

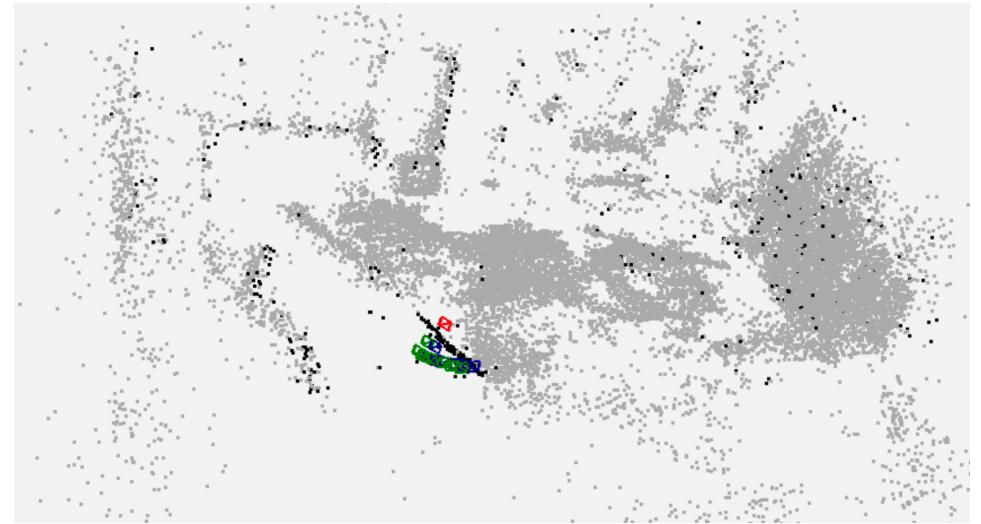
Indirect Visual Odometry with Optical Flow

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Visual Odometry

- What we had after exercises:
 - keypoints and descriptors
 - match keypoints and keypoint – landmark pairs using descriptors
- Goal: use optical flow to track (and match) keypoints



Optical Flow

- Motion from image sequence
- ill posed problem, need assumptions
- Assumption: Brightness constancy
 - $I(x, y, t) = I(x + \Delta x, y + \Delta y, t + \Delta t)$
- Optical Flow Constraint:

$\nabla I^T \vec{v} + I_t = 0$

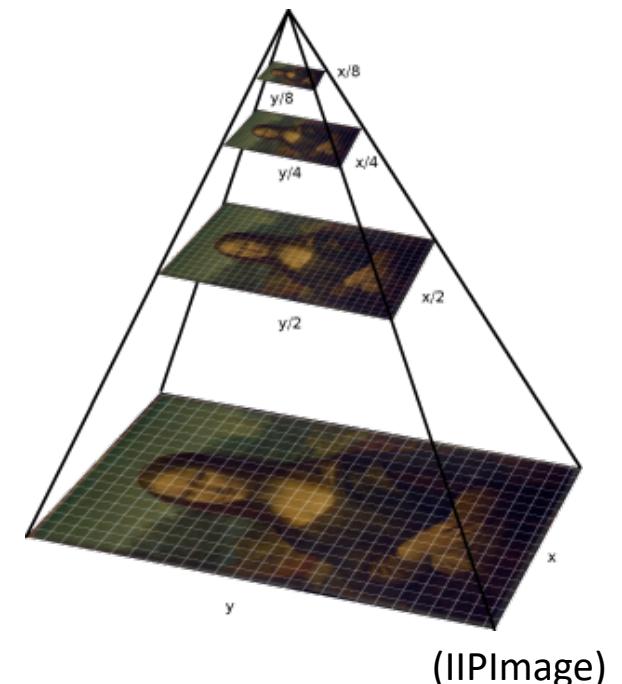
(via Taylor exp)
- need another assumption to solve



(OpenCV)

Lucas – Kanade Method

- Assumption: movement is **constant** in local neighborhoods around pixels
- Solve optical flow for all pixels in local neighborhood via least squares approach
- Use **image pyramids** to track points
 - from level n to level 0 (original image)



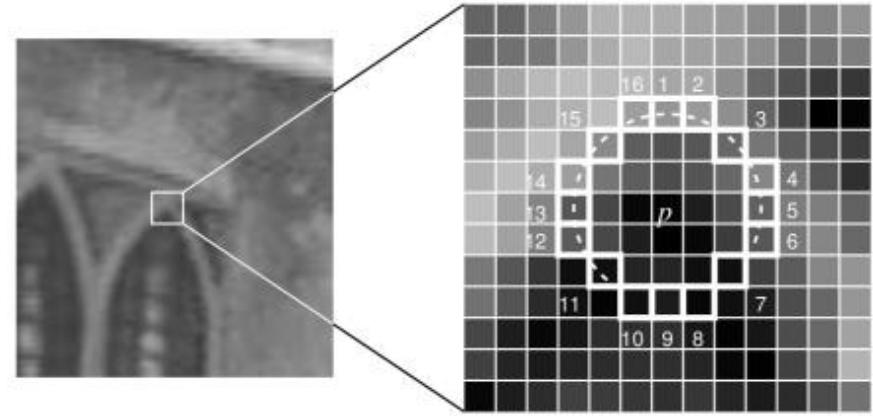
Method

- Select keypoints for first frame
- Track keypoints
- Add new keypoints (if necessary)
- Track keypoints from left image to right image for stereo matches
- Triangulate stereo matches
- Add new landmarks
- Optimize via reprojection error of landmark observations

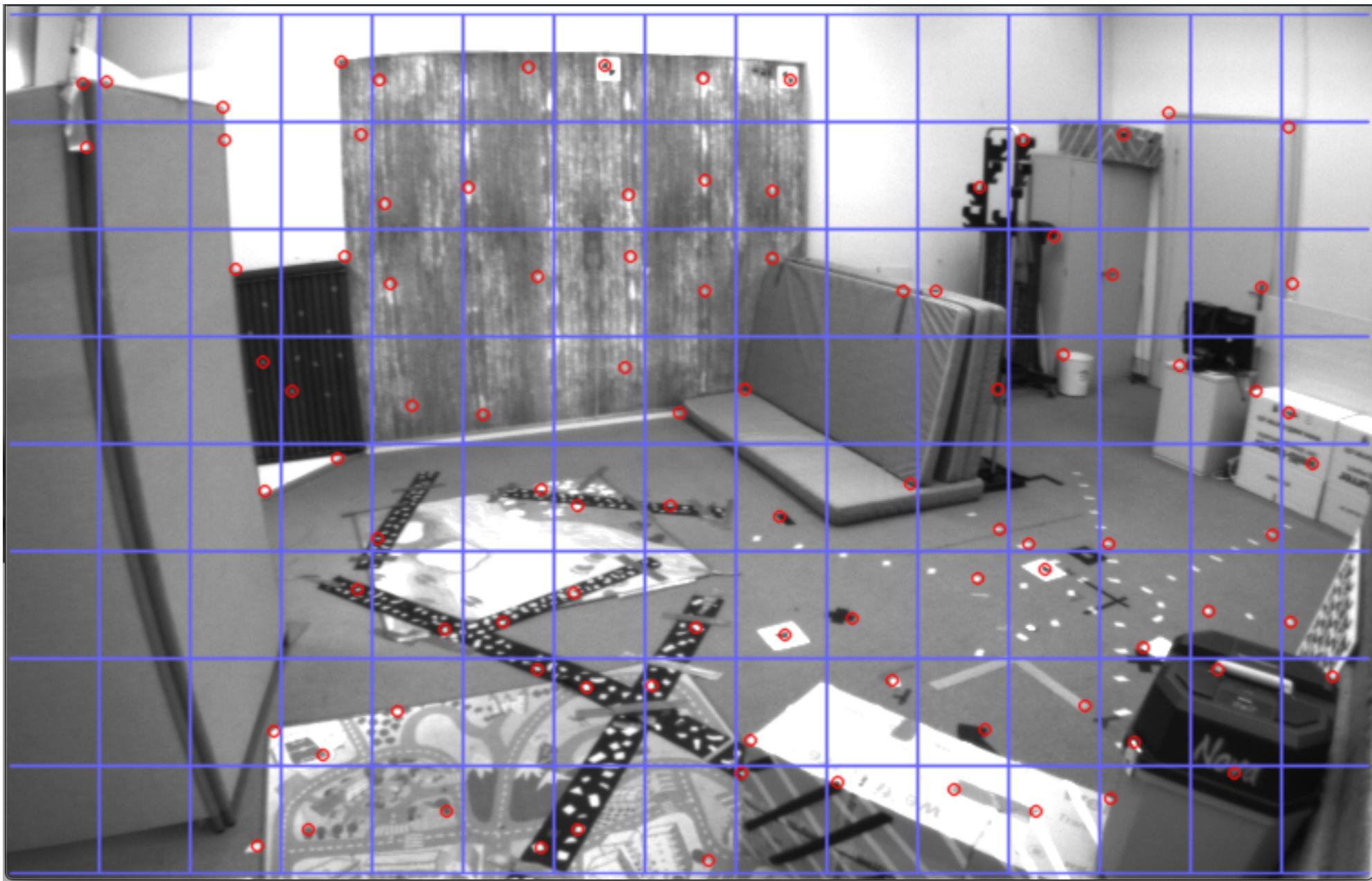
same
logic as
before

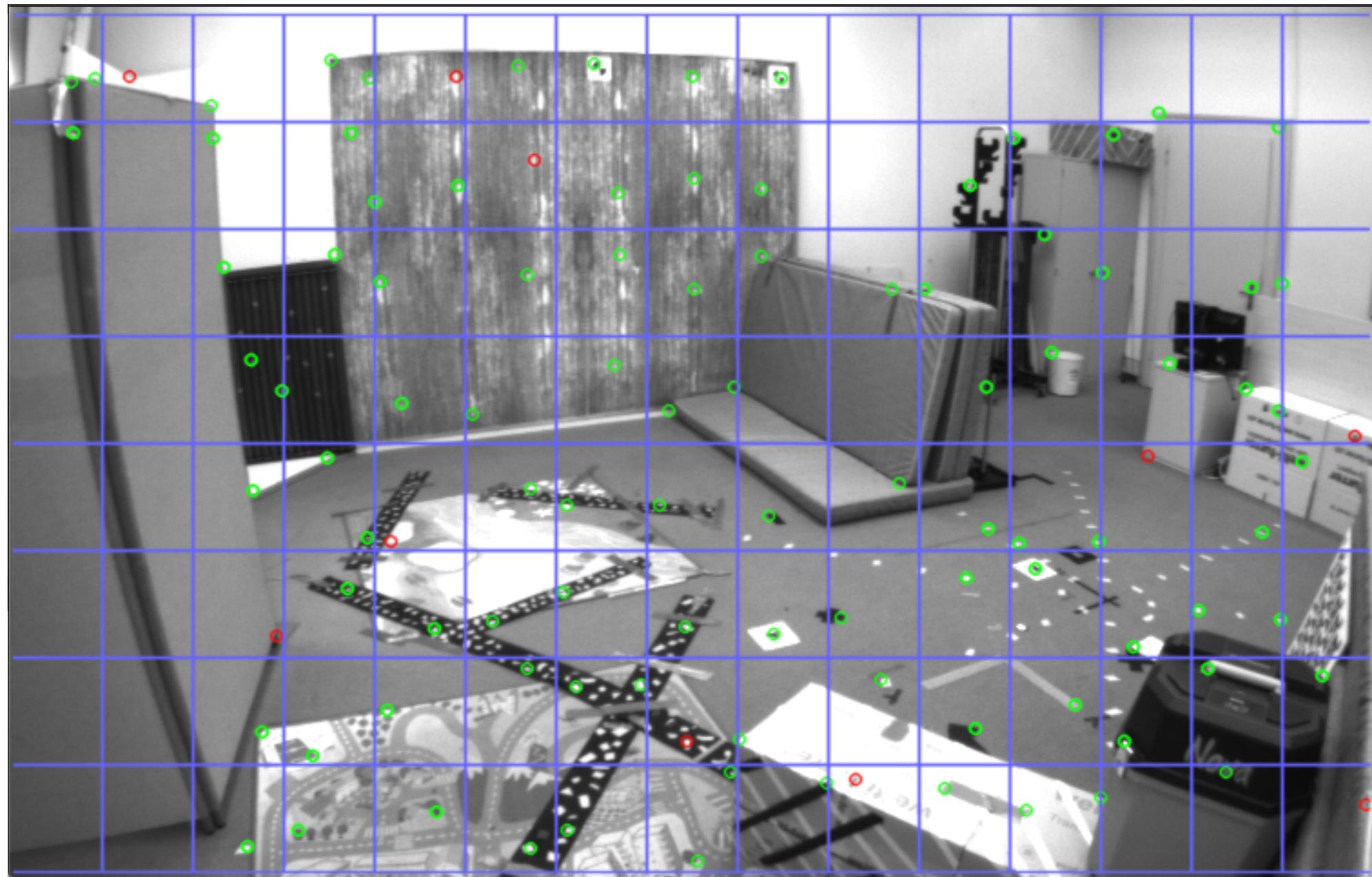
Keypoint Selection

- FAST feature detector
 - cv::FAST
- Consider image as a grid
 - evaluate each cell separately
 - add selected number of keypoints if possible
- Do not pick keypoints very close to each other



n contiguous brighter or darker pixels





Tracking via Optical Flow

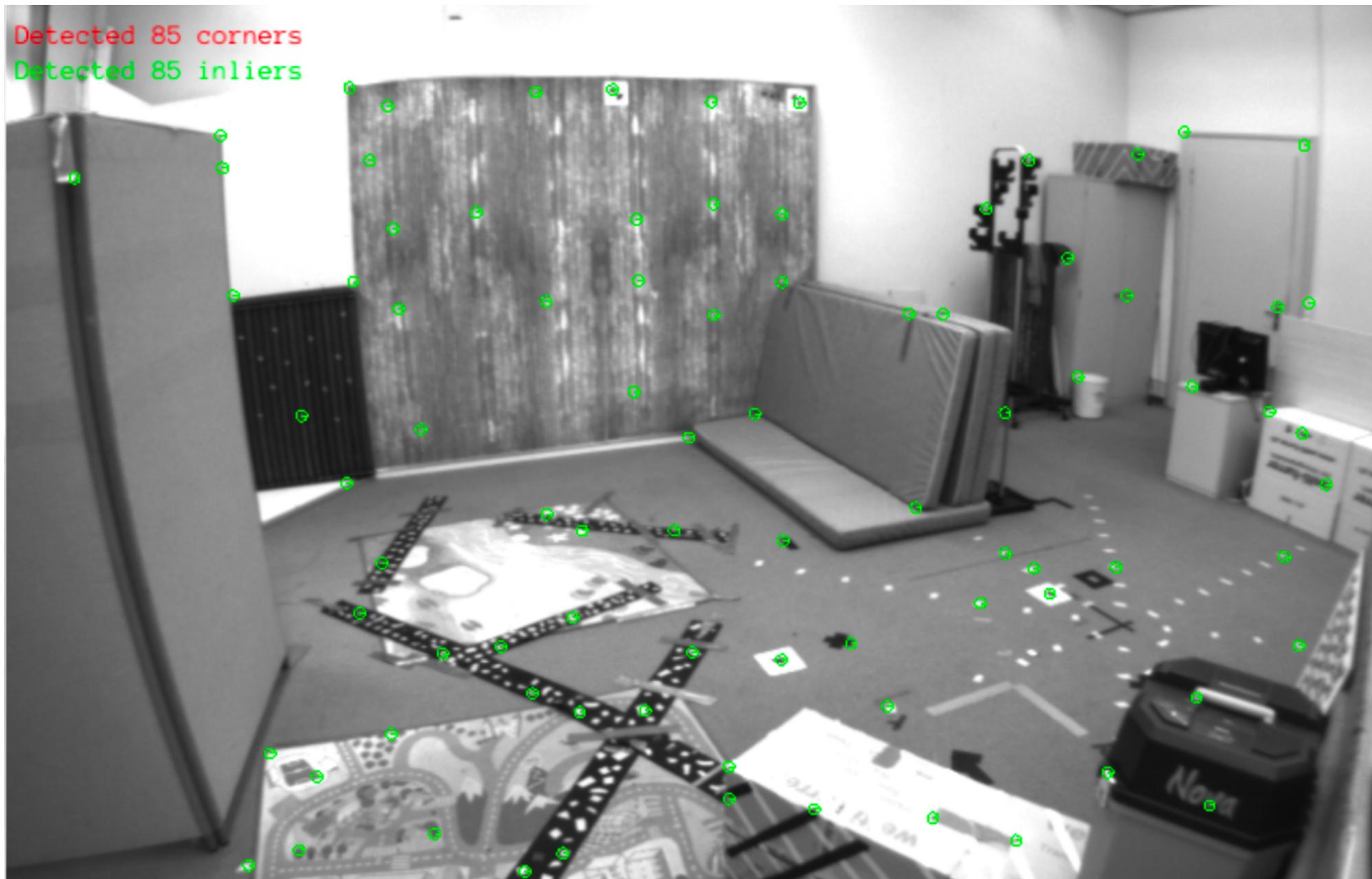
- cv:: calcOpticalFlowPyrLK
 - implementation of Lucas – Kanade method with image pyramids
- Apply optical flow twice to track back points.
Only select points which are near the original point
- Apply epipolar constraint for stereo images

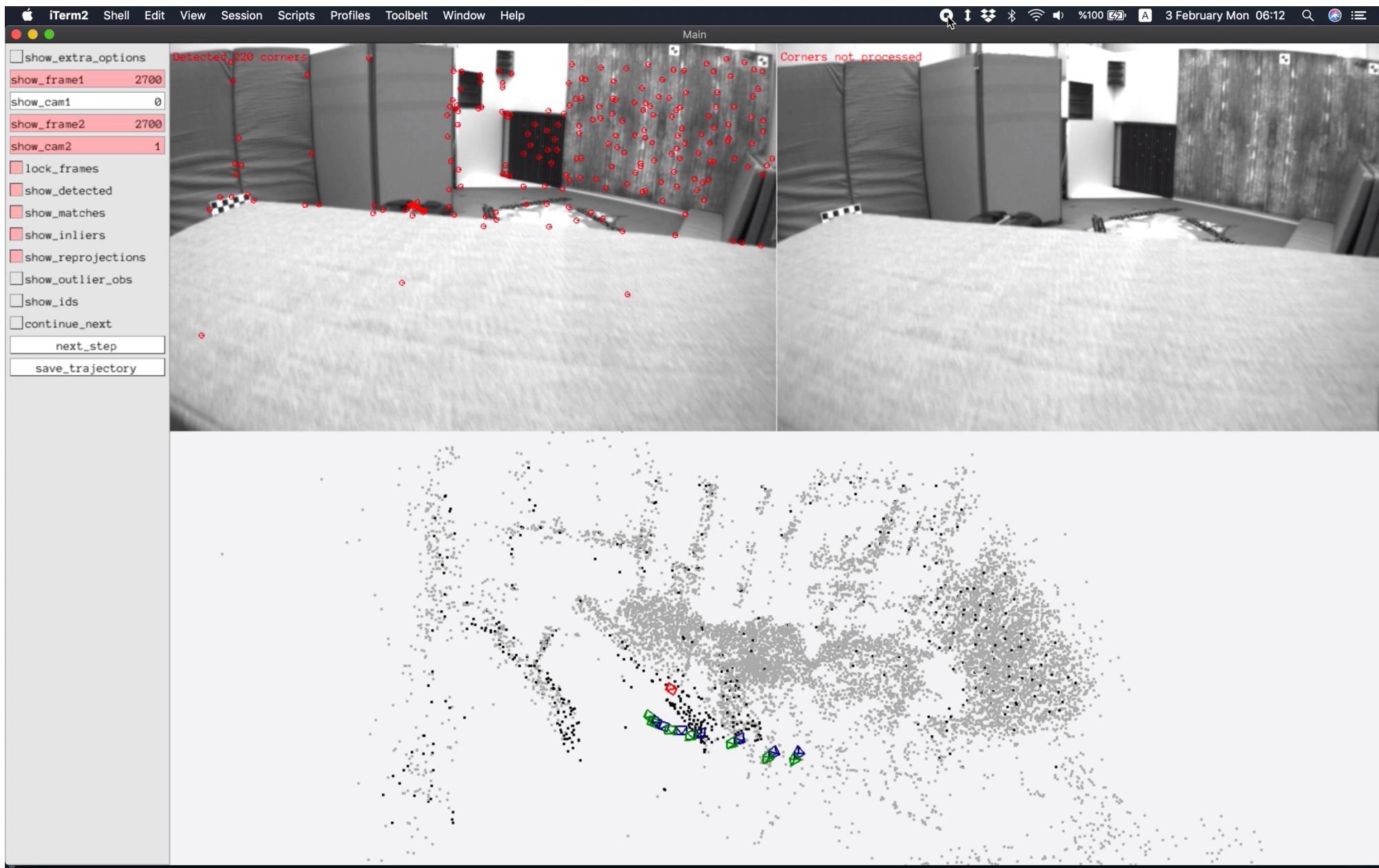
Detected 97 corners

Detected 85 inliers



Detected 85 corners
Detected 85 inliers





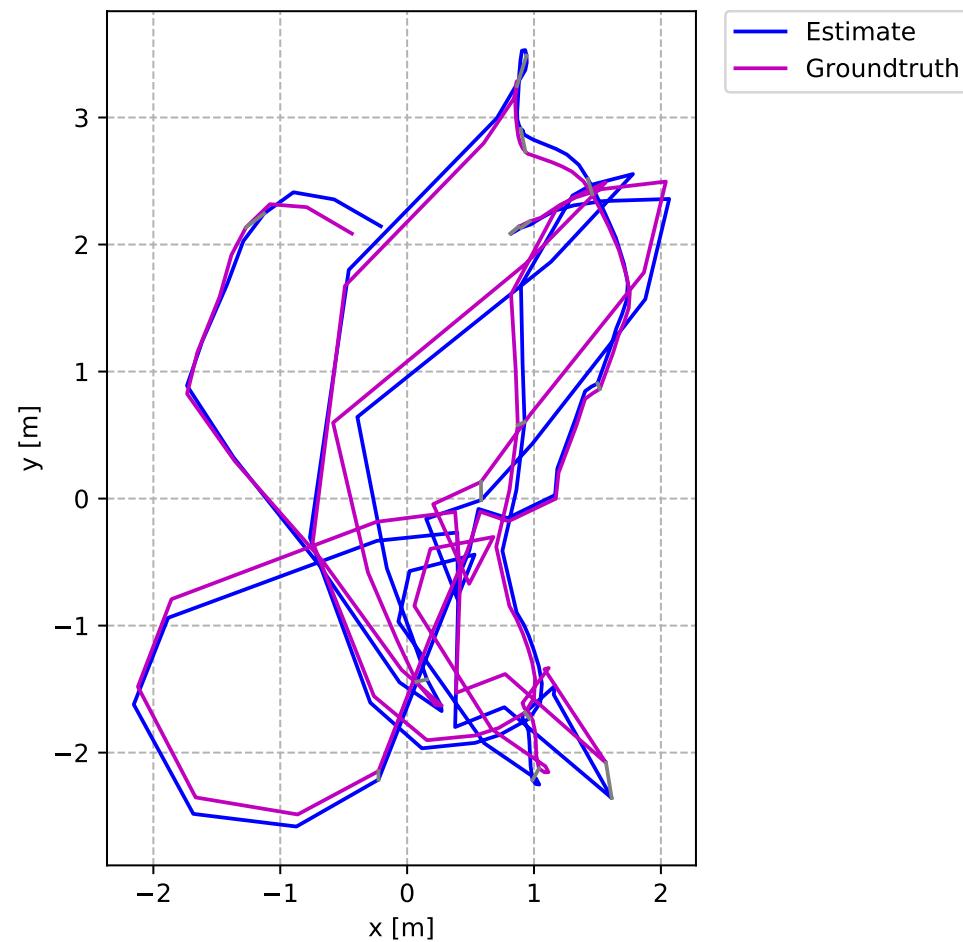
Evaluation

- Runtime
- Absolute Trajectory Error
 - align trajectories
 - compute difference between each pair of camera poses
 - root mean square error
- [github: uzh-rpg/rpg_trajectory_evaluation](#) (visualization and evaluation)
 - Zhang et al. IROS 2018
- [tum rgbd_benchmark_tools](#) (evaluation)
 - Sturm et al. IROS 2012

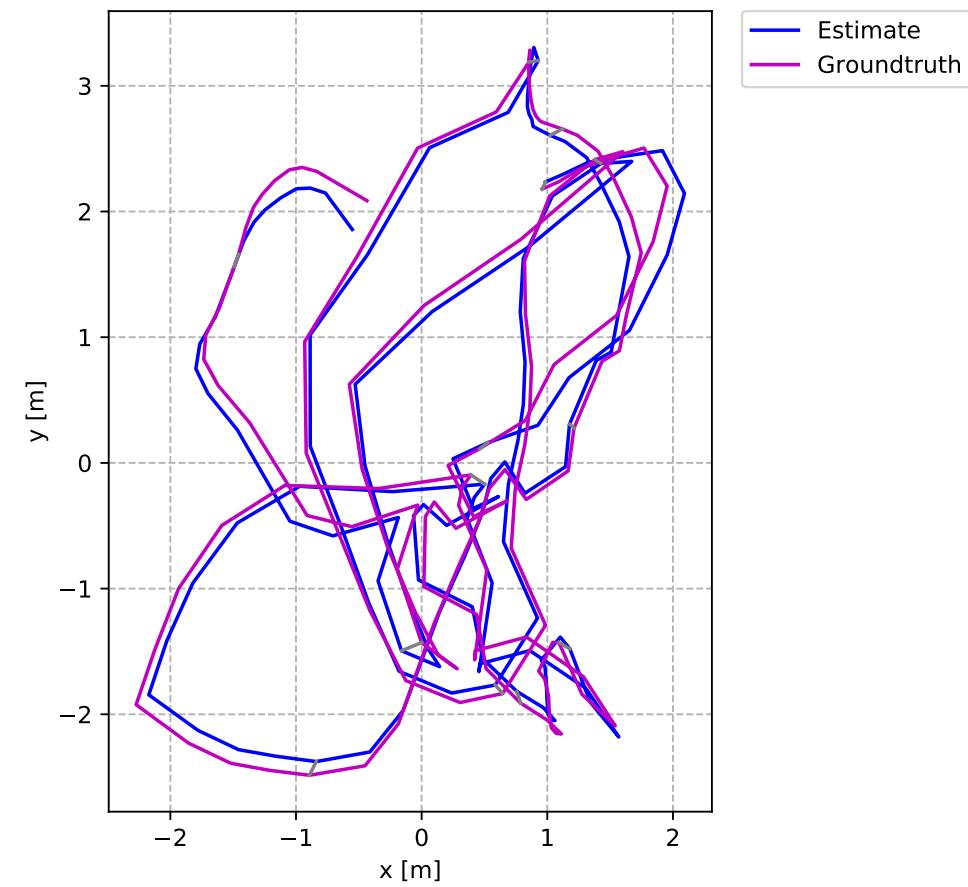
Results

Method	RMS ATE (m)	RMS ATE stddev	Runtime (s)	Runtime stddev
VO optical flow	0.143596	0.043500	47.44	0.41
VO descriptor	0.122272	0.031707	52.76	0.97

Each method is run 5 times



optical flow (0.144867)



descriptor (0.127981)

Conclusion

- No feature descriptor computation
- Relatively faster
- Slightly worse performance than descriptor based VO

Thank you for listening

References

- **Visual-Inertial Mapping with Non-Linear Factor Recovery** (V. Usenko, N. Demmel, D. Schubert, J. Stueckler and D. Cremers), *In IEEE Robotics and Automation Letters (RA-L) & Int. Conference on Intelligent Robotics and Automation (ICRA)*, IEEE, volume 5, 2020.
- **Machine learning for high-speed corner detection** (Rosten, Edward, and Tom Drummond), *European conference on computer vision*, 2006
- **Equivalence and efficiency of image alignment algorithms** (Baker, Simon, and Iain Matthews), *In IEEE Computer Society Conference on Computer Vision and Pattern Recognition*. Vol. 1. IEEE Computer Society; 1999, 2001.
- **A Tutorial on Quantitative Trajectory Evaluation for Visual(-Inertial) Odometry** (Zhang, Zichao and Scaramuzza, Davide), *IEEE/RSJ Int. Conf. Intell. Robot. Syst. (IROS)*, 2018
- **A Benchmark for the Evaluation of RGB-D SLAM Systems** (J. Sturm, N. Engelhard et al.), *Proc. of the International Conference on Intelligent Robot Systems (IROS)*, 2012

Aperture Problem

- Cannot see motion in direction of constant brightness
- Need to make another assumption
 - determines the optical flow method



Same motion both both image
Barberpole illusion (Wikipedia)