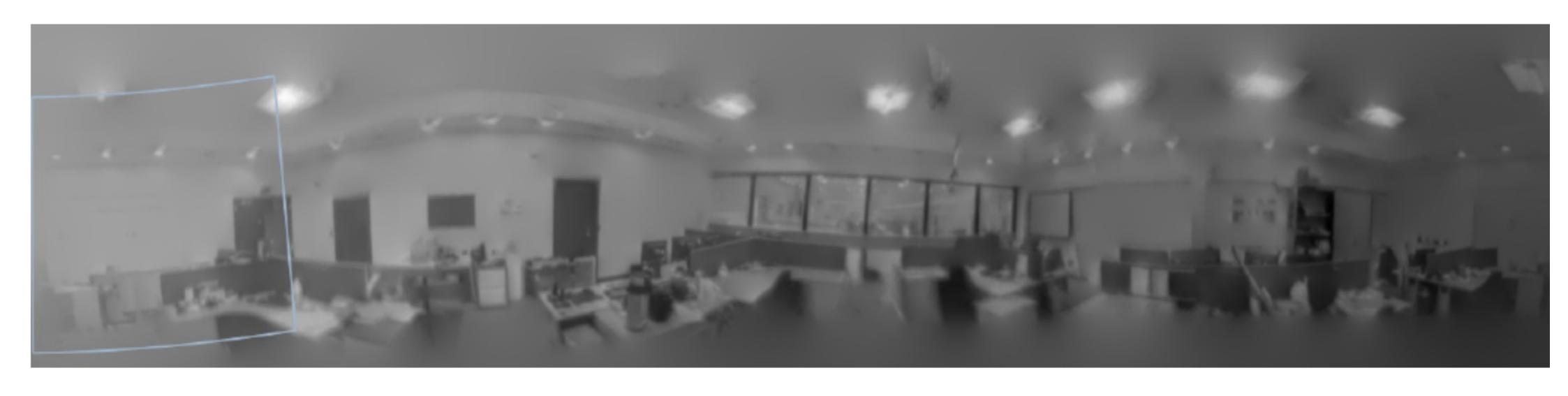
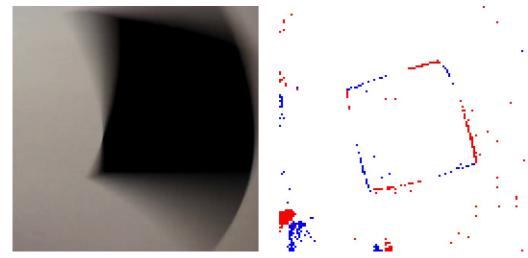
IMAGE RECONSTRUCTION FROM DVS

Marcel Geppert, Samuel Bryner

Petri Tanskanen



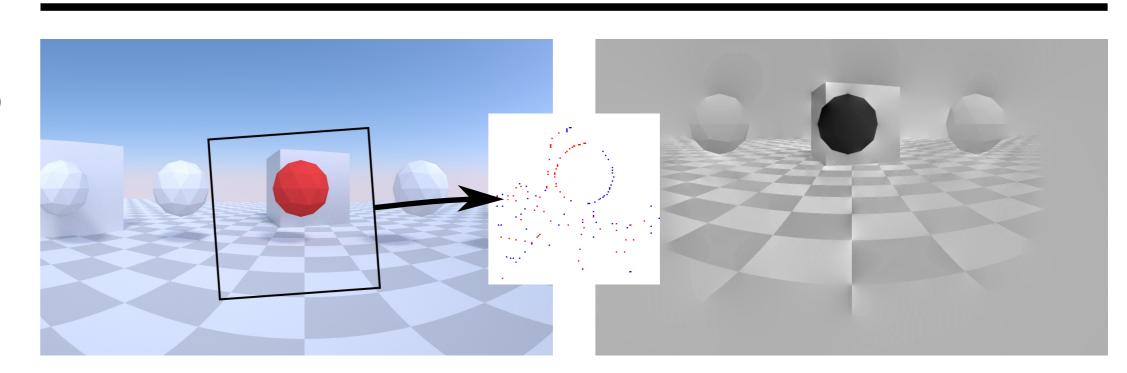
1. DYNAMIC VISION SENSOR



normal camera event camera (DVS)

- a DVS delivers instantaneous changes 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



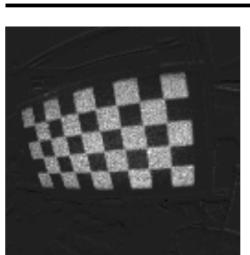
2. MOTIVATION

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

5. REAL DATA

CALIBRATION



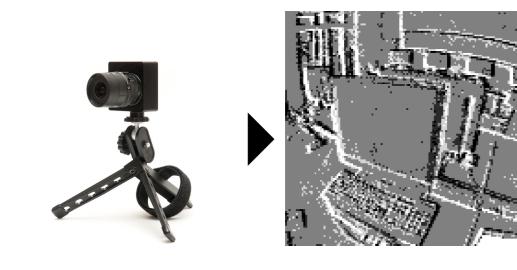
- flickering display of normal checkerboard pattern
- standard camera calibration toolbox

INITIALIZATION



▼ integrate events when removing a dark cover gives an initial image patch

3. CORE ALGORITHM



Jointly track the global rotational motion of a camera and estimate the gradients of the scene around it. The gradient map is then upgraded to a full image-like mosaic.

Each of these components essentially believes that the current estimate from the other is correct.

- a change in brightness is caused by a movement of the camera (static scene)
- camera only rotating, no translation

RECONSTRUCTION

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

- extended Kalmann filter reduces noise in the image
- **▼** Poisson-solver computes grayscale image from gradients.

ASSUMPTIONS

and therefore no parallax displacement

6. CONCLUSION

A dynamic vision sensor is a feasible option for rotational motion tracking. With some optimizations the system is easily real-time capable, especially when combined with other sensors, such as an IMU.

MOVEMENT TRACKING

- rotation tracking with particle filter and constant position motion model
- When an event occurs: 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

7. REFERENCES

- [1] H. Kim, A. Handa, R. Benosman, S. Ieng, A. Davison, 2014 "Simultaneous Mosaicing and Tracking with an Event Camera"
- [2] D. Weikersdorfer, D. Adrian, D. Cremers, Jörg Conradt, ICRA 2014 "Event-based 3D SLAM with a depth-augmented dynamic vision sensor"
- [3] E. Mueggler, B. Huber, D. Scaramuzza, IROS 2014 "Event-based, 6-DOF Pose Tracking for High-Speed Maneuvers"

ETHzürich