

Selecting Keypoint Detector and Descriptor Combination for Augmented Reality Application

Lukáš Bureš Luděk Müller

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NTIS

Motivation

- ▶ Augmented Reality



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- ▶ Augmented Reality
- ▶ Keypoint approach



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- ▶ Augmented Reality
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- ▶ Speed



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- ▶ Stability

- ▶ Augmented Reality
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- ▶ Speed
- ▶ Stability
- ▶ Goal: select the best KP Detector and Descriptor which perform the best on a selected criterion

Keypoint Detectors and

- ▶ SIFT
- ▶ SURF
- ▶ BRISK
- ▶ KAZE
- ▶ AKAZE

Keypoint Detectors

- ▶ FAST
- ▶ AGAST
- ▶ CenSurE

Keypoint Descriptors

- ▶ BRIEF
- ▶ DAISY
- ▶ FREAK
- ▶ LUCID
- ▶ LATCH



NTIS Datasets

- ▶ Oxford dataset

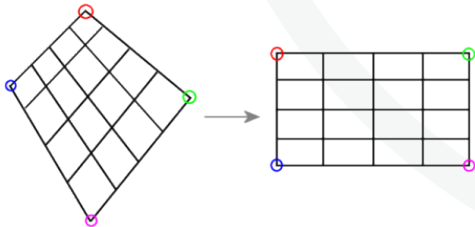


NTIS Datasets

- ▶ Oxford dataset
- ▶ 8 datasets – each contains 6 images

+ NTIS Datasets

- ▶ Oxford dataset
- ▶ 8 datasets – each contains 6 images
- ▶ each set contains 1st to 2nd, 1st to 3rd, ..., 1st to 6th Ground Truth Homography transformation matrix



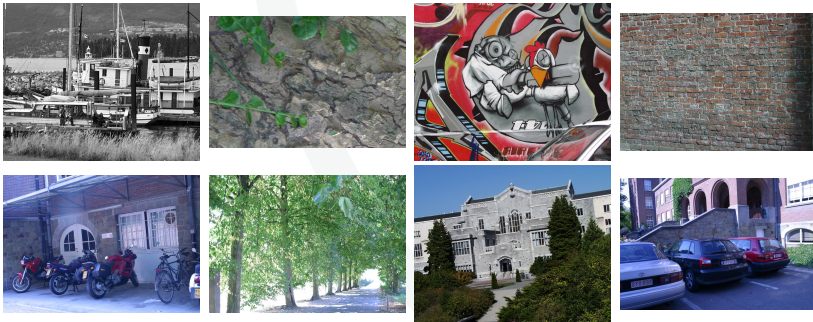


Figure : The example of the datasets (from top left):
 boat: zoom+rotation, bark: zoom+rotation, graffiti: viewpoint change,
 wall: viewpoint change, bikes: image blur, trees: image blur, ubc: JPEG
 compression, leuven: light change.

- ▶ The putative match ratio:

$$\textit{Putative Match Ratio} = \frac{\# \textit{Putative Matches}}{\# \textit{Features}} \quad (1)$$

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- ▶ **The precision:**

$$\textit{Precision} = \frac{\# \textit{Correct Matches}}{\# \textit{Putative Matches}} \quad (2)$$

- ▶ The matching score:

$$\text{Matching Score} = \frac{\# \text{Correct Matches}}{\# \text{Features}} \quad (3)$$

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- ▶ **The recall:**

$$\text{Recall} = \frac{\# \text{Correct Matches}}{\# \text{Correspondences}} \quad (4)$$

Results

Detector	Descriptor	Average Putative Match Ratio							
sift	sift	0.112	0.127	0.116	0.204	0.128	0.050	0.262	0.294
surf	surf	0.077	0.050	0.053	0.092	0.197	0.061	0.325	0.262
brisk	brisk	0.036	0.037	0.030	0.035	0.067	0.020	0.193	0.131
kaze	kaze	0.172	0.074	0.099	0.234	0.364	0.164	0.696	0.297
akaze	akaze	0.060	0.030	0.029	0.083	0.390	0.096	0.573	0.281
orb	orb	0.052	0.021	0.022	0.026	0.136	0.040	0.446	0.098
star	surf	0.044	0.039	0.061	0.022	0.116	0.019	0.307	0.130
star	freak	0.019	0.021	0.018	0.017	0.101	0.017	0.340	0.096
agast	latch	0.000	0.000	0.001	0.009	0.044	0.009	0.137	0.153

Detector	Descriptor	Average Precision							
sift	sift	0.851	0.940	0.405	0.737	0.774	0.698	0.942	0.929
surf	surf	0.678	0.879	0.288	0.675	0.798	0.633	0.908	0.874
brisk	brisk	0.894	0.931	0.583	0.648	0.754	0.708	0.955	0.950
kaze	kaze	0.858	0.923	0.518	0.796	0.838	0.685	0.964	0.905
akaze	akaze	0.503	0.726	0.276	0.514	0.884	0.586	0.948	0.910
orb	orb	0.719	0.600	0.329	0.726	0.850	0.660	0.978	0.920
star	surf	0.486	0.656	0.226	0.466	0.669	0.433	0.908	0.726
star	freak	0.003	0.683	0.000	0.461	0.009	0.007	0.186	0.021
agast	latch	0.000	0.000	0.000	0.001	0.000	0.000	0.007	0.002

Average Matching Score

Detector	Descriptor	Average Matching Score							
sift	sift	0.106	0.119	0.089	0.194	0.107	0.042	0.255	0.276
surf	surf	0.062	0.043	0.031	0.082	0.169	0.047	0.309	0.233
brisk	brisk	0.034	0.034	0.026	0.033	0.056	0.017	0.190	0.126
kaze	kaze	0.161	0.066	0.082	0.218	0.323	0.125	0.677	0.272
akaze	akaze	0.041	0.025	0.020	0.068	0.356	0.065	0.555	0.261
orb	orb	0.048	0.014	0.018	0.024	0.123	0.033	0.442	0.091
star	surf	0.029	0.027	0.024	0.016	0.087	0.011	0.293	0.101
star	freak	0.000	0.017	0.000	0.014	0.002	0.000	0.084	0.002
agast	latch	0.000	0.000	0.000	0.000	0.000	0.000	0.001	0.000

Detector	Descriptor	Average Recall							
sift	sift	0.226	0.358	0.189	0.310	0.580	0.106	0.423	0.652
surf	surf	0.142	0.136	0.070	0.148	0.373	0.101	0.436	0.437
brisk	brisk	0.043	0.073	0.038	0.045	0.259	0.029	0.241	0.225
kaze	kaze	0.237	0.220	0.133	0.316	0.583	0.255	0.764	0.527
akaze	akaze	0.060	0.069	0.027	0.088	0.559	0.125	0.608	0.532
orb	orb	0.062	0.049	0.022	0.040	0.168	0.073	0.482	0.155
star	surf	0.072	0.127	0.055	0.028	0.307	0.031	0.369	0.336
star	freak	0.000	0.075	0.000	0.024	0.004	0.001	0.101	0.007
agast	latch	0.000	0.000	0.000	0.000	0.000	0.000	0.002	0.001

Conclusion

- ▶ The paper presents the problem of AR by using KP
- ▶ Oxford dataset was described
- ▶ A list of existing KP detectors and descriptor combinations was presented
- ▶ Experiments were conducted to determine the best combination based on the chosen criterion

Thank you for your attention