



# Recent Advances in 3D Computer Vision pre-meeting

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#### Outline

General Information
About the Seminar
Organization

Possible Papers
RGB-D SLAM
Stereo and Monocular SLAM
Dense Reconstruction
Dynamic Mapping
3D Deep Learning

## About the Seminar What will you learn in the seminar?

- Get an overview on fundamental techniques to solve problems in tracking and mapping
- Be able to read and understand scientific publications
- Prepare and give a talk
- Write a scientific report

#### About the Seminar

#### How do you need to prepare for the seminar?

- Please do not work on your topic completely alone
  - → meet at least twice with your supervisor
- Schedule:
  - 1 month before the talk: meet supervisor to discuss paper
  - 1 week before the talk: meet supervisor to discuss your slides
  - [optional] shortly after the talk: get feedback of your supervisor
  - 2 weeks after the talk: submit report

#### About the Seminar

#### How will your presentation be?

- ▶ 30 minutes talk, 10 minutes for questions
  - → make sure to finish on time!
- ▶ 1-2 minutes per slide → 15-30 slides, do not put too much information!
- Recommended structure:
  - Introduction, problem motivation, outline
  - Approach
  - Experimental results
  - Discussion
  - Summary of scientific contributions

## About the Seminar What about the final report?

Use LaTeX template and send final pdf via email to supervisor

► Length: 6-10 pages

Language: English

## About the Seminar What do we expect from you?

- Regular attendance is required!
- Active participation in the discussions
- Quality of the talk
- Quality of the report

#### Organization

#### How is the seminar organized?

- Weekly meetings are on Wednesdays from 2pm to 4pm
- Two presentations per week
- ▶ 14 participants → 7 weeks of presentations
- There might be an introductory meeting before the presentations start
- Starting date will be announced on web page once papers are assigned, probably end of May

#### Organization

#### Where do you find the list of papers?

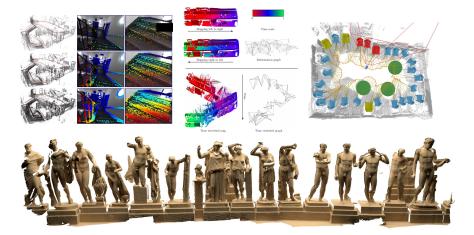
- ► Go to the seminar web page: https://vision.in.tum.de/teaching/ss2018/seminar\_3dcv
- ► Follow the link to the material section (will be added after pre-meeting)
- ► Login with the password seminar\_3dcv\_ss18

### Organization How do we assign papers?

- Register on TUMonline for the seminar and
- Send us an email with your 3 preferred papers
- We will match the TUMonline list with the emails we get and select candidates.
- You will get an email with our decision both in case of acceptance and in case of rejection.



### Paper of Interests

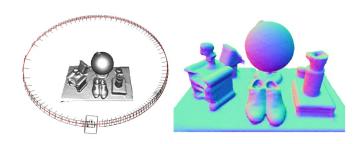


**RGB-D SLAM** 

### RGB-D SLAM static, rigid

### KinectFusion: Real-Time Dense Surface Mapping and Tracking

R. Newcombe, S. Izadi, O. Hilliges, D. Molyneaux D. Kim, A. Davison, P. Kohli, J. Shotton S. Hodges and A. Fitzgibbon





RGB-D SLAM static, rigid

### BundleFusion: Real-time Globally Consistent 3D Reconstruction using On-the-fly Surface Re-integration

A. Dai, M. Nießner, M. Zollhöfer, S. Izadi and C. Theobalt



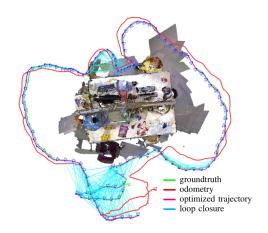




RGB-D SLAM static, rigid

### Dense Visual SLAM for RGB-D Cameras

C. Kerl, J. Sturm and D. Cremers





RGB-D SLAM static, rigid

# Real-Time Camera Tracking and 3D Reconstruction Using Signed Distance Functions

E. Bylow, J. Sturm, C. Kerl, F. Kahl and D. Cremers

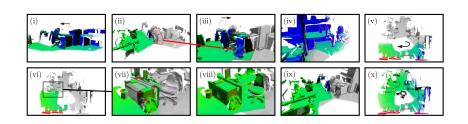




### RGB-D SLAM static, rigid

# ElasticFusion: Real-Time Dense SLAM and Light Source Estimation

T. Whelan, R. F. Salas-Moreno, B. Glocker A. Davison and S. Leutenegger

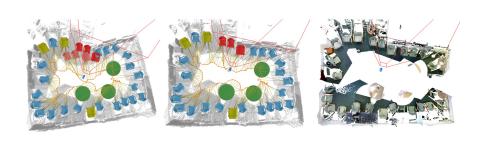


#### RGB-D SLAM

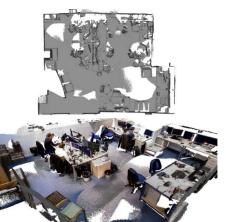
static, rigid, semantic

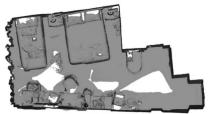
SLAM++: Simultaneous Localisation and Mapping at the Level of Objects

R. F. Salas-Moreno, R. Newcombe, H. Strasdat, P. Kelly and A. Davison



### RGB-D SLAM video demo







Dense Reconstruction

Dense Reconstruction dense, RGB-D, efficient storage

# Real-time 3D Reconstruction at Scale using Voxel Hashing

M. Nießner, M. Zollhöfer, S. Izadi and M. Stamminger





#### Dense Reconstruction color refinement

### Color Map Optimization for 3D Reconstruction with **Consumer Depth Cameras**

Q.-Y. Zhou and V. Koltun







Input

Optimized reconstruction

### Dense Reconstruction dense, RGB-D

### Robust Reconstruction of Indoor Scenes

S. Choi, Q.-Y. Zhou and V. Koltun

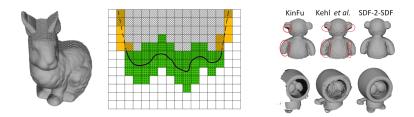




### Dense Reconstruction dense, depth only

SDF-2-SDF: Highly Accurate 3D Object Reconstruction

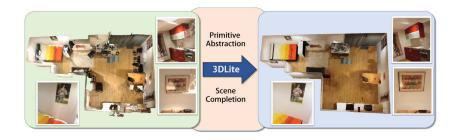
M. Slavcheva, W. Kehl, N. Navab and S. Ilic



## Dense Reconstruction abstraction, efficient storage

# 3DLite: Towards Commodity 3D Scanning for Content Creation

J. Huang, A. Dai, L. Guibas and M. Nießner



Dynamic Mapping

## Dynamic Mapping non-rigid

# VolumeDeform: Real-time Volumetric Non-rigid Reconstruction

M. Innmann, M. Zollhöfer, M. Nießner C. Theobalt and M. Stamminger



### Dynamic Mapping non-rigid

### DynamicFusion: Reconstruction and Tracking of Non-rigid Scenes in Real-Time

R. Newcombe, D. Fox and S. Seitz









(e) Canonical model warped into its live frame





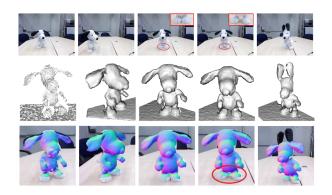


(f) Model Normals

# Dynamic Mapping non-rigid

KillingFusion: Non-rigid 3D Reconstruction without Correspondences

M. Slavcheva, M. Baust, D. Cremers and S. Ilic



## Dynamic Mapping non-rigid

# Fusion4D: Real-time Performance Capture of Challenging Scenes

M. Dou, S. Khamis, Y. Degtyarev, P. Davidson S. Fanello, A. Kowdle, S. Escolano, C. Rhemann D. Kim, J. Taylor, P. Kohli, V. Tankovich and S. Izadi



Stereo and Monocular SLAM

# ORB-SLAM: a Versatile and Accurate Monocular SLAM System

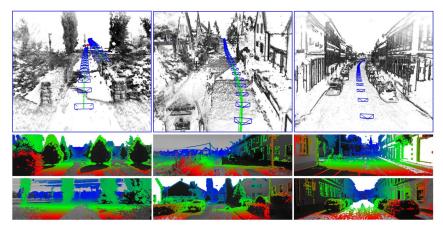
R. Mur-Artal, J. Montiel, and J. Tardós





### Large-Scale Direct SLAM with Stereo Cameras

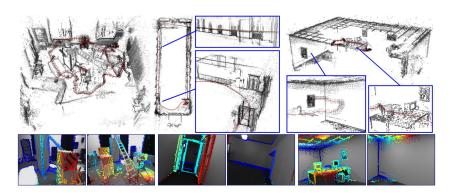
J. Engel, J. Stueckler and D. Cremers



## Stereo and Monocular SLAM monocular, direct, sparse

### **Direct Sparse Odometry**

J. Engel, V. Koltun and D. Cremers





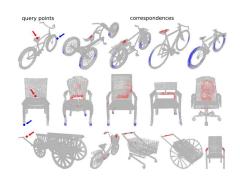
Deep Learning

#### **Deep Learning**

#### 3D feature, correspondences

# 3DMatch: Learning Local Geometric Descriptors from RGB-D Reconstructions

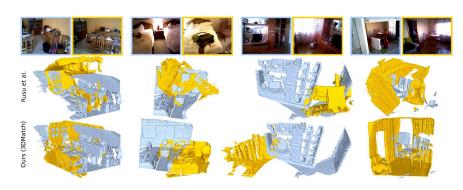
A. Zeng, S. Song, M. Nießner, M. Fisher, J. Xiao and T. Funkhouser



#### Deep Learning

#### 3D feature, correspondences

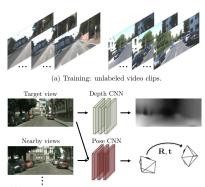
- keypoint matching
- correspondences over 3D meshes
- geometric registration

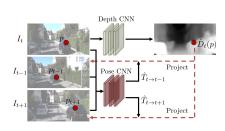


## Deep Learning depth and motion

# Unsupervised Learning of Depth and Ego-Motion from Video

T. Zhou, M. Brown, N. Snavely and D. G. Lowe





(b) Testing: single-view depth and multi-view pose estimation.

## Deep Learning depth and motion

- ► CNN
- unsupervised learning
- adapt canonical algorithms with deep learning

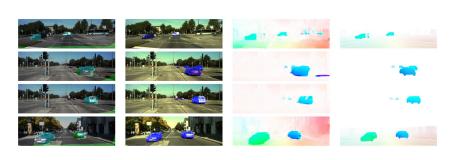




### Deep Learning CNN for SfM

## SfM-Net: Learning of Structure and Motion from Video

S. Vijayanarasimhan, S. Ricco and C. Schmid, R. Sukthankar and K. Fragkiadaki









#### Questions?

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