# IMAGE RECONSTRUCTION FROM DVS

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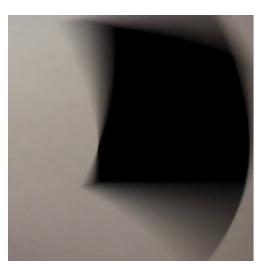
BASED ON WORK BY KIM ET AL. [1]

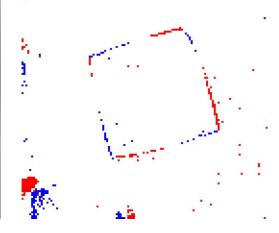
# 1. MOTIVATION

✓ increase robustness and speed of visual odometry / SLAM by replacing normal cameras with event cameras

reduce SLAM problem to camera rotation in a static scene and reconstruction of a complete image

### 2. DYNAMIC VISION SENSOR



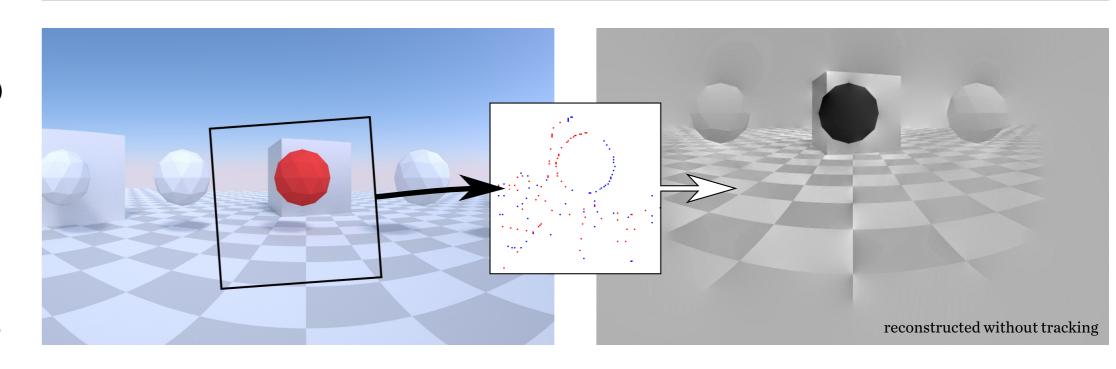


normal camera event camera (DVS)

✓ a DVS delivers instantaneouschanges in image brightness ("events")instead of periodic full frames

- practically no motion blur
- very high dynamic range
- drastically reduced bandwith incurs significantly lower computational costs

# 4. SIMULATION



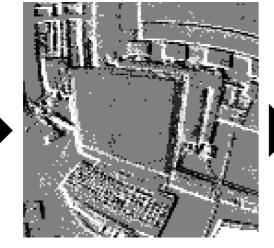
The full system is implemented in simulation. Both tracking and image reconstruction can either work with ground truth data or results from the other component.

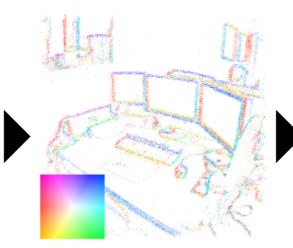
#### 3. CORE ALGORITHM



image-like mosaic.

the other is correct.





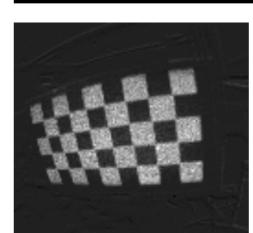


#### **ASSUMPTIONS**

- a change in brightness is caused by a movement of the camera (static scene)
- only rotation, no translation and therefore no parallax displacement

#### 5. REAL DATA

# **CALIBRATION**



- ▼ flickering display of normal checker-board pattern
- standard camera calibration toolbox

# INITIALIZATION



ving a dark cover results in an initial image patch

✓ another possibilty: the 2nd generation DVS is able to take full pictures

# **MOVEMENT TRACKING**

Jointly track the global rotational mo-

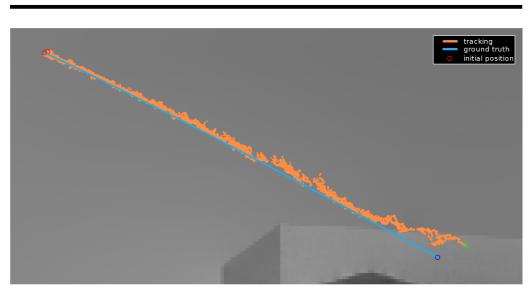
tion of a camera and estimate the gra-

dients of the scene around it. The gra-

dient map is then upgraded to a full

Each of these components essentially

believes that the current estimate from



- ▼ rotation tracking with particle filter and constant position motion model
- For every event, compare intensity at event position for every possible camera rotation to intensity at (assumed) position of last event:

 $z := log(M(\mathbf{p}^t)) - log(M(\mathbf{p}^{t-1}))$ 

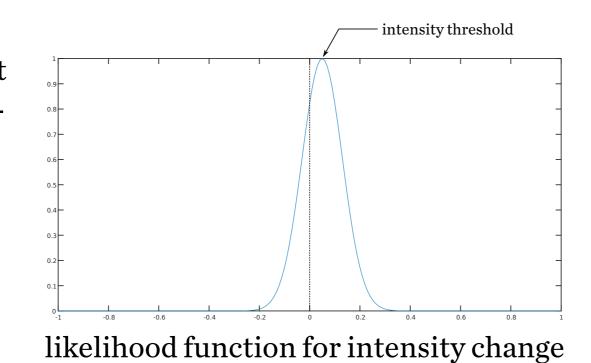
the closer the intensity change to the camera's threshold the more likely is the proposed movement

▶ bootstrap: start with a small known patch at the center. See also initialization in section 5.

# RECONSTRUCTION

Use movement between current and last event of the pixel to estimate gradient (intensity change) at event pixel.

- vextended Kalmann filter reduces noise in the image
- Poisson-solver computes grayscale image from gradients.



### 6. CONCLUSION

A dynamic vision sensor is a feasible option for rotational motion tracking. With some optimizations the system is conceiveably real-time capable - especially when combined with other sensors such as an IMU. This might even lead to full 3D-SLAM with a dynamic vision sensor.

#### 7. REFERENCES

- [1] H. Kim, A. Handa, R. Benosman, S. Ieng, A. Davison, 2014 "Simultaneous Mosaicing and Tracking with an Event Camera"
- [2] E. Mueggler, B. Huber, D. Scaramuzza, IROS 2014 "Event-based, 6-DOF Pose Tracking for High-Speed Maneuvers"

