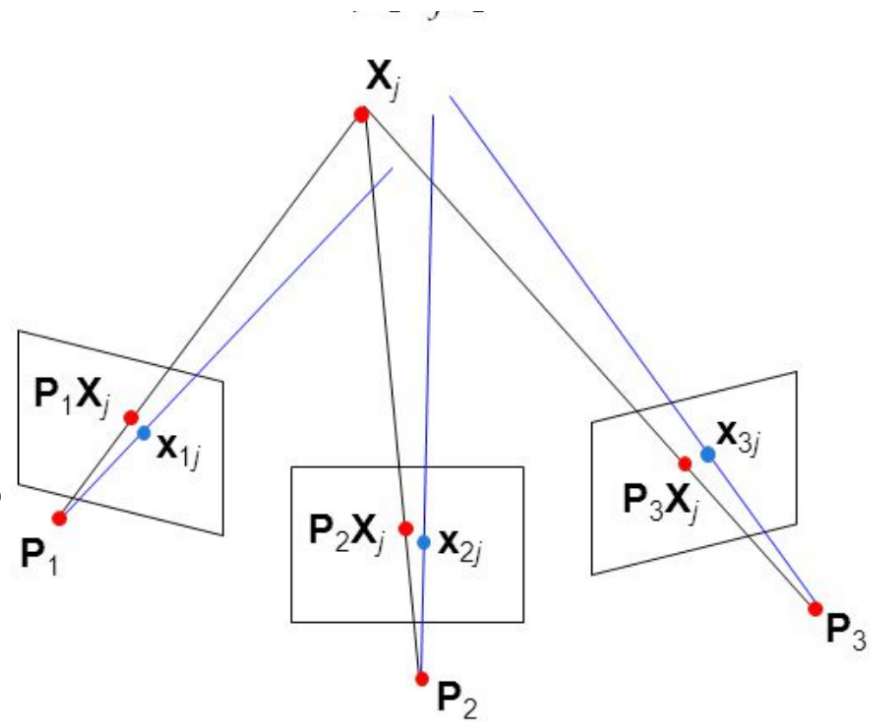
Camera Parameters

- Camera tracks points over multiple frames
- Approximate camera position and surrounding



Reprojection Error

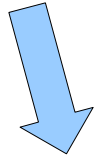
- X_j = predicted point
- x_j = point as seen from Camera P_i
- Modifying Intrinsic and Extrinsic Parameters decreases Error (distance X_j to x_j)



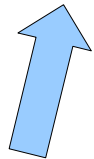
Camera Position & Depth

$$K = \begin{bmatrix} \alpha_x & \gamma & u_0 \\ 0 & \alpha_y & v_0 \\ 0 & 0 & 1 \end{bmatrix}$$

Intrinsic parameters
(focal length, pixel size, principal point)

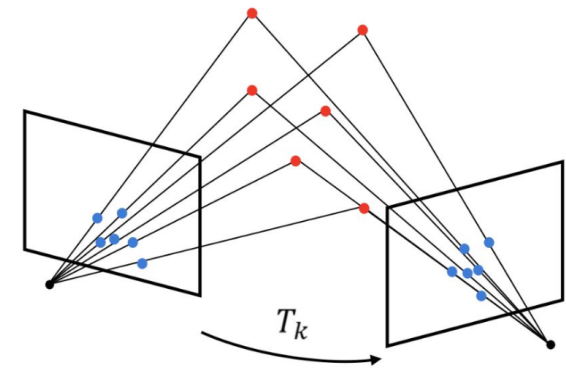
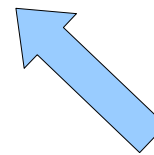


Linear System with Error metric to minimize estimates parameters



$$\begin{bmatrix} X_c \\ Y_c \\ Z_c \\ 1 \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

Extrinsic parameters
(rotation, translation)



Feature Points from
multiple Frames/Cameras

ORB-SLAM 2 Results



Universidad
Zaragoza



Instituto Universitario de Investigación
en Ingeniería de Aragón
Universidad Zaragoza

ORB-SLAM2: an Open-Source SLAM System
for Monocular, Stereo and RGB-D Cameras

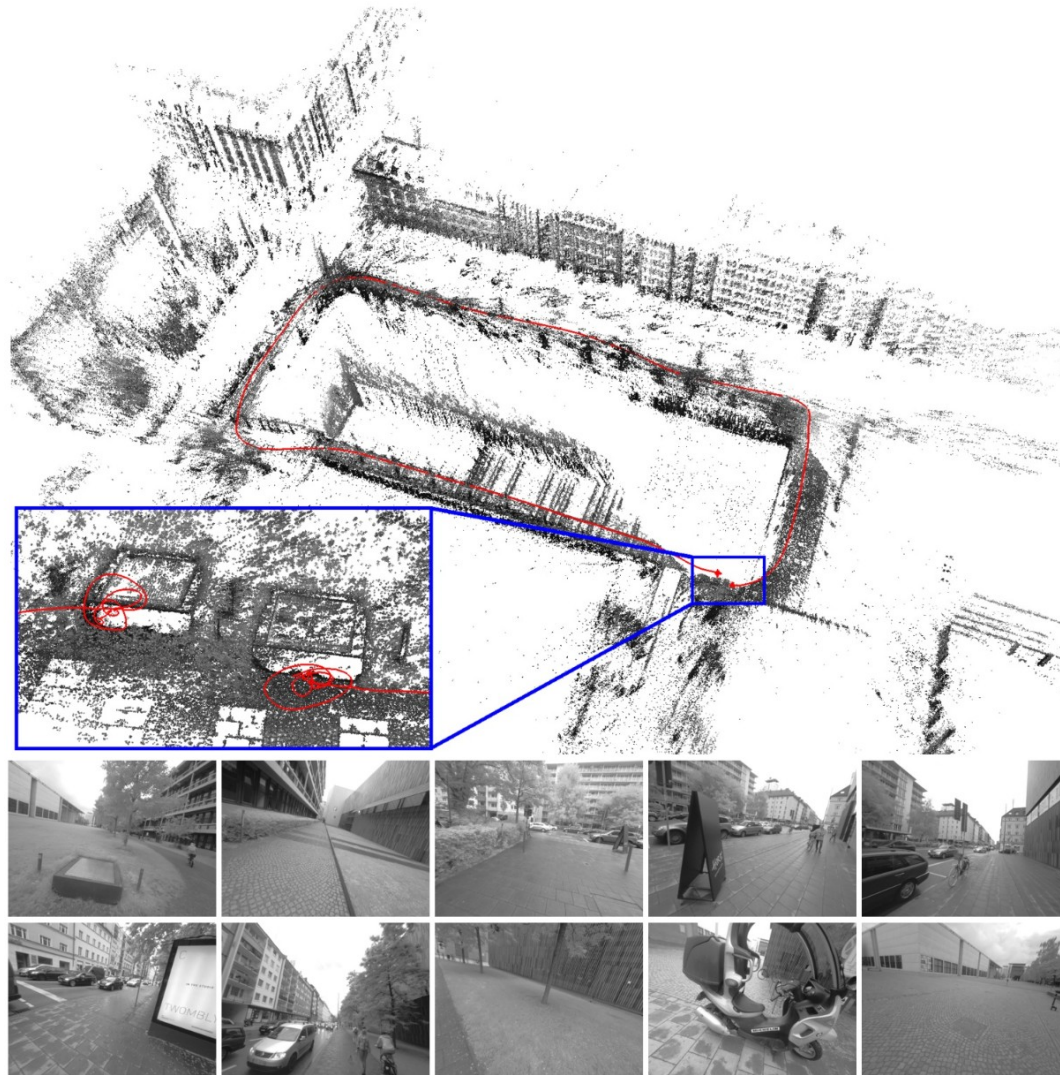
Raúl Mur-Artal and Juan D. Tardós

raulmur@unizar.es

tardos@unizar.es

Direct Sparse Odometry

Jakob Engel, Vladen Koltun, Daniel Cremers
TU Munich 2016



Direct Methods

- Instead of Feature Points, Image Intensities are tracked



-4.8 s



-3.9 s



-3.1 s



-2.2 s



-1.2 s



-0.8 s



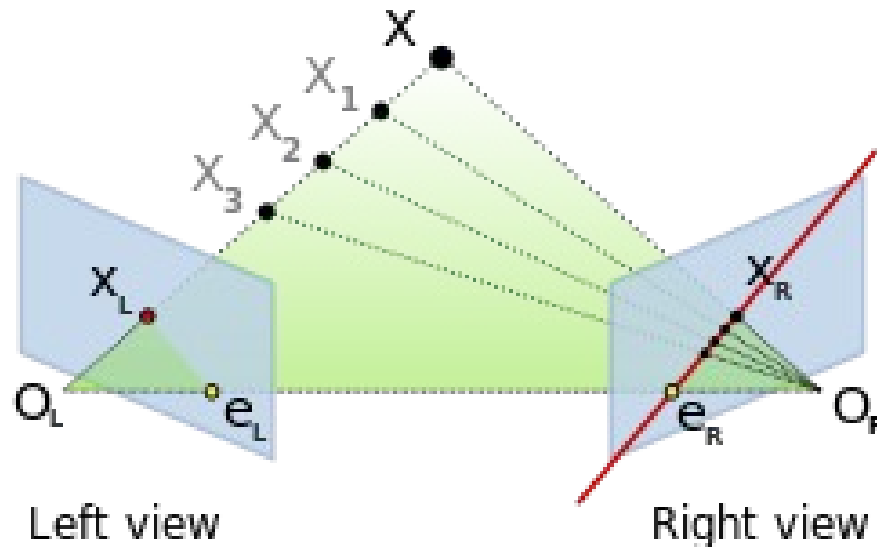
-0.5 s



-0.4 s

Direct Methods

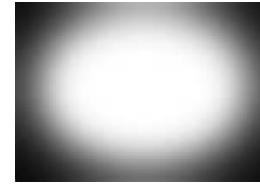
- Point X_L projected into space must be along the epipolar line X_R



Camera Position & Depth

$$K = \begin{bmatrix} \alpha_x & \gamma & u_0 \\ 0 & \alpha_y & v_0 \\ 0 & 0 & 1 \end{bmatrix}$$

Intrinsic parameters
(focal length, pixel size, principal point)



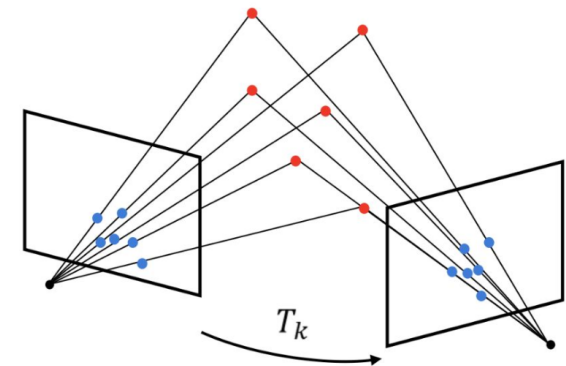
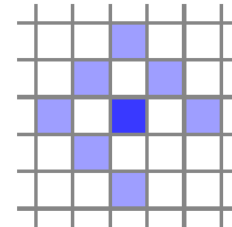
Vignetting

Exposure Time

Linear System with Error metric to minimize estimates parameters

$$\begin{bmatrix} X_c \\ Y_c \\ Z_c \\ 1 \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

Extrinsic parameters
(rotation, translation)



Feature Keypoints from
multiple Frames/Cameras

Camera Position

- Instead of Reprojection Error:
Minimizing Photometric Error
- Results in a System of Linear equations to
be solved by Gauss-Newton

Semi-Dense Visual Odometry for AR on a Smartphone

Semi-Dense Visual Odometry for AR on a Smartphone

Thomas Schöps, Jakob Engel, Daniel Cremers
ISMAR 2014, Munich

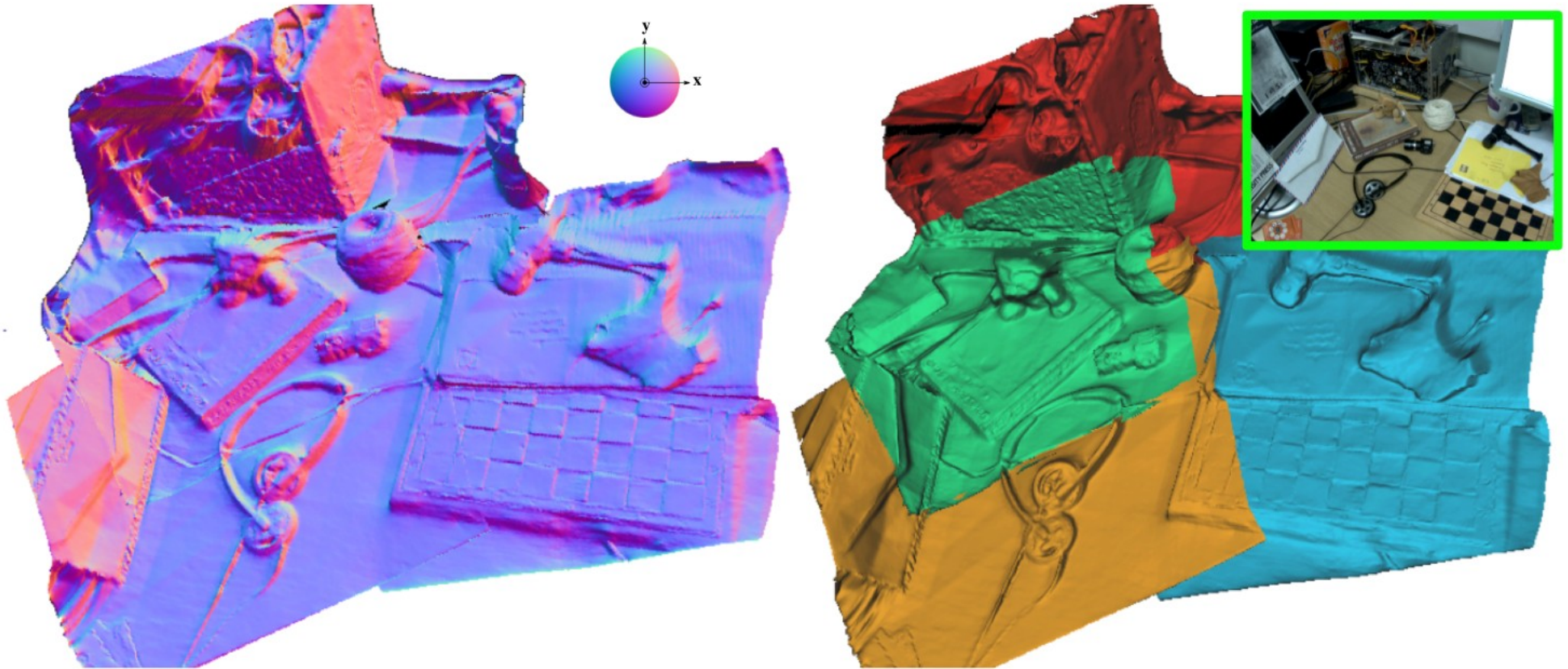


Computer Vision Group
Department of Computer Science
Technical University of Munich



Live Dense Reconstruction with a Single Moving Camera

Richard A. Newcombe and Andrew J. Davison
Imperial College London - 2010

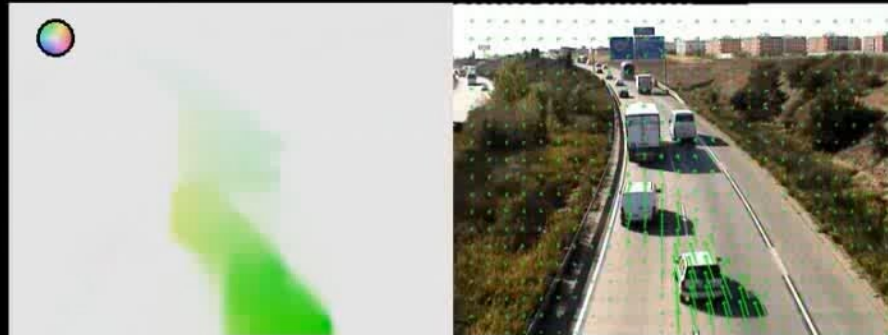


Live Dense Reconstruction with a Single Moving Camera

- Simple Mesh triangulation unsatisfactory
- Small movements of Camera add accuracy
- Refine model over time

Optical Flow

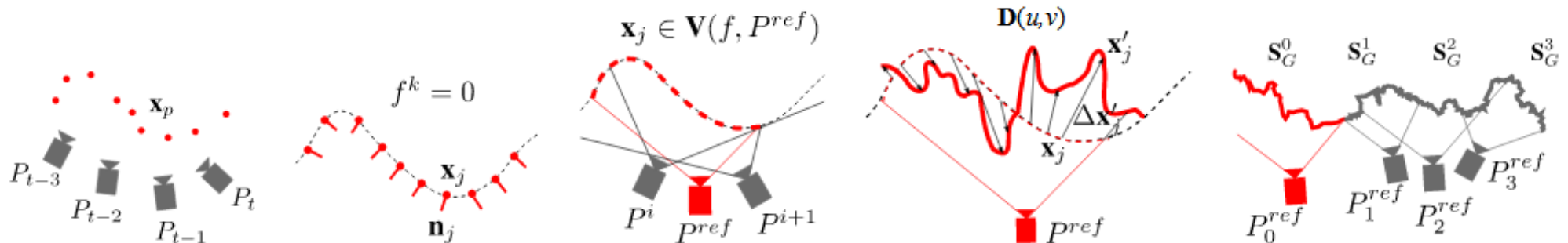
OpenCV CUDA Dense Optical Flow



Code: <http://www.robosafe.com/personal/pablo.alcantarilla/code.html>

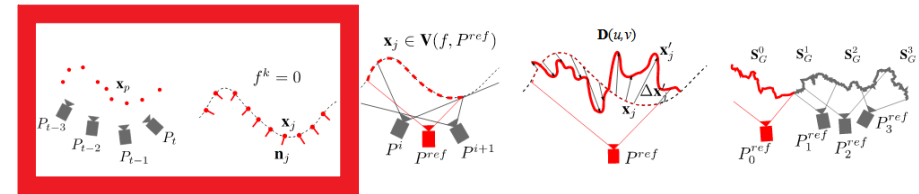
Live Dense Reconstruction with a Single Moving Camera

- Approximate Base Surface
- Refine Depth Estimations
- Fuse Depth Maps



Live Dense Reconstruction with a Single Moving Camera

- Find Feature Points (PTAM)
- Remove outliers
- Reconstruct coarse Base Mesh



Point Cloud



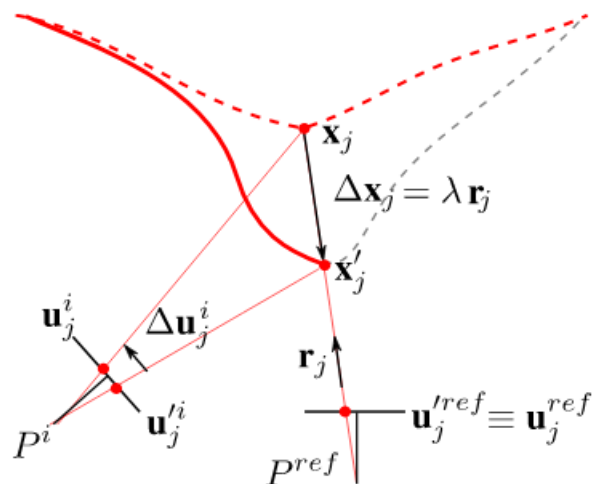
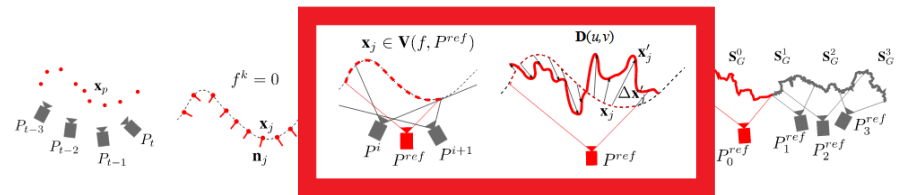
Polygon &
Landmarks



Homologous
shape model

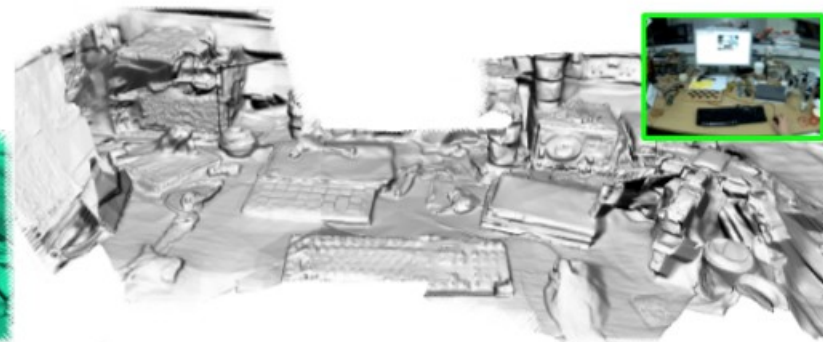
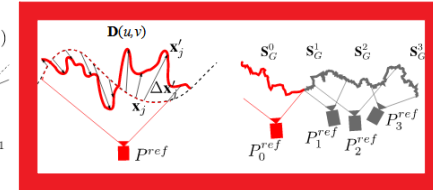
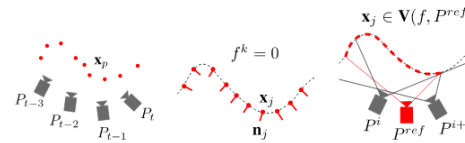
Live Dense Reconstruction with a Single Moving Camera

- Select Bundles of cameras with overlapping surface visibility
- For each pixel: Intersect ray with base model, predict position in bundle



Live Dense Reconstruction with a Single Moving Camera

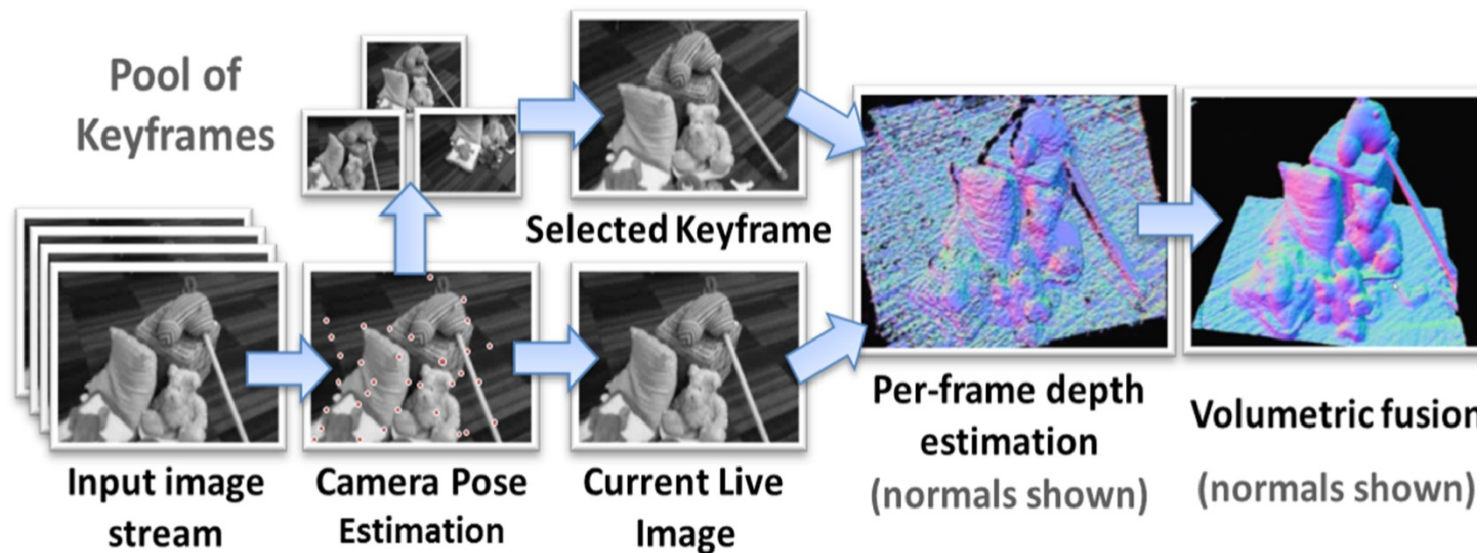
- Form linear system of predictions
- Find least squares solution and update surface
- Link up surface points to form depth map
- Integrate into Model



Monofusion

Vivek Pradeep, Christoph Rhemann, et al.
Cambridge 2013

- Depth estimation by baseline-matching
- Volumetric Fusion - SDF



Summary

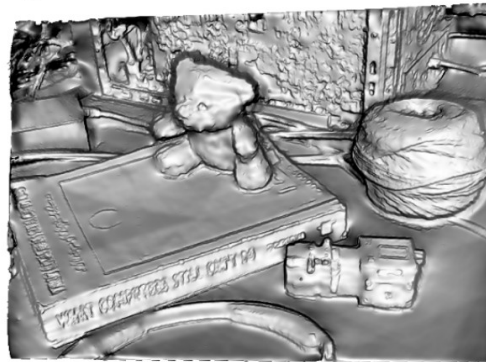
	Camera Type	Results in	Heavily Utilizes
Kinect Fusion	RGBD	Voxel Grid	GPU, RAM
Dynamic Fusion	RGBD	Voxel Grid + Warp field	GPU, RAM
Live Dense Reconstruction with a single moving camera	RGB	Mesh	CPU, GPU(optical flow)
Monofusion	RGB	Voxel Grid	

Comparison

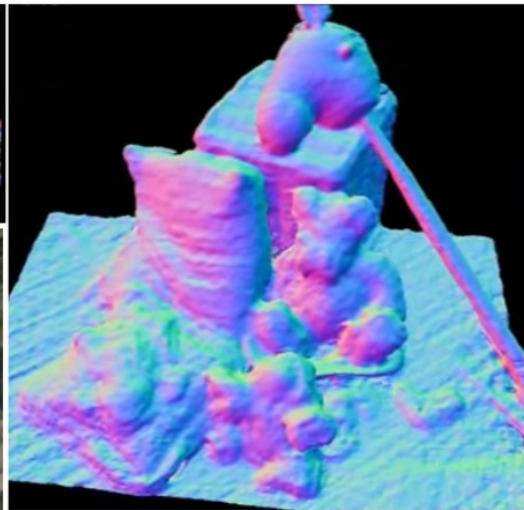
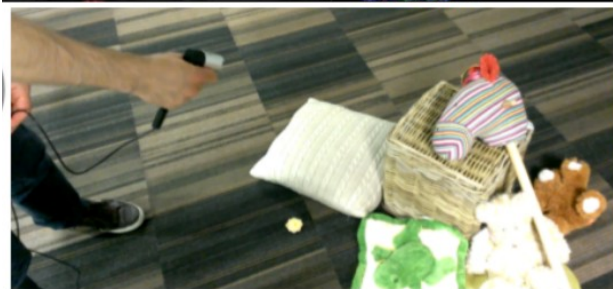
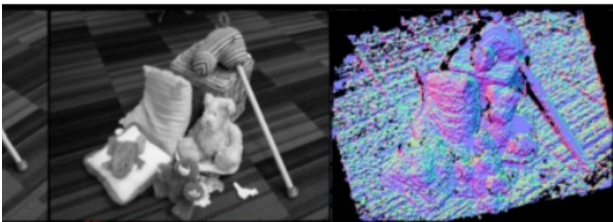
Kinect Fusion:



Live Dense Reconstruction:



Monofusion:



References

- Shahram Izadi, David Kim, Otmar Hilliges, David Molyneaux, Richard Newcombe, Pushmeet Kohli, Jamie Shotton, Steve Hodges, Dustin Freeman, Andrew Davison, and Andrew Fitzgibbon. 2011. KinectFusion: real-time 3D reconstruction and interaction using a moving depth camera. In Proceedings of the 24th annual ACM symposium on User interface software and technology (UIST '11). ACM, New York, NY, USA, 559-568. DOI: <https://doi.org/10.1145/2047196.2047270>
- R. A. Newcombe, D. Fox and S. M. Seitz, "DynamicFusion: Reconstruction and tracking of non-rigid scenes in real-time," 2015 IEEE Conference on Computer Vision and Pattern Recognition (CVPR), Boston, MA, 2015, pp. 343-352. doi: 10.1109/CVPR.2015.7298631
- Matthias Nießner, Michael Zollhöfer, Shahram Izadi, and Marc Stamminger. 2013. Real-time 3D reconstruction at scale using voxel hashing. ACM Trans. Graph. 32, 6, Article 169 (November 2013), 11 pages. DOI: <https://doi.org/10.1145/2508363.2508374>
- Mur-Artal, Raul; Tardos, Juan D. ORB-SLAM2: an Open-Source SLAM System for Monocular, Stereo and RGB-D Cameras, eprint arXiv:1610.06475
- Jakob Engel and (2016). Direct Sparse Odometry. CoRR, abs/1607.02565,
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