



Vision Algorithms for Mobile Robotics

Lecture 01 Introduction

Davide Scaramuzza

<http://rpg.ifi.uzh.ch>

Today's Outline

- About me and my research lab
- What is Computer Vision?
- Why study Computer Vision?
- Example of vision applications
- Organization of the course
- Start: Visual Odometry overview

Who am I?

Current position



- Professor of Robotics & computer vision since 2012
- Dep. of Informatics (UZH) and Neuroinformatics (UZH & ETH)
- Director of the Master program in Artificial Intelligence at the Dep. of Informatics (UZH)
- Adjunct Professor of the ETH Master in Robotics, Systems and Control and Associate faculty of the ETH AI Center

Education



- Master in Electronics Engineering at the University of Perugia, Italy, 2004
- PhD in Robotics and Computer Vision at ETH Zurich, Switzerland, 2008
- Post-doc at the University of Pennsylvania, USA
- Visiting professor at Stanford University, 2019



Book

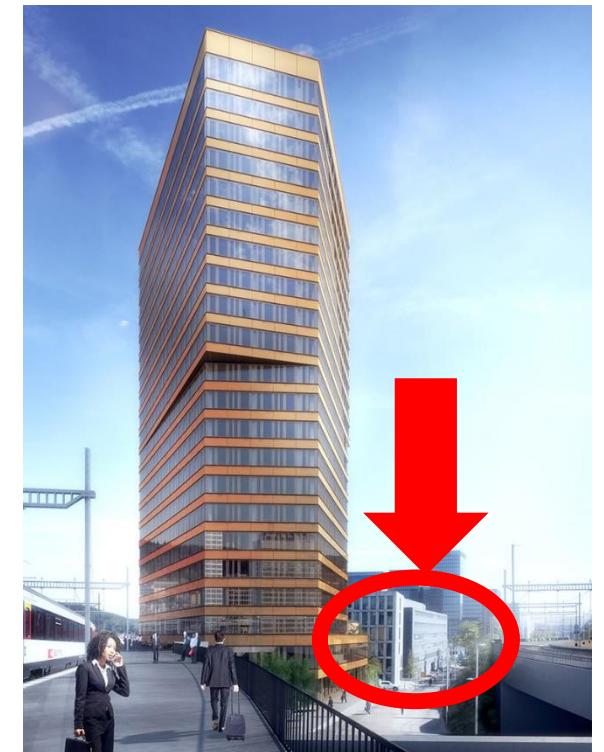
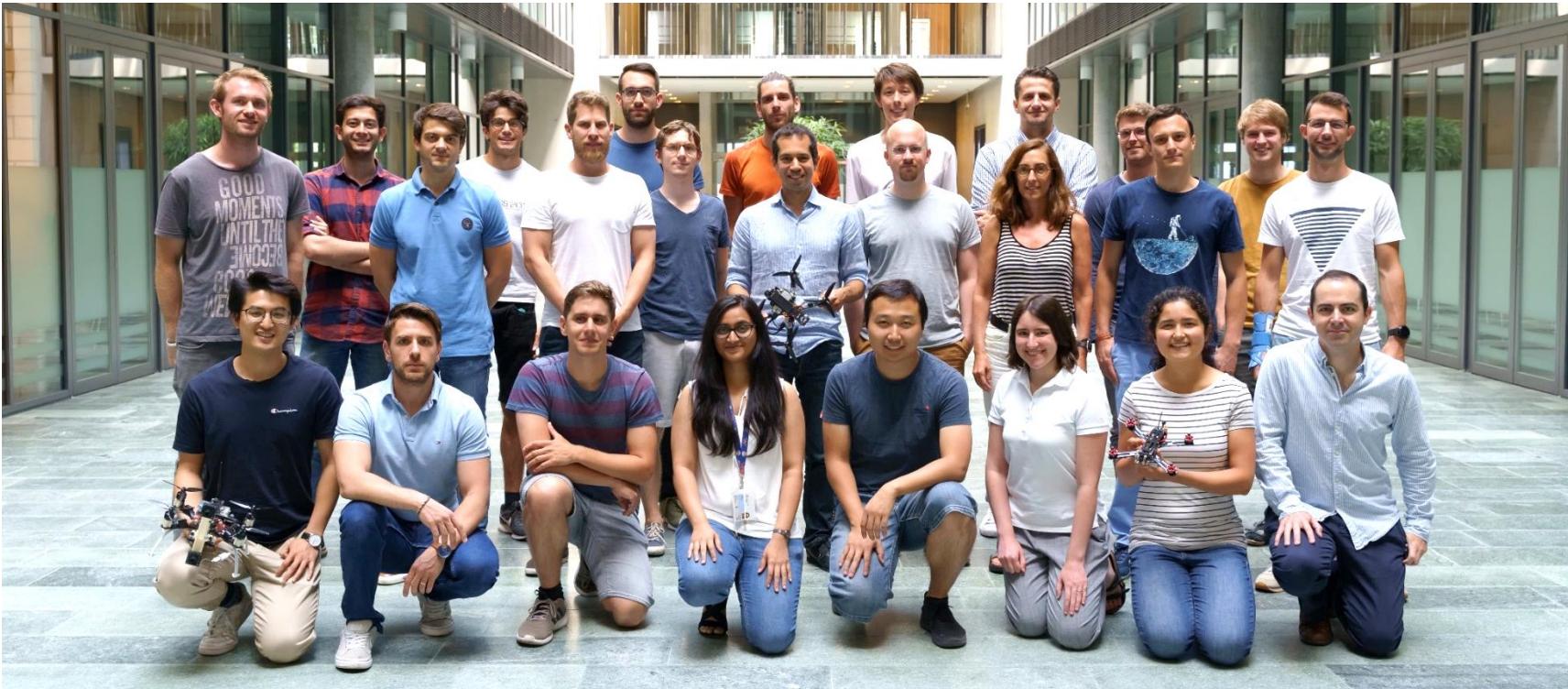
- "Autonomous Mobile Robots," MIT Press, 2011

Hobbies

- Running, piano, magic tricks

My lab: the Robotics and Perception Group

- **Address:** Andreasstrasse 15, 2nd floor, next to **Zurich Oerlikon** train station
- **Webpage:** <http://rpg.ifi.uzh.ch>



Research Topics

Real-time, Onboard Computer Vision and Control for Autonomous Drone Navigation:

- Robot Learning
- Robot Vision
- Motion Planning & Control

Motivation:

- Search and rescue applications

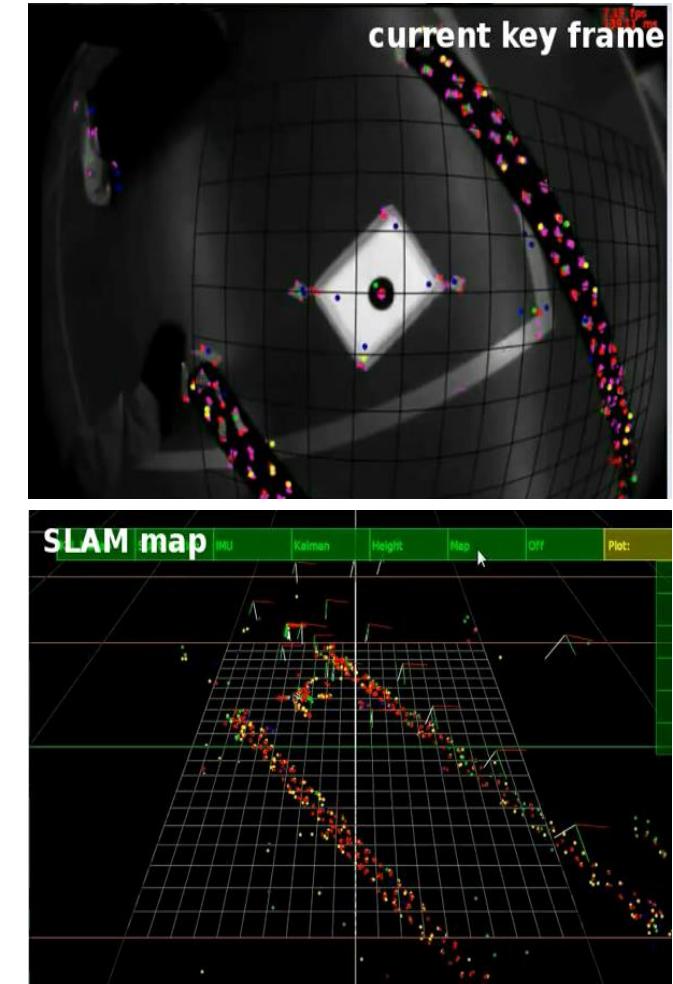


For an overview on our research, watch this keynote at the MIT Robotics Today Seminar Series:

<https://youtu.be/LhO5WSFH7ZY>



13-years ago: First Vision-based Autonomous Flight



European Micro Aerial Vehicle competition, Sep. 9, 2009

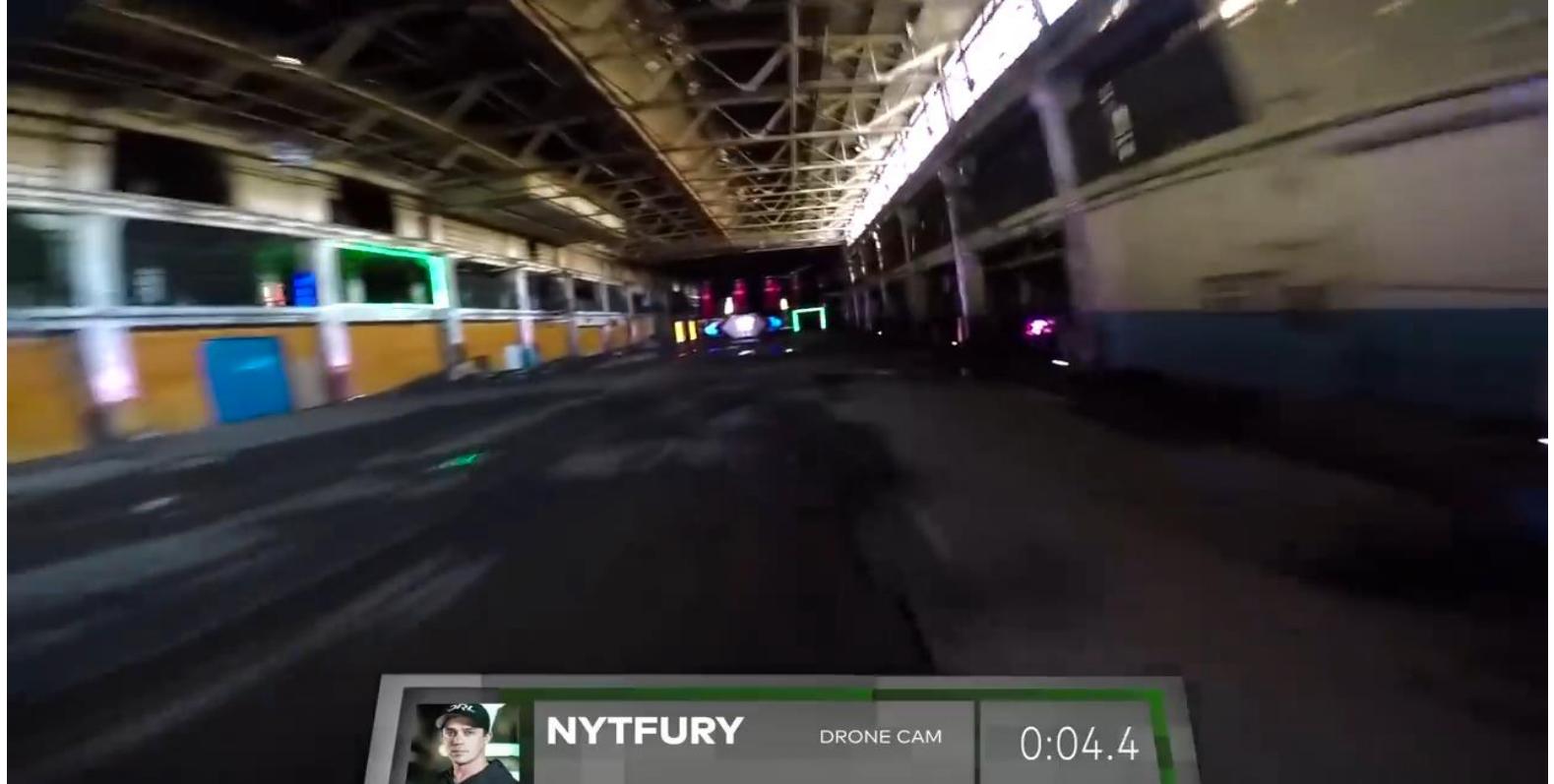
Bloesch, Weiss, Scaramuzza, Siegwart, *Vision Based MAV Navigation in Unknown and Unstructured Environment*, ICRA'10 [[PDF](#)]

Today



NASA Ingenuity helicopter performing autonomous vision-based flight on Mars

What does it take to fly as good as or better than human pilots?



WARNING! This drone is NOT autonomous; it is operated by a human pilot.
Human pilots take years to become agile!

Vision-based High-Speed Flight in the Wild



This AI-controlled drone is fully autonomous and uses onboard vision and computation

Loquercio, Kaufmann, Ranftl, Mueller, Koltun, Scaramuzza, *Learning High Speed Flight in the Wild*, **Science Robotics**, 2021

[PDF](#). [Video](#). [Code & Datasets](#)



Autonomous Drone Racing



Foehn et al., AlphaPilot: Autonomous Drone Racing, RSS 2020, Best System Paper Award. [PDF](#) [Video](#)

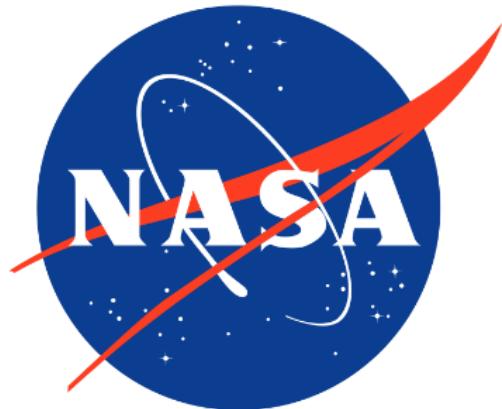
Tech Transfer and Spin-offs



SONY

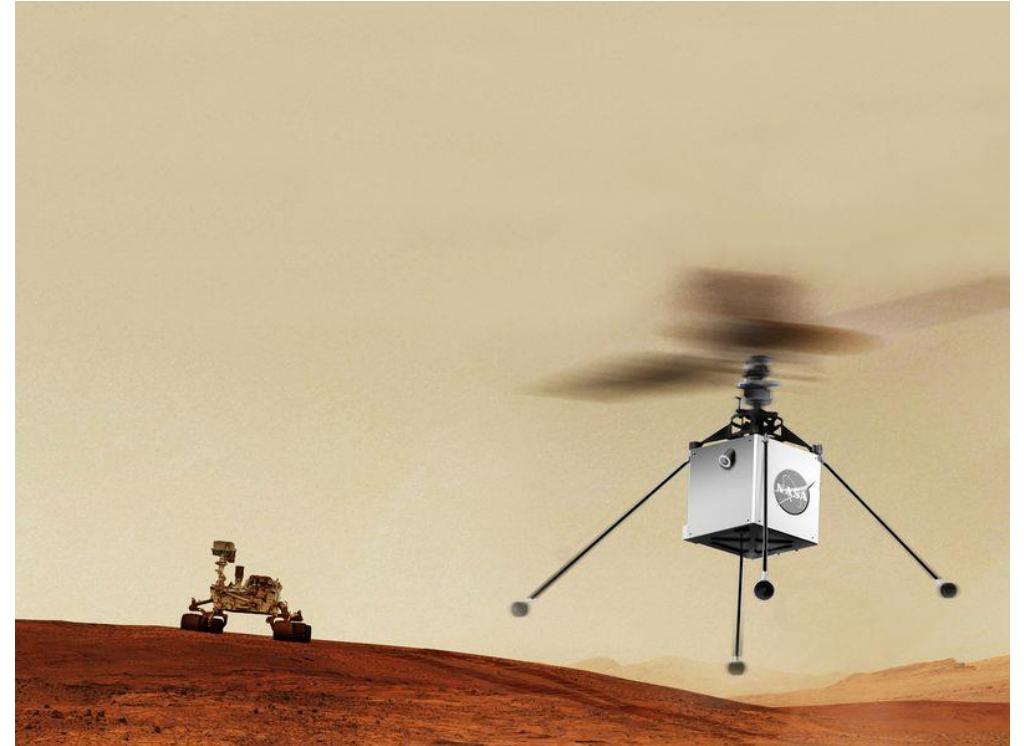


Collaboration with NASA/JPL for future Mars missions



JPL

Jet Propulsion Laboratory
California Institute of Technology



Read the details on [this Swissinfo article](#)



SUIND Making Commercial Drones Able to See

Precise navigation, safe landing, and obstacle avoidance with onboard cameras and computation.



Fotokite A tethered drone for first response

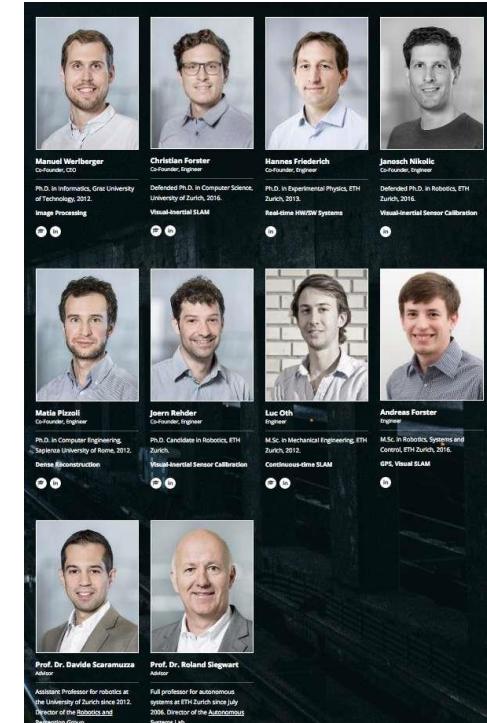
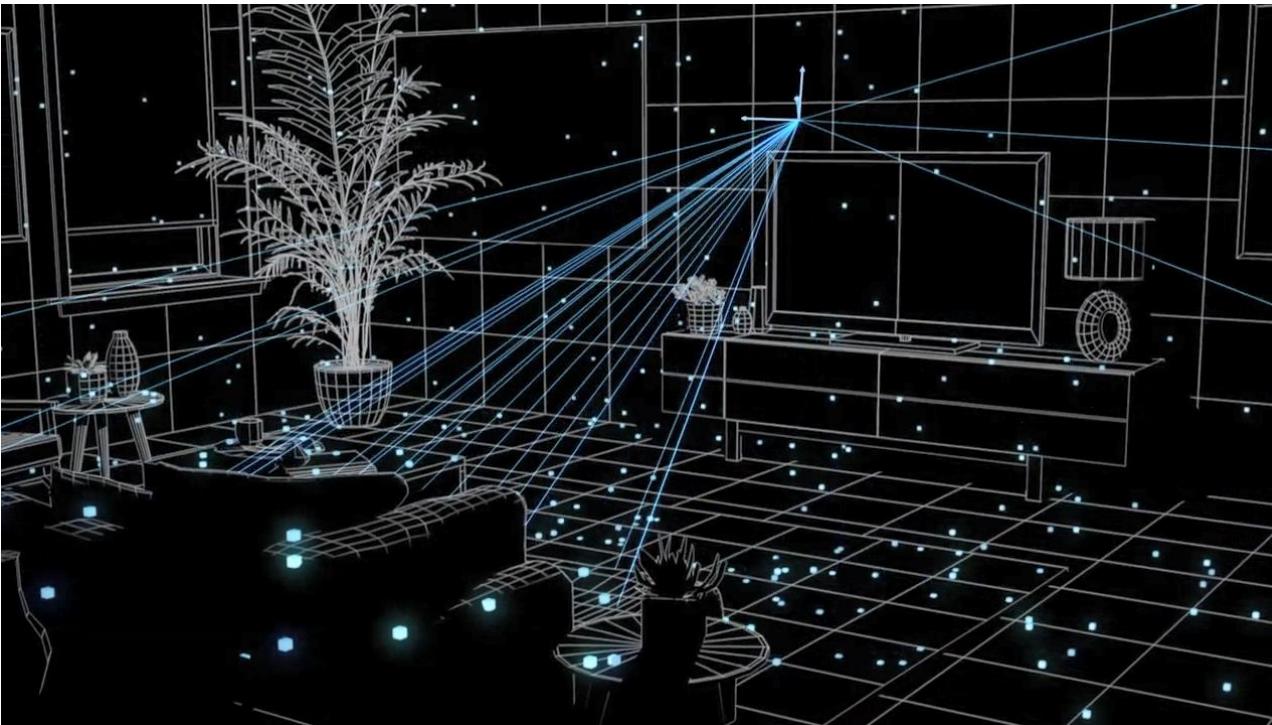
The drone receives electrical power over-tether from the ground so that it can fly “forever”





“Zurich-Eye” – Today: Meta Zurich

- **Vision-based Localization and Mapping** systems for mobile robots
- Born in Sep. 2015, became **Facebook-Oculus Zurich** in Sep. 2016. Today, **200 employees**.
- In 2018, Zurich-Eye launched **Oculus Quest** (10 million units sold so far)



“From the lab to the living room”: The story behind Facebook’s Oculus Insight technology from Zurich-Eye to Oculus Quest:
<https://tech.fb.com/the-story-behind-oculus-insight-technology/>

Student Projects: http://rpg.ifi.uzh.ch/student_projects.php

- **Topics:** machine learning, computer vision, control, planning, system integration
- **Highlights:** many of our Master students have published their thesis in international conferences, won prestigious awards (e.g., ETH Medal, Fritz-Kutter Award, conference paper awards), got a PhD at prestigious institutions (MIT), worked at NASA/JPL, etc.

The screenshot shows a webpage for the University of Zurich (ETH Zurich) Robotics & Perception Group. The header includes the university's logo, the text "University of Zurich ETH Zurich", and the "ROBOTICS & PERCEPTION GROUP". Below the header, there are links for "Department of Informatics", "Institute of Neuroinformatics", and "Robotics and Perception Group". The main content area is titled "Student Projects". On the left, there is a sidebar with links for "News", "People", "Research", "Publications", "Software/Datasets", "Open Positions", "Student Projects" (which is currently selected and highlighted in blue), "Teaching", "Media Coverage", "Awards", "Gallery", and "Contact". To the right of the sidebar, there is a section titled "How to apply" with instructions for sending CVs and transcripts. Below this, there is a section titled "Bayesian Optimization for Racing Aerial Vehicle MPC Tuning" which is marked as "Available". This section includes a thumbnail image of a racing drone, a brief description of the project goal, contact details for Angel Romero Aguilar, and a link to see the project on SIROP. Further down, there are sections for "Generating High-Speed Video with Event Cameras" and "Learning an Event Camera", each with their own descriptions, contact details, and links to SIROP.

Student Projects

How to apply

To apply, please send your CV, your Ms and Bs transcripts by email to all the contacts indicated below the project description. Do not apply on SIROP. Since Prof. Davide Scaramuzza is affiliated with ETH, there is no organizational overhead for ETH students. Custom projects are occasionally available. If you would like to do a project with us but could not find an advertised project that suits you, please contact Prof. Davide Scaramuzza directly to ask for a tailored project. (davide@ifi.uzh.ch)

Upon successful completion of a project in our lab, students may also have the opportunity to get an **internship** at one of our numerous industrial and academic partners worldwide (e.g., NASA/JPL, University of Pennsylvania, UCLA, MIT, Stanford, ...).

Bayesian Optimization for Racing Aerial Vehicle MPC Tuning - Available

Description: In recent years, model predictive control, one of the most popular methods for controlling constrained systems, has benefited from the advancements of learning methods. Many applications showed the potential of the cross fertilization between the two fields, i.e., autonomous drone racing, autonomous car racing, etc. Most of the research efforts have been dedicated to learn and improve the model dynamics, however, controller tuning, which has a crucial importance, have not been studied much.

Goal: The objective of this project is to implement an auto-tuning learning based algorithm for Model Predictive Contouring Control in a racing drone setting. The controller will learn how to tune the controller weights by using specialized Bayesian optimization algorithms [1] that can explore the large dimensional controller parameters space. The learning algorithm will be first tested in simulation and then validated with hardware experiments on a racing aerial vehicle. Your project would include: - A literature research on the current state of the art about Bayesian optimization [1] for controller tuning and on MPCCC literature [2] - Implementation of the state-of-the-art algorithms identified in the previous point - Development of a tailored automatic controller parameters adaptation based on Bayesian optimization - Simulation of the developed algorithms on a racing drone - Test of the algorithms on a real racing drone. The thesis will be in collaboration between UZH Robotics and Perception group and ETH IDS-C Intelligent Control Systems group. [1] Fröhlich, Lukas P., Melanir N. Zeilinger, and Edgar D. Kierszen. "Cautious Bayesian optimization for efficient and scalable policy search." Learning for Dynamics and Control. PMLR, 2021. [2] A. Romero, S. Sun, P. Foehr, and D. Scaramuzza, "Model predictive contouring control for time-optimal quadrotor flight," IEEE Trans. Robot., doi: 10.1109/TRO.2022.3173711.

Contact Details: Angel Romero Aguilar roaguil@ifi.uzh.ch, Andrea Carron, carrona@ethz.ch, Kim Wabersich wkim@ethz.ch

Thesis Type: Master Thesis
[See project on SIROP](#)

Generating High-Speed Video with Event Cameras - Available

Description: Event cameras have shown amazing capabilities in slowing down video as was shown in our previous work, TimeLens (<https://www.youtube.com/watch?v=dVlyia-evo>). This is because, compared to standard cameras, event cameras only capture a highly compressed representation of the visual signal, and do this with high dynamic range and very low latency. It is this signal that can be decoded into intermediate frames. In this project we want to push the limits of what is possible using such a method and explore new extensions.

Goal: In this project we want to explore new extensions of video frame interpolation using an event camera.

Contact Details: Daniel Gehrig (dgehrig@ifi.uzh.ch), Mathias Gehrig (mgehrig@ifi.uzh.ch)

Thesis Type: Semester Project / Master Thesis
[See project on SIROP](#)

Learning an Event Camera - Available

Description: Event cameras such as the Dynamic Vision Sensor (DVS) are recent sensors with a lot of potential for high-speed and high dynamic range robotic applications. They have been successfully applied in many applications, such as high speed video and high speed visual odometry. In spite of this success, the exact operating principle of event cameras, that is, how events are generated from a given visual signal and how noise is generated, is not well understood. In this work we want to explore new techniques for modelling the generation of events in an event camera, which would have wide implications for existing techniques.

Applicants should have a background in C++ programming and low-level vision. In addition, familiarity with learning frameworks such as pytorch or tensorflow are required.

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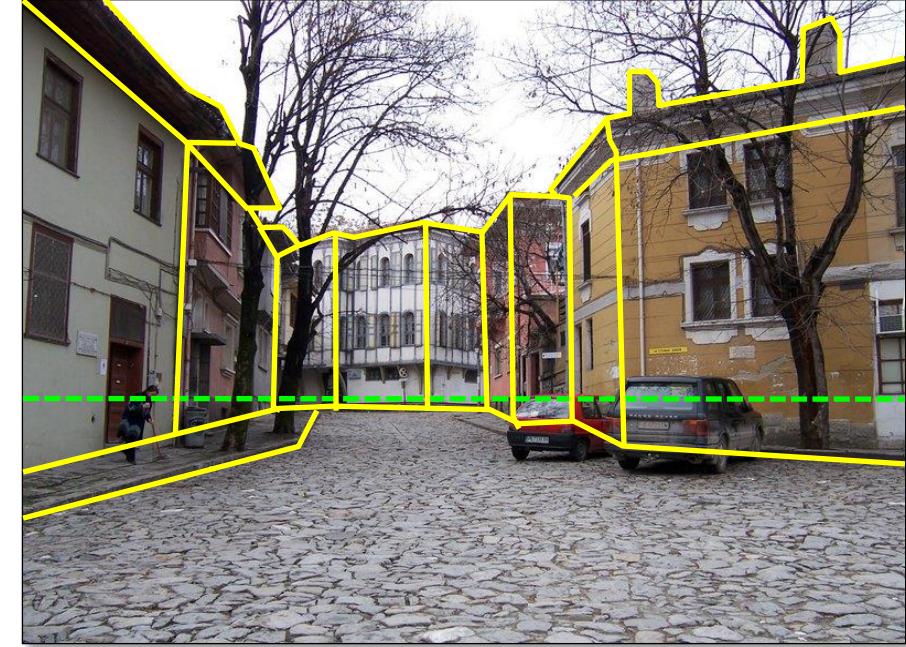
What is computer vision?

Automatic extraction of “meaningful” information from images and videos



Semantic information

(“Image Analysis and Computer Vision” course)



Geometric information

(this course)

Vision Demo?

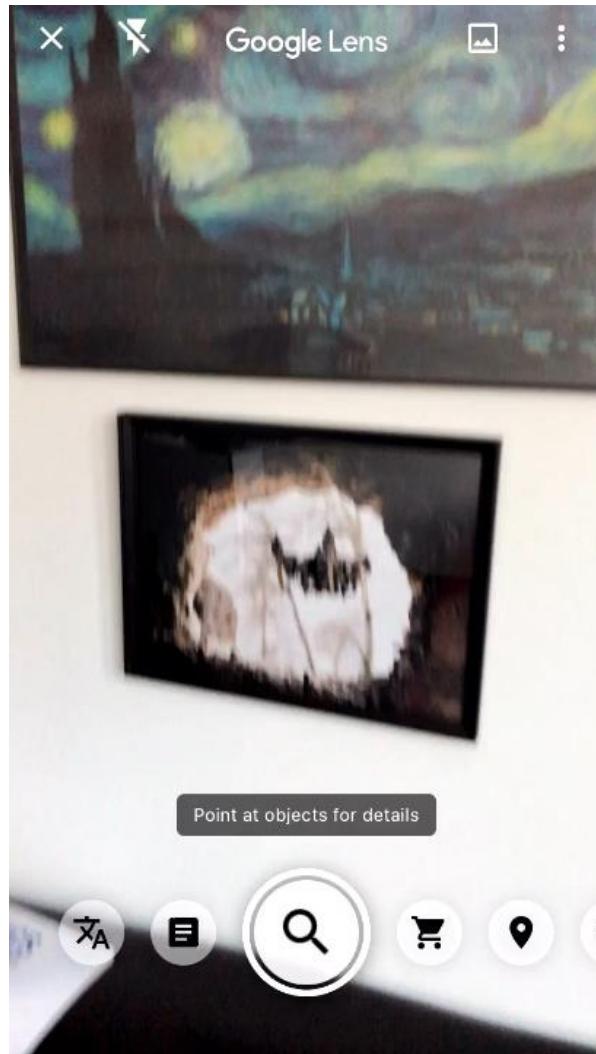


Terminator 2



Are we there? Almost!

Google App



Today's Outline

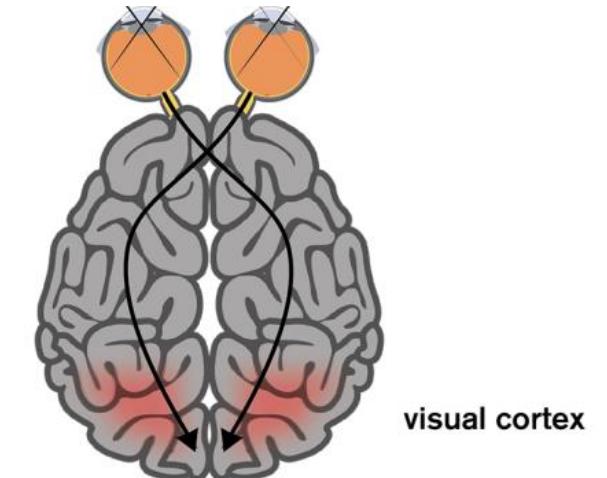
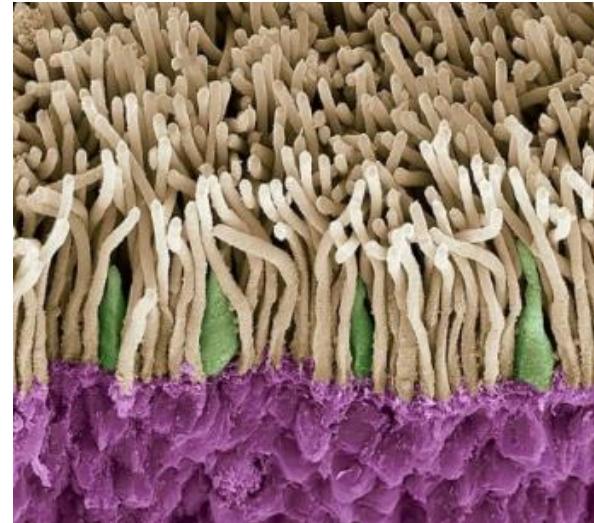
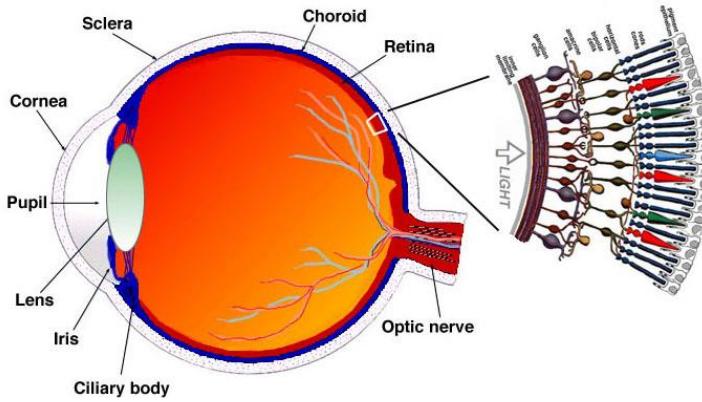
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Why study computer vision?

- **Relieve** humans of boring, easy tasks
- **Enhance** human abilities: human-computer interaction, visualization, augmented reality (AR)
- Perception for autonomous **robots**
- **Organize** and give access to visual **content**
- Lots of computer-vision **companies and jobs in Switzerland** (Zurich & Lausanne):
 - Meta-Facebook-Oculus (Zurich): AR/VR, Instagram
 - Huawei (Zurich): automotive, autonomous cars, event cameras, computational photography
 - Verity (Zurich): SLAM engineer
 - Perspective Robotics (Zurich): Computer vision engineer
 - Fixposition (Zurich): Sensor fusion engineer
 - Magic-Leap (Zurich): AR/VR
 - Microsoft Research (Zurich): Robotics and Hololens AR
 - Google (Zurich): Brain, Positioning Services, Street View, YouTube
 - Apple (Zurich): Autonomous Driving, face tracking
 - NVIDIA (Zurich): simulation, autonomous driving
 - Logitech (Zurich, Lausanne)
 - Disney-Research (Zurich)
 - VIZRT (Zurich): sport broadcasting, 3D replay
 - Pix4D (Lausanne): 3D reconstruction from drones
 - More on [glassdoor.ch](https://www.glassdoor.ch)

Vision in humans

- **Vision** is our most powerful sense. **Half of primate cerebral cortex** is devoted to visual processing
- The retina is $\sim 1,000 \text{ mm}^2$. Contains **130 million photoreceptors** (120 mil. rods for low light vision and 10 mil. cones for color sampling) covering a **visual field of 220×135 degrees**
- Provides enormous amount of information: **data-rate of $\sim 3\text{GBytes/s}$**
- To match the eye resolution, we would need a **500 Megapixel** camera. But in practice the acuity of an eye is **8 Megapixels** within a **18-degree field of view** (5.5 mm diameter) around a small depression called **fovea**

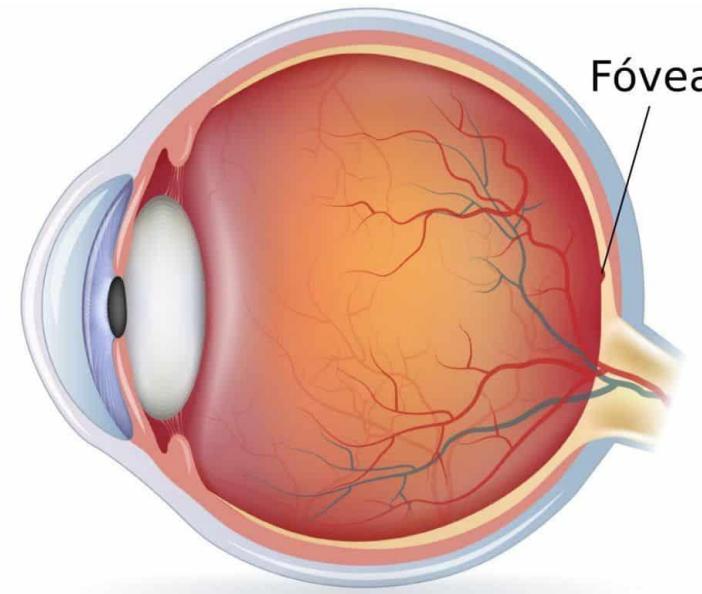


Vision in humans: how we see

- The **area we see in focus** and in **full color** represents the part of the visual field that is covered by the **fovea**
- The **fovea** is 0.35 mm in diameter, covers a visual field of **1-2 degrees**, has **high density of cone cells**
- Within the rest of the **peripheral visual field**, the image we perceive becomes more **blurry (rod cells)**



How we actually see. This principle is used in
[foveated rendering](#)



If you are interested to study human perception, check out the
UZH course “Computational Vision” (this semester)

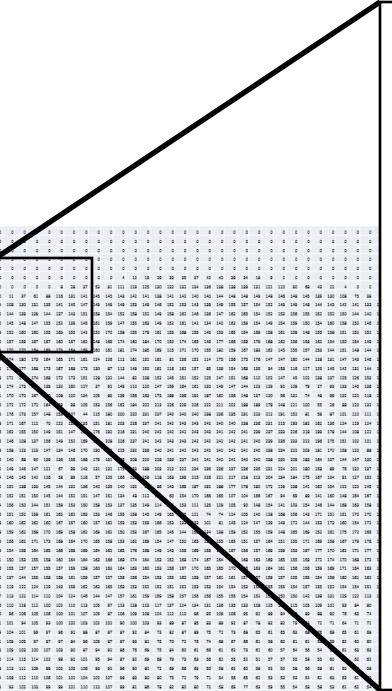
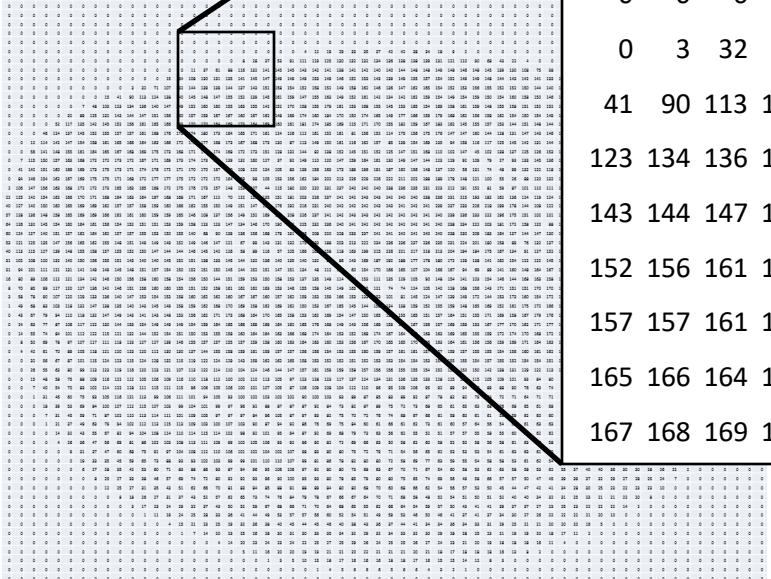
What a newborn sees every month in the first year

"Your baby sees things best from 15 to 30 cm away. This is the perfect distance for gazing up into the eyes of mom or dad. Any farther than that, and the newborn sees mostly blurry shapes because they're nearsighted. At birth, a newborn's eyesight is between 20/200 and 20/400."



Why is vision hard?

How do we go from an array of numbers to recognizing a fruit?



0	0	0	0	0	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	0	8	28	
0	0	0	0	0	0	0	11	37	61	88	116	132
0	0	0	0	15	64	108	130	131	135	141	145	
0	3	32	71	107	132	144	139	139	144	137	143	
41	90	113	124	138	140	145	148	147	155	152	139	
123	134	136	140	147	149	152	160	160	155	163	155	
143	144	147	151	156	160	157	159	167	167	160	167	
152	156	161	165	166	169	170	170	164	169	173	164	
157	157	161	168	176	175	174	180	173	164	165	171	
165	166	164	163	166	172	179	177	168	173	167	168	
167	168	169	175	173	168	171	177	174	168	172	173	

What a computer sees

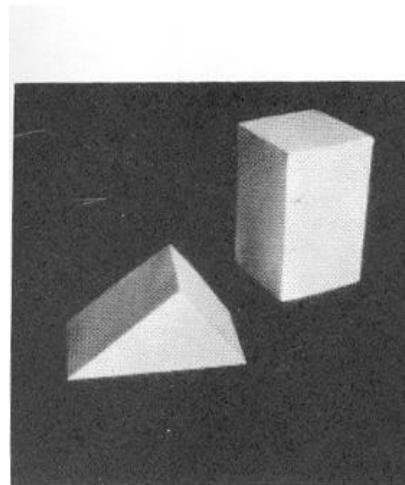


What we see

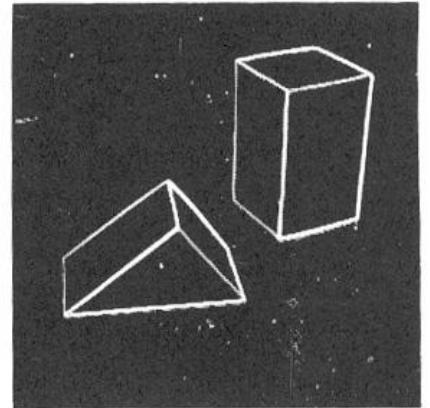
Origins of computer vision

[- 23 - 4445(a-d)]

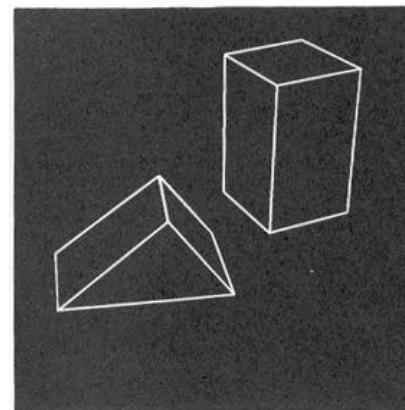
- **1963** - [L. G. Roberts](#) publishes his PhD thesis on [*Machine Perception of Three Dimensional Solids*](#), thesis, MIT Department of Electrical Engineering
- He is the **inventor of ARPANET, the current Internet**



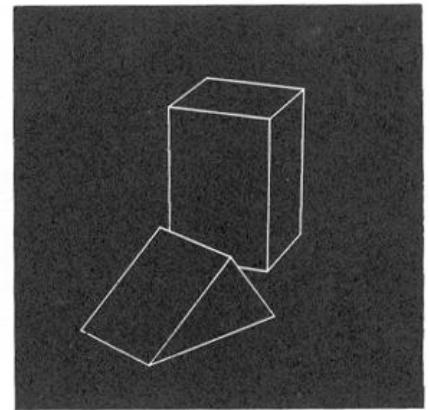
(a) Original picture.



(b) Differentiated picture.



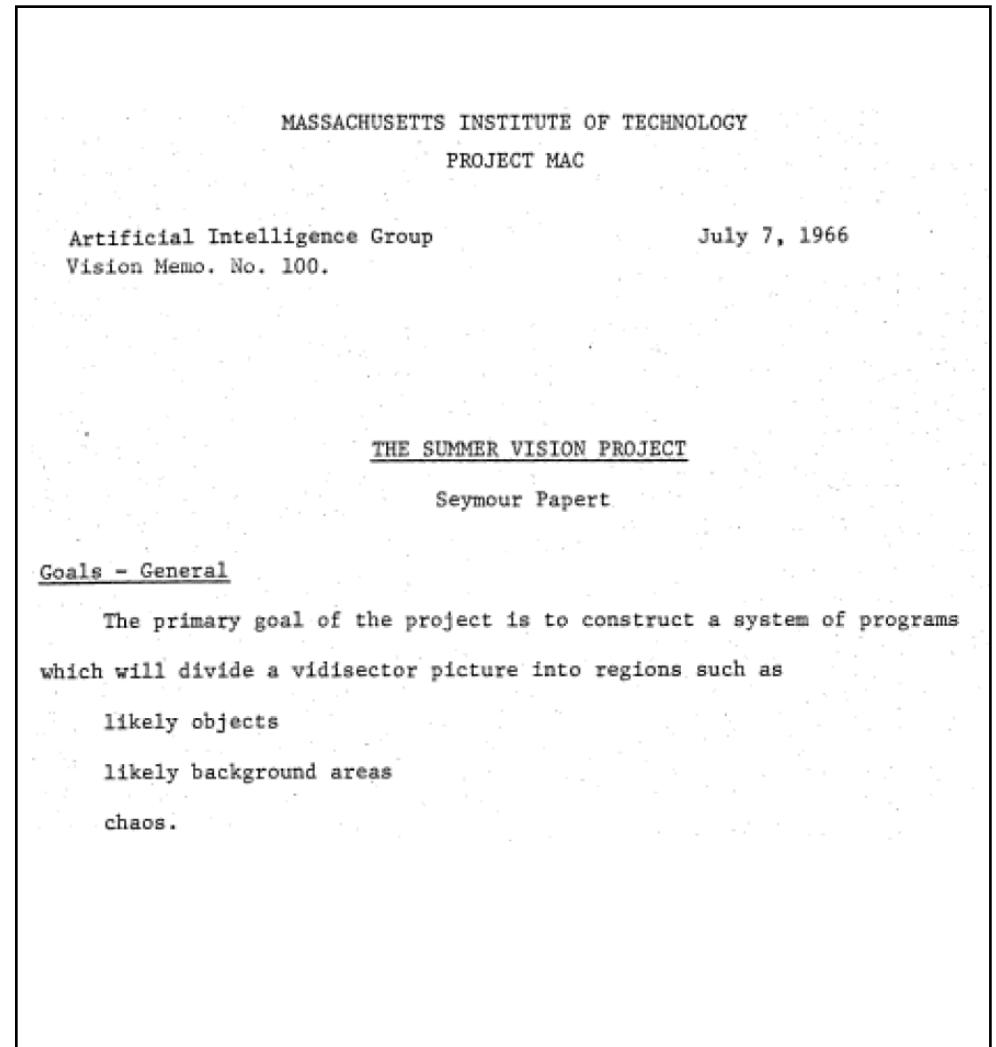
(c) Line drawing.



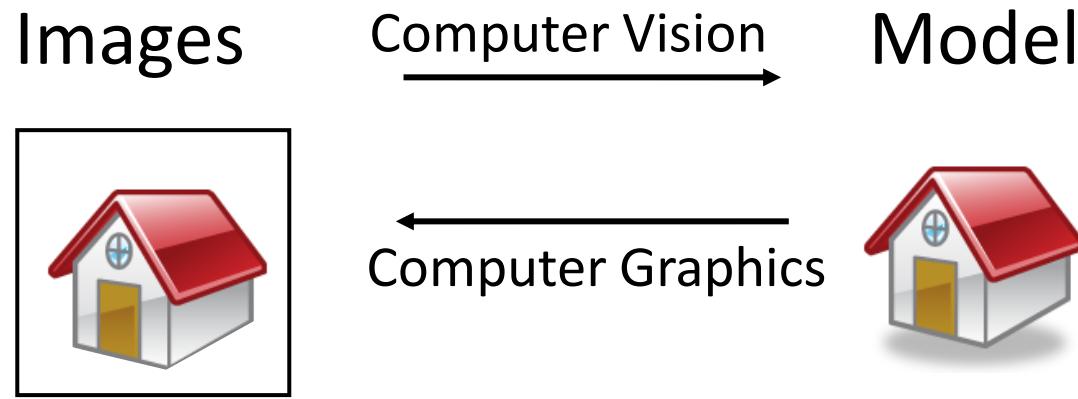
(d) Rotated view.

Origins of computer vision

- **1963** - [L. G. Roberts](#) publishes his PhD thesis on [*Machine Perception of Three Dimensional Solids*](#), thesis, MIT Department of Electrical Engineering
- He is the **inventor of ARPANET, the current Internet**
- **1966** – [Seymour Papert](#), MIT, publishes the [Summer Vision Project](#) asking students to design an algorithm to segment an image into objects and background... within summer!
- **1969** – [David Marr](#) starts developing a [framework for processing visual information](#)



Computer Vision vs Computer Graphics

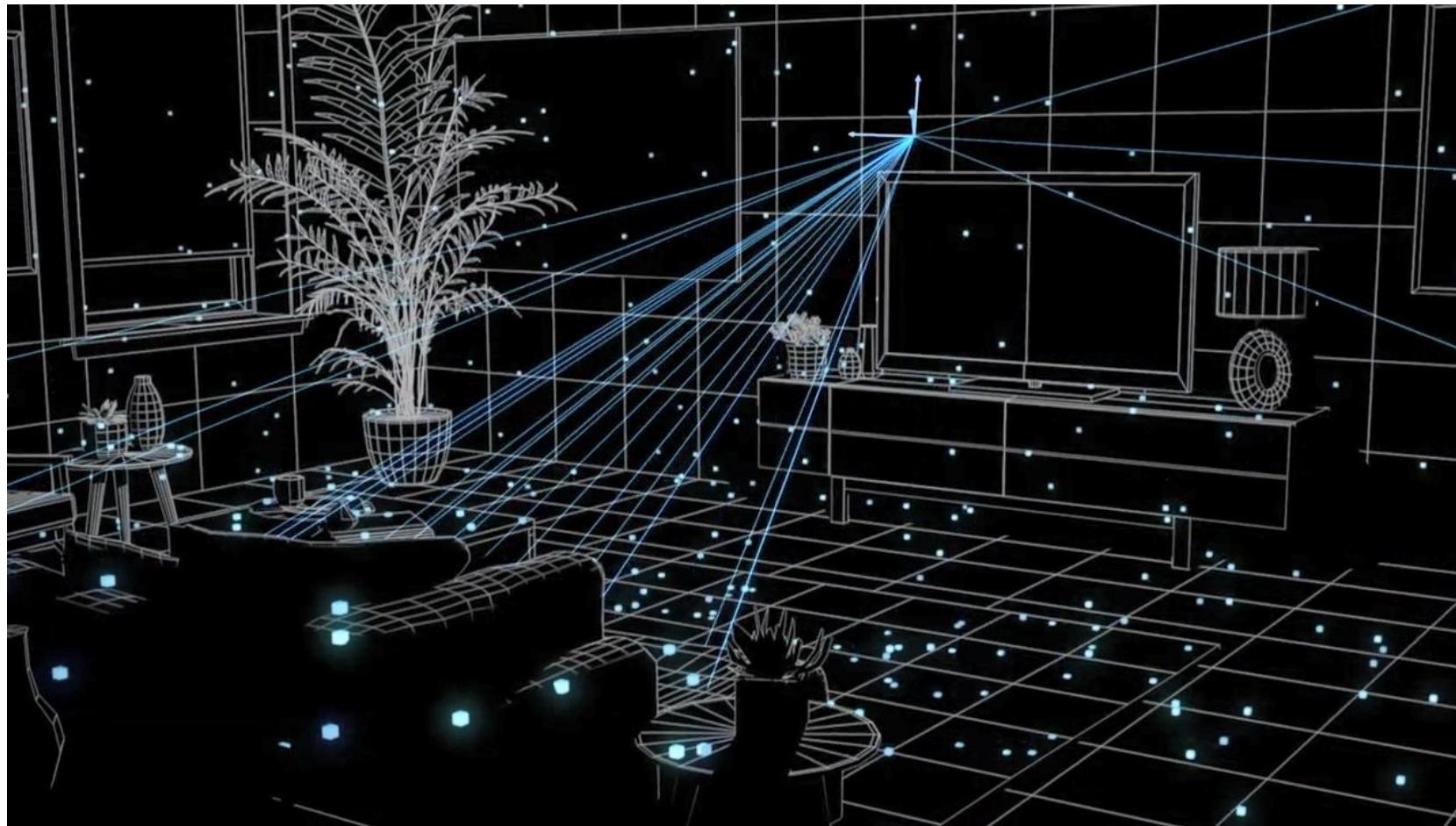


Inverse problems: analysis and synthesis.

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VR/AR



Oculus Quest uses four cameras to track the pose of the head and the controllers

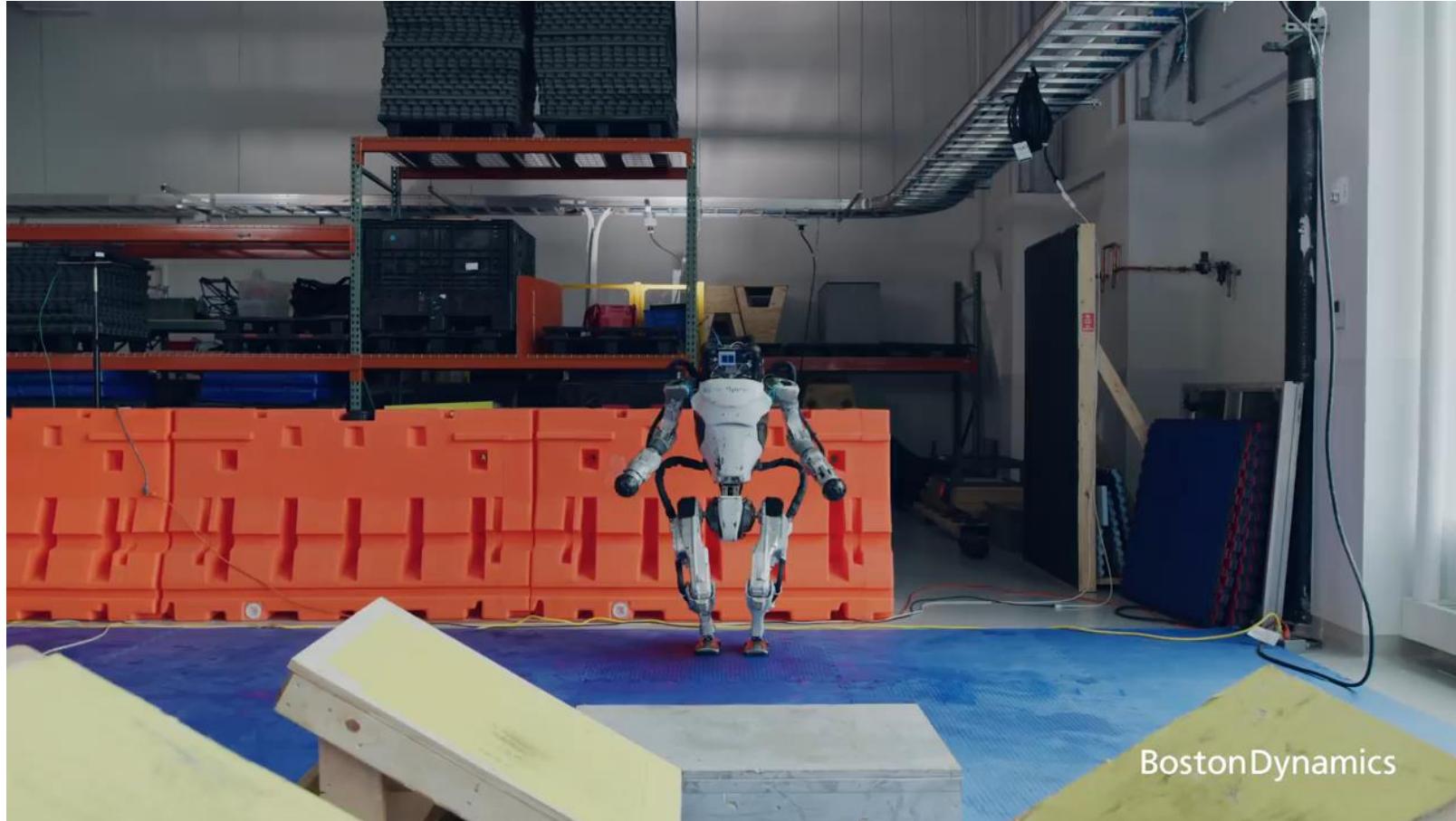
Advanced Driving Assistance Systems (ADAS)



[Mobileye](#): Vision system used at **BMW, GM, Volvo** models. Bought by **Intel in 2017 for 15 billion USD!**

- **Pedestrian & car** collision warning
- **Lane departure** warning
- **Safety distance** monitoring and warning

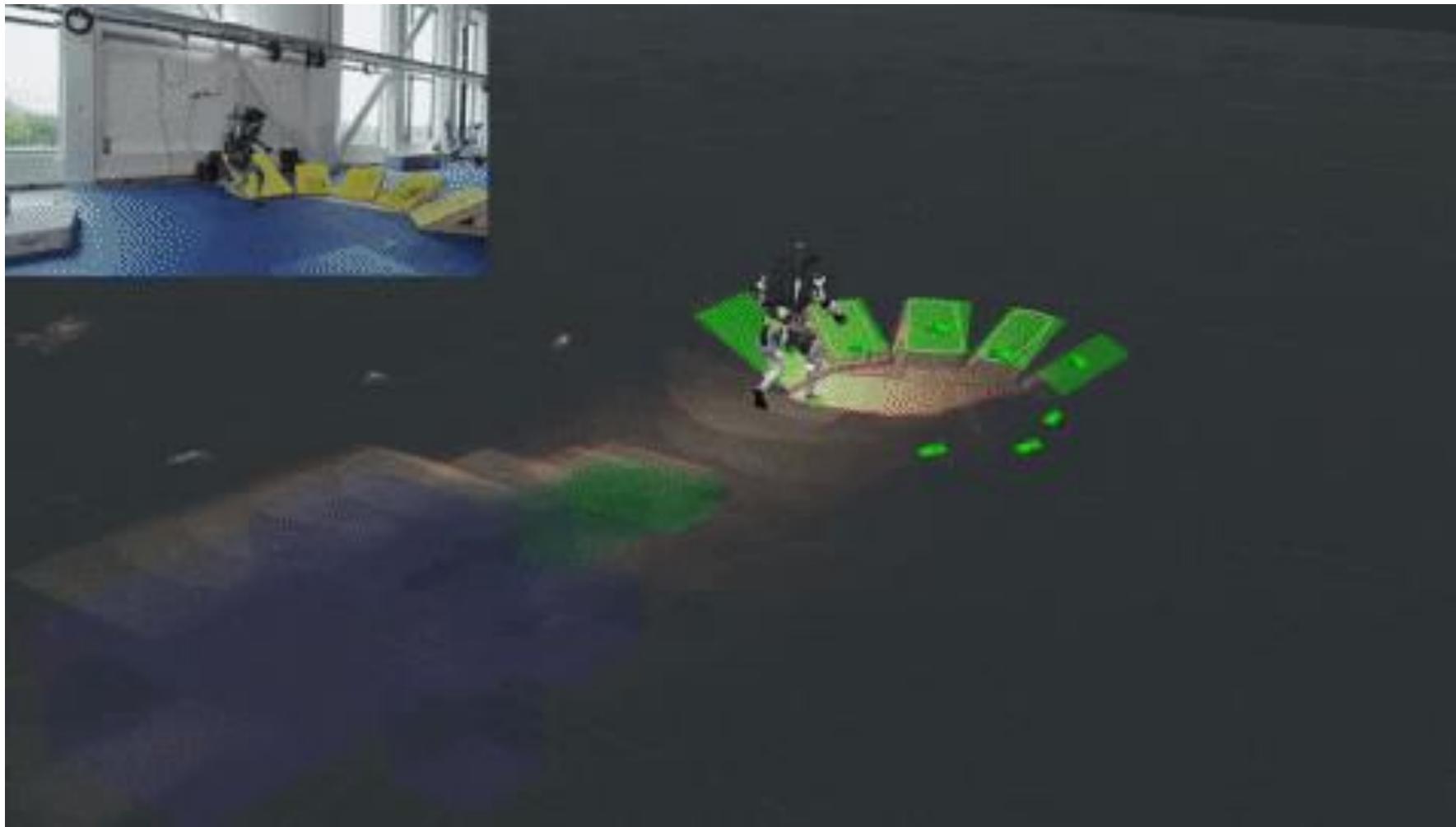
Boston Dynamics ATLAS Robot



<https://youtu.be/tF4DML7FIWk>

Watch Boston Dynamics keynote at the MIT Robotics Today Seminar Series: <https://youtu.be/EGABAx52GKI>

Boston Dynamics ATLAS Robot



<https://blog.bostondynamics.com/flipping-the-script-with-atlas>

Roomba Vacuum Cleaner

- Introduced in 2002 by iRobot
- More than 20 million Roombas sold till 2020
- Fully autonomous, uses camera to recognize places

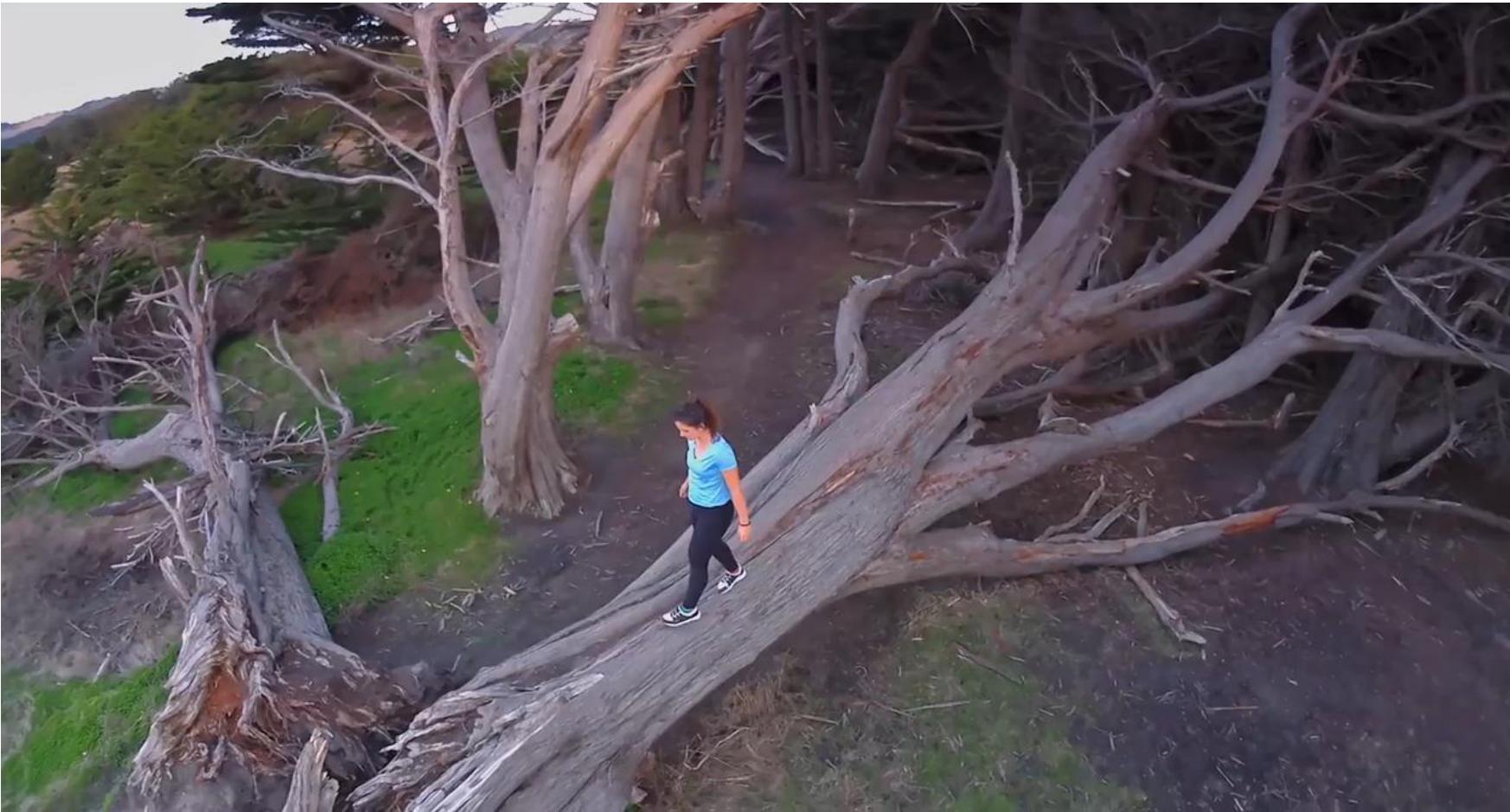


The Roomba 880 navigates by zigzagging randomly across the room. Average cleaning time: 20 minutes



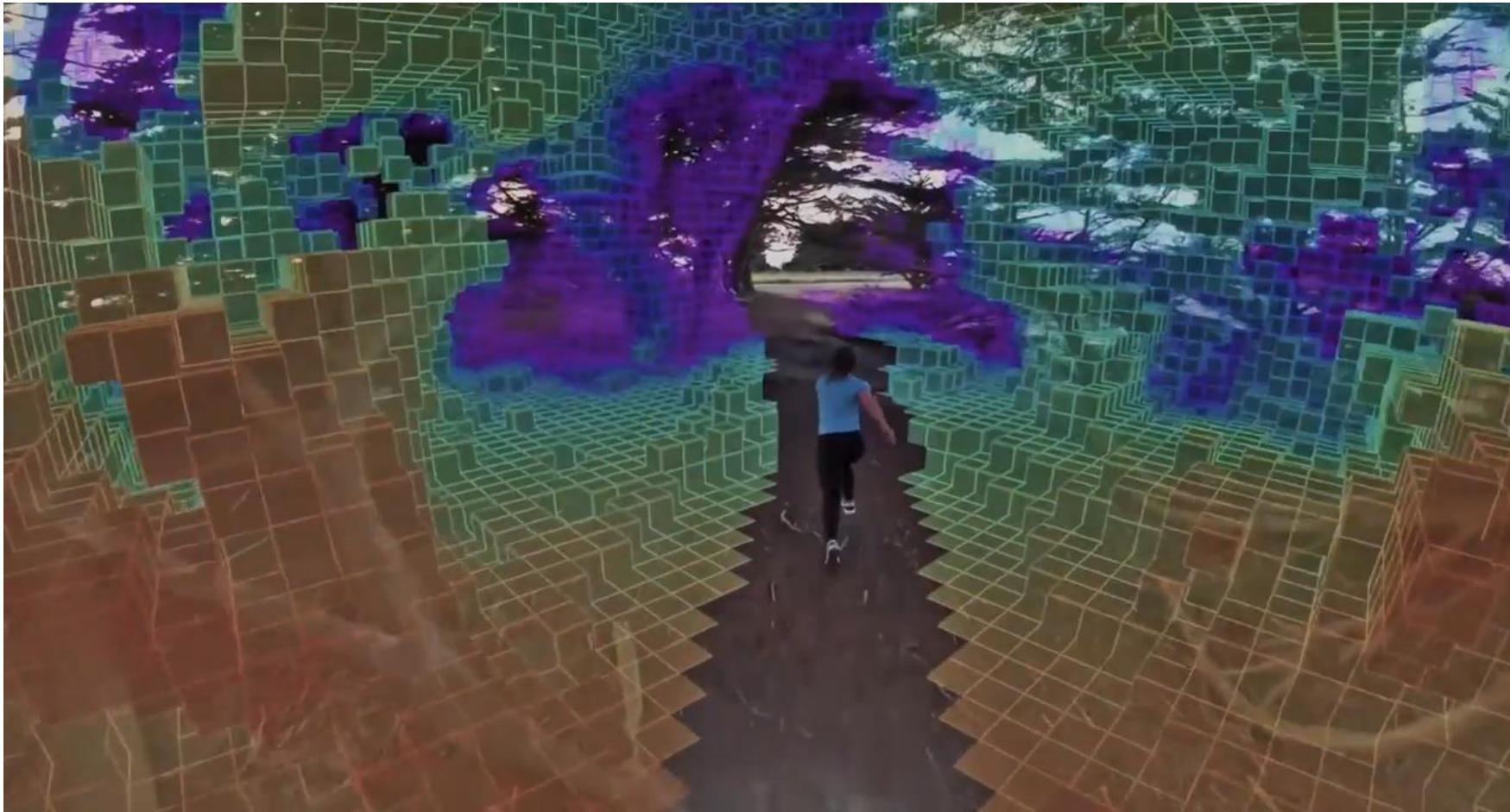
The Roomba 980 navigates by following a pre-defined path optimized thanks to visual SLAM - Average cleaning time: 5 minutes

Skydio and DJI Drones



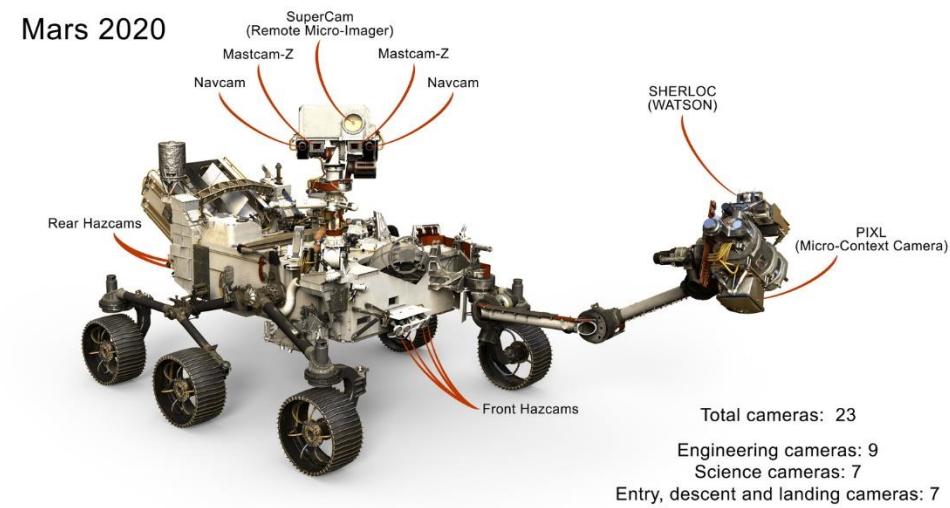
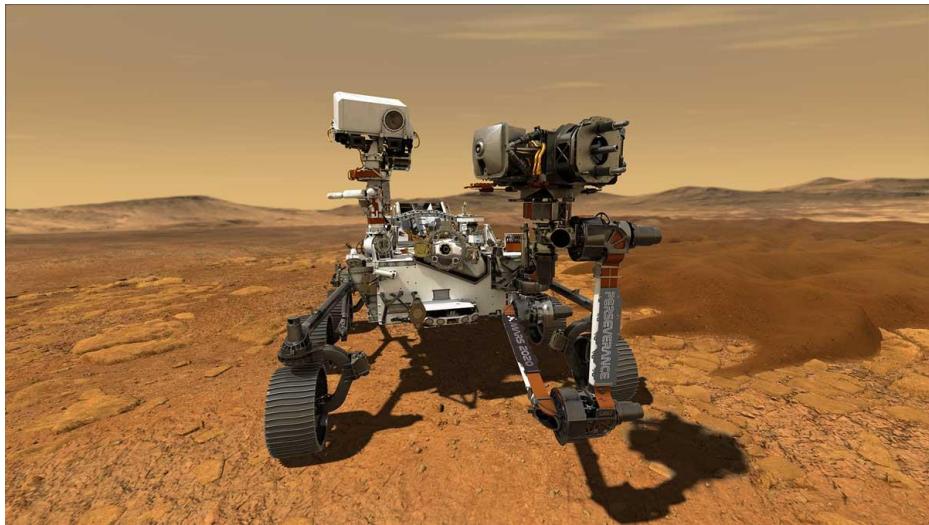
The Skydio-2 drone features 7 cameras for obstacle avoidance, visual odometry, and person following
Watch Skydio keynote at the MIT Robotics Today Seminar Series: <https://youtu.be/ncZmnfIRIWE>

Skydio and DJI Drones



The Skydio R2 drone features 7 cameras for obstacle avoidance, visual odometry, and person following
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NASA Mars Rovers

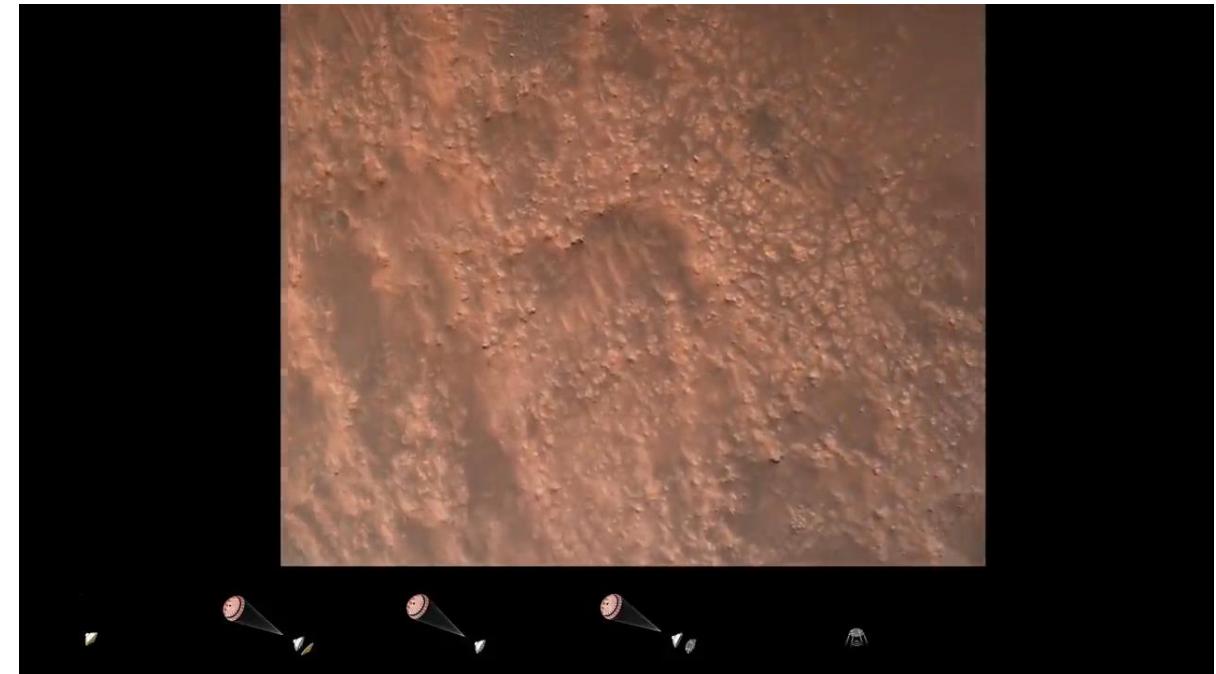
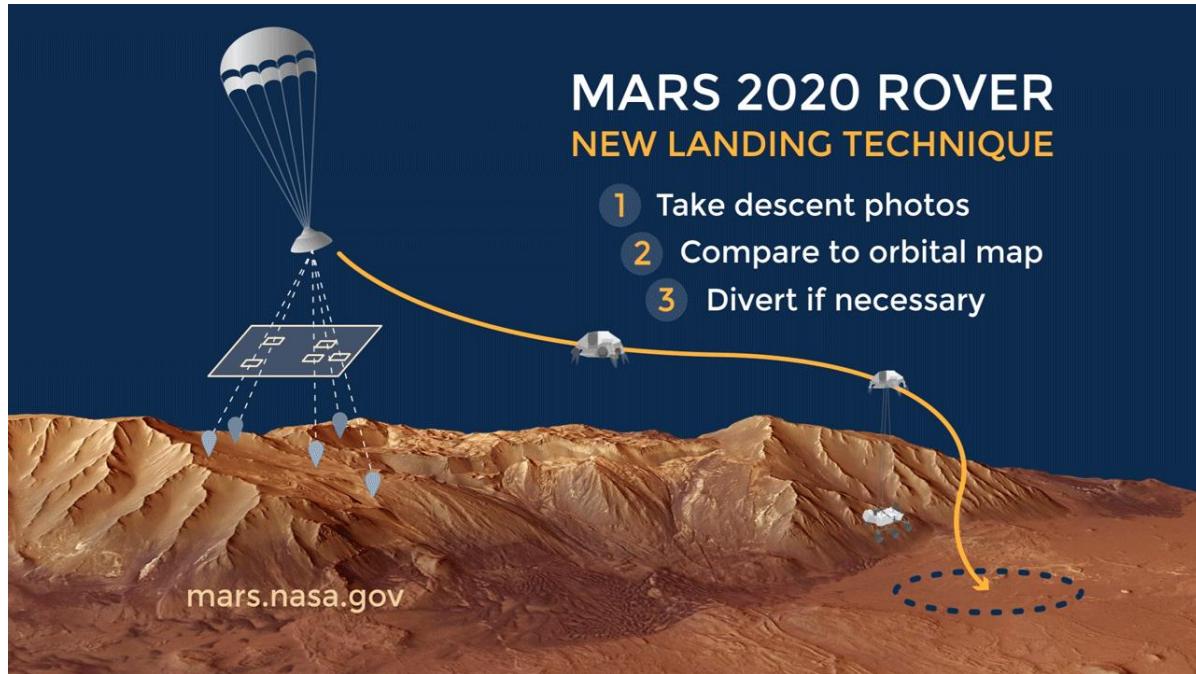


NASA'S Perseverance Rover landed in 2021 features 23 cameras used for:

- Autonomous landing on Mars (Terrain Relative Navigation)
- Autonomous navigation and positioning of the robotic arm
- Science: 3D photos and videos, analysis of chemical components of the soil

For more info, watch the RSS'21 keynote by Larry Matthies, head of the Computer Vision Group at NASA/JPL:
<https://youtu.be/Ncl6fJOzBsU>

Perseverance Descent via Vision-based Terrain Relative Navigation



<https://mars.nasa.gov/mars2020/mission/technology/#Terrain-Relative-Navigation>

Real footage recorded by Perseverance during descent <https://youtu.be/4czjS9h4Fpg>

Vision-based Flight on Mars



The NASA Ingenuity helicopter performs the first autonomous vision-based flight on April 19, 2021, on Mars.

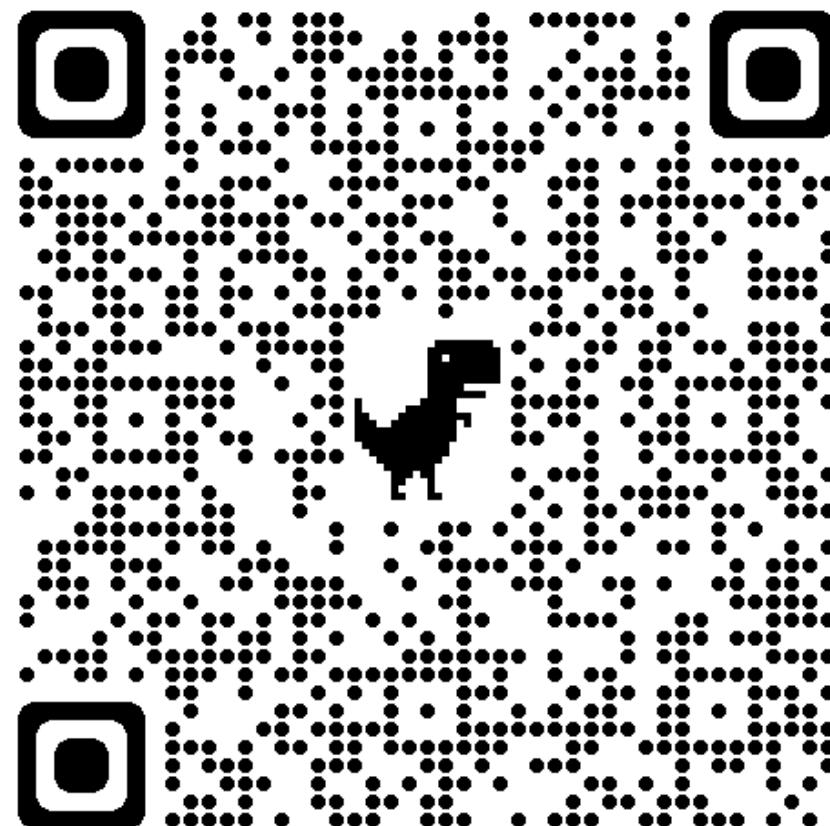
<https://youtu.be/p1KolyCqlCI?t=2502>

<https://mars.nasa.gov/technology/helicopter/#>

December 1, we will have a lecture by [Jeff Delaune](#), from NASA/JPL,
who developed the visual navigation of Ingenuity

Before the Break

Please fill this quick survey: <https://tinyurl.com/jjnvye3b>



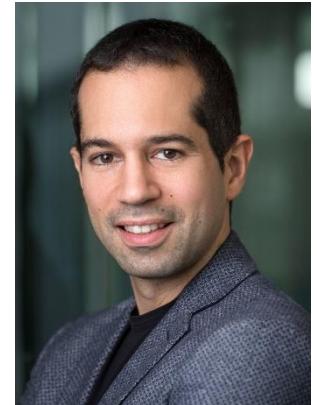
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Instructors

Lecturer: Davide Scaramuzza

- **Contact:** sdavide (at) ifi (dot) uzh (dot) ch
- **Office hours:** every **Thursday from 16:00 to 18:00**
both in person or via ZOOM possible (**please announce yourself by email**)
- **Teaching Assistants:** Nico Messikommer and Leonard Bauersfeld



Nico Messikommer
nmessi (at) ifi (dot) uzh (dot) ch



Leonard Bauersfeld
bauersfeld (at) ifi (dot) uzh (dot) ch

Lectures and Exercises

Lectures:

- **08:00 to 09:45** every week. After class, I usually stay for 10 more minutes for questions
- Breaks: always but can vary between 5-15 min
- Room: SOC-F-106, Rämistrasse 69, 8001 Zurich.

Exercises:

- **12:15 to 13:45**: Starting from Sep. 29 (2nd week). Then roughly every week.
- Room: same as above

Course & Exam Registration and Cancellation

- Registration and exam cancellation deadline for the course is **October 11, 23:59 hrs**
- NB: at UZH, when you register for a course, you are also automatically registered for the exam! If you want to cancel the exam, you need to unbook the course by October 11. Afterward, it will no longer be possible to cancel the exam. No show at the exam will be graded as 1.0. If you cannot take the exam because you fell ill, you can submit a petition with medical certificate within 5 business days from the exam. If you are a student with a disability, you must request assistance in due time. Further info [here](#)

Tentative Course Schedule

For updates, slides, and additional material:
<http://rpg.ifi.uzh.ch/teaching.html>

22.09.2022	Lecture 01 - Introduction to Computer Vision and Visual Odometry No Exercise today
29.09.2022	Lecture 02 - Image Formation: perspective projection and camera models Exercise 01- Augmented reality wireframe cube
06.10.2022	Lecture 03 - Camera Calibration Exercise 02 - PnP problem
13.10.2022	Lecture 04 - Filtering & Edge detection No Exercise today
20.10.2022	Lecture 05 - Point Feature Detectors, Part 1 Exercise 03 - Harris detector + descriptor + matching
27.10.2022	Lecture 06 - Point Feature Detectors, Part 2 Exercise 04 - SIFT detector + descriptor + matching
03.11.2022	Lecture 07 - Multiple-view geometry 1 Exercise 05 - Stereo vision: rectification, epipolar matching, disparity, triangulation
10.11.2022	Lecture 08 - Multiple-view geometry 2 Exercise 06 - Eight-Point Algorithm
17.11.2022	Lecture 09 - Multiple-view geometry 3 Exercise 07 - P3P algorithm and RANSAC
24.11.2022	Lecture 10 - Multiple-view geometry 4 Exercise session: Intermediate VO Integration
01.12.2022	Lecture 11 – NASA Talk + Lecture on Optical Flow and KLT Tracking Exercise 08 - Lucas-Kanade tracker
08.12.2022	Lecture 12 – Reconstruction, Recognition, Deep Learning Exercise hours replaced by Deep Learning lecture
15.12.2022	Lecture 13 - Visual inertial fusion Exercise 09 - Bundle Adjustment
22.12.2022	Lecture 14 - Event based vision Exercise session: Final VO Integration

Study Material

- **Schedule, lecture slides, exercise download, mini projects, course info** on the official course website: <http://rpg.ifi.uzh.ch/teaching.html>
- **Video Recordings** of lectures and exercises will be uploaded to **OLAT**:
<https://lms.uzh.ch/auth/RepositoryEntry/17268801963>
- Post any **questions** related to lectures or exercises in the **OLAT Forum**

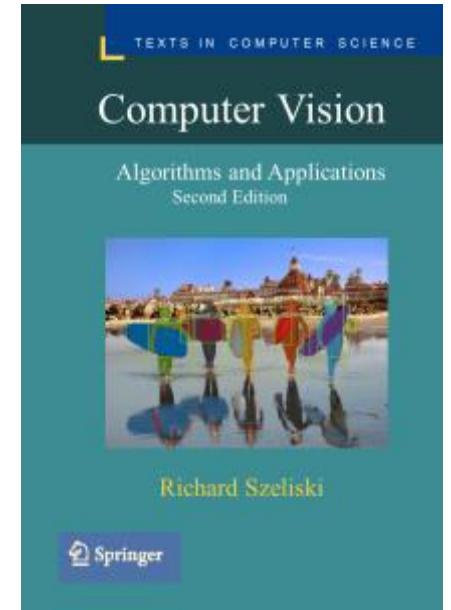
Reference Textbooks

- **Computer Vision: Algorithms and Applications, 2nd Edition**, by Richard Szeliski.
Freely downloadable from the author webpage: <http://szeliski.org/Book/>

- **Chapter 4 of Autonomous Mobile Robots**, 2nd edition, 2011, by R. Siegwart, I.R. Nourbakhsh, D. Scaramuzza. [PDF](#)

- **Additional readings** (i.e., optional and not requested at the exam) for interested students will be provided along with the slides and linked directly from the course website

- Further readings:
 - *Robotics, Vision and Control: Fundamental Algorithms*, 2nd edition, by Peter Corke
 - *An Invitation to 3D Vision*: Y. Ma, S. Soatto, J. Kosecka, S.S. Sastry
 - *Multiple view Geometry*: R. Hartley and A. Zisserman



Prerequisites

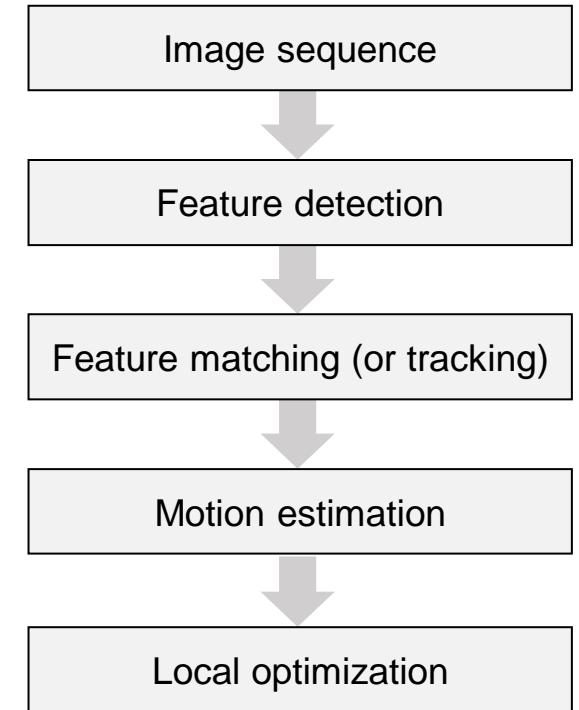
- **Linear algebra**
- **Matrix calculus:** matrix multiplication, inversion, singular value decomposition
 - Check out this [Linear Algebra Primer](#) from Stanford University
 - Check out this [Immersive Linear Algebra](#) interactive tool by Ström, Åström, and Akenine-Möller
 - Check out this [tutorial](#) on camera pose notation
- **No prior knowledge of computer vision and image processing is required**

Learning Objectives

- **High-level goal:** learn to implement the visual-inertial odometry algorithms used in current mobile robots (drones, cars, planetary robots), AR/VR products (Oculus Quest, Microsoft HoloLens, Magic Leap), and Google Visual Positioning Service (e.g., Google Map Live View).
- You will also learn **to implement the fundamental computer vision algorithms** used in mobile robotics, in particular:
 - image formation,
 - filtering,
 - feature extraction,
 - multiple view geometry,
 - dense reconstruction,
 - feature and template tracking,
 - image retrieval,
 - event-based vision,
 - visual-inertial odometry, Simultaneous Localization And Mapping (SLAM),
 - and some basics of deep learning.

Exercises

- **Learning Goal** of the exercises: **Implement a full visual odometry pipeline** (like the one running on Mars rovers).
- **Each week** you will learn how to implement a **building block** of visual odometry.
- Two exercises will be dedicated to **system integration**.
- **NB: Questions about the implementation details of each exercise can be asked at the exam.**



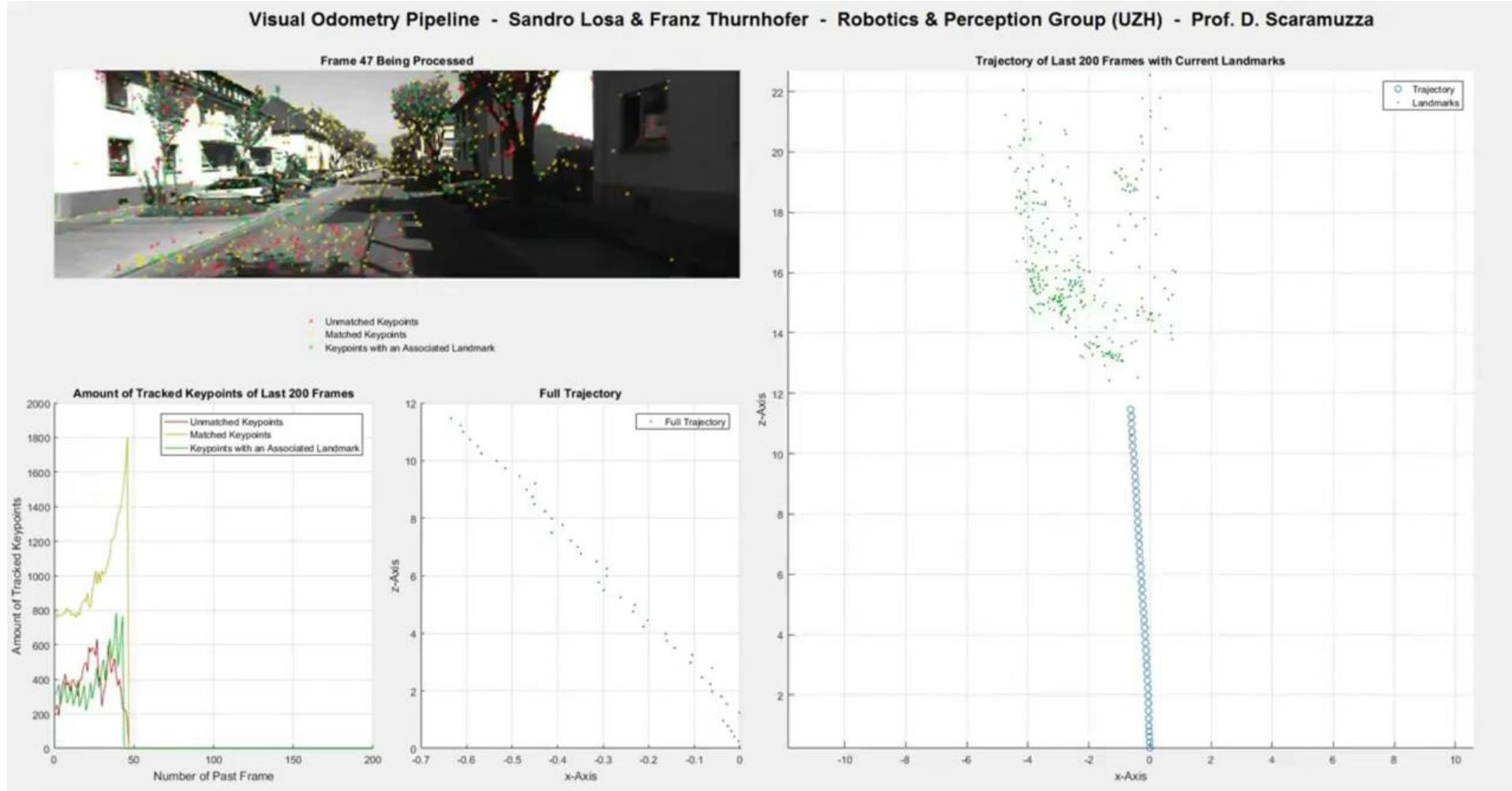
Building blocks of visual odometry
along with information flow

Exercises

- **Bring your own laptop**
- Exercises in **Python (first time for this course) or Matlab**. However, the exercise statements and material presented during the exercise sessions will only cover the implementations in Matlab. Additionally, to the Matlab solutions, we will upload Python scripts as template solutions for the exercise statements. You will need to have Matlab or Python already pre-installed on your machine for the exercises.
- Python can be downloaded from [here](#).
- You can download Matlab from:
 - **ETH:** Download: <https://itshop.ethz.ch/EndUser/Items/Home>
 - **UZH:** Download: <https://www.zi.uzh.ch/de/students/software-elearning/campussoft.html>
 - An introductory tutorial on Matlab can be found here: <http://rpg.ifi.uzh.ch/docs/teaching/2022/MatlabPrimer.pdf>
 - **Please install all the toolboxes included in the license.** If you don't have enough space in your PC, then install at least the Image Processing, Computer Vision, and Optimization toolboxes



Outcome of last year exercises



Grading and Exam

- The **final grade is based on a written exam** (2 hours). Example exam questions will be posted during the course.
 - Exam date: **January 12, 2023, from 08:00 to 10:00 in person** unless changed by UZH
 - **Closed-book exam**
 - **Details about the exam will be provided during the course**
- **Optional mini project:**
 - you have the **option** (i.e., not mandatory) to do a **mini project**, which consists of implementing a working visual odometry algorithm in **Matlab** (C++ or Python are also accepted)
 - If the algorithm runs smoothly, producing a reasonable result, you will be rewarded with an **up to 0.5 grade increase on the final grade**. However, notice that the mini project can be very time consuming!
 - The **deadline** to hand in the mini project is **08.01.2023**.
 - **Group work: minimum 2, max 4 people.**

Class Participation

- **Strong class participation is encouraged!**
- Class participation includes
 - **ask and answer questions**
 - **being able to articulate key points from last lecture**

Today's Outline

- About me and my research lab
- What is Computer Vision?
- Why study computer vision?
- Example of Vision Applications
- Organization of the course
- Start: Visual Odometry overview

What is Visual Odometry (VO) ?

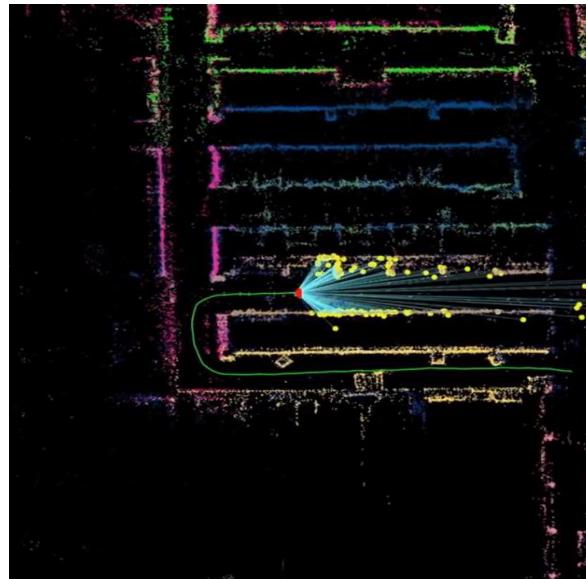
VO is the process of incrementally estimating the pose of the vehicle by examining the changes that motion induces on the images of its onboard cameras

input



Image sequence (or video stream)
from one or more cameras attached to a moving vehicle

output

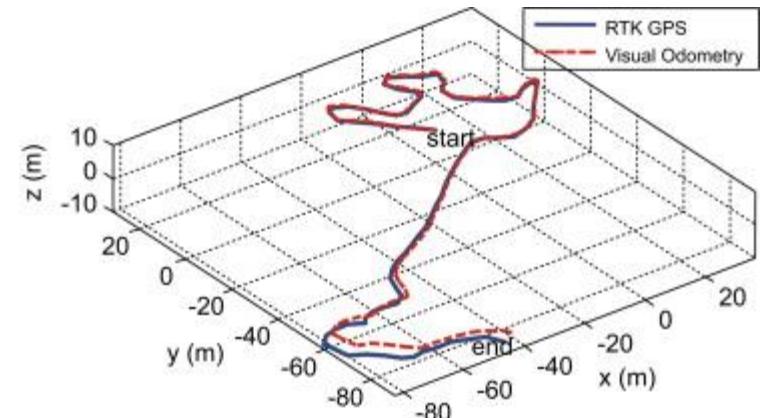


Camera trajectory (3D structure is a plus)

$$R_0, R_1, \dots, R_i$$
$$t_0, t_1, \dots, t_i$$

Why VO?

- VO is crucial for **flying**, **walking**, and **underwater** robots
- Contrary to wheel odometry, VO is **not affected by wheel slippage** (e.g., on sand or wet floor)
- Very accurate:
relative position error is 0.1% – 2% of the travelled distance
- VO can be used as a complement to
 - wheel encoders (wheel odometry)
 - GPS (when GPS is degraded)
 - Inertial Measurement Units (IMUs)
 - laser odometry



Assumptions

- **Sufficient illumination** in the environment
- **Dominance of static scene** over moving objects
- **Enough texture** to allow apparent motion to be extracted
- Sufficient **scene overlap** between consecutive frames



Is any of these scenes good for VO? Why?



A Brief history of VO

- **1980:** First known VO real-time implementation on a robot by **Hans Moraveck** PhD thesis (**NASA/JPL**) for Mars rovers using one sliding camera (*sliding stereo*).

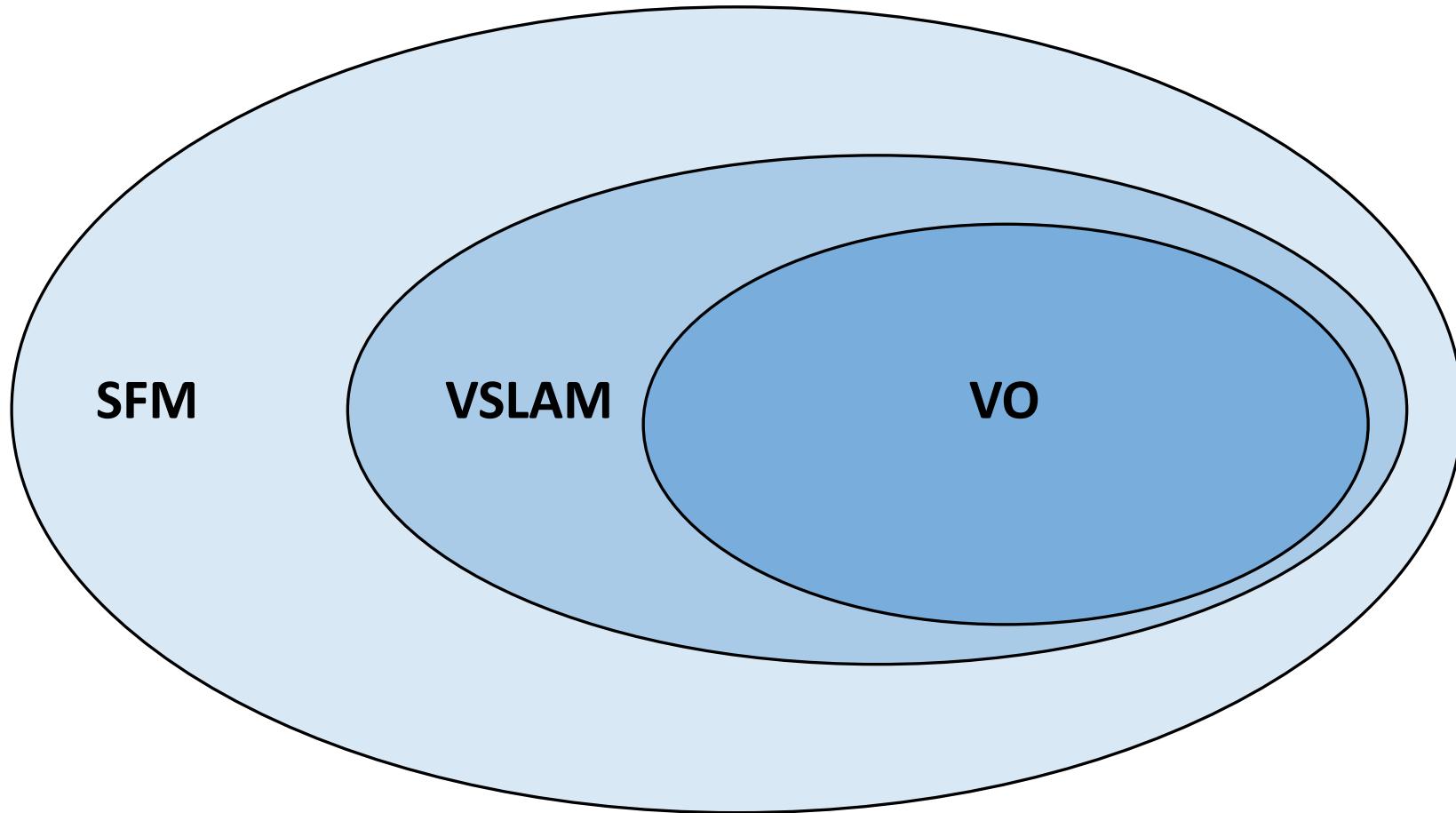


A Brief history of VO

- **1980:** First known VO real-time implementation on a robot by **Hans Moraveck** PhD thesis (**NASA/JPL**) for Mars rovers using one sliding camera (*sliding stereo*).
- **1980 to 2000:** The VO research was dominated by **NASA/JPL** in preparation of the **2004 mission to Mars**
- **2004:** VO was used on a robot on another planet: Mars rovers Spirit and Opportunity (see seminal paper from [NASA/JPL, 2007](#))
- **2004.** VO was revived in the academic environment by **David Nister's** «[Visual Odometry](#)» paper. The term VO became popular.
- **2015-today:** VO becomes a **fundamental tool of several products:** VR/AR, drones, smartphones
- **2021.** VO is used on the **Mars helicopter**

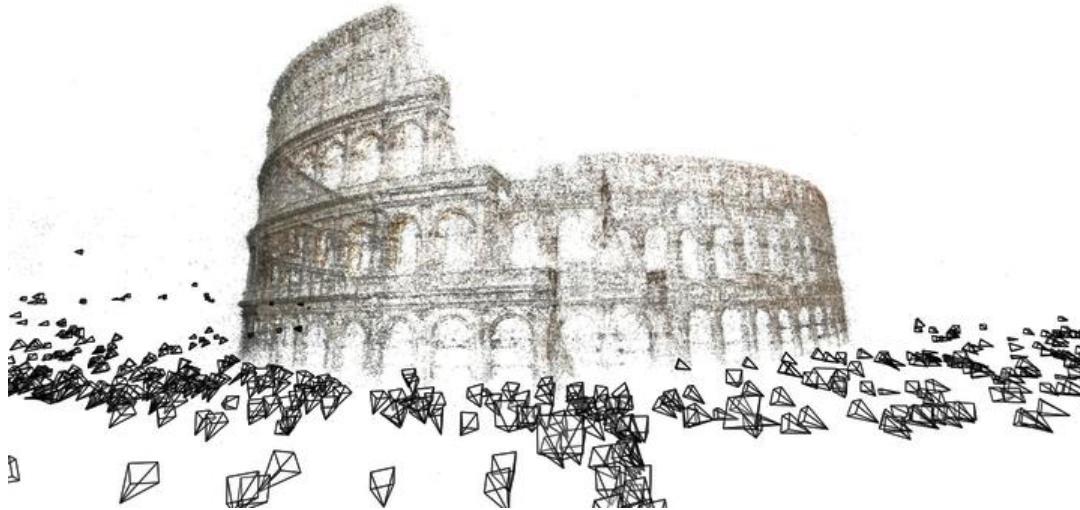


VO vs VSLAM vs SFM



Structure from Motion (SFM)

SFM is more general than VO and tackles the problem of 3D reconstruction and 6DOF pose estimation from **unordered image sets**



Reconstruction from 3 million images from Flickr.com on a cluster of 250 computers, 24 hours of computation

Paper: "[Building Rome in a Day](#)", ICCV'09.

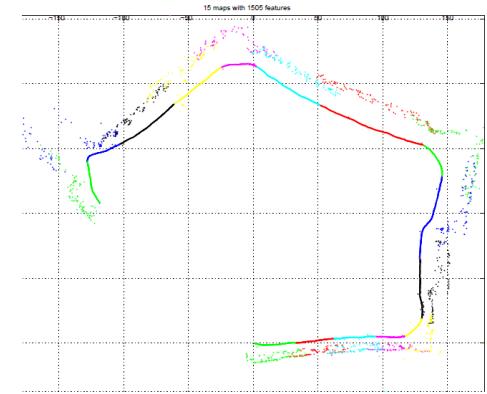
State of the art software: [COLMAP](#)

VO vs SFM

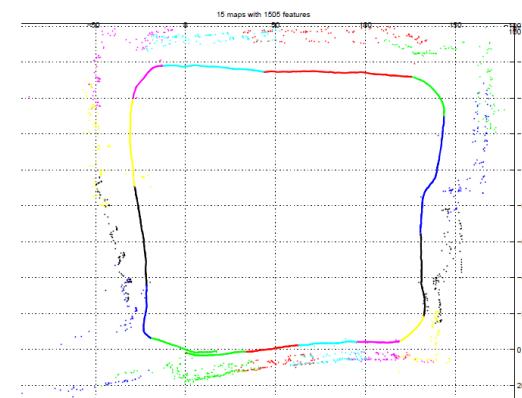
- VO is a **particular case** of SFM
- VO focuses on estimating the 6DoF motion of the camera **sequentially** (as a new frame arrives) and in **real time**
- Terminology: sometimes **SFM** is used as a **synonym** of **VO**

VO vs. Visual SLAM

- **Visual Odometry**
 - Focus on incremental estimation
 - **Guarantees local consistency** (i.e., estimated trajectory is locally correct, but not globally, i.e. from the start to the end)
- **Visual SLAM (Simultaneous Localization And Mapping)**
 - **SLAM = visual odometry + loop detection & closure**
 - **Guarantees global consistency** (the estimated trajectory is globally correct, i.e. from the start to the end)



Visual odometry

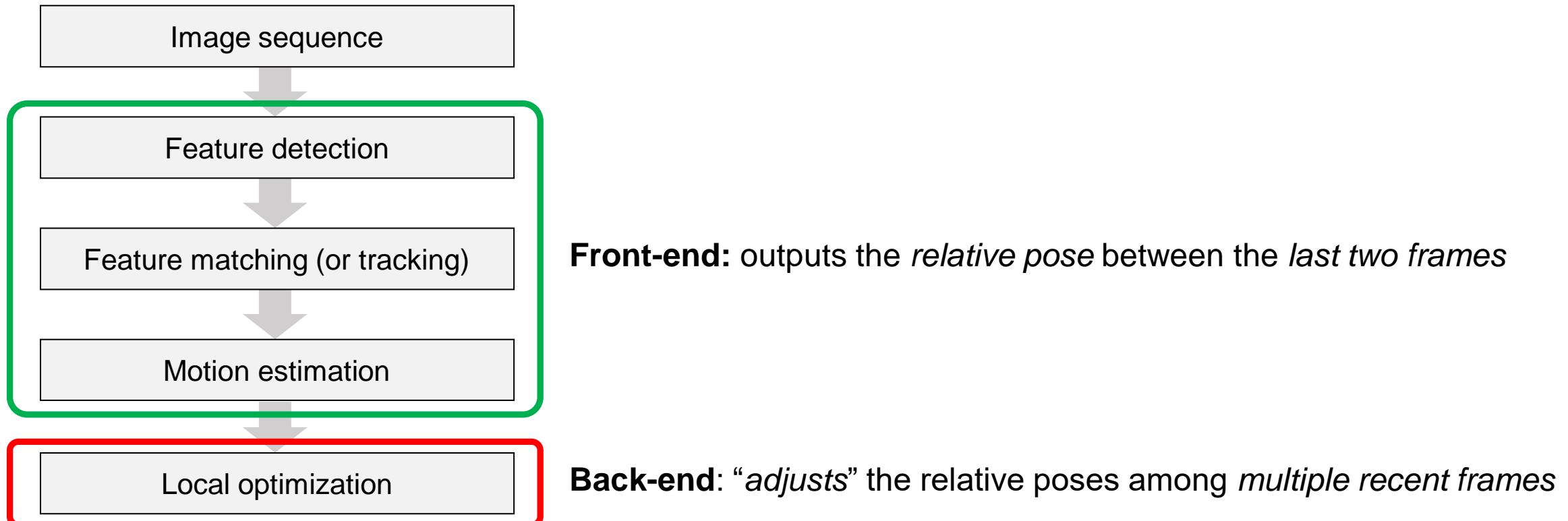


Visual SLAM

Image courtesy of [Clemente et al., RSS'07]

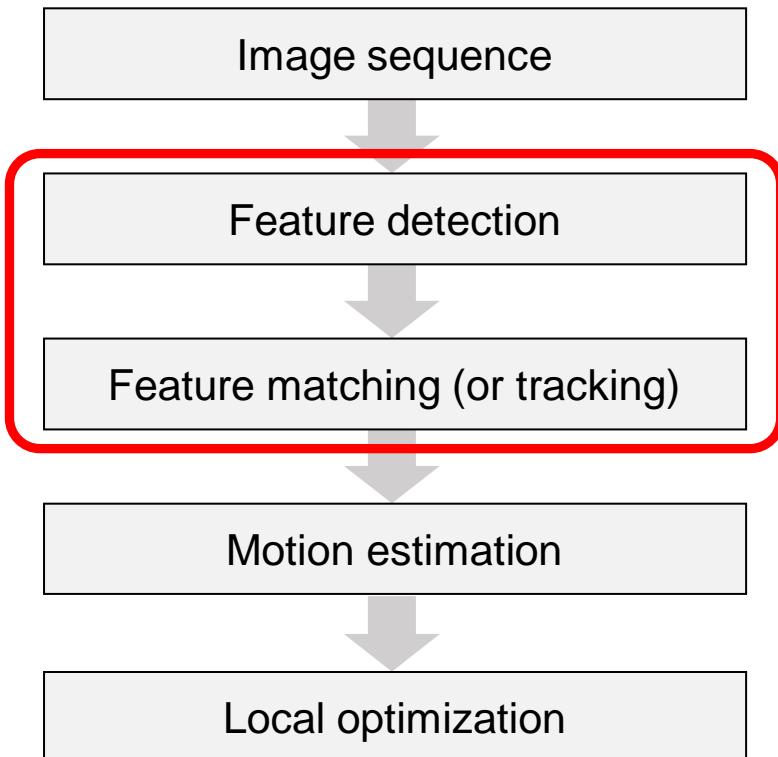
VO Flow Chart

VO computes the camera path incrementally (pose after pose)



VO Flow Chart

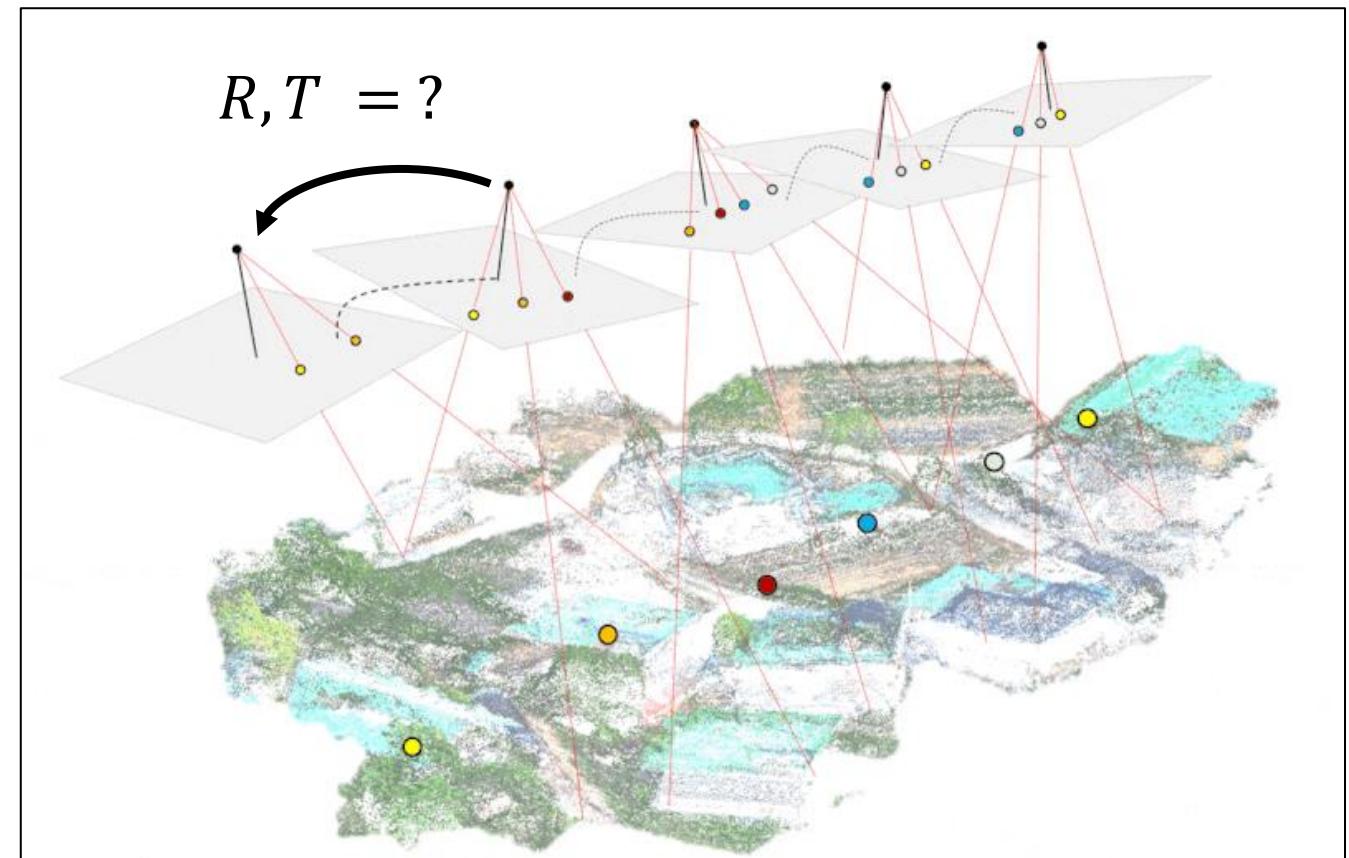
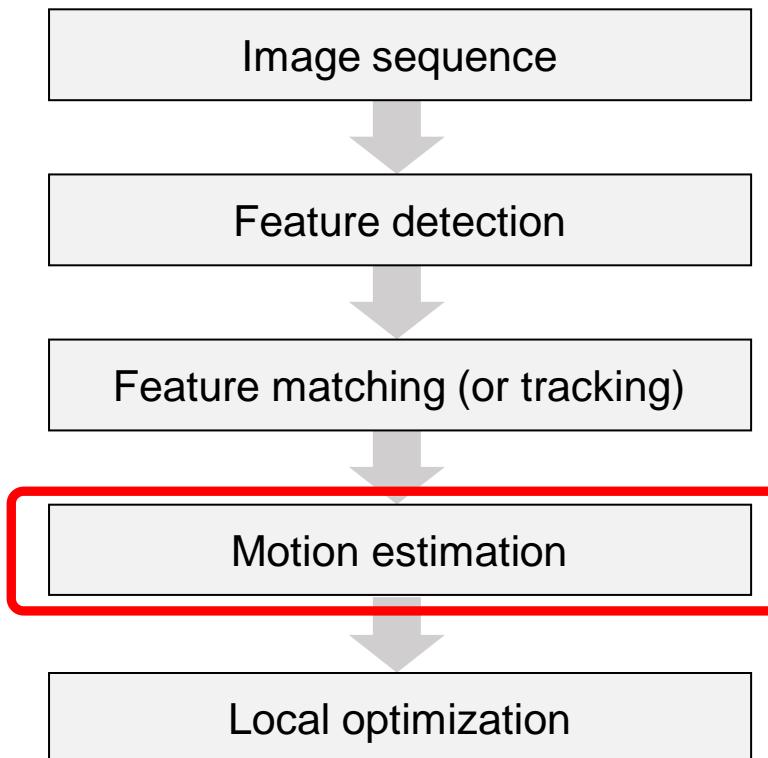
VO computes the camera path incrementally (pose after pose)



Features tracked over multiple recent frames
overlaid on the last frame

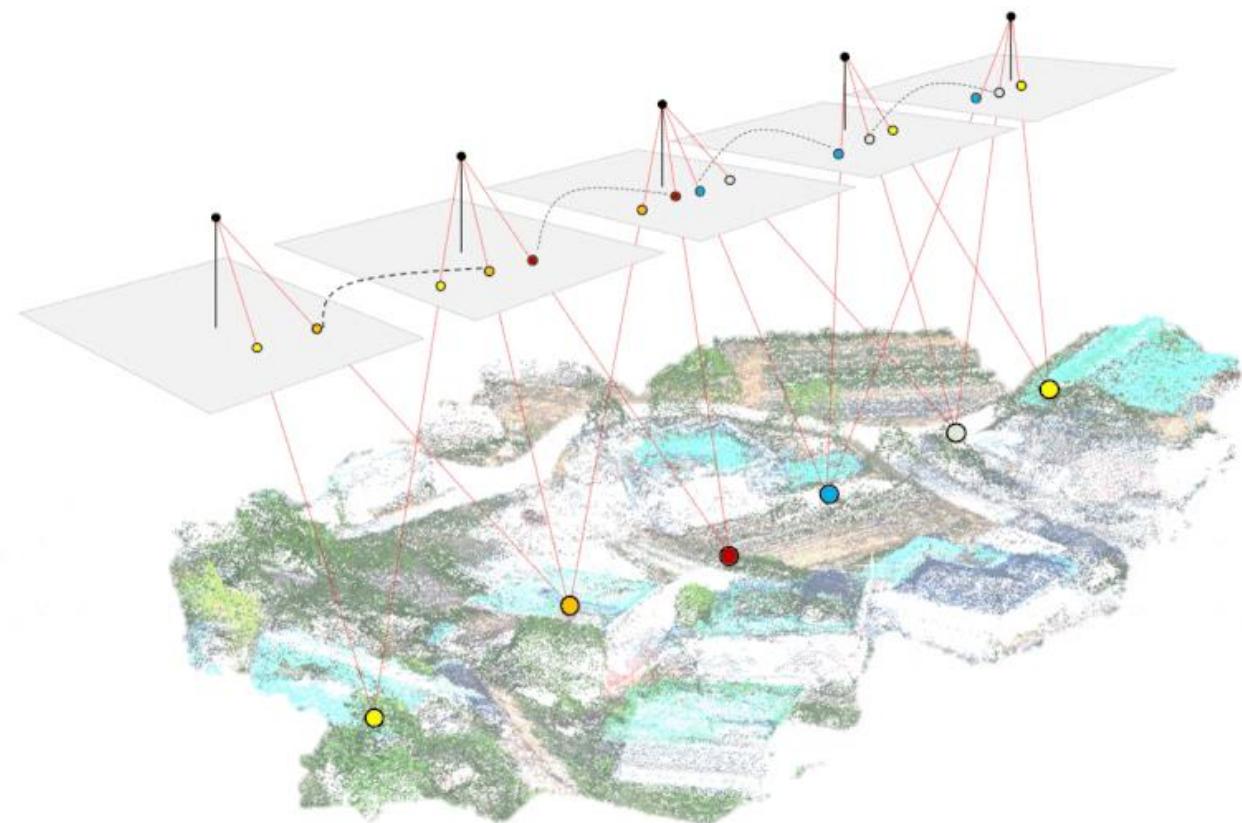
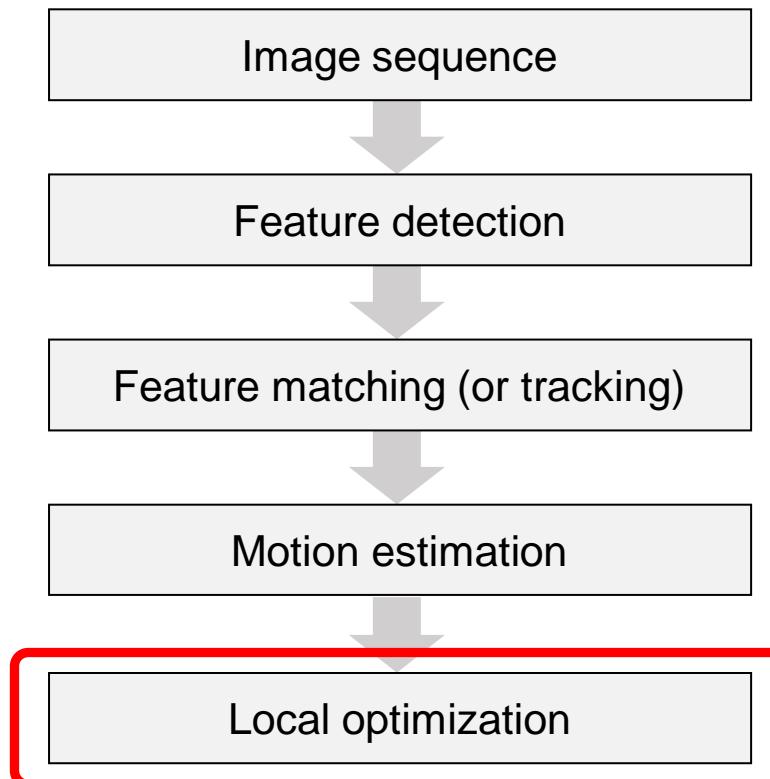
VO Flow Chart

VO computes the camera path incrementally (pose after pose)



VO Flow Chart

VO computes the camera path incrementally (pose after pose)

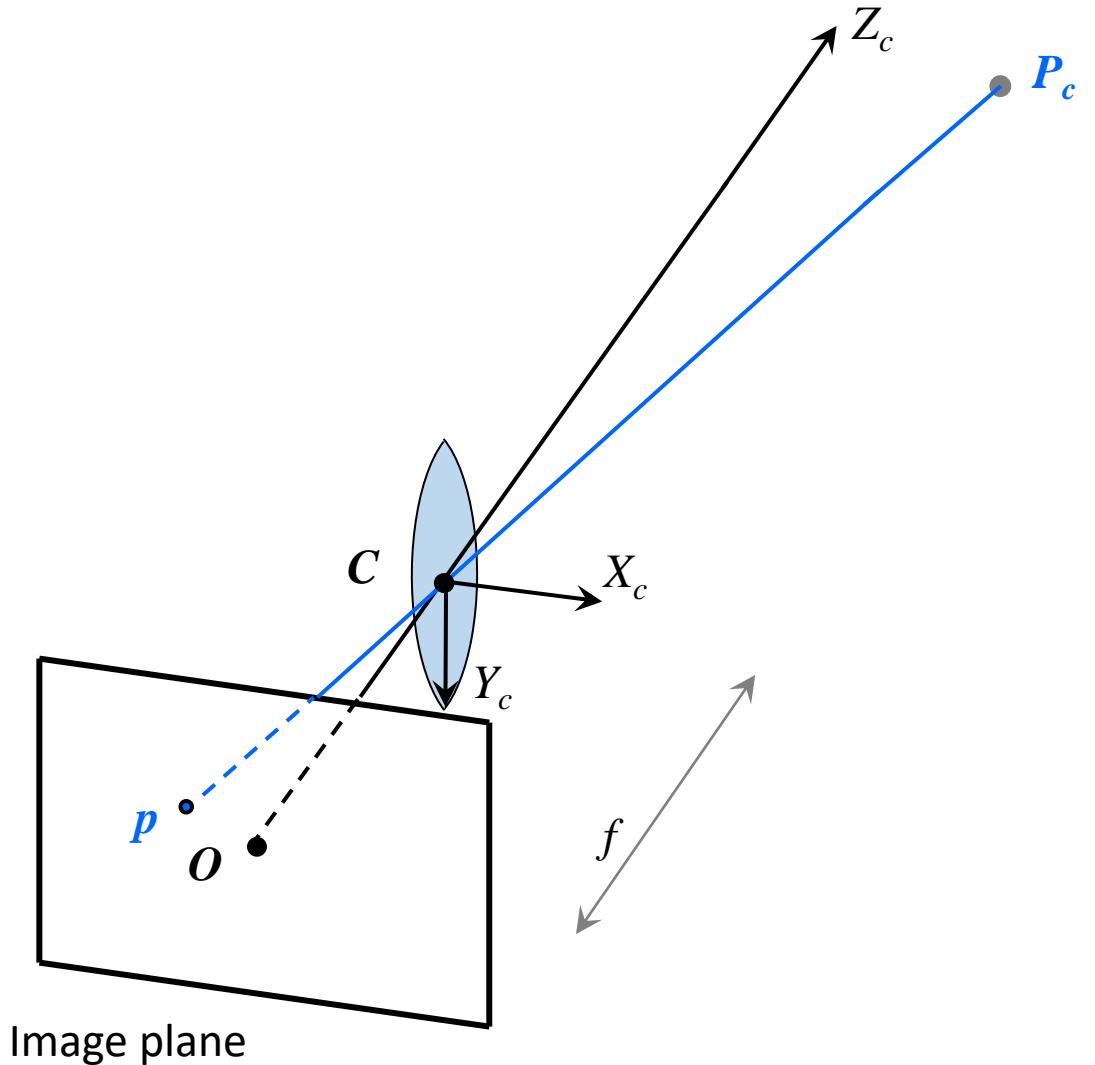
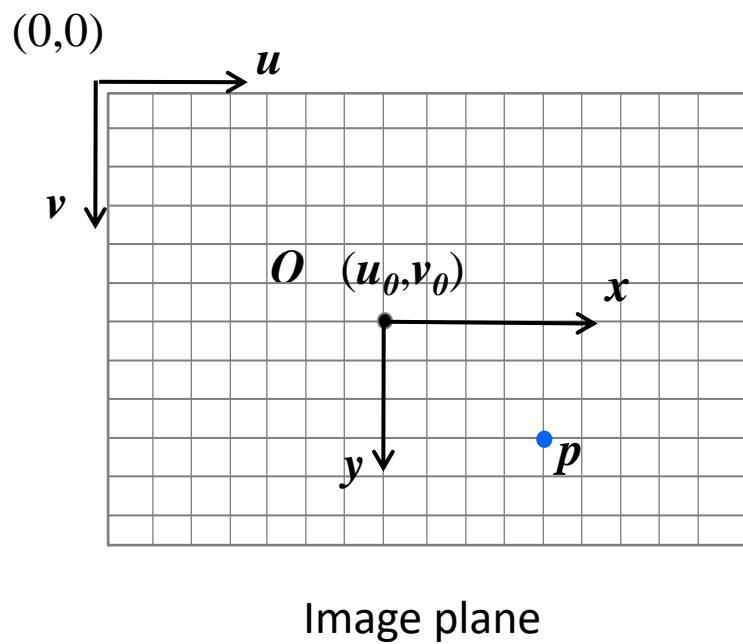


Course Topics

- Principles of image formation
- Image filtering
- Feature detection and matching
- Multi-view geometry
- Dense reconstruction
- Visual place recognition
- Deep learning
- Visual inertial fusion
- Event-based Vision

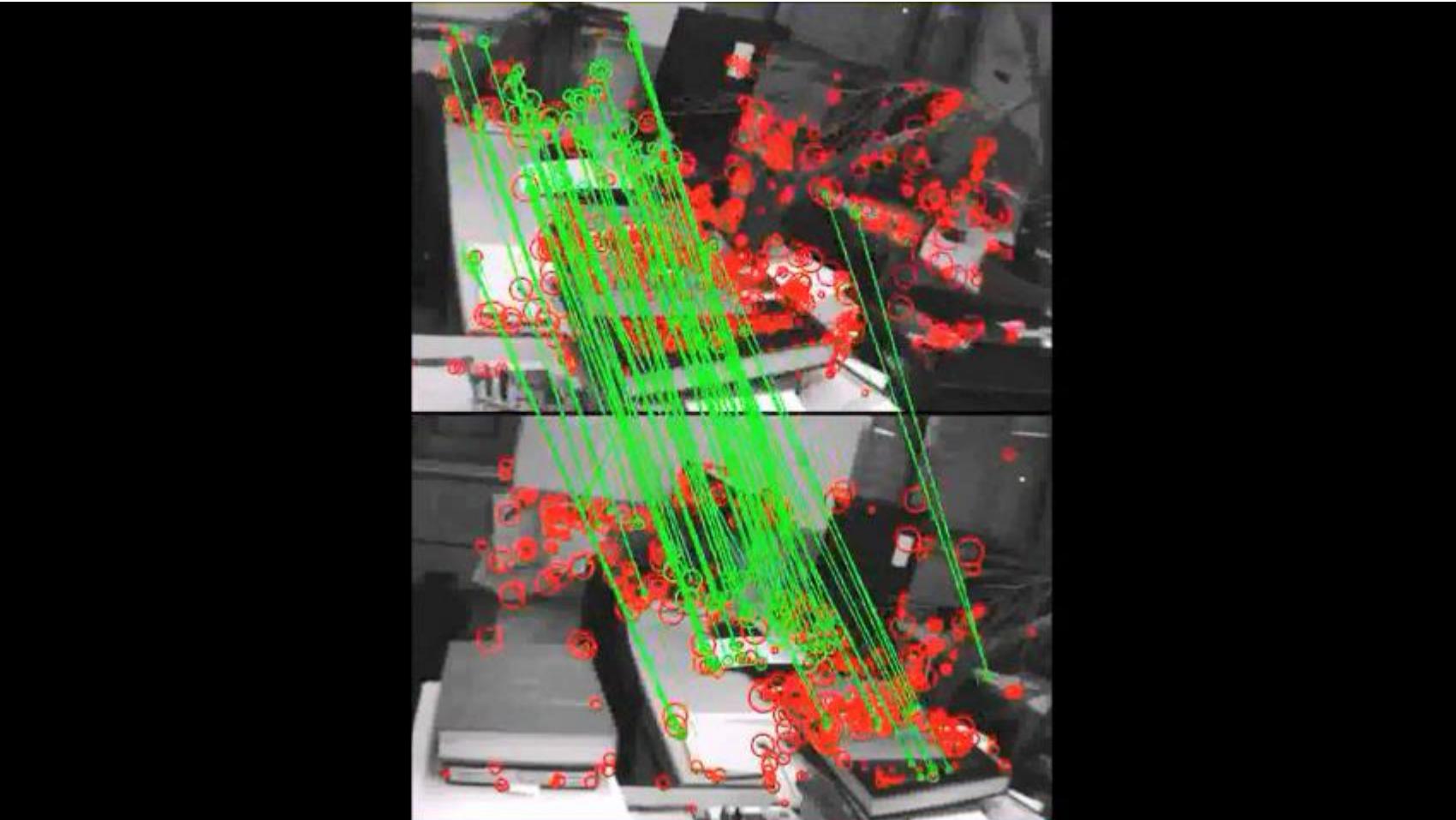
Course Topics

- Principles of image formation
 - Perspective projection
 - Camera calibration



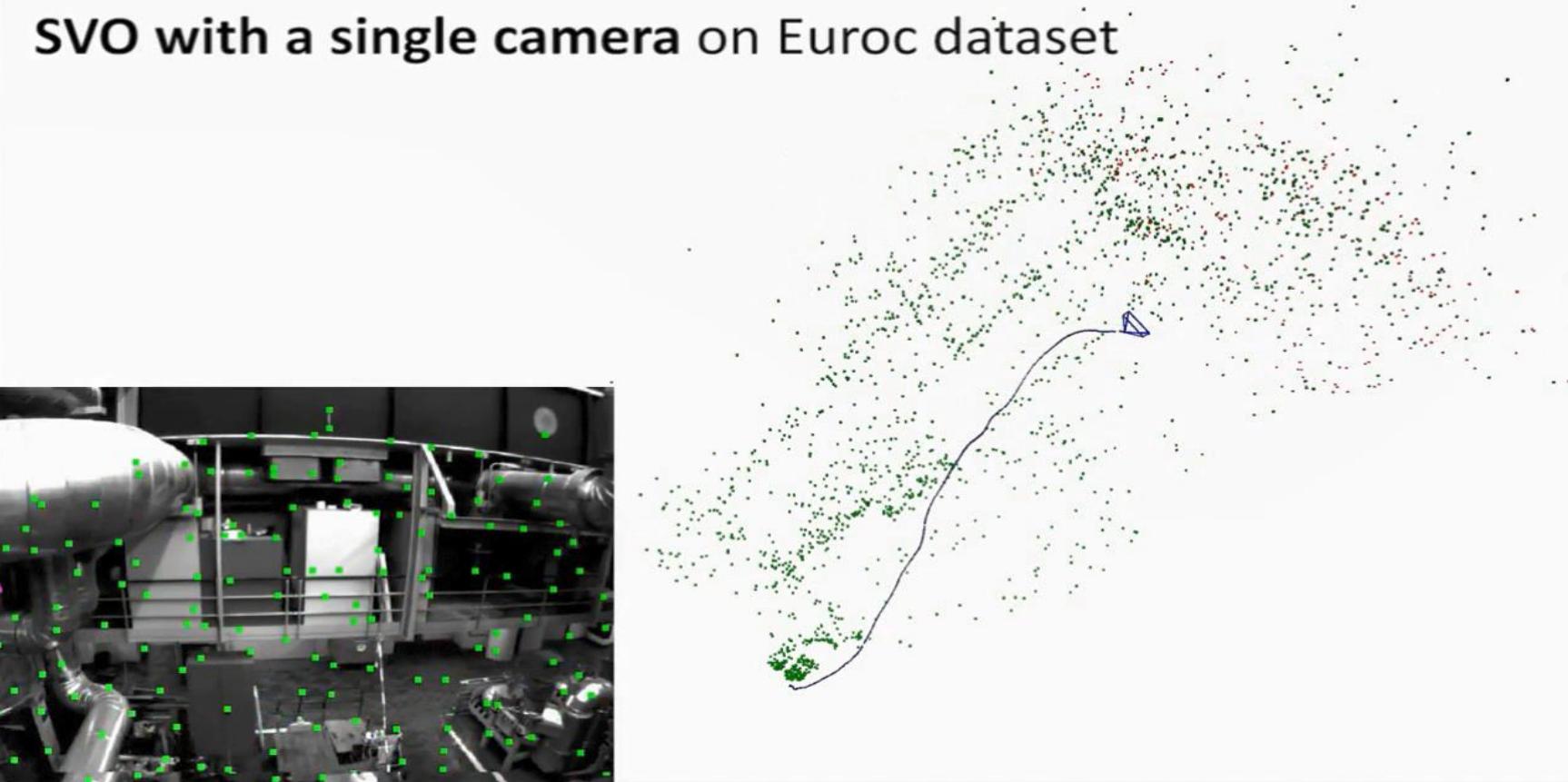
Course Topics

- Feature detection and matching



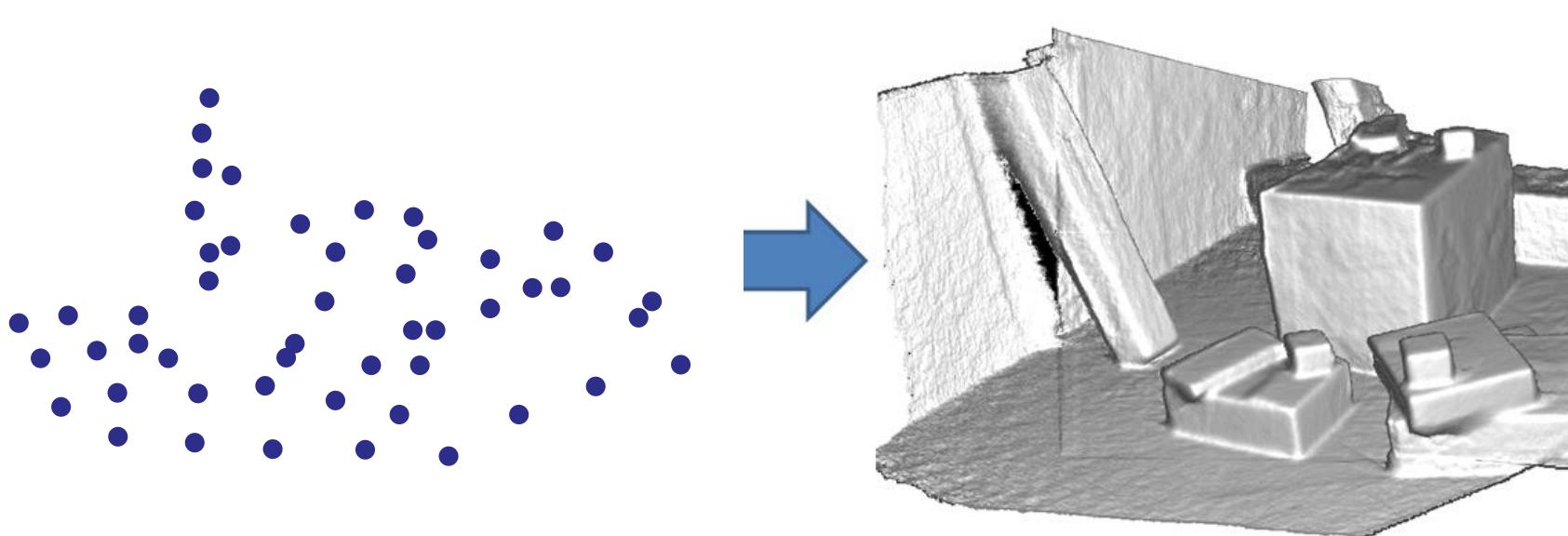
Course Topics

- Multi-view geometry and sparse 3D reconstruction



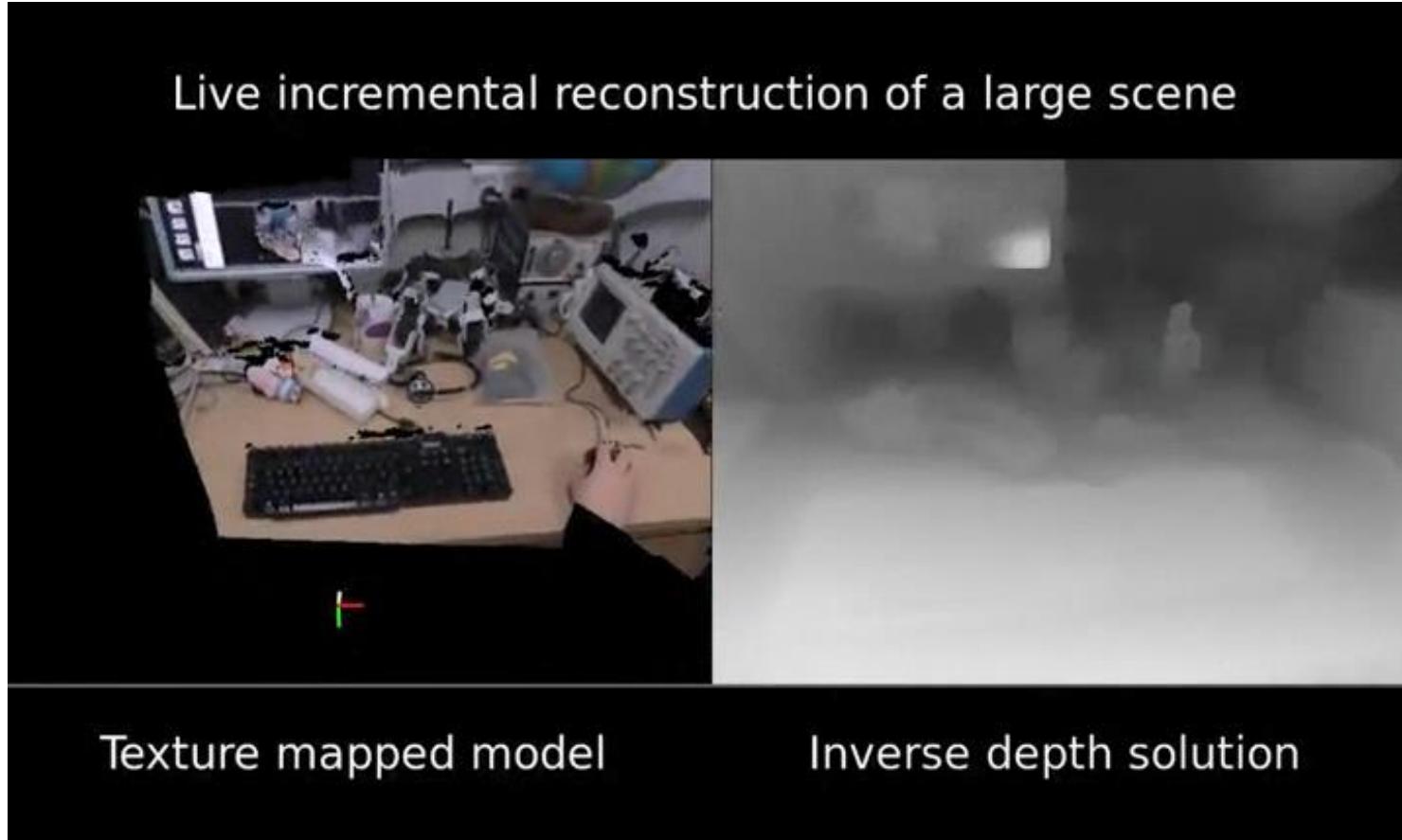
Course Topics

- Dense 3D reconstruction



Course Topics

- Dense 3D reconstruction

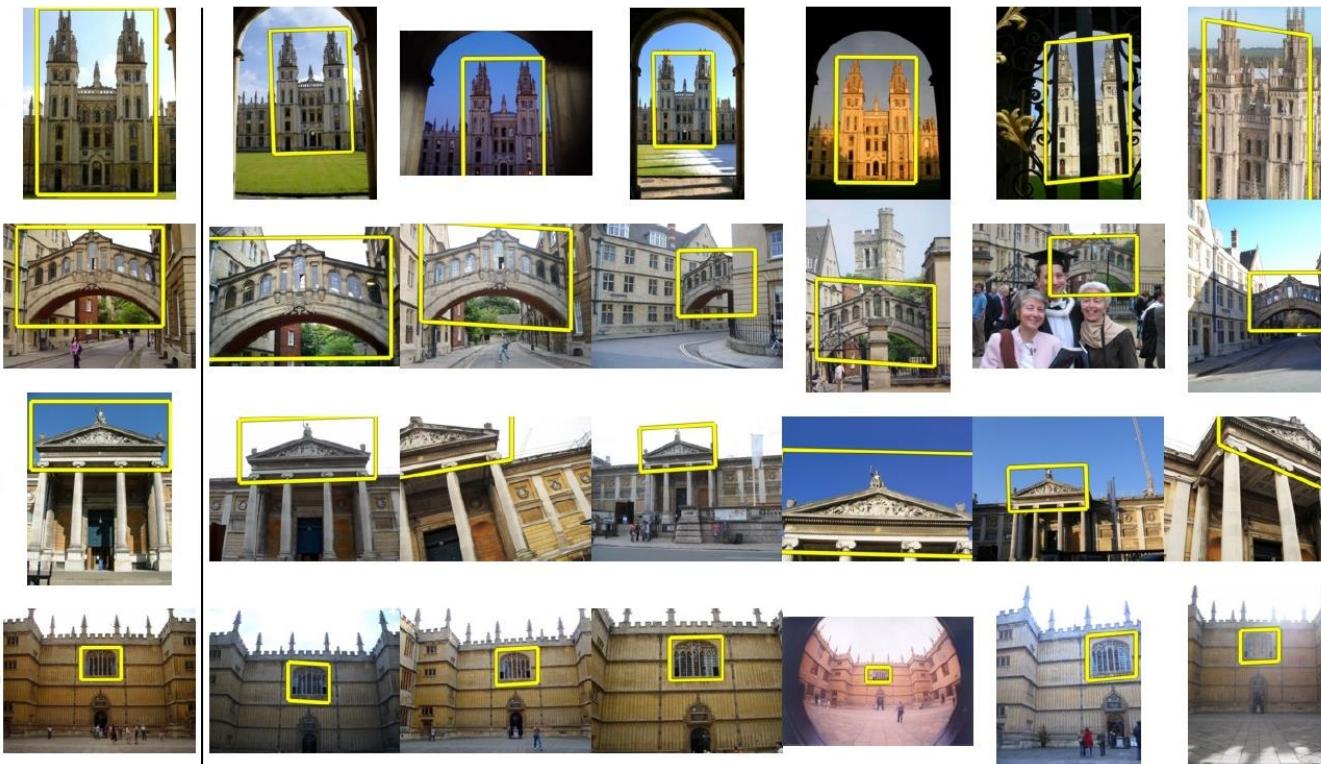


Course Topics

- Place recognition and deep learning

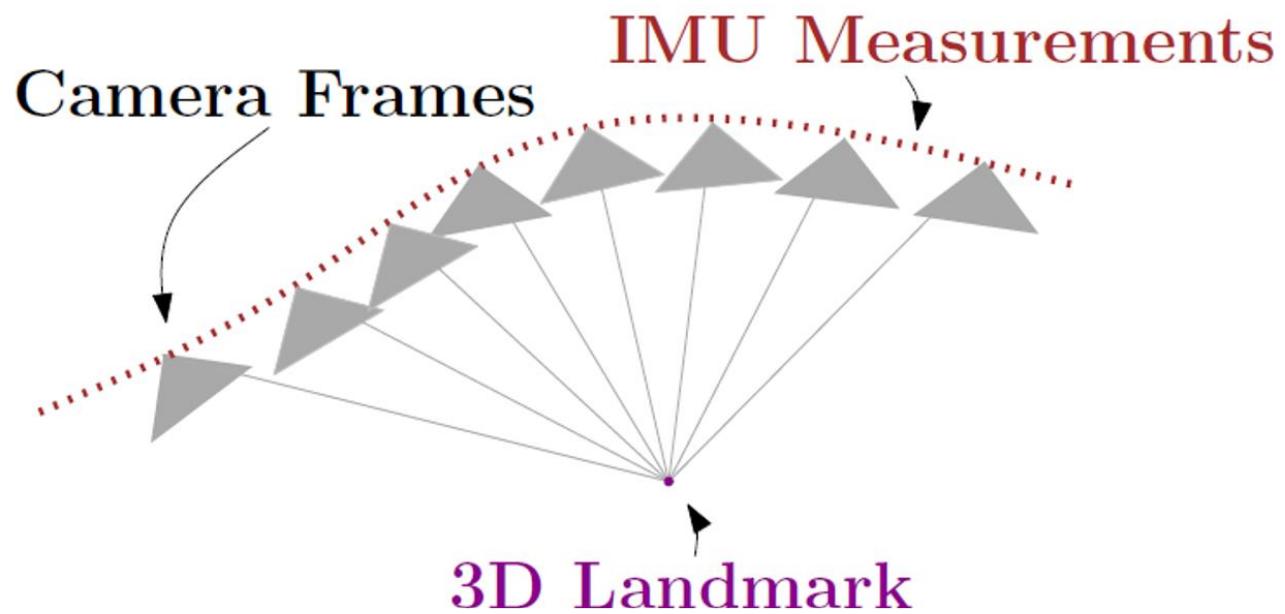
Query
image

Most similar places from a database of millions of images



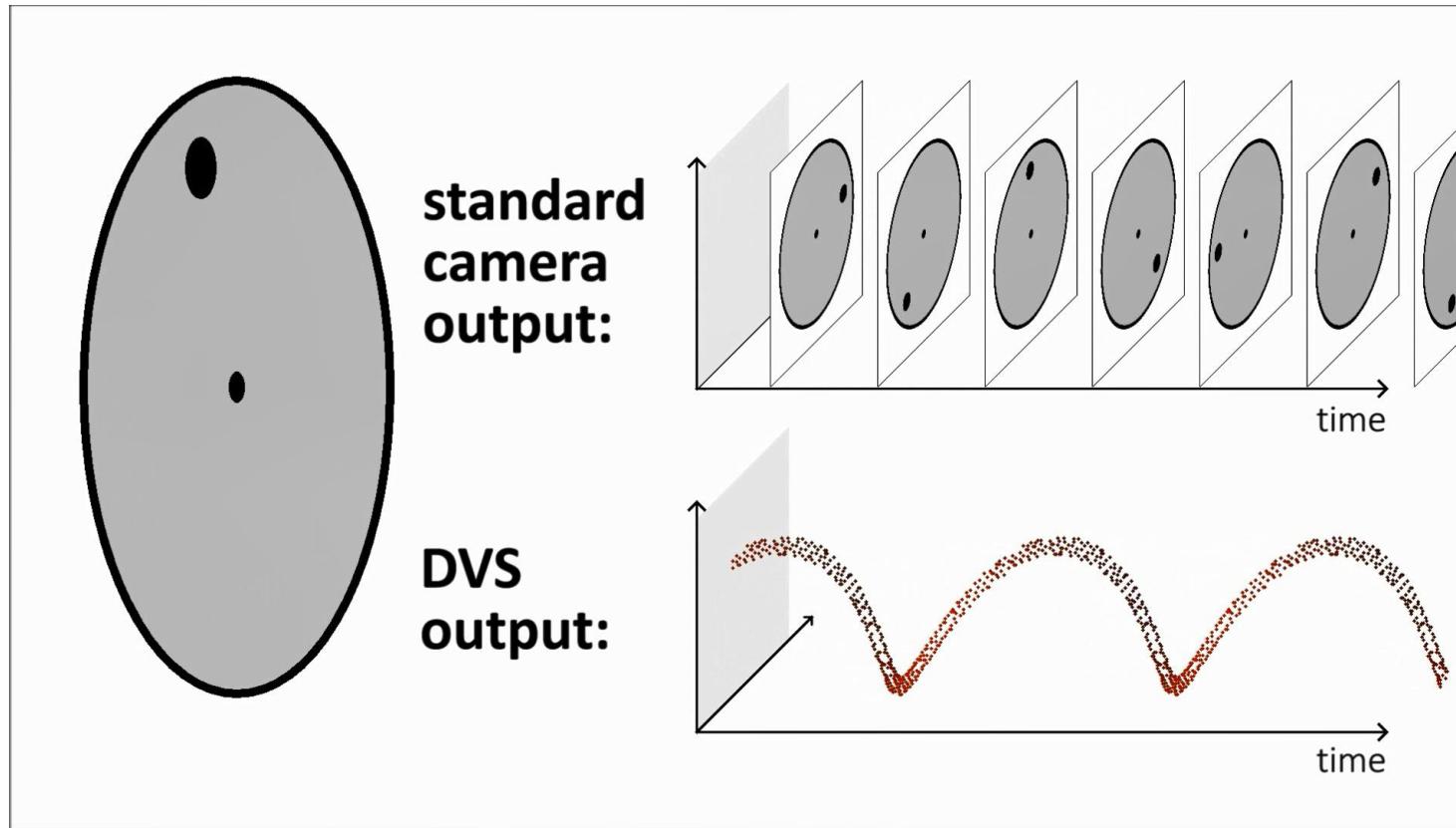
Course Topics

- Visual-inertial fusion



Course Topics

- Event cameras



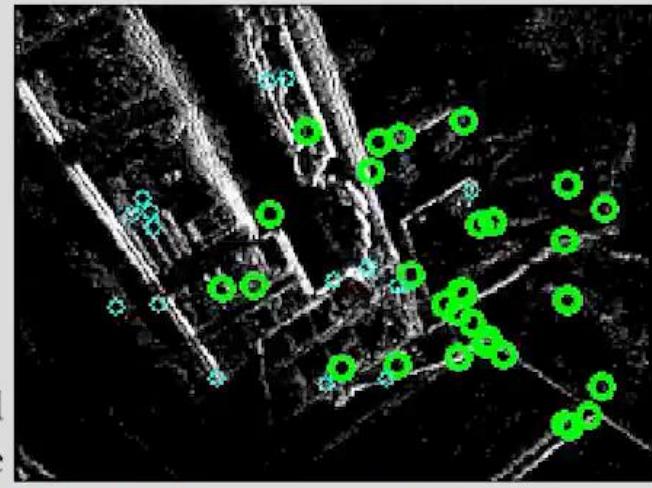
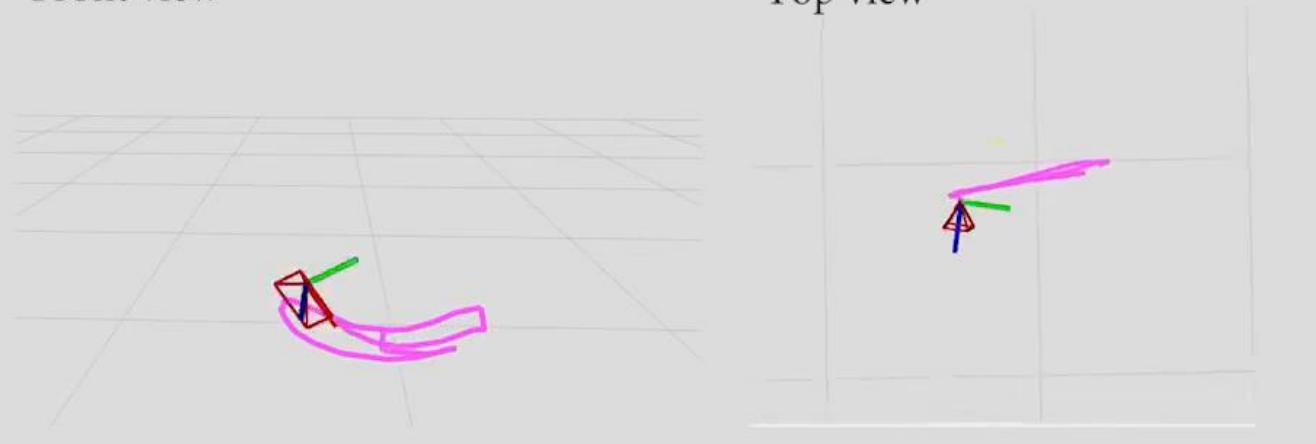
Application of event cameras: high-speed VO



Standard camera
Global shutter,
Auto-exposure on



Front view



Top view

Candidate features Persistent features

Reading

- Scaramuzza, D., Fraundorfer, F., **Visual Odometry: Part I - The First 30 Years and Fundamentals**, *IEEE Robotics and Automation Magazine*, Volume 18, issue 4, 2011. [PDF](#)
- Fraundorfer, F., Scaramuzza, D., **Visual Odometry: Part II - Matching, Robustness, and Applications**, *IEEE Robotics and Automation Magazine*, Volume 19, issue 1, 2012. [PDF](#)
- C. Cadena, L. Carlone, H. Carrillo, Y. Latif, D. Scaramuzza, J. Neira, I.D. Reid, J.J. Leonard, **Past, Present, and Future of Simultaneous Localization and Mapping: Toward the Robust-Perception Age**, *IEEE Transactions on Robotics*, Vol. 32, Issue 6, 2016. [PDF](#)

Understanding Check

Are you able to:

- Provide a definition of Visual Odometry?
- Explain the most important differences between VO, VSLAM, and SFM?
- What assumptions does VO rely on?
- Illustrate the flow chart of VO?