

# 3D Computer Vision

Radim Šára    Martin Matoušek

Center for Machine Perception  
Department of Cybernetics  
Faculty of Electrical Engineering  
Czech Technical University in Prague

<https://cw.fel.cvut.cz/wiki/courses/tdv/start>

<http://cmp.felk.cvut.cz>

<mailto:sara@cmp.felk.cvut.cz>

phone ext. 7203

rev. September 26, 2023

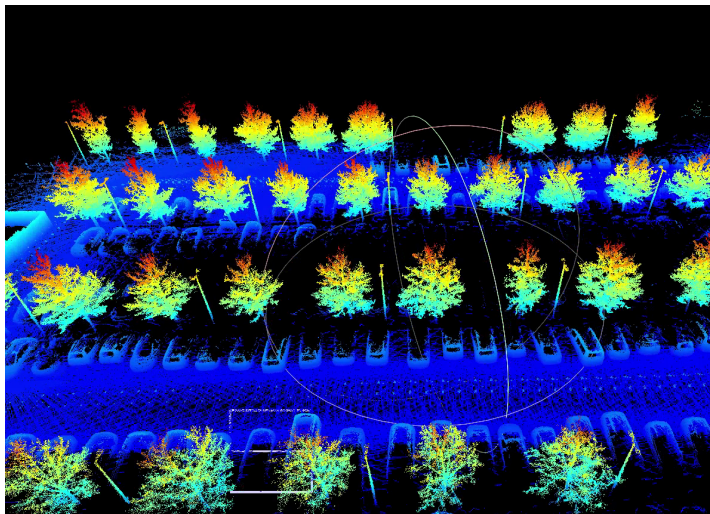


Open Informatics Master's Course

# Module I

## Course Overview

## 3D Vision is Not Just about 3D Point Clouds



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- today, we have laser-based rangefinders (eg. LiDARs)
- figure: point cloud obtained from a vehicle with 4 LiDARs on its roof
- **this course focuses on obtaining such results (and more) by means of passive sensors**

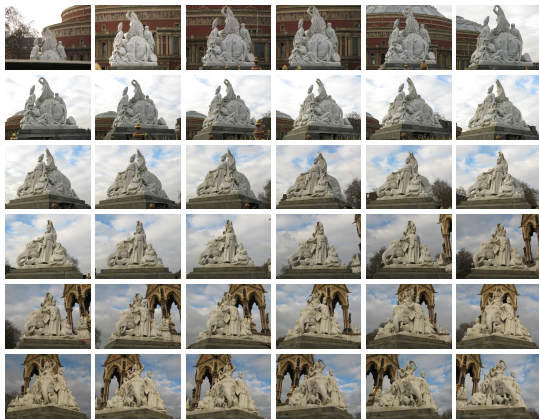
active sensors  
color = height

# This Course: Structure from Motion & Dense Point Clouds

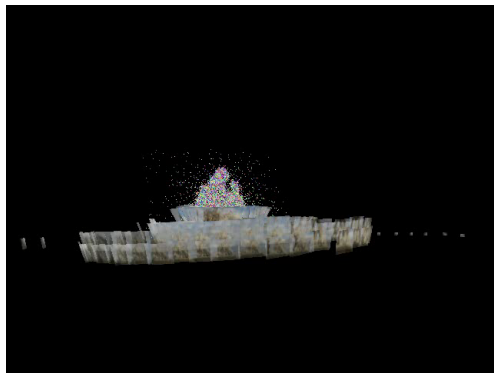
images + some knowledge about cameras



cameras in 3D + sparse 3D points



36 of 237 images of a memorial



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all camera poses, closest 2m, farthest 40m away

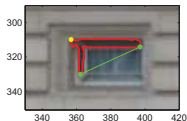
## Typical phases of a processing pipeline:

1. finding sparse image matches (Matching)
2. determining correspondences and camera poses (Structure from Motion)
3. finding dense correspondences → 3D point clouds (Stereovision)
4. (optional: occupancy mapping or surface reconstruction)

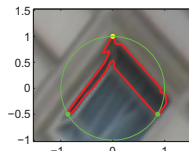
# Phase 1: Sparse Image Matches

image features, their descriptors, matches and correspondences

MSEDS



bitangent + distant pt

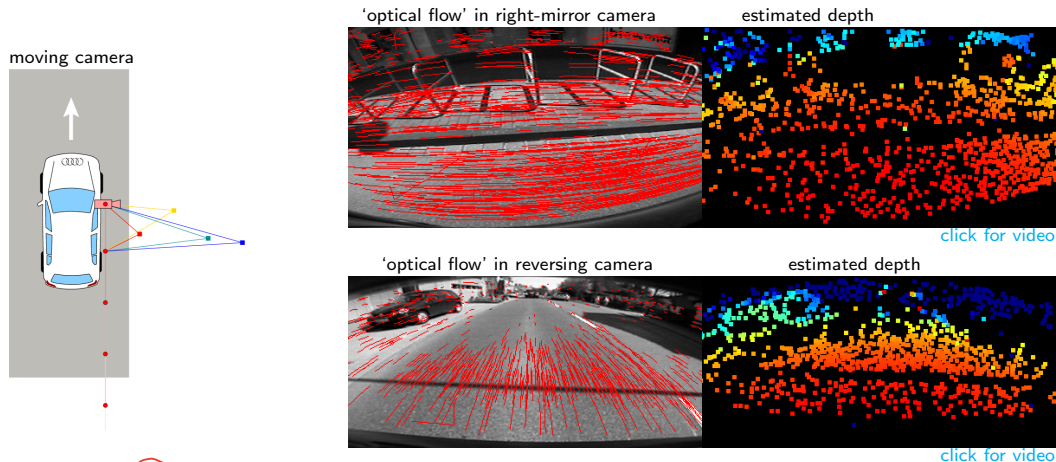


normalized feature

- matches  $\sim$  visually similar
- correspondences  $\sim$  visually similar and geometrically consistent (yellow)
- finding matches must cope with ambiguity
- 5 correspondences determine the relative angular orientation and translation direction between calibrated cameras  
calibrated = we know their internal parameters like the focal length etc

## Phase 2: From Matches To Correspondences (“Structure from Motion”)

- **Example:** Sensing depth from a single moving camera, 30 fps data stream  
standard automotive wide-angle sensor – reversing camera
- moving videocamera  $\sim$  time constraint on image match evolution  $\sim$  ‘optical flow’



- standard term: **SfM** (Structure from Motion should rather be Structure and Motion)
- single camera: problems with moving objects (wrong depth)

# Phase 2: Recovering Camera Poses ( = "Motion" )

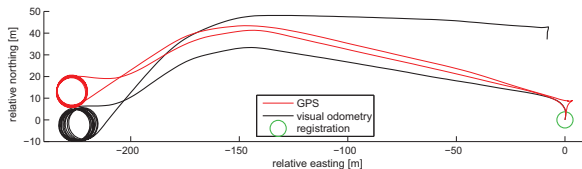
- reversing camera on a car, 30fps; error against RT 3000 GPS system (red)

no fusion with GPS!

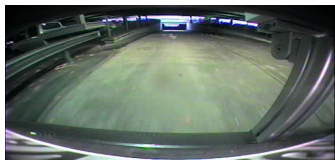
## Scene I



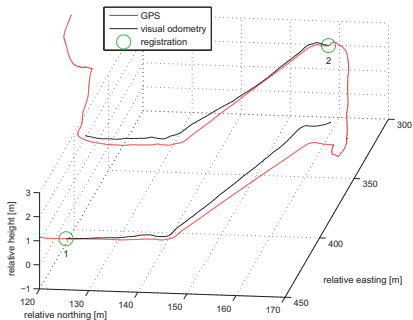
[click for video](#)



## Scene II



[click for video](#)



- 1 km, 5% accumulated drift
- measures elevation
- bad lighting conditions
- difficult scene

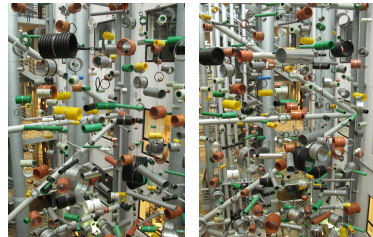
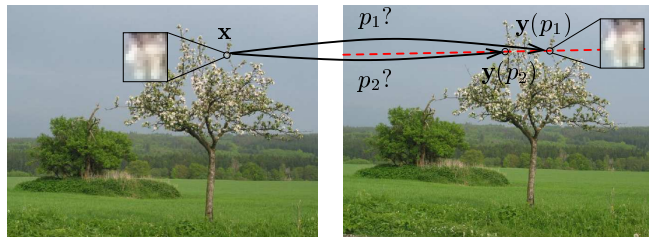
## Some applications:

- visual odometry
- SLAM
- the drift is reduced if the correspondences linking camera pairs form a dense graph, not a chain like here

# Phase 3: Dense Correspondences by Stereovision

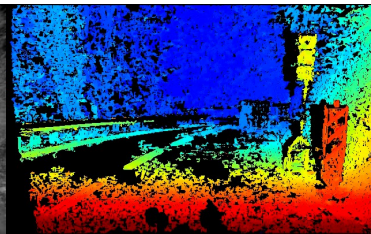
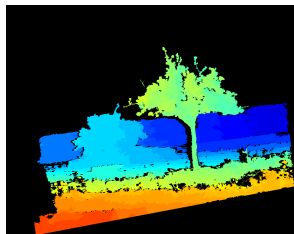
stereo seems easy in familiar scenes

unfamiliar scenes are difficult



input images

Malmö Högskola, Centrum för teknikstudier



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- the result is a dense 3D point cloud (color = range)

typically  $10^6 - 10^9$  3D points

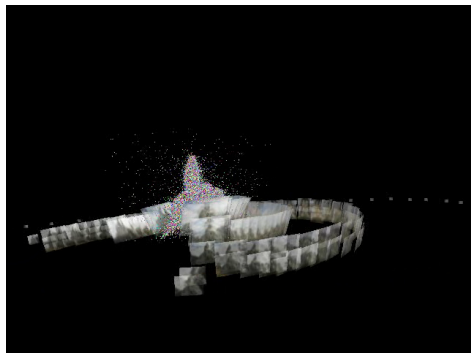


## Phase 4: Point Cloud and Surface Reconstruction

cameras + point cloud + images



triangulated surface



[click for video](#)



[click for video](#)

- we will not cover surface reconstruction in this course
- (but you may be able to use one of the popular algorithms)

Thank You



