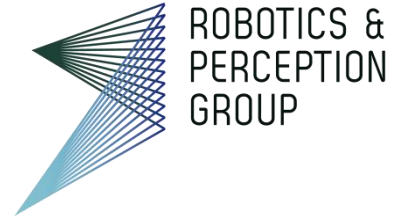




University of  
Zurich <sup>UZH</sup>

**ETH** zürich

Institute of Informatics – Institute of Neuroinformatics



# Autonomous, Agile, Vision-controlled Drones:

## From Frame to Event Vision

Daive Scaramuzza

- My lab homepage: <http://rpg.ifi.uzh.ch/>
- Publications: <http://rpg.ifi.uzh.ch/publications.html>
- Software & Datasets: [http://rpg.ifi.uzh.ch/software\\_datasets.html](http://rpg.ifi.uzh.ch/software_datasets.html)
- YouTube: <https://www.youtube.com/user/ailabRPG/videos>

# Research Overview

## Real-time, Onboard Computer Vision and Control for **Autonomous, Agile Drone Flight**

### Visual-Inertial State Estimation (~SLAM)

[http://rpg.ifi.uzh.ch/research\\_vo.html](http://rpg.ifi.uzh.ch/research_vo.html)



### Autonomous Flight

[http://rpg.ifi.uzh.ch/research\\_mav.html](http://rpg.ifi.uzh.ch/research_mav.html)



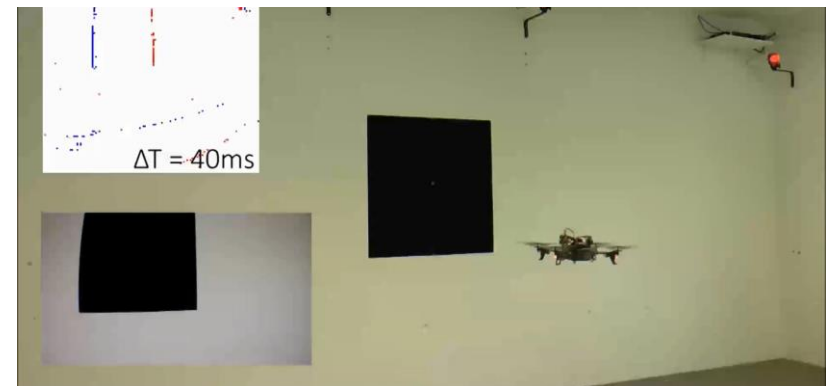
### Learning-aided Autonomous Navigation

[http://rpg.ifi.uzh.ch/research\\_learning.html](http://rpg.ifi.uzh.ch/research_learning.html)



### Event-based Vision for Low-latency Control

[http://rpg.ifi.uzh.ch/research\\_dvs.html](http://rpg.ifi.uzh.ch/research_dvs.html)



# Today's Main Applications of Drones (PWC report 2017)

Infrastructure Inspection



Agriculture



Aerial photography



Transportation



Security & Search and rescue

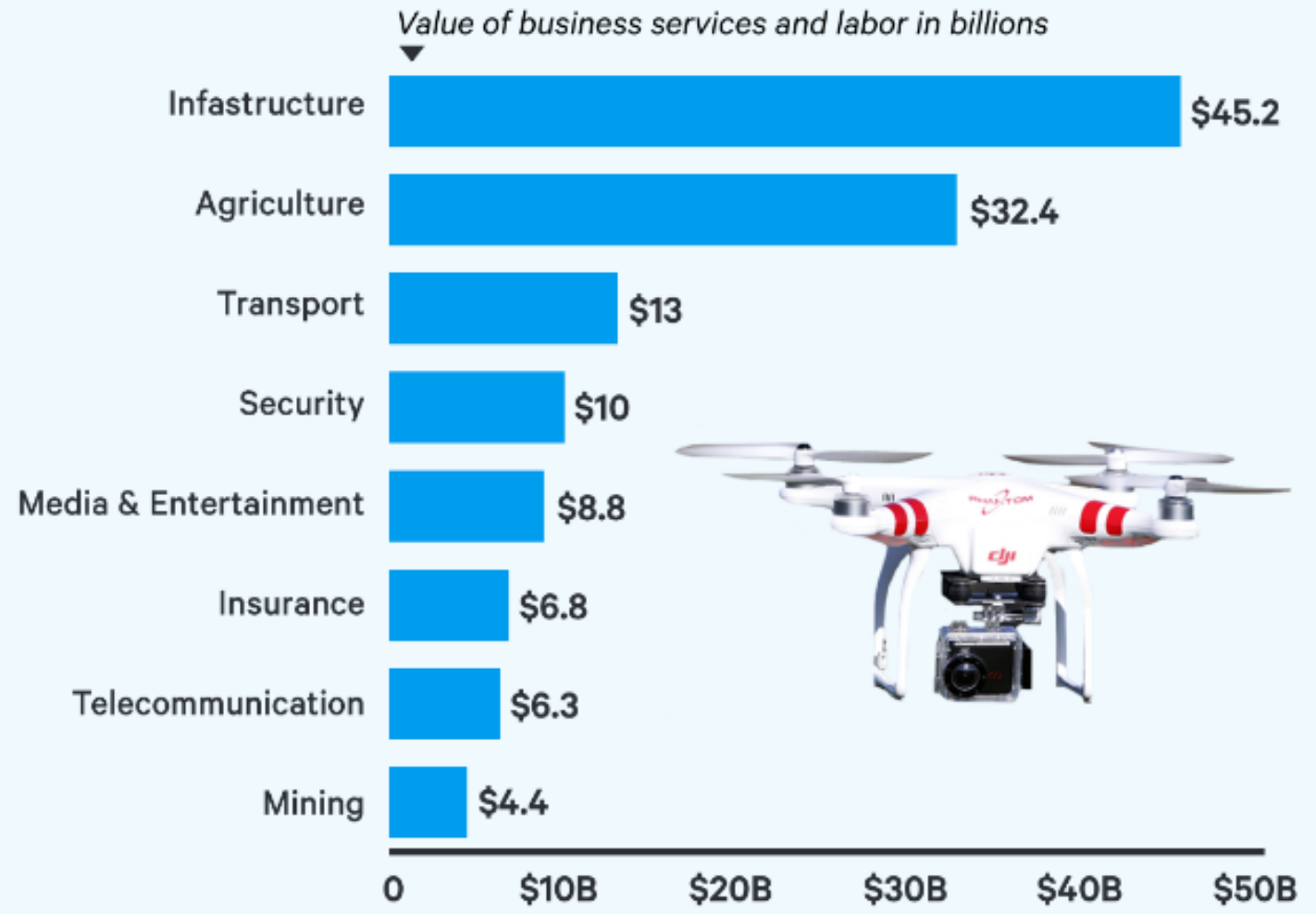


Law enforcement



# The Drone Market (current value: 130B\$) [PWC report 2017]

## Predicted value of drones by industry



SOURCE: PwC

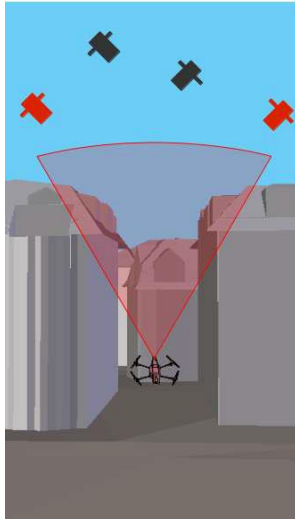
# How to fly a drone

## ➤ Remote control

- Requires **line of sight** or **communication link**
- Requires **skilled pilots**

## ➤ GPS-based navigation

- Does not work indoors!
- Can be unreliable outdoors (e.g., low altitude in urban canyons, under bridges, etc.)

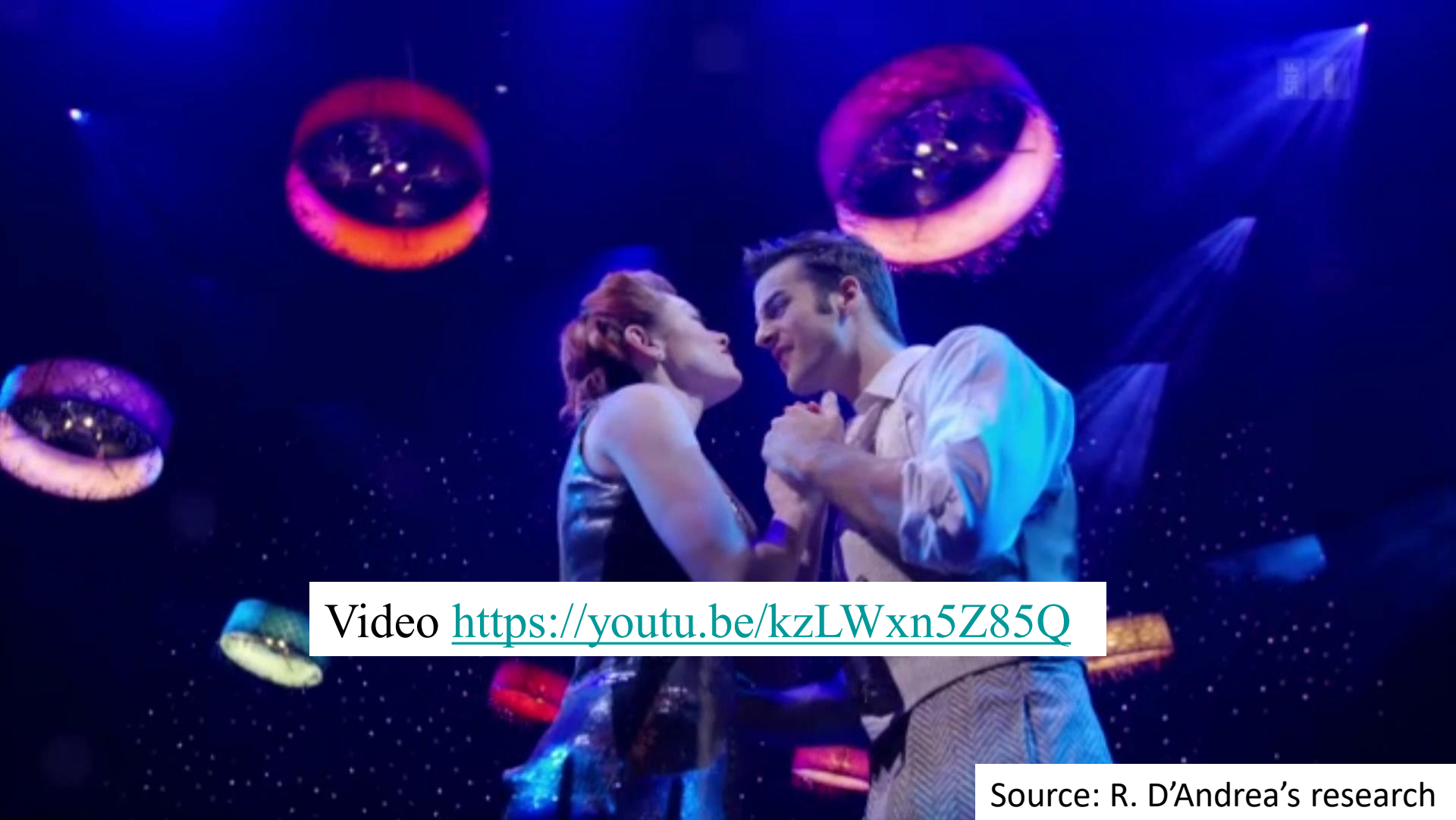


Drone crash during soccer match, Brasilia, 2013

# My Goal: Flying Robots to the Rescue!



# State of the Art on Autonomous Drone Navigation

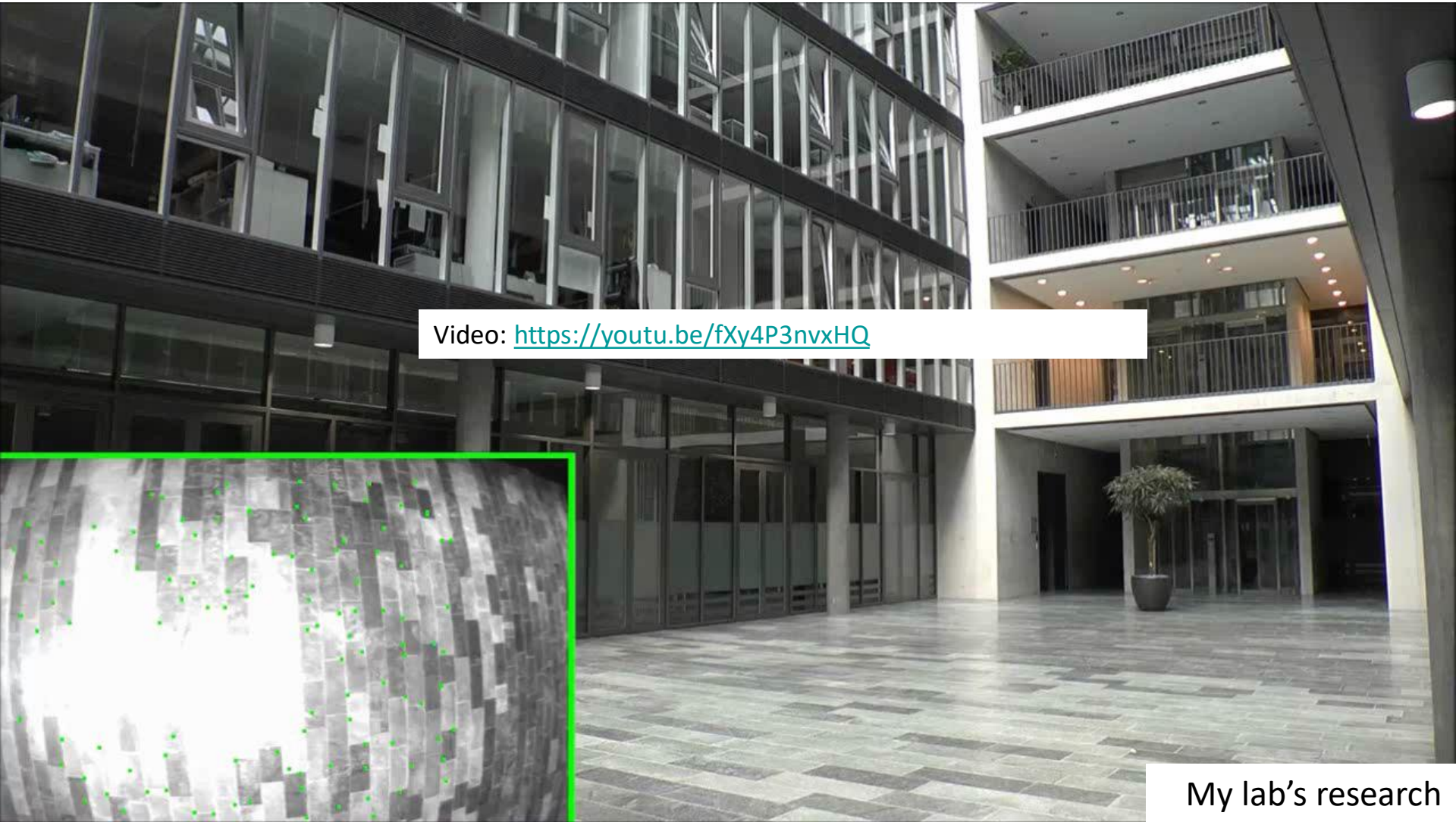


Video <https://youtu.be/kzLWxn5Z85Q>

Source: R. D'Andrea's research

... but these robots are completely “**blind**”

# State of the Art on Autonomous Drone Navigation



Video: <https://youtu.be/fXy4P3nvxHQ>

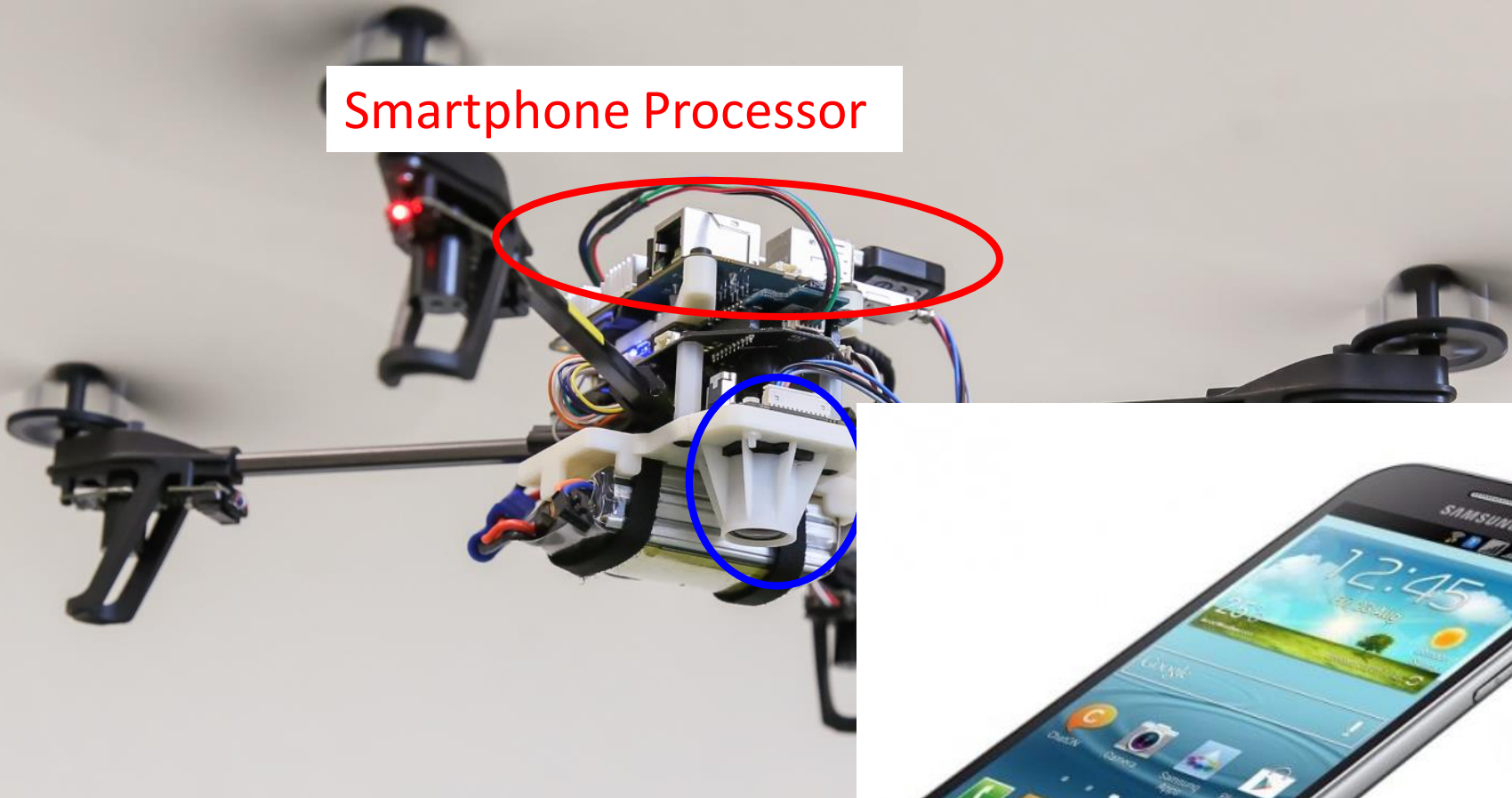
My lab's research

...while this robot can “see”



# System Overview

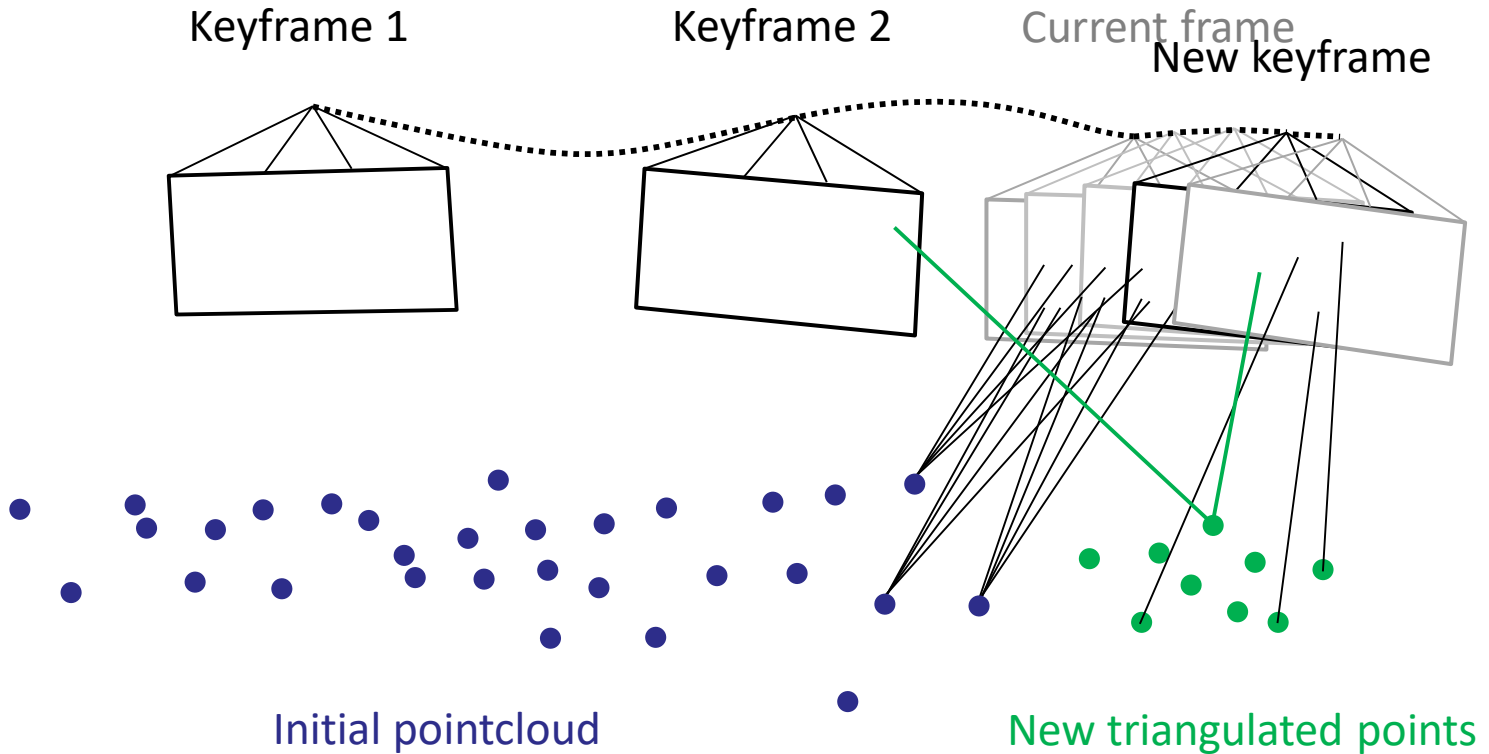
Smartphone Processor



450 grams



# Simultaneous Localization And Mapping (SLAM)



Also used in several open-source monocular systems:  
PTAM, LSD-SLAM, ORBSLAM, OKVIS, DSO, SVO

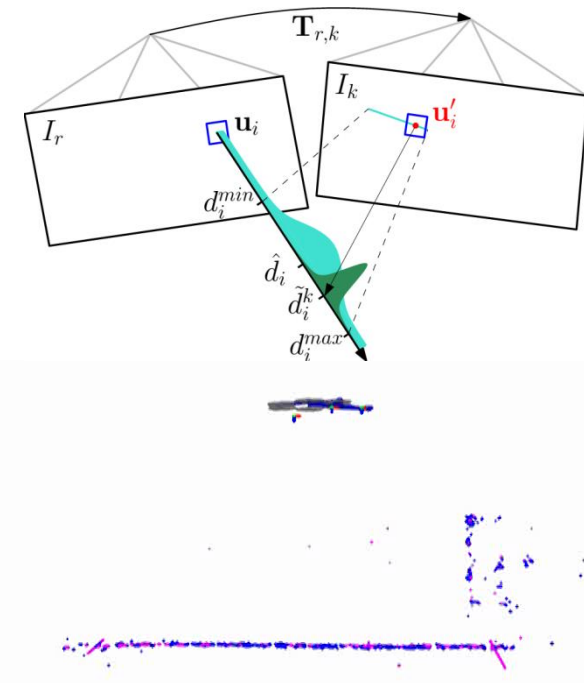
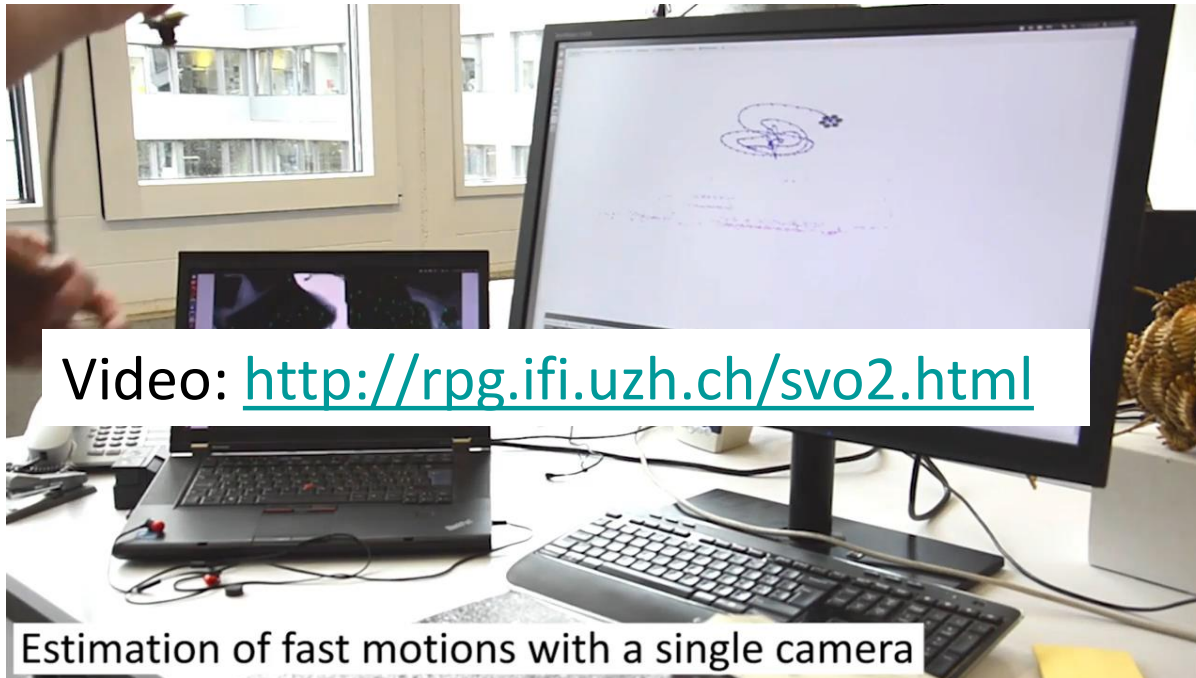
# SVO: Semi-direct Visual Odometry [ICRA'14, [TRO'17](#)]

- Jointly tracks features and 6DoF motion under mostly-rigid world assumption
- Minimizes both photometric and geometric error

## Achieves lowest latency & CPU load

- 2.5ms (400 fps) on i7 laptops
- 10ms (100 fps) on smartphones

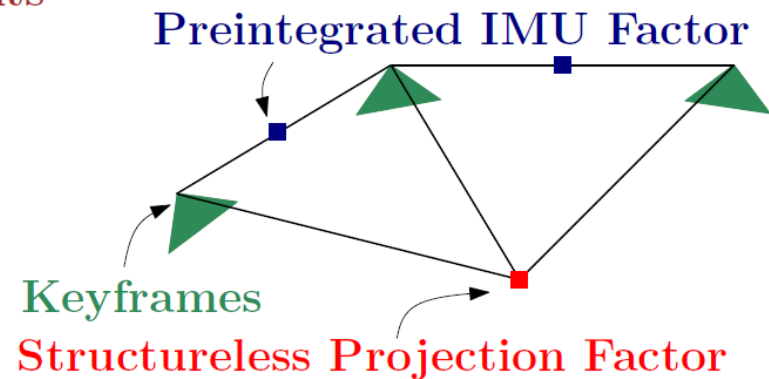
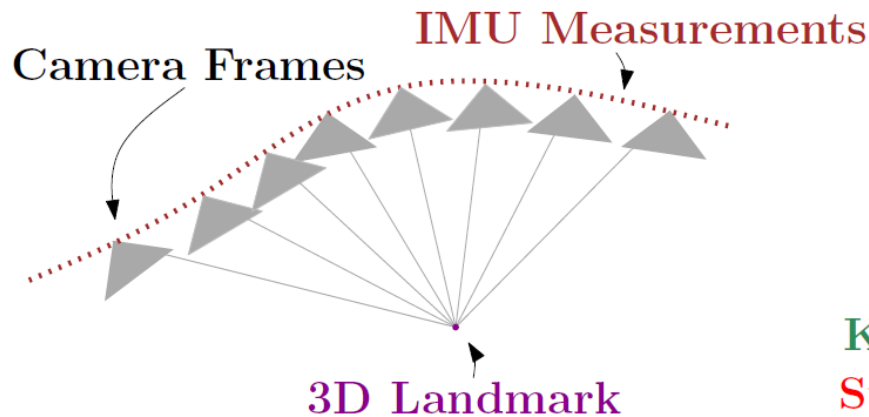
	Mean	St.D.	CPU@20 fps
SVO Mono	2.53	0.42	55 ±10%
ORB Mono SLAM (No loop closure)	29.81	5.67	187 ±32%
LSD Mono SLAM (No loop closure)	23.23	5.87	236 ±37%



Download from : <http://rpg.ifi.uzh.ch/svo2.html>

# Visual-Inertial Odometry via Full-Smoothing

- Full smoothing methods **estimate the entire history of the states** (camera trajectory and 3D landmarks), by solving a large nonlinear optimization problem
- **Superior accuracy over filtering** methods, which only update the last state
- Solved using **phactor graphs** (iSAM): only update variables affected by a new measurement



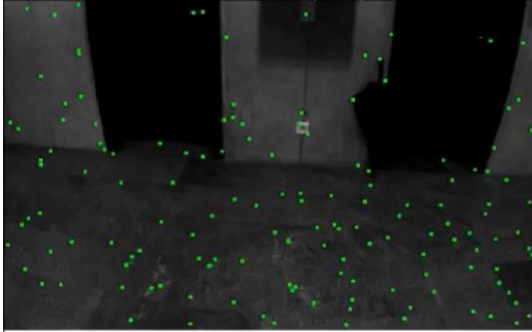
$$\sum_{(i,j) \in \mathcal{K}_k} \|\mathbf{r}_{\mathcal{I}_{ij}}\|_{\Sigma_{ij}}^2 + \sum_{i \in \mathcal{K}_k} \sum_{l \in \mathcal{C}_i} \|\mathbf{r}_{\mathcal{C}_{il}}\|_{\Sigma_c}^2$$


*IMU residuals*      *Reprojection residuals*

**Open Source**  
<https://bitbucket.org/gtborg/gtsam>

1. Forster, Carlone, Dellaert, Scaramuzza, *On-Manifold Preintegration for Real-Time Visual-Inertial Odometry*, IEEE Transactions on Robotics 2017, **TRO'17 Best Paper Award**. [PDF](#), [Video](#)
2. Delmerico, Scaramuzza, *A Benchmark Comparison of Monocular Visual-Inertial Odometry Algorithms*, ICRA'18, [PDF](#), [Video](#)

# Comparison to Google Tango and OKVIS

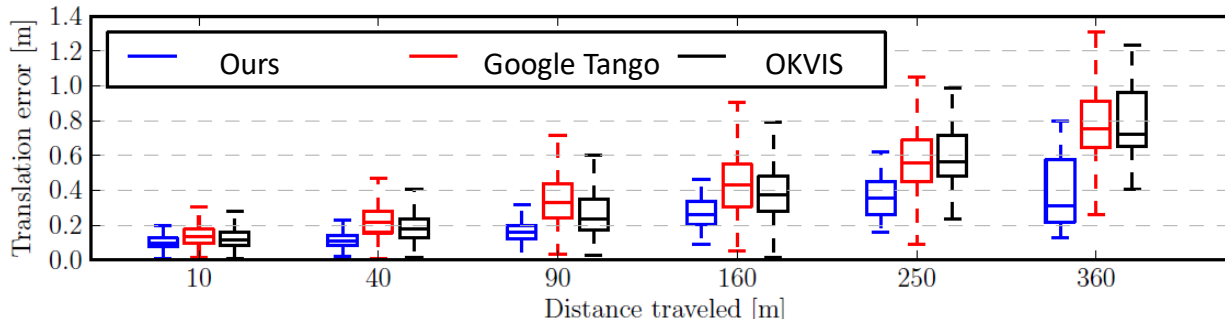
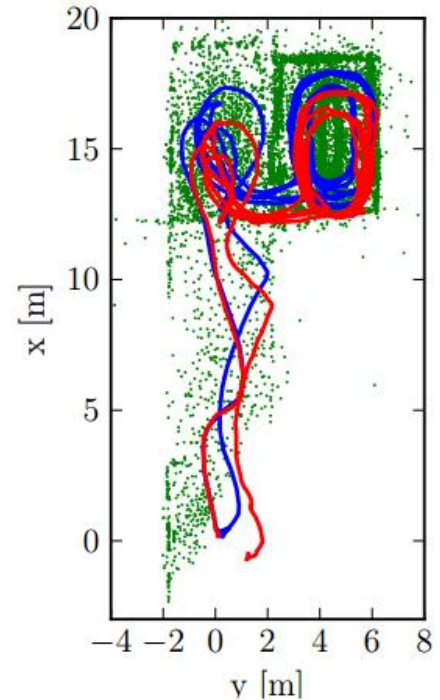


Video: 

<https://youtu.be/CsJkci5lfco>

5x

Accuracy: 0.1% of the travel distance



1. Forster, Carlone, Dellaert, Scaramuzza, *On-Manifold Preintegration for Real-Time Visual-Inertial Odometry*, IEEE Transactions on Robotics 2017, **TRO'17 Best Paper Award**. [PDF](#), [Video](#)
2. Delmerico, Scaramuzza, *A Benchmark Comparison of Monocular Visual-Inertial Odometry Algorithms*, ICRA'18, [PDF](#), [Video](#)

# Autonomous Visual Navigation



Video: [https://youtu.be/\\_-p08o\\_oTO4](https://youtu.be/_-p08o_oTO4)

Scaramuzza, Fraundorfer, Pollefeys, Siegwart, Achtelick, Weiss, et al., *Vision-Controlled Micro Flying Robots: from System Design to Autonomous Navigation and Mapping in GPS-denied Environments*, RAM'14, [PDF](#)

# DARPA FLA Program (June 2015 – June 2018)



Video: <https://youtu.be/6eeetSVHXPk>



Mohta, Loiano, Scaramuzza, Daniilidis, Taylor, Kumar, Fast, Autonomous Flight in GPS-denied and Cluttered Environments, *Journal of Field Robotics*, 35 (1), 2018, [PDF](#), [Video](#)

# Robustness to “Strong Disturbances”!



Video: <https://youtu.be/pGU1s6Y55JI>

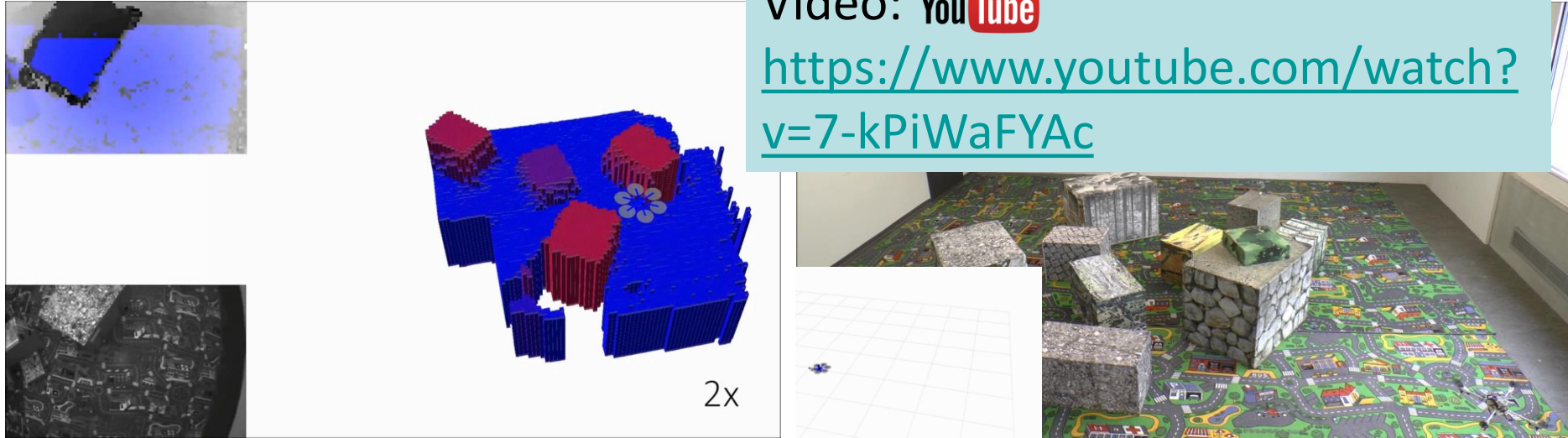
View from the onboard camera

Faessler, Fontana, Forster, Scaramuzza, [Automatic Re-Initialization and Failure Recovery for Aggressive Flight with a Monocular Vision-Based Quadrotor](#), ICRA'15. [Featured in IEEE Spectrum.](#)



# Autonomus, Live, Dense Reconstruction

REMODE: probabilistic, REgularized, MOnocular DEnse reconstruction in real time [ICRA'14]  
State estimation with SVO 2.0



Running at 25Hz onboard (Odroid U3) - Low res.

Running live at 50Hz on laptop GPU – HD res.

Open Source

[https://github.com/uzh-rpg/rpg\\_open\\_remode](https://github.com/uzh-rpg/rpg_open_remode)

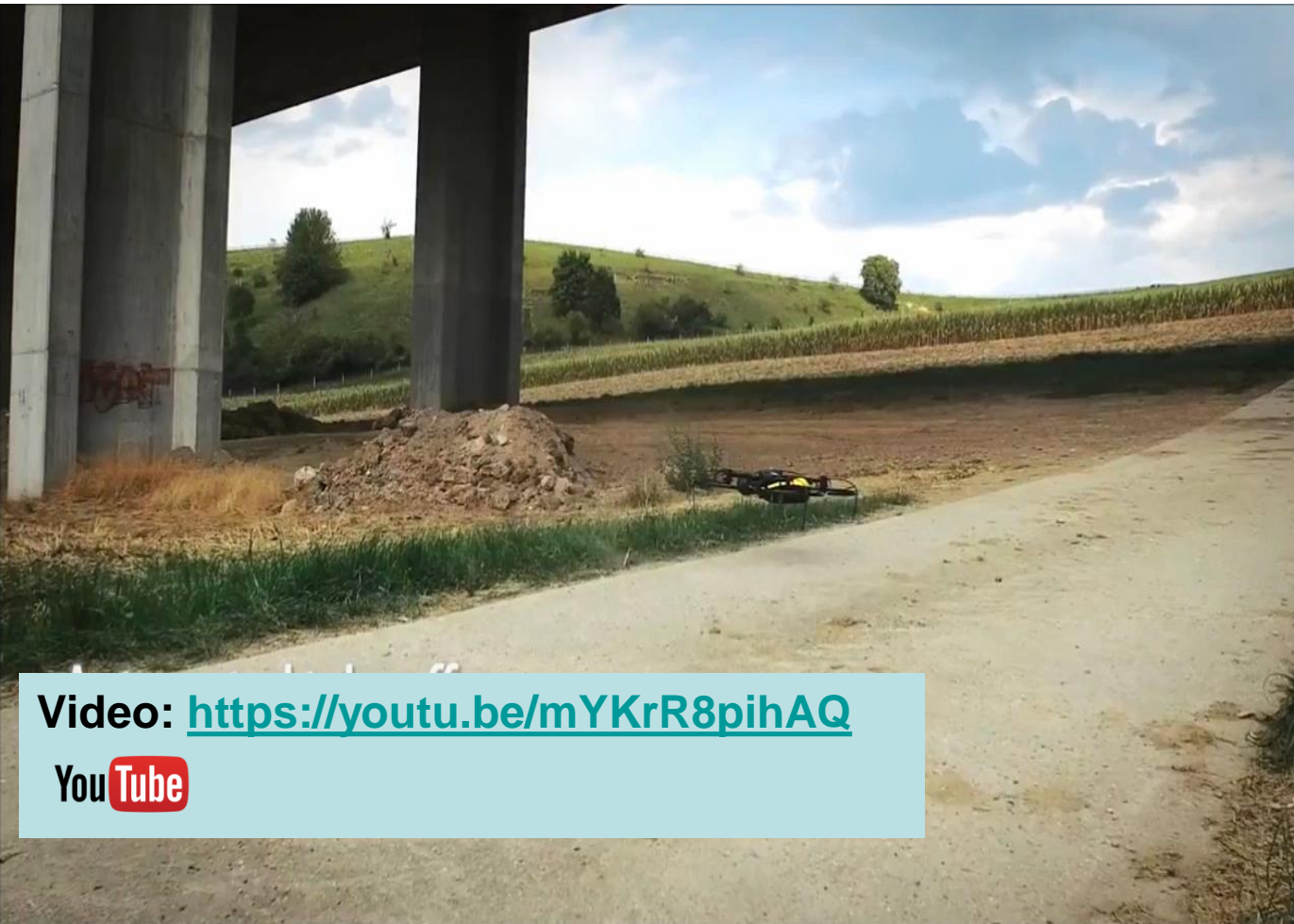
1. [Pizzoli et al., \*REMODE: Probabilistic, Monocular Dense Reconstruction in Real Time\*, ICRA'14\]](#)
2. [Forster et al., \*Appearance-based Active, Monocular, Dense Reconstruction for Micro Aerial Vehicles\*, RSS' 14](#)
3. [Forster et al., \*Continuous On-Board Monocular-Vision-based Elevation Mapping Applied ...\*, ICRA'15.](#)
4. [Faessler et al., \*Autonomous, Vision-based Flight and Live Dense 3D Mapping ...\*, JFR'16](#)

# Industrial Applications of my Research

# Parrot Albris (2016)

Parrot senseFly

- 5 wide angle cameras (one is pan-tilt) + one thermal camera
- “Tap and Fly” Tablet Interface



Video: <https://youtu.be/mYKrR8pihAQ>

You Tube



Powered by SVO 2. (Download: <http://rpg.ifi.uzh.ch/svo2.html>)

# Dacuda 3D (now Magic Leap Zurich)

➤ Fully immersive VR (running on iPhone)



Dacuda's  
3D division



magic  
leap

Video: <https://youtu.be/mYKrR8pihAQ>

You 



# Zurich-Eye (now Oculus Zurich)

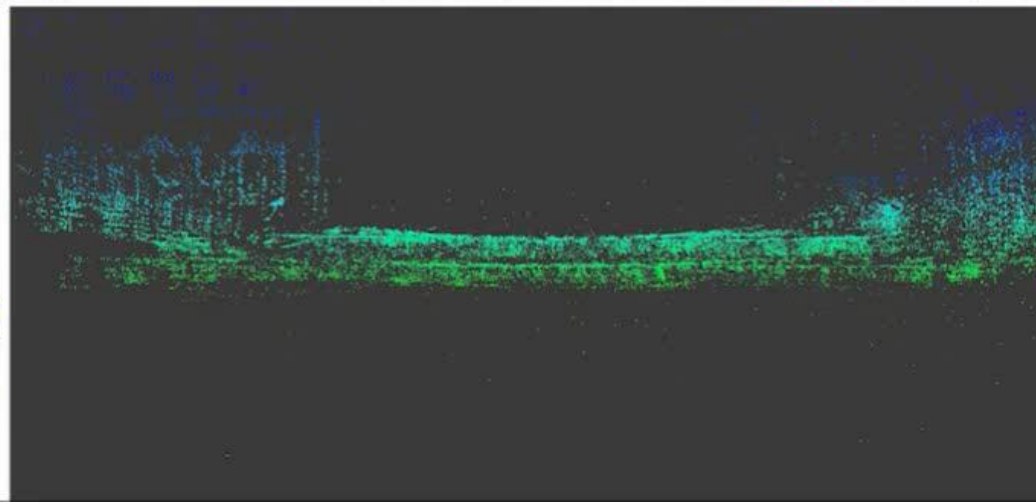
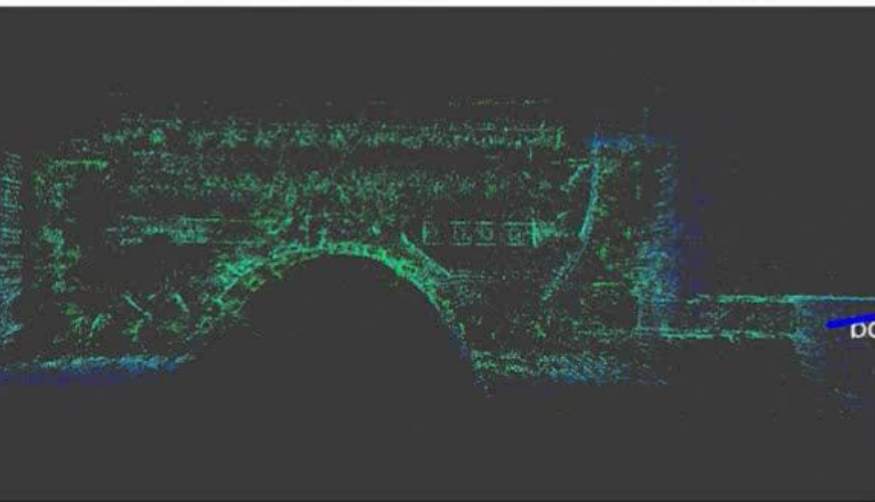
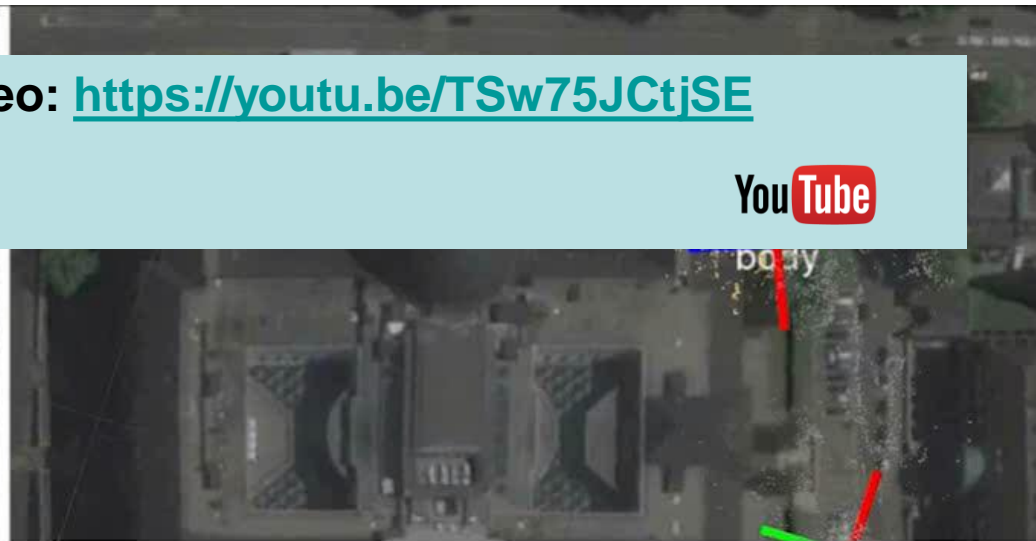
Vision-based Localization and Mapping Solutions for Mobile Robots

Created in Sep. 2015, **became Facebook-Oculus Zurich in Sep. 2016**



Video: <https://youtu.be/TSw75JCtjSE>

YouTube



What's next?

# My Dream Robot: Fast, Lightweight, Autonomous!



Video: <https://youtu.be/JDvcBuRSDUU>

LEXUS commercial, 2013 – Created by Kmel, now Qualcomm

**WARNING!**

There are 50 drones in this video but 40 are CGI and  
10 are controlled via a Motion Capture System 😊

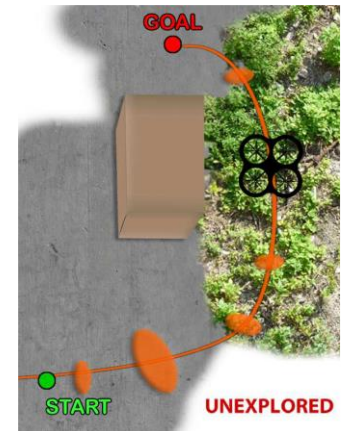
But this is just a vision!  
How to get there?



# Open Challenges

Perception algorithms are **mature but not robust**

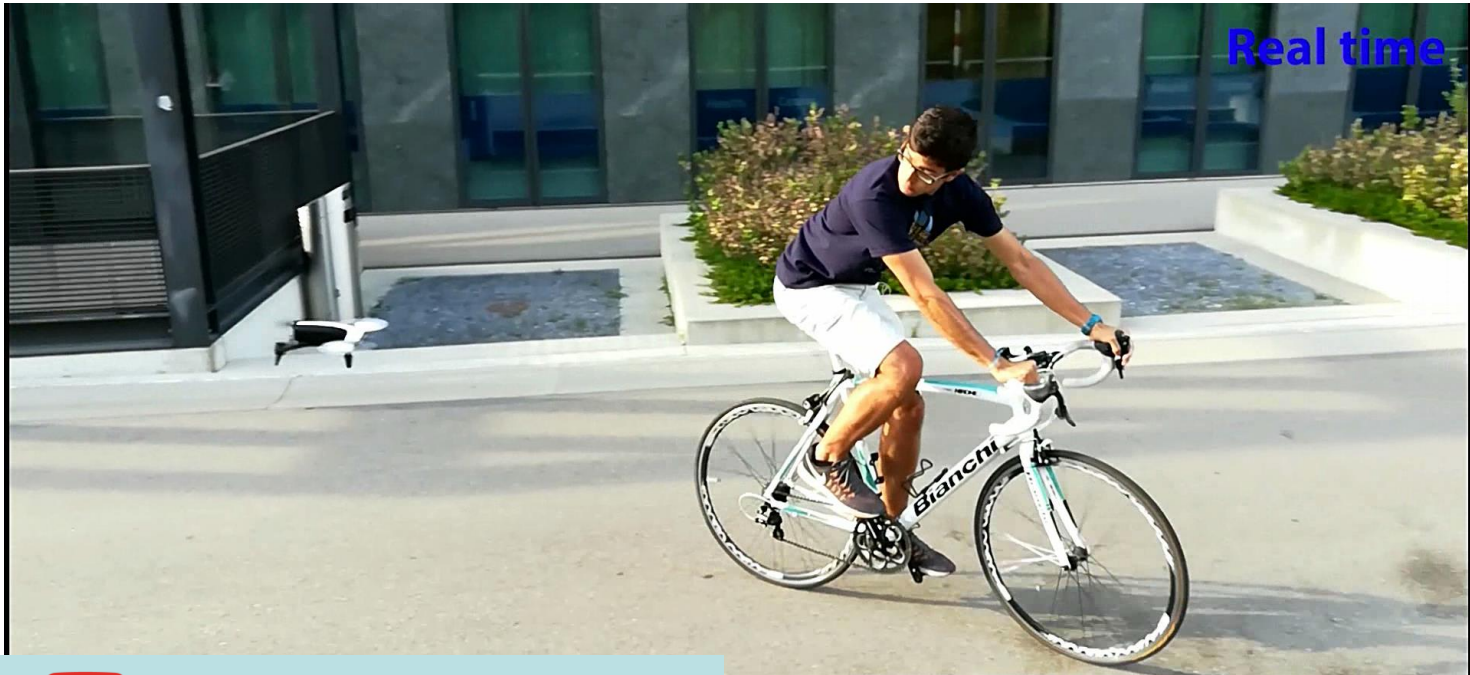
- Problems with **low texture, HDR scenes, motion blur**
- Algorithms and sensors have **big latencies** (50-200 ms) → need faster sensors
- Need **accurate models** of the **sensors** and the **environment**
- **Control & Perception** are often **considered separately** (e.g., perception, state estimation, and planning are treated as separate blocks)




# Deep-Learning based Navigation

# DroNet: Learning to Fly by Driving

- DroNet learns to follow streets autonomously, without interaction
- The hardest problem in Machine Learning is **data collection**
- **Our idea: learn to fly autonomously by mimicking cars and bicycles!**



Video: 

<https://youtu.be/ow7aw9H4BcA>

, promptly reacting to them.

[Loquercio, DroNet: Learning to Fly by Driving, IEEE RAL'18  
[PDF](#). Featured on [IEEE Spectrum](#), [MIT Technology Review](#),  
and [Discovery Channel Global](#)

Code & Datasets:

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

# Low-latency, Event-based Vision

# Latency and Agility are tightly coupled!

Current flight maneuvers achieved with onboard cameras are still too slow compared with those attainable by **birds**. We need **faster sensors and algorithms!**



A sparrowhawk catching a garden bird (National Geographic)

# What does it take to fly like an eagle?

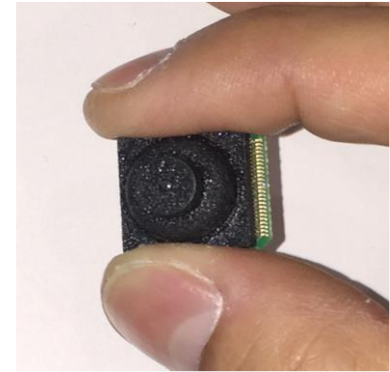
- Standard cameras are not good enough!



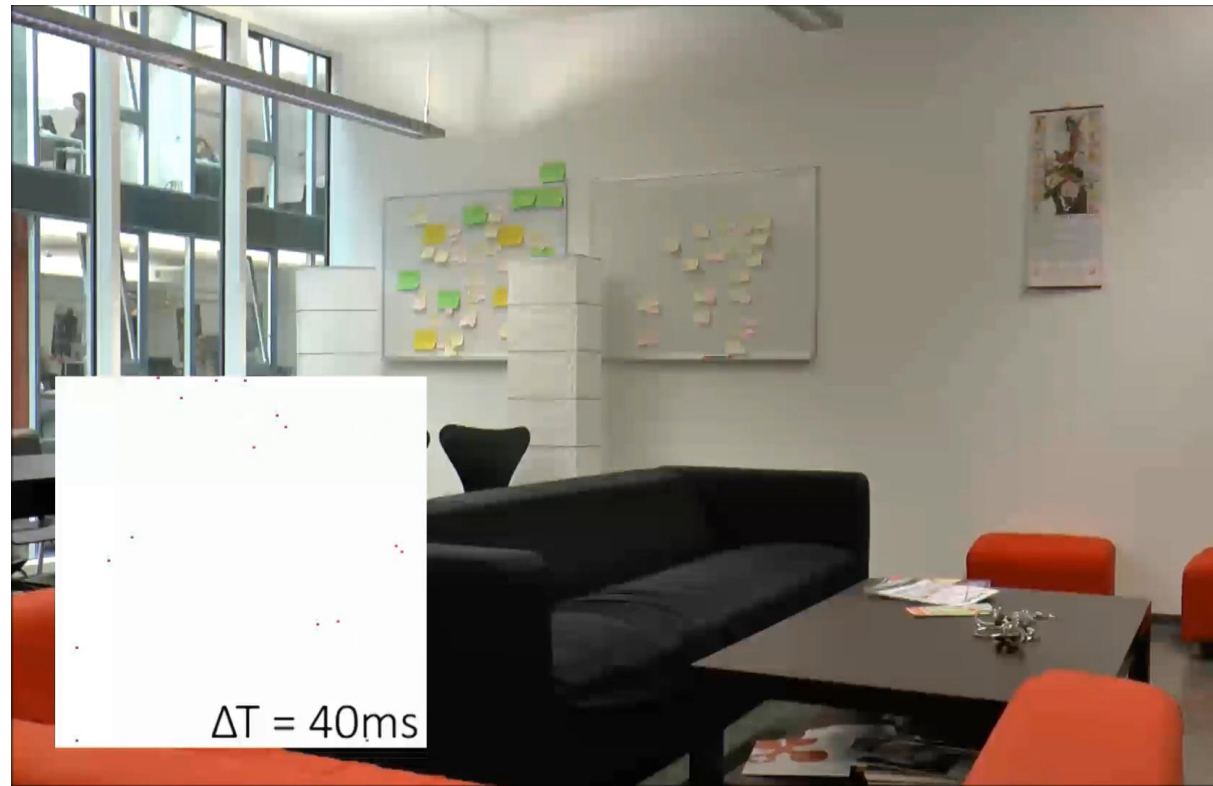
- Tasks that need to be done *reliably*, and with *low latency*:
  - Visual odometry (for control)
  - Obstacle detection
  - Recognition
- **Event cameras** promise to solve these three problems!

# What is an event camera?

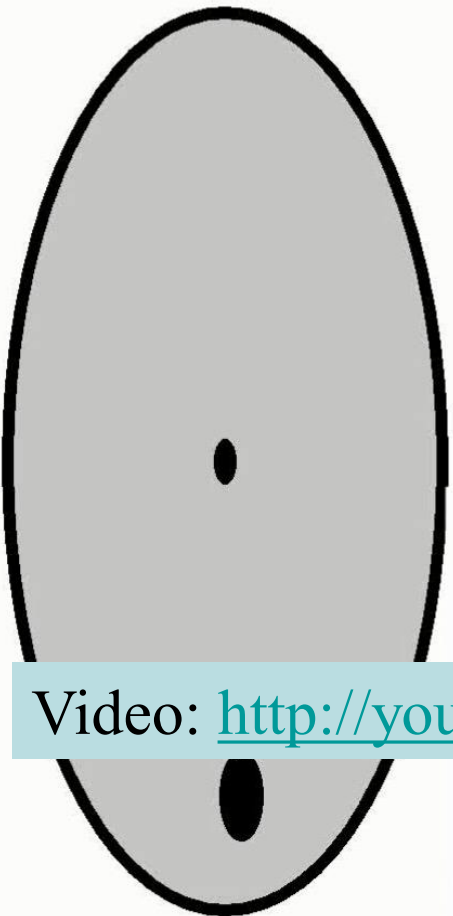
- Novel sensor that measures only motion in the scene
- **Low-latency ( $\sim 1 \mu\text{s}$ )**
- **No motion blur**
- **High dynamic range (140 dB instead of 60 dB)**
- Well-suited for visual odometry
- **But traditional vision algorithms for standard cameras cannot be used!**



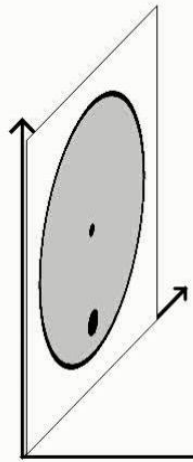
Mini DVS sensor from IniVation.com  
Check out their booth in the exhibition hall



# Camera vs Event Camera



**standard  
camera  
output:**



time

Video: <http://youtu.be/LauQ6LWTkxM> YouTube



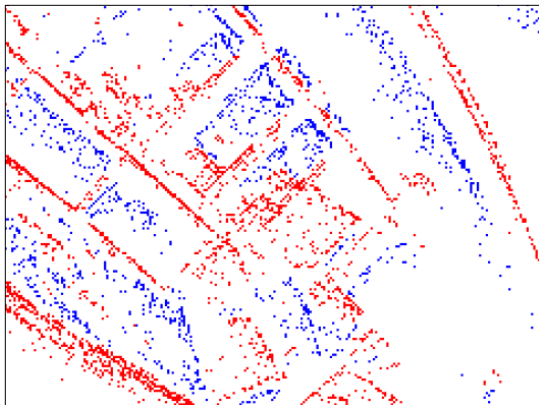
## Event Camera

## Standard Camera

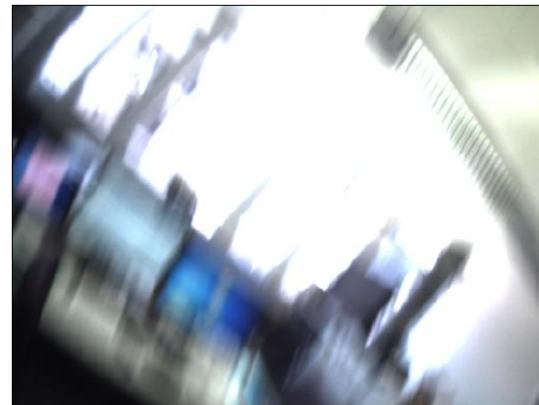
Update rate	1MHz and asynchronous	100Hz (synchronous)
Dynamic Range	High (140 dB)	Low (60 dB)
Motion Blur	No	Yes
Absolute intensity	No	Yes
Contrast sensitivity	Low	High

## Our idea: combine them!

< 10 years research



> 60 years of research!

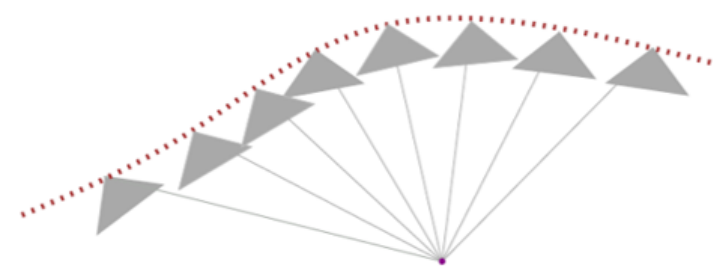
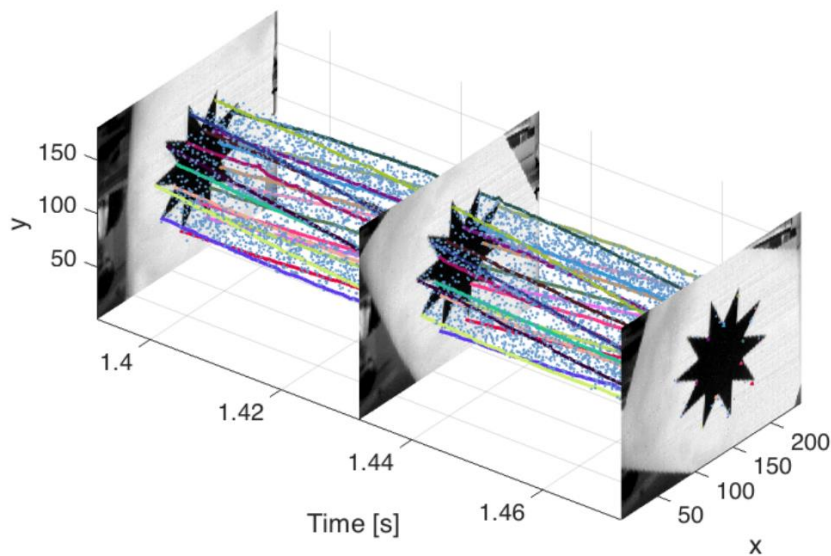


# UltimateSLAM: Visual-inertial SLAM with Events + Frames + IMU

Feature tracking from  
Events and Frames



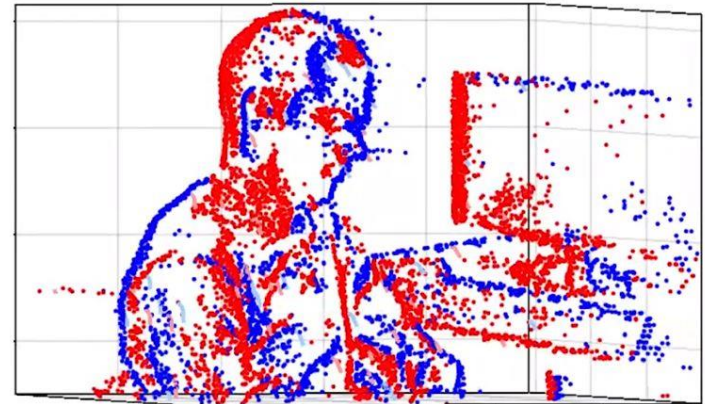
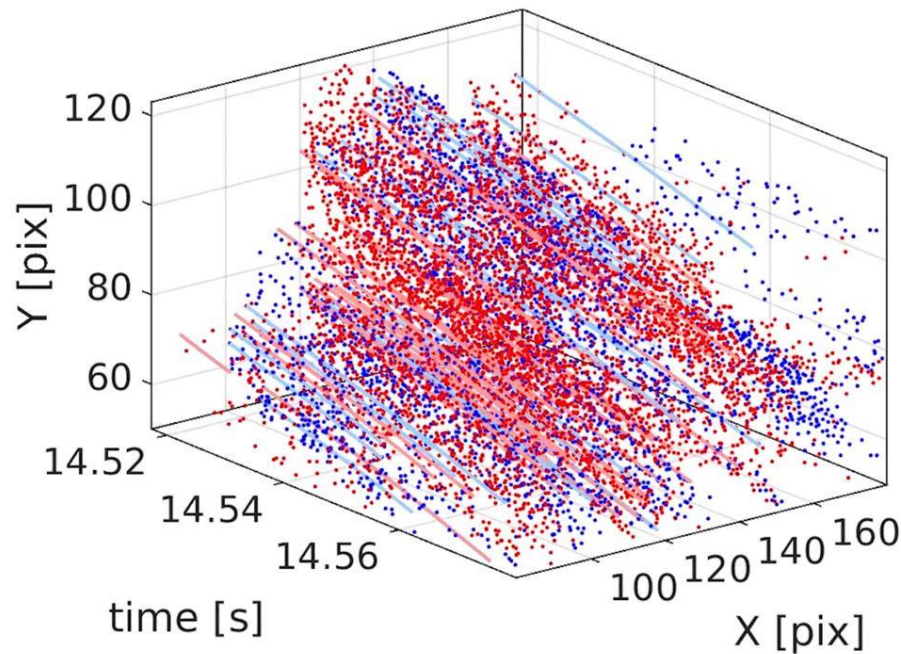
Visual-inertial Fusion



# Tracking by Contrast Maximization [CVPR'18]

- Directly estimate the motion curves that align the events

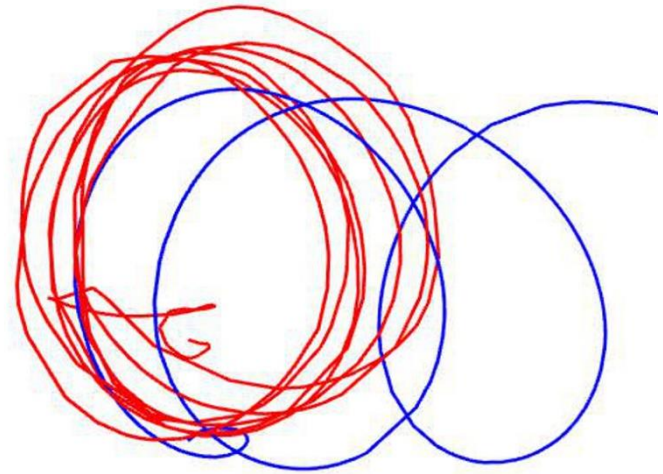
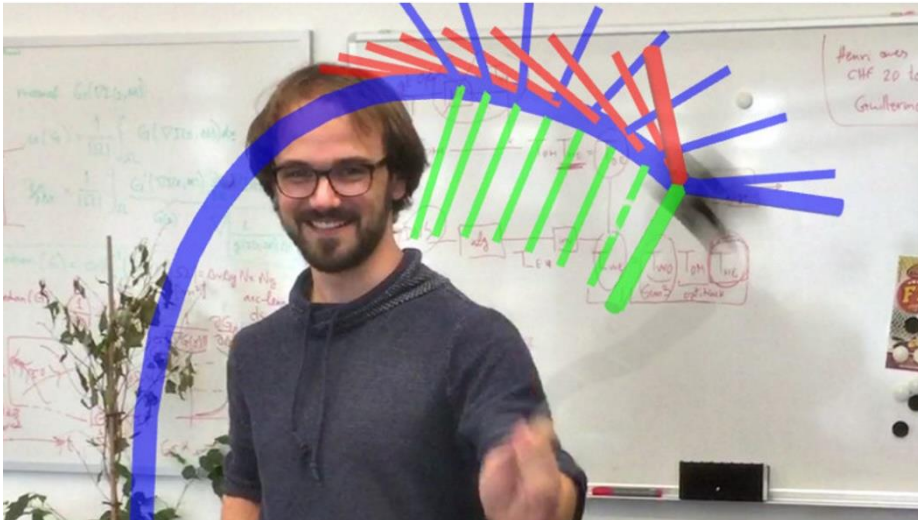
Our framework seeks for the point trajectories that are best aligned with the events



Gallego, Rebecq, Scaramuzza, A Unifying Contrast Maximization Framework for Event Cameras, with Applications to Motion, Depth, and Optical Flow Estimation, CVPR'18, Spotlight talk, [PDF](#), [YouTube](#)

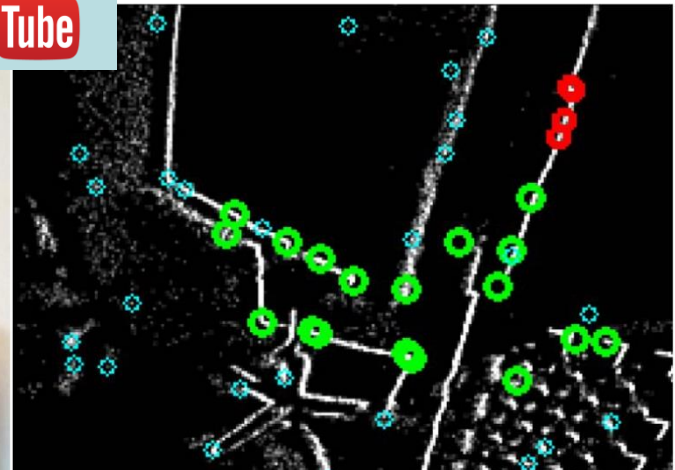
# UltimateSLAM: Events + Frames + IMU

Tightly coupled fusion. Runs in real time on a smartphone processor.



Video: <https://youtu.be/DyJd3a01Zlw>

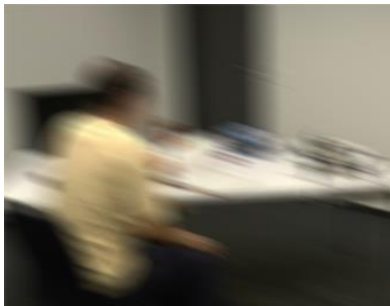
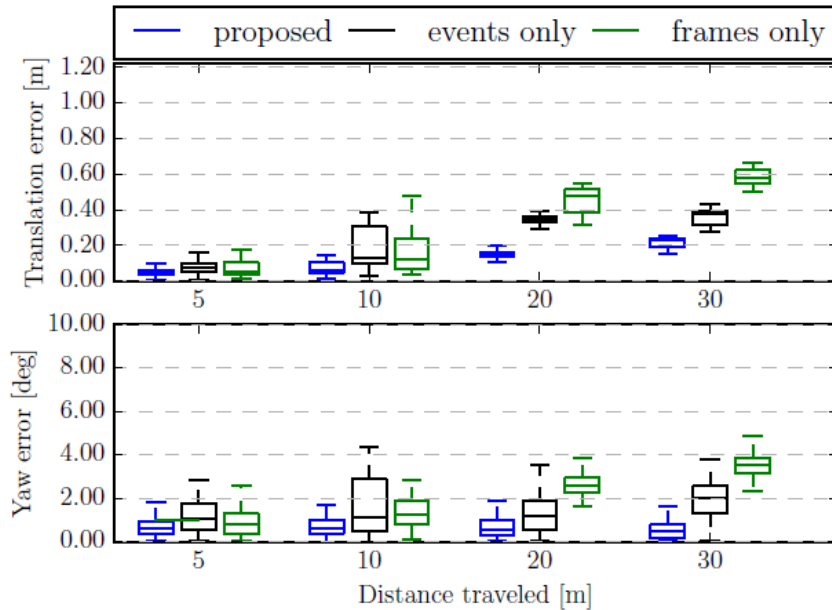
YouTube



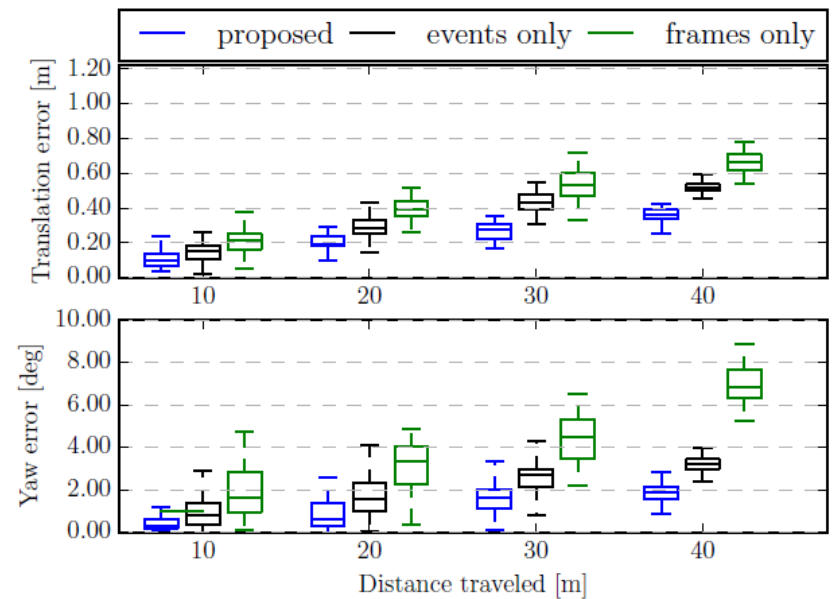
Rosinol et al., Ultimate SLAM? Combining Events, Images, and IMU for Robust Visual SLAM in HDR and High Speed Scenarios, IEEE RAL'18, [PDF](#)

# 85% accuracy gain over frame-based visual-inertial odometry

## High-speed sequence

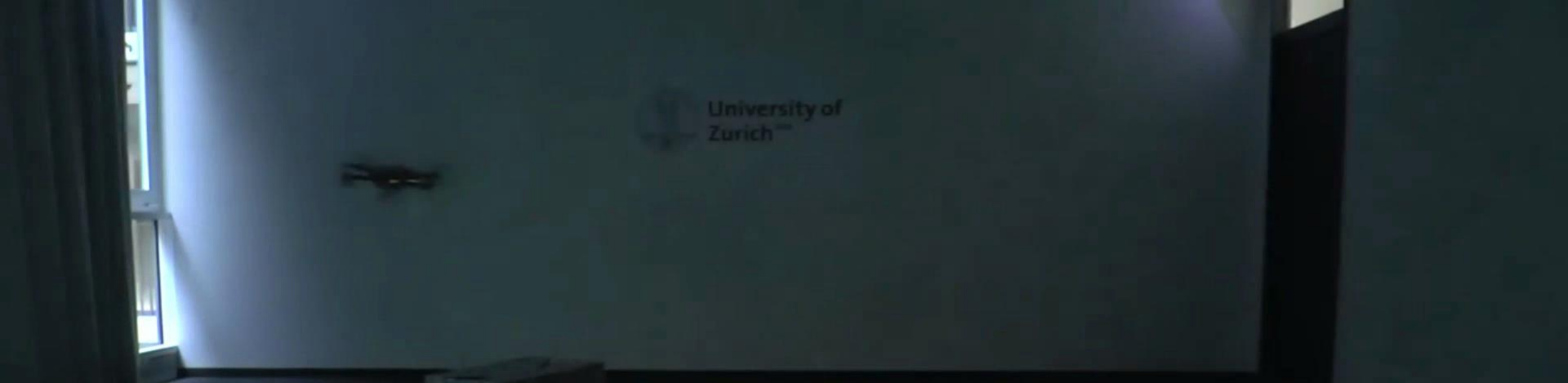


## HDR sequence



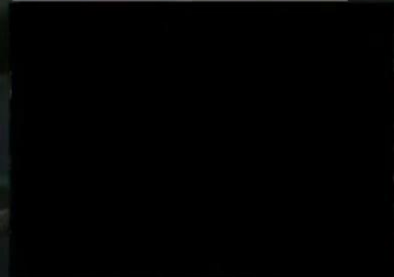
# UltimateSLAM: Autonomous Navigation in Low Light

Fully onboard (Odroid), event camera + IMU, tightly coupled

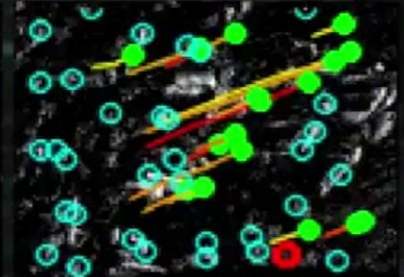


Video: [https://youtu.be/DN6PaV\\_kht0](https://youtu.be/DN6PaV_kht0) 

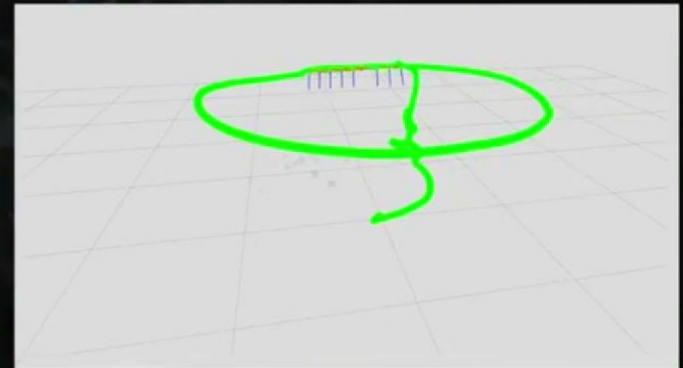
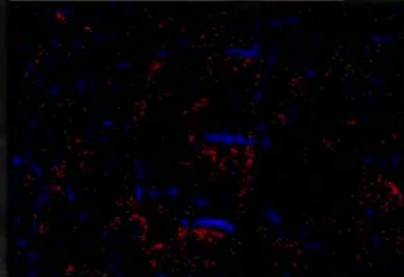
Standard frames



Event frames



Events only



Rosinol et al., Ultimate SLAM? Combining Events, Images, and IMU for Robust Visual SLAM in HDR and High Speed Scenarios, IEEE RAL'18, [PDF](#)

# Low-latency Obstacle Avoidance (ongoing work)

In collaboration with [Insightness](#) company (makes event cameras and collision avoidance systems for drones)



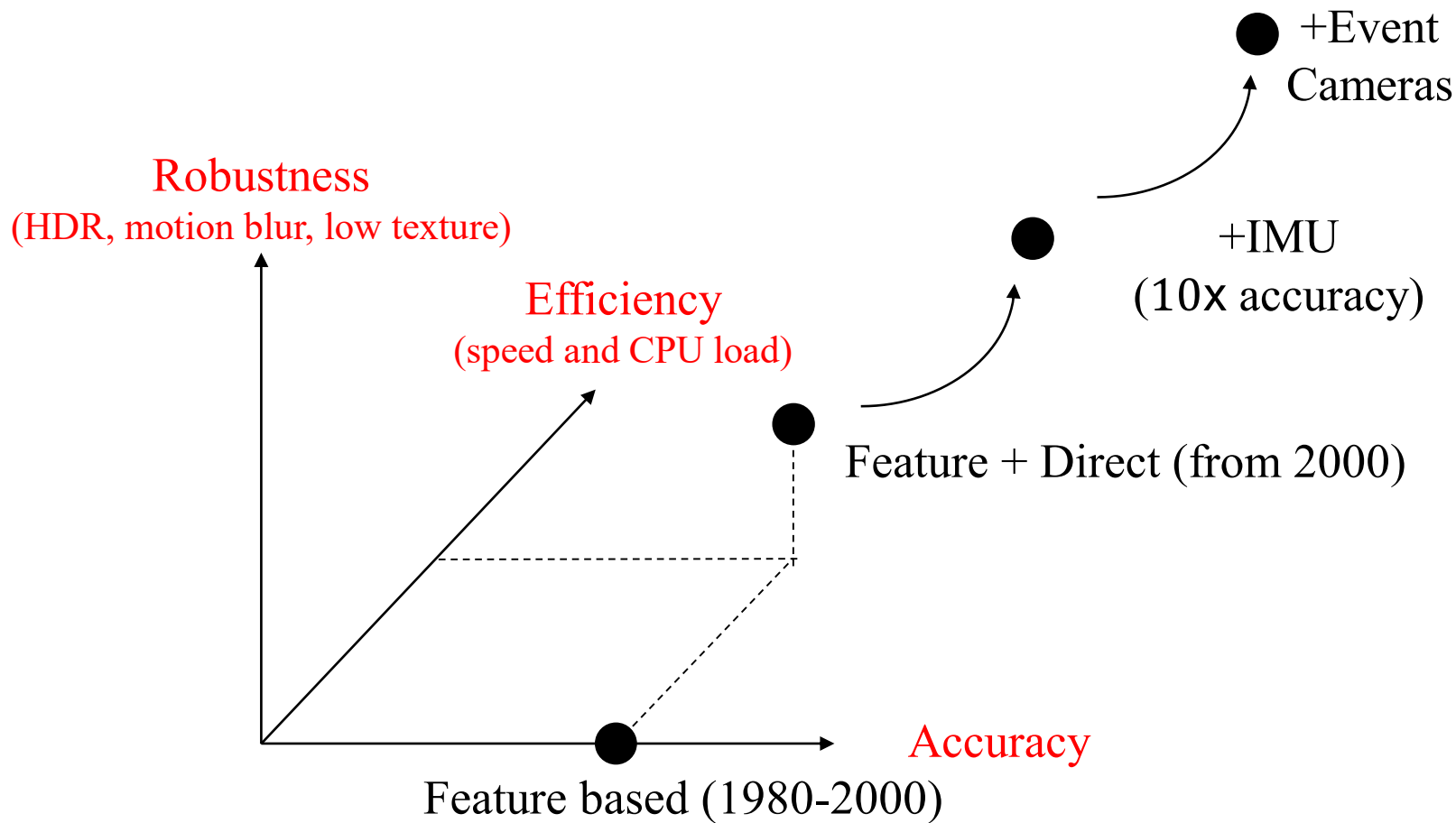
Video: <https://youtu.be/6aGx-zBSzRA>

# Conclusions

- Agile flight (**like birds**) is still far (10 years?)
- **Perception and control** need to be considered **jointly!**
- SLAM theory is **well established**
  - Biggest challenges today are **reliability and robustness** to:
    - High-dynamic-range scenes
    - High-speed motion
    - Low-texture scenes
    - Dynamic environments
- **Machine Learning can exploit context & provide robustness and invariance to nuisances**
- **Event cameras** are revolutionary and provide:
  - Robustness to **high speed motion** and **high-dynamic-range scenes**
  - Allow **low-latency** control (ongoing work)
  - **Intellectually challenging**: standard cameras have been studied for 50 years! → need of a change!



# A Short Recap of the last 30 years of Visual Inertial SLAM

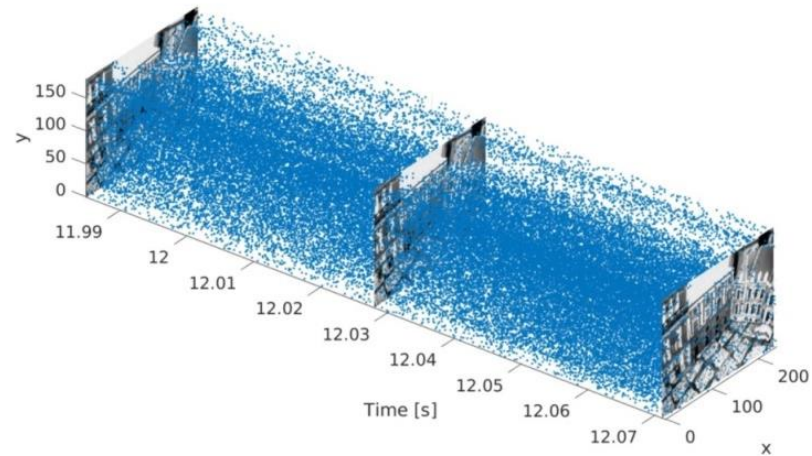


# Event Camera Dataset and Simulator [IJRR'17]

- **Publicly available:** [http://rpg.ifi.uzh.ch/davis\\_data.html](http://rpg.ifi.uzh.ch/davis_data.html)
- **First event camera dataset** specifically made for **VO and SLAM**
- **Many diverse scenes:** HDR, Indoors, Outdoors, High-speed
- **Blender simulator of event cameras**
- Includes
  - **IMU**
  - **Frames**
  - **Events**
  - **Ground truth** from a motion capture system

➤ Complete of code, papers, videos, companies:

- [https://github.com/uzh-rpg/event-based\\_vision\\_resources](https://github.com/uzh-rpg/event-based_vision_resources)



[Mueggler, Rebecq, Gallego, Delbruck, Scaramuzza,](#)

[The Event Camera Dataset and Simulator: Event-based Data for Pose Estimation, Visual Odometry, and SLAM, International Journal of Robotics Research, IJRR, 2017.](#)

# Thanks!

Code, datasets, publications, videos: <http://rpg.ifi.uzh.ch>



- My lab homepage: <http://rpg.ifi.uzh.ch/>
- Publications: <http://rpg.ifi.uzh.ch/publications.html>
- Software & Datasets: [http://rpg.ifi.uzh.ch/software\\_datasets.html](http://rpg.ifi.uzh.ch/software_datasets.html)
- YouTube: <https://www.youtube.com/user/ailabRPG/videos>