

Department of Informatics - **Institute of Neuroinformatics**

SIPs: Succinct Interest Points from Unsupervised Inlierness Probability Learning <u>Titus Cieslewski</u>, Konstantinos G. Derpanis, Davide Scaramuzza



Input images

Given a detector, how many interest points are enough? Can we train a detector to require as little interest points as possible?

Architecture



- Input: Image \rightarrow Output: **Per-pixel** score
- Non-maxima suppression to obtain *n* interest points
- Match using a descriptor; here, SURF
- Loss: **Probabilistic classification of inliers**
- Self-supervised training, random initial weights, converges to self-consistency
- Trained on pairs of images
- Results in **peaked response** without explicitly training for this



SIPs CNN output, interest points (circles), inliers (lines)



Peaked response without explicit training for peakedness

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Self-supervised Training

• From **uncalibrated** image sequences

• Select image pairs based on visual overlap (KLT track densely sampled points) • Ground truth labels: KLT track interest point of one image into the other image • Any other labeling would also work (e.g. inliers after RANSAC)

k-succinctness

• A novel metric to benchmark interest point detectors

• "How many points need to be detected to result in k inliers after matching and RANSAC?" Plot cumulative distribution over set of image pairs; summarize with area under curve

Results

• Evaluated on **KITTI and EuRoC**, some **HPatches** results

• Relative pose estimation: Accuracy plateau reached with **10 inliers** or more \rightarrow k = 10

• Our detector typically **requires less points** than baselines: 50 – 100 points often enough Point score predicts "inlierness" probability





Open source code:











Pose quality vs succinctness, as more inliers are required: KITTI (top), EuRoC (bottom)

https://github.com/uzh-rpg/sips2_open