



# Computer Vision II: Multiple View Geometry (IN2228)

Chapter 00 Introduction

Dr. Haoang Li

19 April 2023 12:00-13:30





# Outline

- Course Information
- Concepts of Multiple View Geometry
- > Applications of Multiple View Geometry
- Overview of Course Content



> Instructors

Lecturer: Haoang Li

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### Teaching Assistants:



For an office visit, making an appointment beforehand is recommend.



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Lectures

### First Class (90 minutes)

- Time: every Wednesday from 12:00 to 13:30
- Room: 102, Hörsaal 2, "Interims I" (5620.01.102)
- 5-minute break

### Second Class (45 minutes)

- Time: every Thursday from 11:00 to 11:45
- Room: 00.02.001, MI HS 1, Friedrich L. Bauer Hörsaal (5602.EG.001)

### Exercises (120 minutes)

- Time: Wednesday from 16:00 to 18:00 (not every week)
- Room: 102, Hörsaal 2, "Interims I" (5620.01.102)
- Detailed schedule will be provided later





Lecture (90+45 minutes)

Ask questions or provide suggestions via Moodle <a href="https://www.moodle.tum.de/mod/forum/">https://www.moodle.tum.de/mod/forum/</a>

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Lemplatform Moodle Technische Universität München	τυπ
Welcome back, Haoang! 👏	
Meine Kursübersicht	
Semester	Einrichtung
SoSe 2023 🗸	Computation, Information and Technology
Computer Vision II: Multiple View Geometry (IN (SoSe 2023   Computation. Information and Technology)	12228)



Tentative Lecture Schedule

For updates, slides, and additional materials: https://cvg.cit.tum.de/teaching/ss2023/cv2

#### 90-minute course; 45-minute course

Foundation

Wed 19.04.2023 Chapter 00: Introduction Thu 20.04.2023 Chapter 01: Mathematical Backgrounds

Wed 26.04.2023 Chapter 02: Motion and Scene Representation (Part 1) Thu 27.04.2023 Chapter 02: Motion and Scene Representation (Part 2)

Wed 03.05.2023 Chapter 03: Image Formation (Part 1) Thu 04.05.2023 Chapter 03: Image Formation (Part 2)

Wed 10.05.2023 Chapter 04: Camera Calibration Thu 11.05.2023 Chapter 05: Correspondence Estimation (Part 1)

Wed 17.05.2023 Chapter 05: Correspondence Estimation (Part 2) Thu 18.05.2023 No lecture

Wed 24.05.2023 Chapter 06: 2D-2D Geometry (Part 1) Thu 25.05.2023 Chapter 06: 2D-2D Geometry (Part 2)

#### Core part

Wed 31.05.2023 Chapter 06: 2D-2D Geometry (Part 3) Thu 01.06.2023 Chapter 07: 3D-2D Geometry (Part 1)

Wed 07.06.2023 Chapter 07: 3D-2D Geometry (Part 2) Thu 08.06.2023 No lecture

Wed 14.06.2023 Chapter 08: 3D-3D Geometry (Part 1) Thu 15.06.2023 Chapter 08: 3D-3D Geometry (Part 2)

Wed 21.06.2023 Chapter 09: Single-view Geometry (Part 1) Thu 22.06.2023 Chapter 09: Single-view Geometry (Part 2)

Wed 28.06.2023 Chapter 10: Bundle Adjustment and Optimization (Part 1) Thu 29.06.2023 Chapter 10: Bundle Adjustment and Optimization (Part 2)

Wed 05.07.2023 Chapter 11: Robot Estimation Thu 06.07.2023 Chapter 12: Photometric Error

Wed 12.07.2023 Chapter 13: SLAM and SFM (Part 1) Thu 13.07.2023 Chapter 13: SLAM and SFM (Part 2)

Wed 19.07.2023 Chapter 14: Advanced Topics (Part 1) Thu 20.07.2023 Chapter 14: Advanced Topics (Part 2) Advanced topics and high-level task



### Tentative Exercise Schedule

- ✓ Please note that exercise sessions may be re-scheduled depending on the course progress.
- ✓ Details of content will be provided by our teaching assistants.

 Wed 26.04.2023
 Exercise 1: Mathematical Background

 Wed 03.05.2023
 Exercise 2: Mathematical Background

 Wed 10.05.2023
 Exercise 3: Representing a Moving Scene

 Wed 24.05.2023
 Exercise 4: Perspective Projection

 Wed 31.05.2023
 Exercise 5: Lucas-Kanade Method

 Wed 14.06.2023
 Exercise 6: Reconstruction from two views

 Wed 21.06.2023
 Exercise 7: Reconstruction from multiple views

 Wed 05.07.2023
 Exercise 8: Direct Image Alignment

 Wed 12.07.2023
 Exercise 9: Direct Image Alignment

Teaching assistants will design 5-6 coding assignments on C++ or Python.

➢ References



### Course

"Computer Vision II" provided by Prof. Daniel Cremers
Materials: <u>https://cvg.cit.tum.de/teaching/ss2022/mvg2022</u>
Video: <u>https://www.youtube.com/playlist?list=PLTBdjV\_4f-EJn6udZ34tht9EVIW7lbeo4</u>

Prof. Daniel Cremers (on sabbatical at Oxford University)

	Knowledge	Formal Mathematical Definitions	Intuitive Illustrations
Prof. Cremers	$\checkmark$	$\checkmark$	
This year	$\checkmark$		✓

• "Vision Algorithms for Mobile Robotics" provided by Prof. Davide Scaramuzza Materials: <u>https://rpg.ifi.uzh.ch/teaching.html</u>



References 

### Book

"Multiple View Geometry in Computer Vision": R. Hartley and A. Zisserman ٠ Link: https://www.robots.ox.ac.uk/~vgg/hzbook/

"An Invitation to 3D Vision": Y. Ma, S. Soatto, J. Kosecka, S.S. Sastry ٠ Link: https://www.eecis.udel.edu/~cer/arv/readings/old\_mkss.pdf

Treat them as further reading for

- Formal definition and rigorous derivation ٠
- Additional knowledge ٠















Acknowledgement

### Source of figures, tables, and videos

- ✓ Internet, e.g., Youtube
- ✓ Academic papers
- $\checkmark$  This course partly refers to the slides of the following courses:
- University of Zurich "Vision Algorithms for Mobile Robotics"
- University of Washington "Computer Vision"

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Prerequisite Course and Knowledge

### There is not a mandatory requirement that you should take a certain prerequisite course.

Some recommendations

- ✓ TUM course "Computer Vision I: Variational Methods"
- Desirable but not essential

Link: <u>https://cvg.cit.tum.de/teaching/online/cvvm</u>

✓ Any "Advanced Mathematics" and "Linear Algebra" courses



Prerequisites Knowledge

### ✓ Linear Algebra and Matrix Calculus

- Necessary (will be reviewed)
- Reference: "Linear Algebra Primer" from Stanford University

Link: http://vision.stanford.edu/teaching/cs131\_fall1617/lectures/lecture2\_linalg\_review\_cs131\_2016.pdf

### ✓ Image Processing

- Desirable but not essential
- Reference: "Digital Image Processing": R. C. Gonzalez and R. E. Woods

Link: <u>http://sdeuoc.ac.in/sites/default/files/sde\_videos/Digital%20Image%20Processing%203rd%20ed.%20-%20R.%20Gonzalez%2C%20R.%20Woods-ilovepdf-compressed.pdf</u>



- Learning Objectives
- $\checkmark$  To understand the concepts and knowledge for recovering 3D shape from images
- ✓ Explore classical and state-of-the-art approaches in multiple view geometry
- ✓ Implement fundamental multiple view geometry algorithms in person





Grading

### The final grade is based on a final exam and assignment bonus.

• For example, for a student whose grade on the final exam is 1.3. If he/she obtained the bonus, his/her final grade is 1.0.

### Exam

- Written and closed-book exam
- Details about the exam and example exam questions will be provided during the course

### Assignment bonus (optional)

- There will be 5-6 coding assignments.
- If all the assignment codes submitted by a student 1) pass the automatic checking system, and also 2) meet the evaluation standard, the student will obtain a bonus of 0.3. There is not a partial bonus, e.g. 0.1.
- Evaluation standard is defined by our teaching assistants.



Computer Vision

Automatic extraction of "meaningful" information from images and videos



Semantic information

Course "Computer Vision III: Detection, Segmentation and Tracking" provided by Dr. Nikita Araslanov







Computer Vision vs Computer Graphics



[1] Ben Mildenhall\*, Pratul P. Srinivasan\*, Matthew Tancik\*, Jonathan T. Barron, Ravi Ramamoorthi, and Ren Ng, "NeRF Representing Scenes as Neural Radiance Fields for View Synthesis", In ECCV, 2020. (Best Paper Honorable Mention)



Journals and Conferences in Computer Vision/Robotics

What papers should you read? Why do you need to read papers?

#### Journals:

- International Journal of Computer Vision (IJCV)
- IEEE Transactions on Pattern Analysis and Machine Intelligence (TPAMI)
- International Journal of Robotics Research (IJRR)
- IEEE Transactions on Robotics (TRO)

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#### Conferences:

- IEEE International Conference on Computer Vision (ICCV)
- IEEE Computer Vision and Pattern Recognition Conference (CVPR)
- European Conference on Computer Vision (ECCV)
- IEEE International Conference on Robotics and Automation (ICRA)
- IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)



#### Relatively high quality in general



- Multi-view Geometry
- ✓ A field studying the relationship between cameras, 3D structure, and image features when there are feature correspondences between images taken from varying viewpoints.
- ✓ The image features are usually feature points or lines.





Multi-view Geometry

Terminology Definition or Differentiation

✓ 2D-2D, 3D-2D, and 3D-3D correspondences



2D-2D correspondences



3D-2D correspondences

3D-3D correspondences



Multi-view Geometry

Terminology Definition or Differentiation

✓ Camera position, camera pose, and camera trajectory









Virtual Reality (VR)/Augmented Reality (AR)





Estimate the pose of the head/glasses

Tracked 2D-2D correspondences



Virtual Reality (VR)/Augmented Reality (AR)





### Autonomous Driving



3D map and camera trajectory

Tracked 3D-2D correspondences

Localize cars: 1) car collision warning, 2) lane departure warning, and 3) safety distance monitoring and warning



Autonomous Driving

KITTI dataset (Karlsruhe Institute of Technology and Toyota Technological Institute)





### Humanoid Robot



Dense 3D map



**Object-level completion** 

Localize the robot and reconstruct the 3D scene



Humanoid Robot

A demo video of robot dog





### Vacuum Cleaner



#### Navigate vacuum cleaner for path planning and obstacle avoidance



Vacuum Cleaner

A demo video





### Mars Rovers



Landing

2D-2D correspondence

navigation

Improve the reliability of autonomous landing on Mars and autonomously navigate the robot.



Mars Rovers

A demo video for navigation



From High Level

We will provide necessary knowledge used in Visual Odometry (VO), Visual Simultaneous Localization and Mapping (VSLAM), and Structure from Motion (SFM).



Relationship between VO, VSLAM, and SFM



### From High Level

VO is the process of incrementally estimating the pose of the vehicle by examining the changes that motion induces on the images of its onboard cameras





Matched/tracked features



Sequential images from one or more cameras attached to a moving vehicle

Camera trajectory (red) and 3D map (green)



From High Level

### VSLAM = VO + loop detection & closure

- VO is affected by noise
- SLAM guarantees global consistency







### An example of loop detection and correction

What if the loop does not exist?



From High Level

SFM is more general than SLAM and tackles the problem from unordered image sets.





#### Sparse reconstruction



#### From sparse to dense reconstruction



From High Level



Flow Chart of VO/VSLAM/SFM





#### Camera calibration



From High Level



Flow Chart of VO/VSLAM/SFM



Feature detection



Feature matching



From High Level



Flow Chart of VO/VSLAM/SFM







From High Level



Flow Chart of VO/VSLAM/SFM



Depth from Disparity



From High Level



Flow Chart of VO/VSLAM/SFM



**Bundle Adjustment** 



Optimization



Knowledge to Learn

### **Chapter 01: Mathematical Backgrounds**

- Vector operations, e.g., cross product and dot product
- Vector space, e.g., linear independence
- Matrix property, e.g., rank, trace, eigenvalue and eigenvectors
- Matrix decomposition, e.g.,., SVD



2w





Knowledge to Learn

### **Chapter 02: Motion and Scene Representation**

- Motion representation, e.g., rotation matrix
- Motion computation/operation, i.e., Lie group and Lie algebra
- 3D scene representation











(c) Implicit Surface

(d) Mesh



Knowledge to Learn

### **Chapter 03: Image Formation**

- Pinhole camera model and perspective projection
- Image coordinates and homogenous coordinates





Knowledge to Learn

### **Chapter 04: Camera Calibration**

- From 3D objects
- From planar grids









Knowledge to Learn

### **Chapter 05: Correspondence Estimation**

- Small motion: Optical flow, e.g., Lucas-Kanade Method
- Wide baseline: Descriptor computation and matching, e.g., SIFT and ORB







Knowledge to Learn

### Chapter 06: 2D-2D Geometry

- Epipolar geometry, e.g., 5-point method, 8-point method
- Stereo vision, e.g., stereo matching, relationship between disparity and depth





Knowledge to Learn

### Chapter 07: 2D-3D Geometry

- Perspective-n-Points
- Perspective-n-Lines (not presented in most multi-view geometry courses)





Knowledge to Learn

### Chapter 08: 3D-3D Geometry

- Iterative closest point (ICP) algoritm
- Closed-form algorithms, e.g., Horn's method and Umeyama algorithm





Knowledge to Learn

### **Chapter 09: Single-view Geometry**

- Vanishing point
- Single-view reconstruction







Knowledge to Learn

### **Chapter 10: Bundle Adjustment and Optimization**

- Re-projection error
- Gradient descent algorithms







Knowledge to Learn

### **Chapter 11: Robot Estimation**

- RANSAC
- M-Estimator







Knowledge to Learn

### **Chapter 12: Photometric Loss**

- Definition
- Applications







Knowledge to Learn

**Chapter 13: SLAM and SFM** 

Demo video of VO from our Computer Vision Group, TUM

This method uses the photometirc loss



Knowledge to Learn

**Chapter 13: SLAM and SFM** 

Demo video of VIO from Hong Kong University of Science and Technology

This system relies on the inertial measurement unit (IMU)



Knowledge to Learn

**Chapter 13: SLAM and SFM** 

Demo video of SLAM from University of Zaragoza

This method relies on the point features



Knowledge to Learn

**Chapter 13: SLAM and SFM** 

Demo video of SLAM from Shanghai Jiao Tong University

This method relies on the line features



Knowledge to Learn

**Chapter 13: SLAM and SFM** 

Demo video of SFM from Cornell University "Building Rome in a Day"

This method relies on the point features



Knowledge to Learn

**Chapter 13: SLAM and SFM** 

Demo video of SFM from ETH Zurich

This method relies on the line features



Knowledge to Learn

### **Chapter 14: Advanced Topics**

- Deep learning for geometric problems, e.g., deep feature-based matching featuremetric Loss
- Additional robust estimation algorithms, e.g., branch and bound
- Additional optimization problems, e.g., quadratically constrained quadratic program (QCQP)







# Summary

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- Concepts of Multiple View Geometry
- > Applications of Multiple View Geometry
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# Thank you for your listening! If you have any questions, please come to me :-)