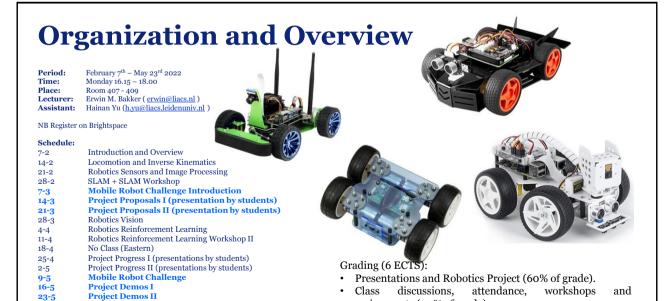
Robotics

Erwin M. Bakker| LIACS Media Lab

28-2 2022



Bij ons leer je de wereld kennen



assignments (40% of grade).

workshop and assignment.

It is necessary to be at every class and to complete every

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Website: http://liacs.leidenuniv.nl/~bakkerem2/robotics/



Period: February 7th - May 23rd 2022 Time: Monday 16.15 – 18.00 Place:

Room 407 - 409 Erwin M. Bakker (<u>erwin@liacs.nl</u>) Lecturer: Hainan Yu (h.yu@liacs.leidenuniv.nl) Assistant:

NB Register on Brightspace

Schedule:

Introduction and Overview 14-2 Locomotion and Inverse Kinematics 21-2 Robotics Sensors and Image Processing SLAM + SLAM Workshop 28-2 Mobile Robot Challenge Introduction 7-3

Project Proposals I (presentation by students) 21-3 Project Proposals II (presentation by students)

28-3 Robotics Vision

Robotics Reinforcement Learning 4-4 Robotics Reinforcement Learning Workshop II 11-4

18-4 No Class (Eastern)

Project Progress I (presentations by students) 25-4 Project Progress II (presentations by students)

Mobile Robot Challenge 2-5

Project Demos I Project Demos II

Website: http://liacs.leidenuniv.nl/~bakkerem2/robotics/



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- discussions, attendance, assignments (40% of grade).
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SLAM: Simultaneous Localization And Mapping

Mapping

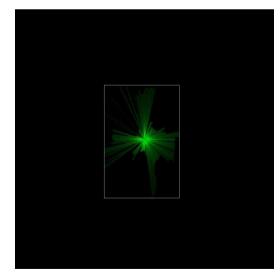
- Use sensor data: US, LIDAR, Camera, Structured Light, etc. to make a map of the environment of the robot.

Localization

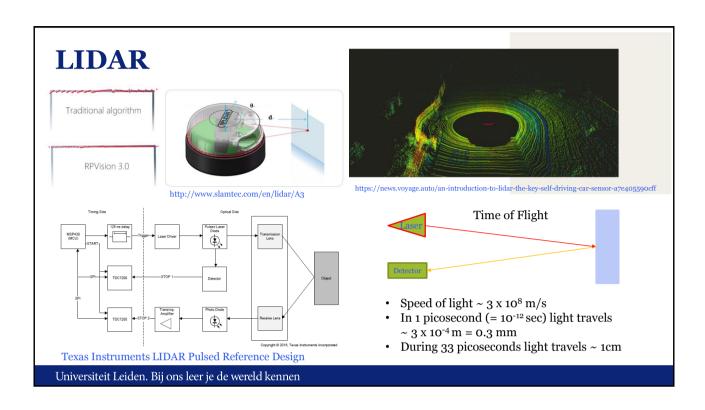
- Determine the pose of the robot relative to the map.
- Initial pose can be given: pose tracking problem
- No initial pose given: global localization problem

SLAM: do both Mapping and Localization at the same time.

VSLAM: use camera (mono, stereo, multi-, depth, etc.) to solve SLAM



MonsterBorg SLAM by E. van der Zande



CES2019

BMW Self Driving Car

InnovizOne Solid-state Lidar (goal: sub \$1000 sensor)

- Angular resolution $0.1^{\circ} \times 0.1^{\circ}$
- FOV 120° x 25°
- 25 FPS
- Range 250m



Perception Capabilities

- · Object detection and classification
- · Lane detection
- · Object Tracking
- SLAM



SLAM (Simultaneous Localization And Mapping)





IMU, or Inertial Measurement Unit

IMU consist of Sensors for

- orientation, by measuring the earths magnetic field and a gyroscope.
- · acceleration,
- · (angular) velocity.

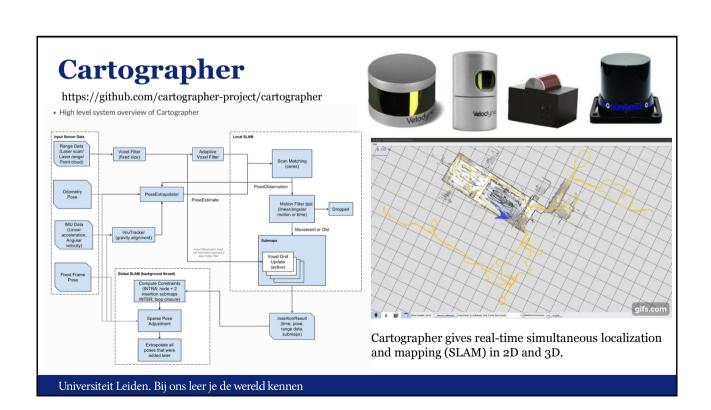
In SLAM

- · Orientation is very useful for scan matching: a good initial guess.
- Acceleration is calculated by subtracting the earth gravity vector from the raw accelerometer data.
- Velocity is calculated by time integration of the acceleration (tends to accumulate large errors)



Accelerometer: 1 KHzGyroscope: 523 HzMagnetometer: 30Hz





2D and 3D Lidar SLAM

Front-end:

- · Matching new measurements against the current belief of the environment
- · Measurements:
- typically point clouds of lidar data
- Movement and heading data: IMU, wheelencoders, etc.
- · Filtering of the data
 - Missing or invalid data (zero length rays)
 - Noisy data: measured too far away
 - Distorted data: e.g., as a result of robot acceleration
- Transform
 - Scan matching tries to find a rigid body transform: translation and rotation
 - IMU can give heading information: more reliable than deriving it from scan matching

(a) Robot scans at the first position Y_R and rotation
ag it from scan

Riadh Dhaoui, Ruhr-Universität Bochum

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2D and 3D Lidar SLAM: Scan Matching

Point to Point Matching

- Between the 2 scans point associations are found
- The translation from points to associated points determines a rigid body transform
- E.g found by Singular Value Decomposition
- Fails in case of degenerate clouds (coplanar)
- Point association is difficult to obtain

Iterative closest point (ICP) determination

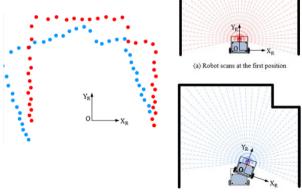
- Compute closest points
- Compute and apply the transformation
- Check if it is good enough, otherwise repeat

Tangent Matching

- In each of the 2 scans find Tangent lines between neighboring points
- Instead of point matching now line matching between 2 scans
- Movement and heading data: IMU, wheelencoders, etc.

Point to Grid Map

• Etc.



Riadh Dhaoui, Ruhr-Universität Bochum

SLAM

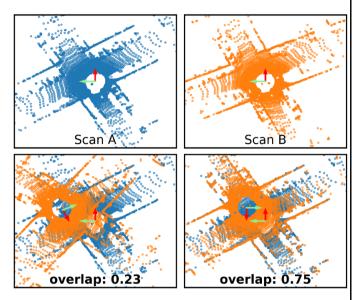
- 1) Pose estimation relative to recent poses: incremental scan matching, odometry
- Loop closing: correcting accumulated drift, maintaining consistency between measurements

Loop closure detection:

- Estimate lidar point clouds overlap (note: also used in photogrammetry):
 - Yaw transformation estimate
 - Score, e.g., using Iterative Closest Point
- Can also be used for global scan matching.

OverlapNet a DNN method that does an estimate without transformation guessing, uses spherical projection of 3D Lidar data.

Performance on Kitty-odometry dataset.



Example: Scan A and B overlap detection: scored after transformation.

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SLAM: Loop Closing OverlapNet

- 1) Pose estimation relative to recent poses: incremental scan matching, odometry
- 2) Loop closing: correcting accumulated drift

Loop closure detection:

- In general overlap scores based on pointcloud differences (see Eq. 3 of X Chen et al., May 2021) will be high at many locations of the map.
- OverlapNet a DNN method that does an estimate without transformation guessing, uses spherical projection of 3D Lidar data.

Example: Scan A and B overlap detection: scored after transformation.

(a) Exhaustive evaluation of Eq. (b) OverlapNet estimates

(c) Ground truth

scan B

scan B

overlap: 0.23

overlap: 0.75

OverlapNet

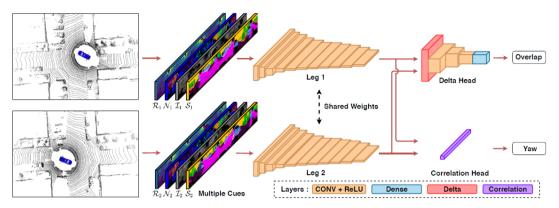


Fig. 3: Pipeline overview of our proposed approach. The left-hand side shows the preprocessing of the input data which exploits multiple cues generated from a single LiDAR scan, including range \mathcal{R} , normal \mathcal{N} , intensity \mathcal{I} , and semantic class probability \mathcal{S} information. The right-hand side shows the proposed OverlapNet which consists of two legs sharing weights and the two heads use the same pair of feature volumes generated by the two legs. The outputs are the overlap and relative yaw angle between two LiDAR scans.

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OverlapNet

X. Chen et al., OverlapNet: Loop Closing for Lidar-based SLAM, May 2021.

TABLE II: Comparison with state of the art.

Dataset	Approach	AUC	F1 score
KITTI	Histogram [26] M2DP [14] SuMa [2] Ours (AllChannel, TwoHeads)	0.83 0.83 - 0.87	0.83 0.87 0.85 0.88
Ford Campus	Histogram [26] M2DP [14] SuMa [2] Ours (GeoOnly)	0.84 0.84 - 0.85	0.83 0.85 0.33 0.84

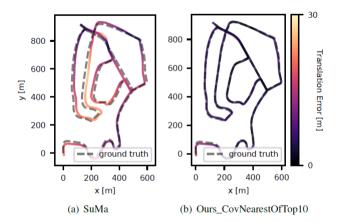


Fig. 6: Qualitative result on KITTI sequence 02.

Visual SLAM

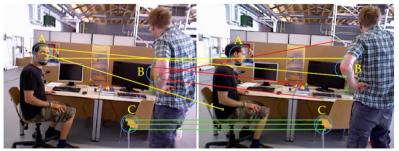
VSLAM: use camera (mono, stereo, multi-, depth, etc.) to solve SLAM

- Readily available on mobile platforms: drones, cars, mobile phones, AR/VR Glasses
- DMS-SLAM, OpenVSLAM, ORBSLAM

DMS-SLAM:

- pose tracking, closed-loop detection and relocalization based on static 3D map points of the local map
- supports monocular, stereo and RGB-D visual sensors in dynamic scenes.





A and B partial Motion, C static => A and B matching errors, C still correct.

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Visual SLAM

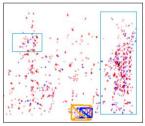
VSLAM: use camera (mono, stereo, multi-, depth, etc.) to solve SLAM

• DMS-SLAM, OpenVSLAM, ORBSLAM



A and B partial Motion, C static => A and B matching errors, C still correct.





(b)



(c)



(d)

(a) ORB-SLAM2.

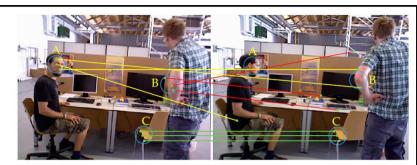
Red-boxes moving

DMS-SLAM

Visual SLAM

VSLAM: use camera (mono, stereo, multi-, depth, etc.) to solve SLAM

• DMS-SLAM, OpenVSLAM, **ORBSLAM**



A and B partial Motion, C static => A and B matching errors, C still correct.





(a) ORB-SLAM2

Blue-boxes wrong feature matching point pairs.

(b) DMS-SLAM

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(a) ORB-SLAM2

Sequence 01

(b) DMS-SLAM







(c) ORB-SLAM2

Sequence 03

(d) DMS-SLAM





(e) ORB-SLAM2

Sequence 09

(f) DMS-SLAM

Figure 15. The pose tracking experiments of ORB-SLAM2 and DMS-SLAM on the 01, 03, 09 sequences in the KITTI dataset. The rectangular box represents the moving object in the scene, and the others are static areas.

OpenVSLAM

A Versatile Visual SLAM Framework. (S. Sumikura et al., 2019)

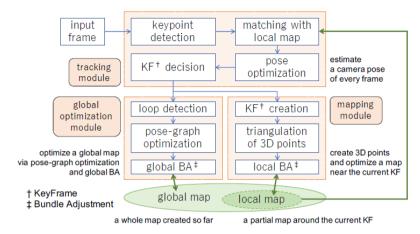
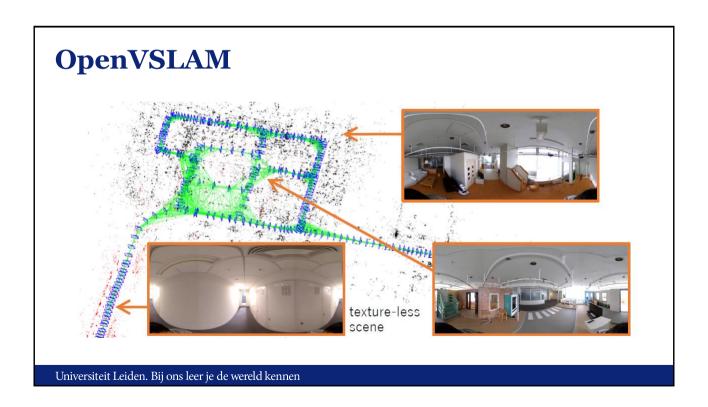


Figure 2: Main modules of OpenVSLAM: tracking, mapping, and global optimization modules.





 Period:
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 Monday 16.15 – 18.00

 Place:
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Assistant: Hainan Yu (h.yu@liacs.leidenuniv.nl)

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 9-5 Mobile Robot Challenge

9-5 Mobile Robot Cha 16-5 Project Demos I 23-5 Project Demos II

Website: http://liacs.leidenuniv.nl/~bakkerem2/robotics/





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Yeti Borg Introduction

YetiBorg

- Raspberry Pi Zero W
- 4 motors
- 1 Camera
- No odometry
- Programming and control through SSH
- Python and OpenCV
- No live camera footage from the YetiBorg
- Live view of environment through Kaltura



YetiBorg URBAN Challenge (previously: 2021)

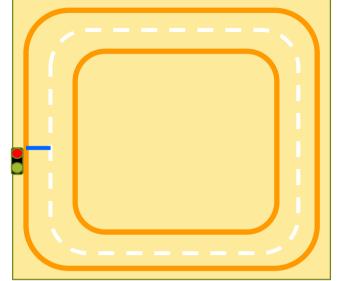
Urban Challenge

- Keep right
- Traffic Lights
- Traffic Signs
- Obstacles



Qualification Track

- Keep Right
- Traffic Lights
- Stop as close as possible before the blue line



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Mobile Robot Challenge Teams

- Form Mobile Robot Challenge Teams of 4-5 members.
- Determine a Team Name and 1 contact person.
- Further details will follow on Brightspace1.



Period: February 7th - May 23rd 2022 Time: Monday 16.15 - 18.00 Place: Room 407 - 409 Erwin M. Bakker (<u>erwin@liacs.nl</u>)

Lecturer: Assistant: Hainan Yu (h.yu@liacs.leidenuniv.nl)

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Mobile Robot Challenge **Project Demos I** Project Demos II

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Robotics Project Proposals Presentations

Monday 14-3 2022 and 21-3 2022

Present your Robotics Project Proposal during a 5 minute (max) talk. Clearly state the title of your project, the team members, your goals, how you will pursue them, what are the challenges and what at least can and should be delivered on the demo days: May 16th 2022 and May 23rd 2022.

Note: Groups of 1-5 members are allowed.

The presentation should contain slides for:

- 1. Title and group members.
- 2. Goal of the project.
- How will you pursue these goals.
- What are the challenges.
- What at least can and should be delivered on the demo days: May 16th 2022 and May 23rd 2022.

The LIACS Media Lab can support your project with some materials for your project. Please clearly state any materials that you would need for your proposal.

Please note that these materials are limited so project goals may need to be adjusted accordingly.

Each presentation will be followed by a short class discussion.

Previous Projects

- 1. Evolutionary Locomotion
- 2. Nao plays Tic-Tac-Toe
- 3. Slam Robot Project.
- 4. Dolphin Drone: Drone Recognition and Maneuvering 4. BorrelBot with Hoops.
- 5. Delivery Drone.
- 6. Programming a NAO to play a tune using a xylophone.
- 7. Floor mapping with Swarm Robotics
- 8. Tootballing Yetiborg
- 9. Cat Flap Opening Based on Audio/Video/RFID
- 10. DrawBot
- 11. Traffic coordination (simulation).
- 12. Plane filling curves (simulation).

- 1. AimBot
- 2. Artificial Muscles
- 3. Ball Tracking Car
- 5. Fetch Bot
- 6. Floor Mapping Robot
- 7. Gesture Control Pachenko
- 8. Hexapod
- 9. Nao Pose
- 10. Position Estimation
- 11. Race Car Training
- 12. Face Touch

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References

SLAM

- M. Sualeh, G.-W. Kim, Simultaneous Localization and Mapping in the Epoch of Semantics: A Survey, International Journal of Control, Automation and Systems 17(3) (2019) 729-742.
- W. Hess, D. Kohler, H. Rapp, D. Andor, Real-Time Loop Closure in 2D LIDAR SLAM, Published in: 2016 IEEE International Conference on Robotics and Automation (ICRA).
- X. Chen et al., OverlapNet: Loop Closing for Lidar-based SLAM, May 2021.
- E.R. van der Zande, An Introduction to 2D Lidar SLAM, December 2021.
- S. Sumikura et al., OpenVSLAM: A Versatile Visual SLAM Framework, 2019.
- G. Liu et al. DMS-SLAM: A General Visual SLAM System for Dynamic Scenes with Multiple Sensors, Sensors, 2019.

Yetiborg Challenge

- 1. https://navoshta.com/detecting-road-features/ by Alex Staravoitau
- 2. OpenCV.org