Hierarchical Localization with hloc and SuperGlue

ECCV 2020 Workshop on Map-based Localization for Autonomous Driving



Super Glue

Paul-Edouard Sarlin Computer Vision and Geometry Lab ETH Zurich



psarlin.com

ETHZÜRICH Computer Vision and Geometry Lab

hloc - a toolbox for visual localization





Mihai Dusmanu Viktor Larsson





Marc Pollefeys

SuperGlue - a graph neural network for feature matching



with

with

Daniel DeTone



Tomasz Malisiewicz





Andrew Rabinovich

First place in 6 localization challenges!

At CVPR 2020: 2 challenges, local features & handheld devices At ECCV 2020, workshops:

- Map-based Localization for Autonomous Driving
- Long-Term Visual Localization under Changing Conditions

this talk

Friday 28th



Hierarchical Localization



From Coarse to Fine: Robust Hierarchical Localization at Large Scale, CVPR 2019

SuperGlue = Graph Neural Nets + Matching



- Extreme wide-baseline image pairs in real-time on GPU
- State-of-the-art indoor+outdoor matching with SIFT & SuperPoint

SuperGlue = Graph Neural Nets + Matching







CVPR 2020

Paper + code + demo: **psarlin.com/superglue**

hloc – a toolbox for SfM & localization



github.com/cvg/Hierarchical-Localization

hloc – reconstruction



hloc – triangulation



hloc – localization



Supported datasets





from https://www.visuallocalization.net/

with 3D SfM model: • (outdoor)

- Aachen Day-Night (v1.0 and v1.1)
- RobotCar Seasons
- CMU Seasons
- SILDa (multicamera coming soon)

with 3D Lidar scans: InLoc (indoor)

Winning the MLAD Challenge hloc



Winning the MLAD Challenge



But what are the baselines?

- 1. Extract features from query and reference images with SuperPoint/D2-Net/R2D2
- 2. 2D-2D matching with Nearest Neighbor or SuperGlue
- 3. Compute disparity map with OpenCV stereo matcher
- 4. Lift ref 2D keypoints to 3D with nearest sampling of disparity
- 5. Pose estimation with PnP+RANSAC and 2D-3D matches

github.com/pmwenzel/mlad-benchmark-baselines

<u>.</u>

But what are the baselines?





KITTI Stereo 2015, OpenCV StereoBM <u>Rank #289</u>

But what are the baselines?

- 1. Extract features from query and reference images with SuperPoint/D2-Net/R2D2
- 2. 2D-2D matching with Nearest Neighbor or SuperGlue
- 3. Compute disparity map with OpenCV stereo matcher
- 4. Lift ref 2D keypoints to 3D with nearest sampling of disparity
- 5. Pose estimation with PnP+RANSAC and 2D-3D matches

vs hloc: triangulation + bundle adjustment

- Is this how GN-Net evaluates SuperPoint and D2-Net?
- Next time: evaluate local features with hloc :)



What should we work on?

1. Going **beyond keypoints**:

combining the **robustness** of keypoint-based localization with the **accuracy** of direct image alignment

- 2. More **powerful image retrieval**: here we have localization priors, what if we don't? image retrieval is **often the bottleneck**
- 3. Leveraging **sequential information**: localize single queries vs sequences

hloc + SuperGlue

hloc = a simple & clean toolbox for visual localization

- 1. Reproduce our CVPR & ECCV 2020 winning results
- 2. Run SfM with SuperGlue to localize with your own datasets
- 3. Evaluate your own local features or image retrieval for localization
- 4. Easily implement & debug new localization pipelines 🔴

github.com/cvg/Hierarchical-Localization



Next talk: Friday 28th @ localization workshop

psarlin.com

github.com/cvg/Hierarchical-Localization