



Visual Inertial Odometry (VIO) based on Event Cameras

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Event cameras



Based on the Dynamic Vision Sensor (DVS), respond to **changes in brightness** at each pixel.



Note: all these models have embedded IMU 🔨



Event cameras, real data examples





Person waving

Objective: use this data for the Visual Inertial Odometry system



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Base tools for VIO using Event Cameras

EKLT (feature tracker) [Gehrig20]

Uses **frames** and **events** to track features

Combines **Visual** and **IMU** information for **pose estimation**.

FUSION (pose estimator) [Brossard17]

Features are tracked by comparing image patches (gradient) with accumulations of events. Has a Lie group structure for the **UKF filter** encompassing pose, velocity and landmark estimation.

[Gehrig20] "EKLT: Asynchronous photometric feature tracking using events and frames", D. Gehrig, H. Rebecq, G. Gallego and D. Scaramuzza, International Journal of Computer Vision, 128(3), 601-618, 2020.

[Brossard17] Brossard, Martin, Silvere Bonnabel, and Axel Barrau. "Unscented Kalman filtering on Lie groups for fusion of IMU and monocular vision." Int. Conf. Robot. Automation (ICRA). 2017. Computer and Robot Vision L

Feature tracking EKLT



- 1) Identify features
- 2) Create patches around features
- 3) Match image template with event template
- 4) Update position estimate

$$\min_{\mathbf{p},\mathbf{v}} \left\| \frac{\Delta L(\mathbf{u})}{\|\Delta L(\mathbf{u})\|_{L^{2}(\mathcal{P})}} - \frac{\Delta \hat{L}(\mathbf{u};\mathbf{p},\mathbf{v})}{\|\Delta \hat{L}(\mathbf{u};\mathbf{p},\mathbf{v})\|_{L^{2}(\mathcal{P})}} \right\|_{L^{2}(\mathcal{P})}^{2}$$

EKLT was designed for the DAVIS event camera

p - warp parameters (position and rotation)v - flow angle





Open loop integration



EKLT tracks visual features

FUSION receives corresponded visual features and combines with IMU to estimate state (camera pose and 3D landmarks).





Open loop integration experiment: setup

Experiment steps:



Render scenario in Unreal Engine

Simulated environment

Process with EKLT the events saved in a ROS bag

Run FUSION on tracked features

Simulating a DAVIS, the events camera for which the EKLT was developed.



Open loop integration experiment: compared results



Error in the estimated current angle



Event camera visual data, added upon IMU data, improves performance

and minimizes the average and absolute errors

Open loop integration experiment: Real world setup



Visual tracking is lost early

EKLT was designed for the DAVIS camera (frames and events always available).

We believe it can be adapted for the DVS by using the estimated ego-motion.





Proposal: Closed Loop Integration of Sensor and Pose



Combining EKLT with FUSION **in a closed loop integration** is expected to:

Filter

- Allow obtaining longer feature tracking with DVS cameras
- Find new features and track them despite periods without data
- Improve the estimation of camera position and orientation



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Thank you for your attention









Open loop integration experiment: Real world setup

Tracking fails after a short while

features get out of FOV
tracking quality degrades



DVS camera - long delay before new frame becomes available (and events are stopped - zero new data)

No simultaneous events and frames to replace features

Problem DVS vs DAVIS: Commutation between modes in the DVS takes too long for the tracking to restart. Sometimes artifacts occur.



Proposal: Closed Loop Integration of Sensor and Pose



Combining EKLT with FUSION **in a closed loop integration** is expected to:

Filter

- Allow obtaining longer feature tracking with DVS cameras
- Find new features and track them despite the long lack of events after acquiring a frame
- Improve the estimation of camera position and orientation

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