

Parallel Tracking and Verifying: A Framework for Real-Time and High Accuracy Visual Tracking

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1. Siamese networks for verification

1.1. Architecture

Figure 1 shows the detailed architecture of siamese networks. Note that there are only two max pooling layers after conv1-2 and conv2-2 because too many max pooling layers may reduce the spatial information in features. Table 1 demonstrates the parameters of each layer. The source code of PTAV will be released for reproducible research.

1.2. Network training

We use the ALOV dataset [14] to train the siamese networks. Note that we exclude all videos that appear in OTB2013 [17], OTB2015 [18] and TC128 [9]. After removing these sequences, the training dataset and the evaluation datasets have no common objects. As in [15], we generate multiple pairs using every two frames in a video. One element in the pair is the groundtruth bounding box in one frame and the other one is a box sampled in the other frame. The pair is considered to be positive if the sampled box has a intersection-over-union overlap larger than 0.7 with the corresponding groundtruth box and considered to be negative if the overlap is smaller than 0.5. The training pairs and validation pairs are generated from different videos, and therefore from different objects. For training, in total we have sampled 60, 000 pairs of frames from ALOV dataset and each pair has 128 pairs of boxes. For validation, we have gathered 2, 000 pairs of frames and the same as for training each pair of frames contains 128 pairs of boxes.

We use the pre-trained network parameters from VGGNet [13] to initialize the networks. The initial learning rate is 0.001 and the weight decay parameter is 0.001. The learning rate is decreased by a factor of 10 after every 2 epochs. Training stops when the validation loss does not decrease any more.

2. Detailed results on OTB2015 [18]

In this supplementary material, detailed results on OTB2015 [18] with 100 videos are provided. Table 2 shows the per-video distance precisions for all trackers in comparison. Note that table 2 also contains the results on OTB2013 [17]. Figure 2 and Figure 3 show the distance precision plots and overlap successful plots for all 11 attributes, respectively.

3. Detailed results on TC128 [9]

We also present detailed results on the TC128 [9] with 128 videos. The per-video distance precision for all trackers in our comparison are reported in table 3. Figure 4 and Figure 5 show the distance precision plots and overlap successful plots for all 11 attributes, respectively.

4. Detailed results on UAV20L [12]

We present detailed results on the UAV20L [12] with 20 long videos. The shortest video contains 1717 frames, and the longest video contains 5527 frames. The per-video distance precision for all trackers in our comparison are reported in table 4. Figure 6 and Figure 7 show the distance precision plots and overlap successful plots for all 12 attributes, respectively.

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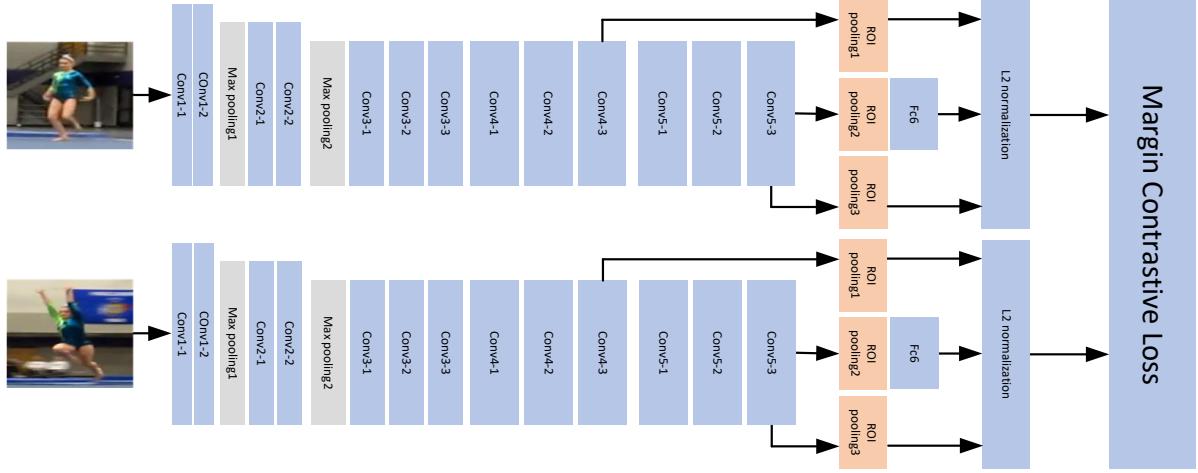


Figure 1. The detailed architecture of siamese networks.

Table 1. Parameters of the siamese networks.

Layer name	Conv1-1	Conv1-2	Max pooling1	Conv2-1	Conv2-2	Max pooling2
Parameters	3x3x64	3x3x64	2x2	3x3x128	3x3x128	2x2
Layer name	Conv3-1	Conv3-2	Conv3-3	Conv4-1	Conv4-2	Conv4-3
Parameters	3x3x256	3x3x256	3x3x256	3x3x512	3x3x512	3x3x512
Layer name	Conv5-1	Conv5-2	Conv5-3	ROI pooling1	ROI pooling2	ROI pooling3
Parameters	3x3x512	3x3x512	3x3x512	7x7	7x7	7x7

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Table 2. A per-video comparison on the OTB2015 [18]. The best two results are highlighted with red and blue fonts (last row of table).

	PTAV	HCF [10]	SRDCF [3]	Staple [1]	MEEM [19]	SINT [15]	LCT [10]	FDSSST [2]	KCF [6]	TGPR [4]	Struck [5]	DLT [16]
CarDark	1.000	1.000	1.000	1.000	1.000	0.964	1.000	1.000	1.000	1.000	1.000	0.715
Car4	1.000	0.997	1.000	1.000	0.686	1.000	0.989	1.000	0.953	1.000	0.992	1.000
David	0.994	1.000	1.000	1.000	0.904	1.000	1.000	1.000	0.987	0.329	0.321	
David2	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.711
Sylvester	0.914	0.852	0.845	0.850	0.954	0.977	0.975	0.837	0.843	0.946	0.995	0.770
Trellis	1.000	1.000	1.000	0.996	0.968	1.000	1.000	1.000	1.000	0.981	0.877	0.339
Fish	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.401
Mhyang	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
Soccer	0.941	0.816	0.934	0.296	0.314	0.531	0.151	0.946	0.791	0.143	0.253	0.138
Matrix	0.390	0.620	0.370	0.150	0.640	0.760	0.360	0.390	0.170	0.110	0.120	0.010
Ironman	0.596	0.645	0.030	0.145	0.506	0.614	0.145	0.078	0.217	0.096	0.114	0.127
Deer	1.000	1.000	1.000	1.000	1.000	1.000	0.817	1.000	0.817	1.000	1.000	0.042
Skating1	1.000	1.000	0.898	1.000	0.693	0.433	1.000	1.000	1.000	0.700	0.465	0.763
Shaking	0.978	0.868	0.014	0.019	0.995	0.981	0.984	0.953	0.019	0.641	0.192	0.926
Singer1	1.000	1.000	1.000	1.000	0.470	0.521	1.000	1.000	0.843	0.219	0.641	1.000
Singer2	0.926	0.041	0.973	1.000	0.038	0.732	0.973	0.904	0.948	0.954	0.036	0.036
Coke	0.966	0.962	0.818	0.897	0.945	0.969	0.914	0.890	0.838	0.942	0.948	0.340
Bolt	0.954	1.000	0.017	1.000	0.966	0.974	1.000	0.017	0.989	0.026	0.020	0.026
Boy	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
Dudek	0.876	0.905	0.833	0.822	0.792	0.930	0.907	0.876	0.877	0.681	0.897	0.918
Crossing	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.950	1.000	1.000
Couple	0.914	0.921	1.000	0.679	1.000	0.871	0.571	0.600	0.257	0.107	0.736	0.307
Football1	1.000	1.000	0.784	1.000	1.000	1.000	0.973	1.000	0.959	0.986	1.000	0.608
Jogging-1	0.974	0.974	0.974	0.228	0.964	0.980	0.971	0.231	0.235	0.225	0.241	0.228
Jogging-2	0.935	1.000	0.997	0.192	0.971	0.977	0.974	0.163	0.163	0.997	0.254	0.173
Doll	0.968	0.978	0.993	0.993	0.985	0.959	0.981	0.994	0.968	0.971	0.919	0.957
Girl	0.918	1.000	0.994	0.868	1.000	1.000	1.000	0.916	0.864	0.904	1.000	0.776
Walking2	0.798	1.000	1.000	1.000	0.392	0.982	0.404	0.774	0.434	0.996	0.982	1.000
Walking	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.748
Fleetface	0.710	0.590	0.597	0.628	0.591	0.714	0.549	0.625	0.460	0.393	0.639	0.434
Freeman1	0.951	0.979	0.948	1.000	0.997	0.402	0.972	0.951	0.402	0.985	0.801	0.380
Freeman3	0.913	0.811	0.996	0.915	0.985	0.935	0.783	0.917	0.911	0.122	0.789	1.000
Freeman4	0.799	0.943	0.996	0.703	0.565	0.339	0.951	0.194	0.534	0.519	0.375	0.346
David3	1.000	1.000	1.000	1.000	0.996	1.000	1.000	1.000	1.000	1.000	0.337	0.698
Jumping	0.974	1.000	1.000	0.307	1.000	0.978	0.978	0.946	0.342	0.109	1.000	0.962
CarScale	0.813	0.627	0.778	0.853	0.651	0.627	0.730	0.813	0.806	0.806	0.647	0.714
Skining	0.086	0.988	0.074	0.160	1.000	1.000	0.136	0.086	0.074	0.111	0.037	0.123
Dog1	1.000	1.000	1.000	1.000	0.982	0.850	1.000	1.000	1.000	1.000	0.996	0.996
Suv	0.978	0.979	0.975	0.978	0.743	0.957	0.980	0.978	0.979	0.531	0.572	1.000
MotorRolling	0.067	0.945	0.043	0.055	0.061	0.610	0.043	0.043	0.049	0.061	0.085	0.043
MountainBike	1.000	1.000	1.000	1.000	0.917	0.921	0.996	1.000	1.000	1.000	0.921	0.811
Lemming	0.939	0.258	0.323	0.273	0.911	0.875	0.856	0.275	0.487	0.275	0.628	0.298
Liquor	0.852	0.816	0.982	0.982	0.925	0.856	0.789	0.975	0.976	0.657	0.390	0.357
Woman	0.938	0.940	0.988	0.998	0.963	0.936	0.940	0.938	0.938	0.940	1.000	0.938
Faceocc1	0.882	0.600	0.831	0.918	0.683	0.858	0.906	0.882	0.730	0.831	0.575	0.462
Faceocc2	0.998	0.994	0.841	0.988	0.986	0.814	0.998	1.000	0.972	0.979	1.000	0.850
Basketball	0.930	1.000	0.996	0.879	0.892	0.866	1.000	0.913	0.923	0.993	0.120	0.086
Football	1.000	1.000	1.000	0.801	0.992	0.166	1.000	1.000	0.796	1.000	0.751	0.296
Subway	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.983	0.023
Tiger1	0.828	0.811	0.957	0.974	0.822	0.837	0.865	0.097	0.851	0.269	0.175	0.433
Tiger2	0.879	0.567	0.940	0.874	0.488	0.685	0.693	0.921	0.356	0.792	0.630	0.329
Biker	0.514	0.521	0.514	0.514	0.535	0.549	0.507	0.514	0.507	0.514	0.556	0.958
Bird1	0.373	0.392	0.071	0.363	0.289	0.453	0.346	0.373	0.069	0.811	0.150	0.588
Bird2	0.848	0.980	0.535	0.960	1.000	0.970	0.778	0.848	0.475	0.737	0.545	0.202
Blurbody	0.997	0.991	0.997	0.988	0.886	0.491	0.967	0.997	0.584	0.790	0.814	0.045
Blurcar1	0.997	0.995	0.999	0.695	0.993	0.410	0.997	0.997	0.995	0.035	0.996	0.026
Blurcar2	1.000	0.961	1.000	1.000	0.959	0.856	0.998	1.000	0.938	0.962	0.916	0.749
Blurcar3	1.000	1.000	1.000	0.986	1.000	0.683	1.000	1.000	0.994	0.042	1.000	0.252
Blurcar4	0.997	1.000	1.000	1.000	0.976	0.800	1.000	0.997	0.997	0.937	0.997	0.563
Blurface	1.000	1.000	1.000	0.998	0.990	0.937	1.000	1.000	1.000	0.990	0.436	0.191
Blurowl1	0.930	0.962	0.984	0.472	0.995	0.678	0.891	0.937	0.228	0.512	0.989	0.070
Board	0.090	0.870	0.761	0.774	0.605	0.529	0.734	0.090	0.656	0.053	0.752	0.570
Bolt2	0.696	0.952	0.017	0.997	0.017	0.014	0.017	0.017	0.020	0.109	0.073	
Box	0.924	0.394	0.415	0.414	0.370	0.960	0.068	0.396	0.415	0.294	0.239	0.396
Car1	1.000	0.391	1.000	1.000	0.196	0.350	0.438	1.000	0.739	0.337	0.334	1.000
Car2	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.074	1.000	1.000
Car24	1.000	1.000	1.000	1.000	1.000	0.942	0.853	1.000	1.000	0.992	0.170	1.000
ClifBar	1.000	0.915	0.945	0.697	0.915	0.572	0.939	1.000	0.445	0.146	0.581	0.464
Coupon	1.000	1.000	1.000	1.000	0.394	0.388	1.000	1.000	1.000	0.388	1.000	0.382
Crowds	1.000	1.000	1.000	1.000	0.991	0.991	1.000	1.000	1.000	1.000	0.911	0.916
Dancer	1.000	1.000	1.000	0.916	0.969	1.000	1.000	1.000	1.000	0.964	0.987	0.964
Dancer2	1.000	1.000	1.000	0.980	1.000	1.000	1.000	1.000	1.000	0.993	1.000	1.000
Diving	0.405	0.753	0.391	0.363	0.209	0.433	0.753	0.400	0.535	0.214	0.521	0.256
Dog	0.992	1.000	1.000	0.992	1.000	1.000	0.756	0.992	0.992	0.945	0.961	
DragonBaby	0.442	0.867	0.336	0.858	0.823	0.850	0.549	0.389	0.336	0.752	0.106	0.372
Gir12	0.955	0.076	0.075	0.087	0.801	0.746	0.076	0.080	0.071	0.577	0.272	0.074
Gym	0.934	0.988	0.983	0.977	0.913	0.952	0.986	0.553	0.795	0.858	0.597	0.146
Human2	0.441	0.540	0.848	0.894	0.180	0.797	0.595	0.160	0.171	0.738	0.432	0.556
Human3	0.975	0.035	0.034	0.034	0.866	0.068	0.006	0.006	0.006	0.010	0.010	0.009
Human4	0.828	0.852	1.000	0.958	0.504	0.597	0.852	0.204	0.534	0.508	0.211	0.205
Human5	0.997	0.245	0.997	1.000	0.997	0.443	0.245	0.997	0.265	0.993	0.990	0.891
Human6	0.855	0.381	0.924	0.980	0.663	0.328	0.274	0.855	0.290	0.295	0.255	0.446
Human7	0.964	1.000	1.000	1.000	0.472	0.308	0.472	0.472	0.472	0.856	1.000	0.436
Human8												

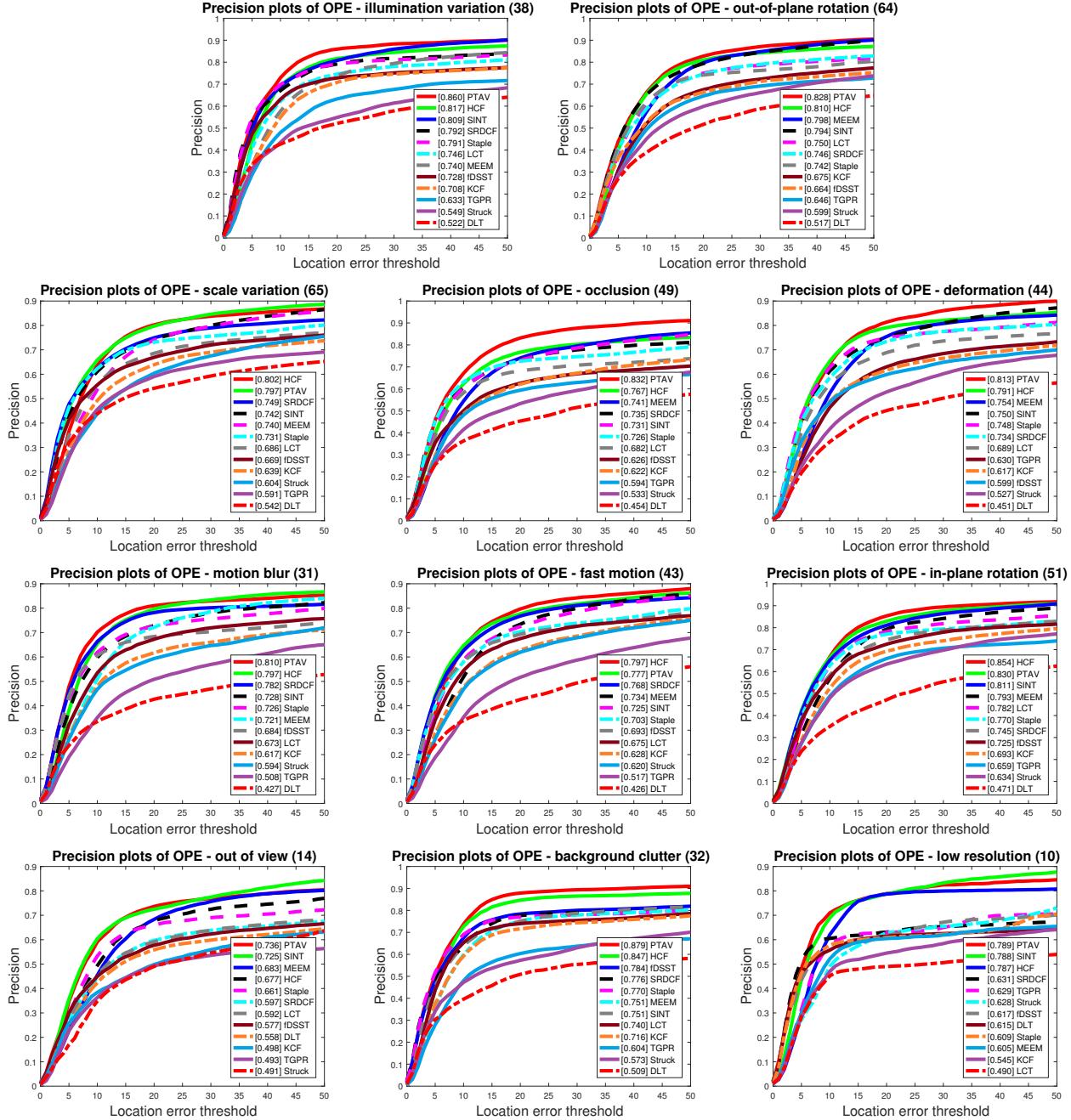


Figure 2. Distance precision plots on OTB2015 [18] for 11 attributes, which are background cluttered (BC), deformation (DEF), fast motion (FM), in-plane rotation (IPR), illumination variation (IV), low resolution (LR), motion blur (MB), occlusion (OCC), out-of-plane rotation (OPR), out-of-view (OV) and scale variation (SV). The distance precision (DP) score is shown in the legend. Our PTAV ranks top 2 on all 11 attributes.

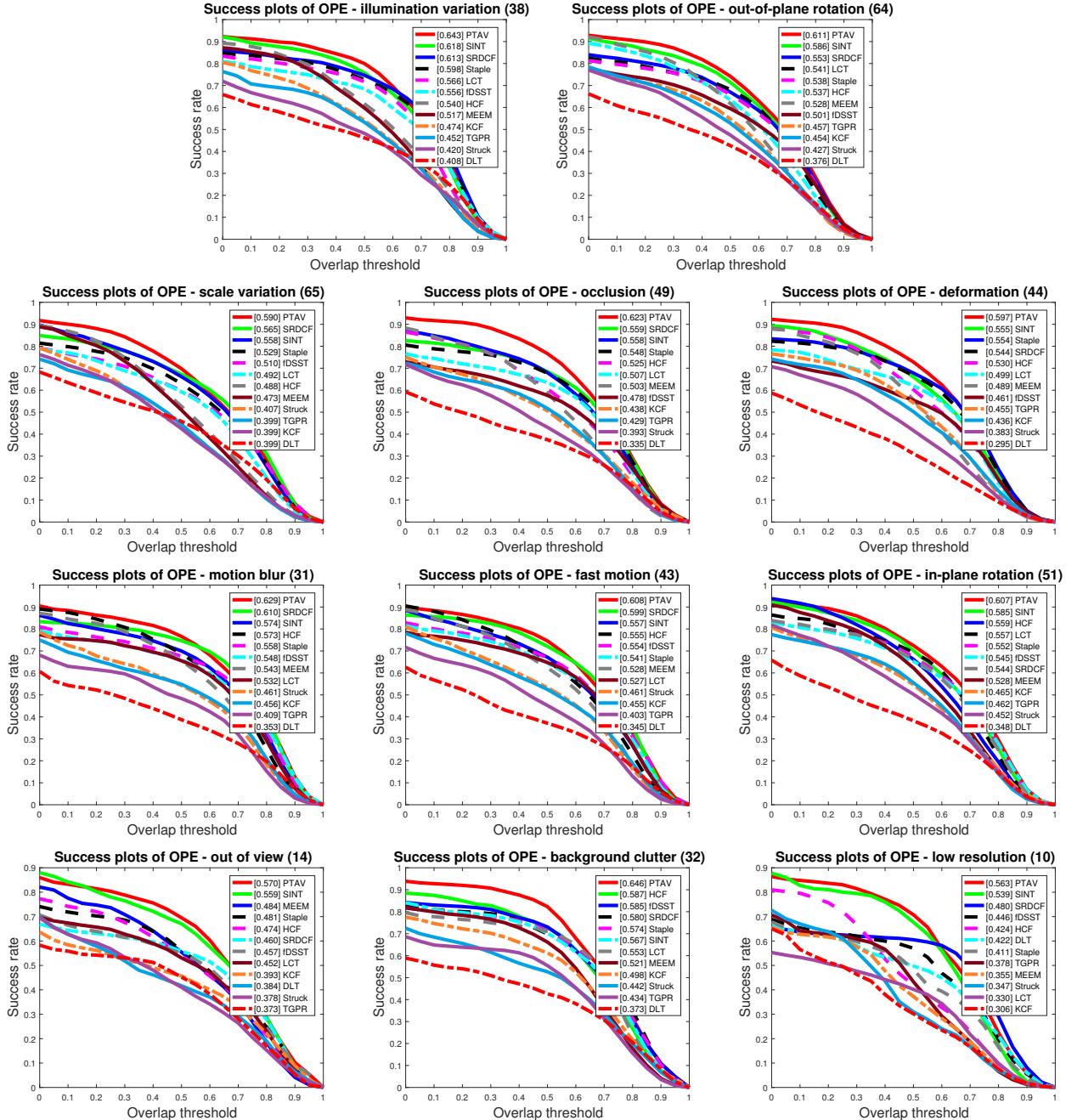


Figure 3. Overlap success plots on OTB2015 [18] for 11 attributes, which are background cluttered (BC), deformation (DEF), fast motion (FM), in-plane rotation (IPR), illumination variation (IV), low resolution (LR), motion blur (MB), occlusion (OCC), out-of-plane rotation (OPR), out-of-view (OV) and scale variation (SV). The title of each attribute plot contains the name of the attribute and the number of videos associated with it. The overlap success (OS) score is shown in the legend. Our PTAV ranks top 2 on all 11 attributes.

Table 3. A per-video comparison on the TC128 [9]. The best two results are highlighted with red and blue fonts (last row of table).

	PTAV	HCF [11]	Staple [1]	SRDCF [3]	MEEM [19]	Struck [5]	LCT [11]	fDSST [2]	KCF [6]		PTAV	HCF [11]	Staple [1]	SRDCF [3]	MEEM [19]	Struck [5]	LCT [11]	fDSST [2]	KCF [6]
David	0.992	1.000	1.000	1.000	0.985	0.975	1.000	1.000	1.000	Kite.cce2	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.600
Airport.cce	0.757	0.392	0.493	0.446	0.378	0.405	0.412	0.459	0.392	Kite.cce3	0.994	1.000	1.000	1.000	1.000	1.000	1.000	0.994	0.994
Baby.cce	1.000	1.000	1.000	1.000	0.956	0.892	0.841	1.000	0.662	Kobe.cce	0.588	0.211	0.215	0.246	0.414	0.275	0.211	0.270	0.220
Badminton.cce1	0.997	0.997	0.915	0.988	0.986	0.998	0.993	0.352	1.000	Lemming	0.939	0.258	0.273	0.323	0.908	0.776	0.856	0.275	0.275
Badminton.cce2	0.918	0.973	0.969	0.787	0.631	0.915	0.929	0.868	0.081	Liquor	0.967	0.816	0.928	0.982	0.551	0.258	0.789	0.974	0.430
Ball.cce1	0.026	0.368	0.031	0.023	0.031	0.192	0.018	0.023	0.028	Logo.cce	0.916	0.290	1.000	1.000	0.464	0.989	0.393	0.993	1.000
Ball.cce2	0.758	0.917	0.481	0.483	0.771	0.599	0.569	0.458	0.479	Matrix	0.390	0.620	0.360	0.370	0.090	0.300	0.360	0.360	0.280
Ball.cce3	0.919	0.762	0.762	0.762	0.762	0.762	0.762	0.762	0.762	Messi.cce	0.717	0.636	0.768	0.989	1.000	0.967	0.533	0.533	0.283
Ball.cce4	0.457	0.054	0.058	0.058	0.059	0.058	0.035	0.058	0.015	Michaeljackson.cce	0.573	0.316	0.575	0.623	0.565	0.656	0.455	0.575	0.776
Basketball	0.930	1.000	0.879	0.996	0.877	0.259	1.000	0.913	0.921	Microphone.cce1	0.863	0.995	0.618	1.000	0.990	1.000	1.000	0.853	1.000
Basketball.cce1	0.522	0.938	0.649	0.643	0.649	0.135	0.978	0.377	0.480	Microphone.cce2	1.000	1.000	1.000	1.000	1.000	1.000	0.602	1.000	1.000
Basketball.cce2	0.549	0.358	0.587	0.798	0.112	0.481	0.582	0.145	0.244	MotorRolling	0.104	0.945	0.055	0.043	0.043	0.159	0.043	0.043	0.049
Basketball.cce3	0.683	0.955	1.000	0.728	0.760	0.776	0.735	0.726	0.726	Motorbike.cce	0.947	0.645	0.236	0.238	0.698	0.236	0.238	0.238	0.238
Bee.cce	0.922	1.000	1.000	0.422	1.000	1.000	0.389	0.400	0.289	MountainBike	1.000	1.000	1.000	0.886	1.000	0.996	1.000	1.000	
Bicycle	1.000	1.000	1.000	1.000	0.646	1.000	0.524	1.000	0.524	Panda	0.033	0.535	1.000	1.000	0.523	1.000	0.382	0.033	0.033
Bike.cce1	1.000	1.000	1.000	1.000	0.999	0.984	1.000	1.000	1.000	Plane.cce2	0.714	0.940	0.980	0.681	0.098	0.300	0.931	0.714	0.914
Bike.cce2	0.180	0.988	0.138	1.000	0.248	0.164	0.472	0.167	0.548	Plate.cce1	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
Biker	0.650	0.594	0.956	0.461	0.589	0.683	0.456	0.467	0.467	Plate.cce2	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
Bikeshow.cce	0.213	0.753	0.050	0.152	0.102	0.579	0.055	0.288	0.258	Pool.cce1	0.952	0.054	0.054	0.072	1.000	1.000	0.048	0.048	0.048
Bird	0.838	0.980	0.909	0.051	0.980	0.485	0.768	0.515	0.566	Pool.cce2	0.940	0.023	0.030	0.030	1.000	1.000	0.023	0.023	0.023
Board	0.186	0.834	0.860	0.855	0.023	0.263	0.831	0.100	0.803	Pool.cce3	0.065	0.065	0.056	0.056	1.000	0.032	0.065	0.065	0.056
Boat.cce1	0.061	0.080	0.066	0.531	0.151	0.074	0.265	0.061	0.130	Railwaystation.cce	0.717	0.785	0.092	0.964	0.111	0.036	0.036	0.036	0.036
Boat.cce2	0.701	0.743	0.745	0.733	0.743	0.745	0.697	0.748	0.745	Ring.cce	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
Bolt	0.954	1.000	1.000	0.017	0.983	0.054	1.000	0.017	0.997	Sailor.cce	0.988	1.000	1.000	0.396	1.000	1.000	0.413	0.582	0.403
Boy	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	Shaking	0.964	0.868	0.019	0.014	0.975	0.770	0.984	0.953	0.022
Busstation.cce1	0.972	0.107	0.107	0.105	0.124	0.116	0.113	0.127	0.113	Singer1	1.000	1.000	1.000	1.000	0.567	1.000	1.000	1.000	1.000
Busstation.cce2	1.000	0.929	0.899	1.000	0.876	0.878	0.896	1.000	0.899	Singer2	0.929	0.041	1.000	0.973	0.036	0.036	0.973	0.904	0.973
CarDark	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	Singer.cce1	0.897	0.963	0.949	0.953	0.813	0.972	1.000	0.897	0.939
CarScale	0.813	0.627	0.841	0.778	0.714	0.690	0.730	0.813	0.806	Singer.cce2	0.257	0.057	0.790	0.053	0.802	0.007	0.393	0.133	0.275
Carchasing.cce1	0.920	0.287	0.283	0.930	0.281	0.279	0.289	0.285	0.285	Skating1	1.000	1.000	0.525	0.898	0.805	0.990	1.000	1.000	1.000
Carchasing.cce3	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	Skating2	0.446	0.871	0.823	0.774	0.226	0.298	0.798	0.038	0.588
Carchasing.cce4	1.000	0.973	1.000	1.000	0.541	0.394	0.229	1.000	0.837	Skating.cce1	0.394	0.337	0.682	0.054	0.570	0.323	0.401	0.059	0.296
Charger.cce	0.117	0.091	0.658	0.577	0.027	0.084	0.087	0.124	0.101	Skating.cce2	0.087	0.284	0.433	0.052	0.151	0.052	0.087	0.099	0.072
Coke	0.966	0.962	0.887	0.818	0.945	0.811	0.914	0.897	0.852	Skiing	0.086	0.988	0.160	0.074	0.123	0.062	0.136	0.086	0.074
Couple	0.729	0.921	0.679	1.000	1.000	0.714	0.571	0.586	0.250	Skiing.cce	0.411	0.581	0.785	0.836	0.755	0.687	0.652	0.528	0.405
Crossing	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	Skyjumping.cce	0.862	0.733	0.934	0.074	0.724	0.638	0.087	0.292	0.361
Cup	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	Soccer	0.941	0.816	0.306	0.934	0.288	0.288	0.151	0.946	0.161
Cup.cce	0.038	0.012	0.021	0.018	0.015	0.015	0.018	0.038	0.041	Spiderman.cce	0.083	0.490	0.353	0.450	0.063	0.014	0.254	0.071	0.060
David3	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	Subway	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
Deer	1.000	1.000	0.831	1.000	1.000	0.901	0.817	1.000	0.887	Suitcase.cce	0.886	0.804	0.815	0.804	0.793	0.391	0.793	0.821	0.799
Diving	0.472	0.801	0.342	0.645	0.182	0.242	0.641	0.394	0.368	Sunshade	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
Doll	0.969	0.978	0.992	0.993	0.984	0.977	0.981	0.994	0.973	SuperMario.cce	0.884	0.815	1.000	1.000	1.000	0.993	0.322	0.404	0.233
Eagle.cce	1.000	1.000	0.536	0.464	1.000	1.000	0.080	1.000	0.080	Surf.cce1	0.153	0.438	0.205	0.062	0.178	0.193	0.248	0.040	0.035
Electricalbike.cce	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	Surf.cce2	0.136	0.020	0.049	0.430	0.069	0.041	0.028	0.013	0.013
FaceOcc1	0.882	0.600	0.918	0.831	0.720	0.706	0.906	0.882	0.731	Surf.cce3	0.158	0.538	0.616	0.237	0.444	0.093	0.398	0.140	0.136
Face.cce	0.913	0.037	0.044	0.042	0.037	0.037	0.037	0.042	0.044	Surf.cce4	0.081	0.044	0.333	0.081	0.193	0.074	0.111	0.170	0.081
Face.cce2	0.338	0.682	0.716	0.588	0.520	0.473	0.088	0.101	0.101	TableTennis.cce	0.864	0.763	0.394	0.106	0.157	0.626	0.657	0.833	0.657
Fish.cce1	0.858	0.698	0.728	0.090	0.085	0.072	0.085	0.077	0.082	TennisBall.cce	0.396	0.038	0.024	0.031	0.024	0.132	0.021	0.021	0.017
Fish.cce2	0.545	0.194	0.147	0.159	0.215	0.387	0.148	0.148	0.161	Tennis.cce1	0.991	1.000	1.000	1.000	1.000	1.000	1.000	0.989	0.993
Football1	1.000	0.973	1.000	0.784	1.000	1.000	0.973	1.000	0.986	Tennis.cce2	0.993	0.990	1.000	1.000	1.000	1.000	1.000	0.993	1.000
Girl	0.882	1.000	0.864	0.994	1.000	1.000	1.000	0.894	0.866	Tennis.cce3	0.951	0.113	1.000	1.000	1.000	1.000	0.108	0.108	0.574
Girlmov	0.962	0.075	0.075	0.075	0.983	0.227	0.075	0.075	0.075	Thunder.cce	1.000	1.000	1.000	1.000	1.000	1.000	0.704	1.000	1.000
Guitar.cce1	0.985	0.963	0.933	1.000	1.000	0.985	0.985	0.985	0.985	Tiger1	0.946	0.856	0.975	0.975	0.935	0.856	0.890	0.949	0.873
Guitar.cce2	0.543	0.840	0.514	0.508	0.454</td														

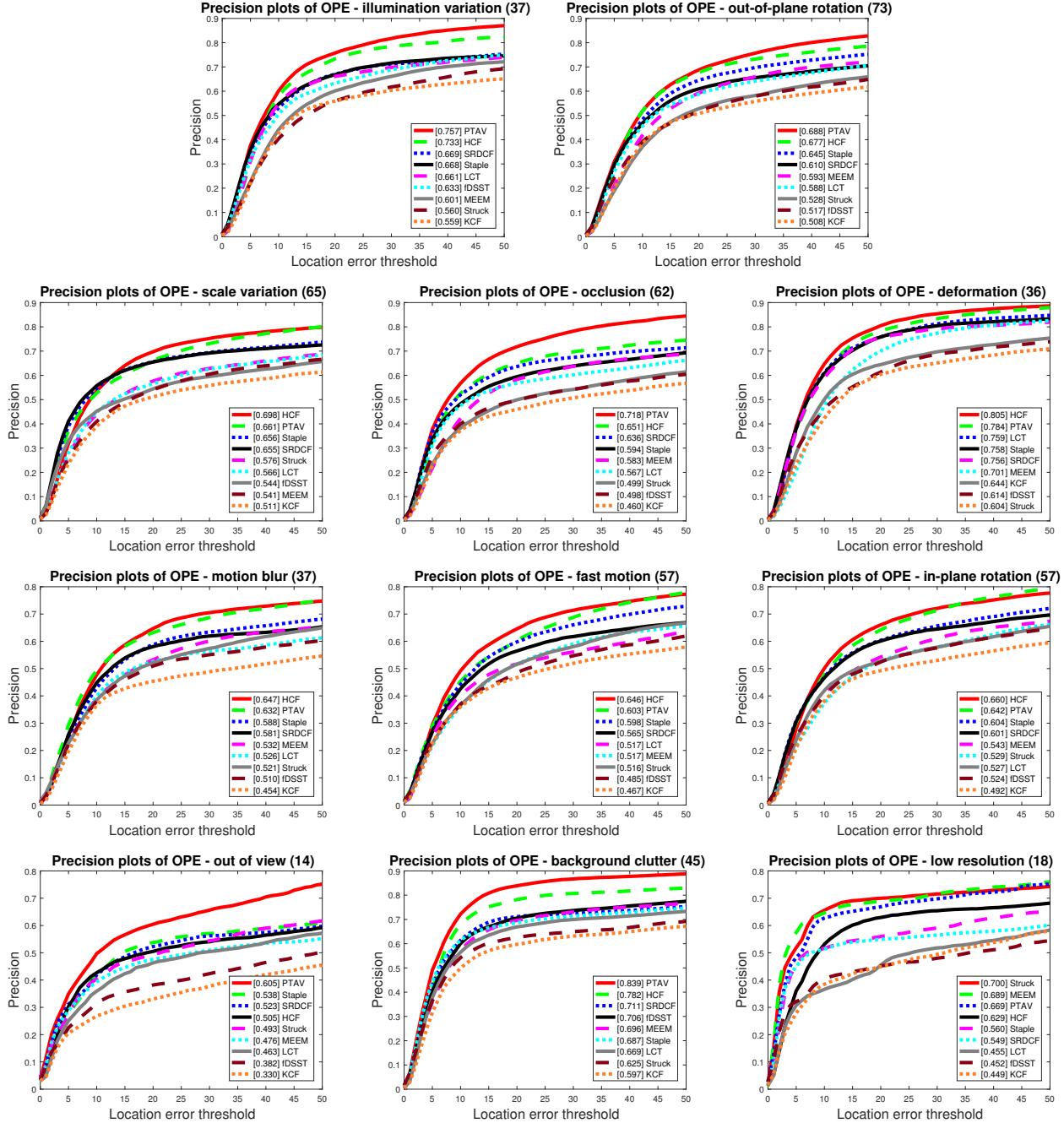


Figure 4. Distance precision plots on TC128 [9] for 11 attributes, which are background cluttered (BC), deformation (DEF), fast motion (FM), in-plane rotation (IPR), illumination variation (IV), low resolution (LR), motion blur (MB), occlusion (OCC), out-of-plane rotation (OPR), out-of-view (OV) and scale variation (SV). The distance precision (DP) score is shown in the legend. Our PTAV ranks top 2 on all 10 of 11 attributes.

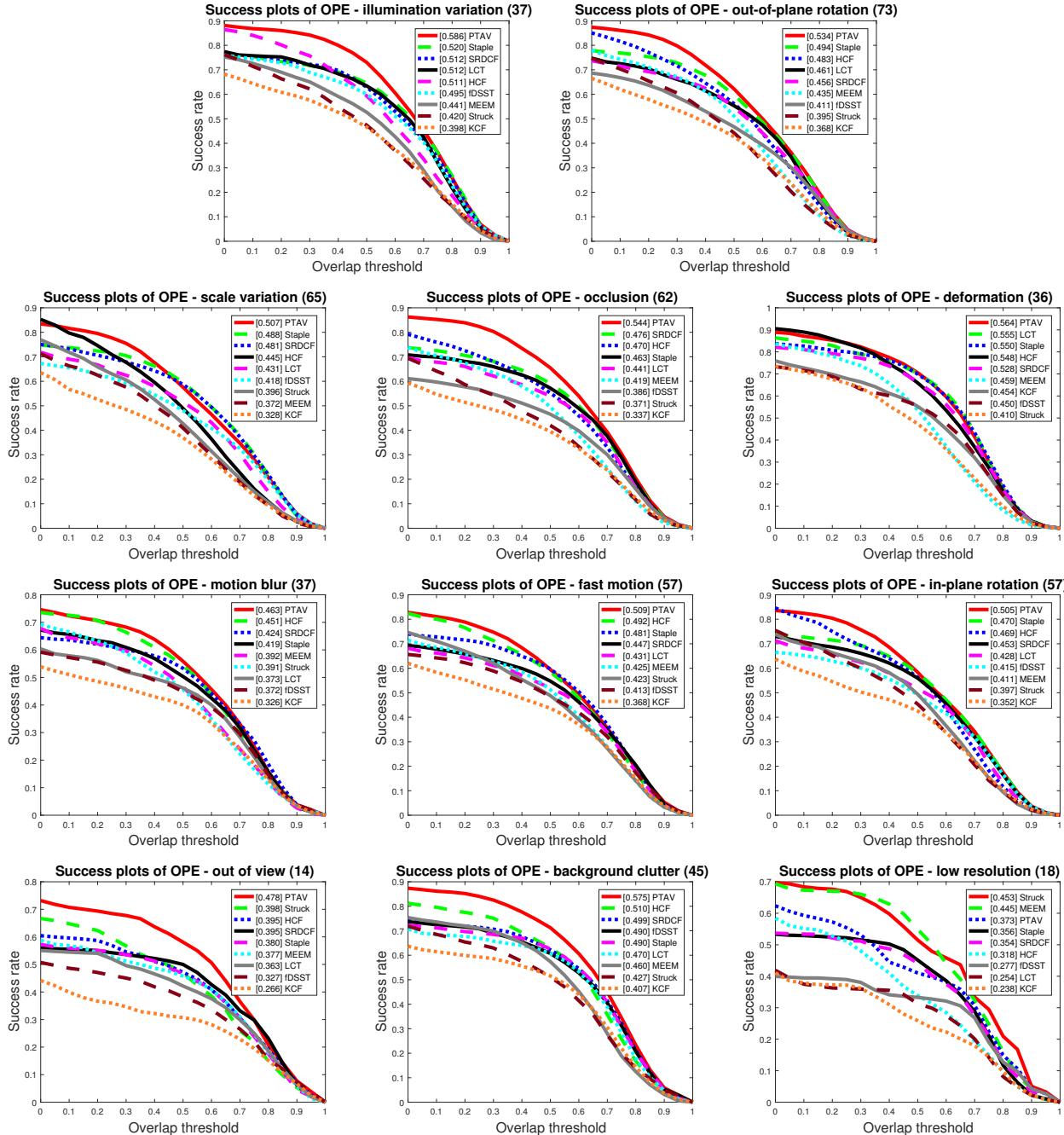


Figure 5. Overlap success plots on TC128 [9] for 11 attributes, which are background cluttered (BC), deformation (DEF), fast motion (FM), in-plane rotation (IPR), illumination variation (IV), low resolution (LR), motion blur (MB), occlusion (OCC), out-of-plane rotation (OPR), out-of-view (OV) and scale variation (SV). The title of each attribute plot contains the name of the attribute and the number of videos associated with it. The overlap success (OS) score is shown in the legend. Our PTAV ranks top 2 on all 10 of 11 attributes.

Table 4. A per-video comparison on the UAV20L [12]. The best two results are highlighted with **red** and **blue** fonts (last row of table).

	PTAV	MUSTer [7]	SRDCF [3]	HCF [10]	MEEM [19]	SAMF [8]	Struck [5]	fDSST [2]	LCT [11]	KCF [6]
Bike1	0.456	0.406	0.330	0.579	0.576	0.395	0.156	0.449	0.349	0.138
Bird1	0.439	0.436	0.436	0.435	0.441	0.436	0.461	0.439	0.436	0.435
Car1	0.830	0.657	0.657	0.618	0.385	0.615	0.569	0.657	0.618	0.382
Car3	1.000	0.981	1.000	1.000	1.000	1.000	0.912	1.000	1.000	0.837
Car6	0.369	0.350	0.117	0.118	0.098	0.175	0.158	0.148	0.124	0.125
Car8	0.434	0.419	0.750	0.087	0.407	0.250	0.388	0.434	0.106	0.078
Car9	0.988	0.979	0.424	0.109	0.396	0.423	0.115	0.423	0.151	0.225
Car16	0.712	0.118	0.356	0.666	0.349	0.125	0.408	0.440	0.093	0.112
Group1	0.356	0.548	0.895	0.939	0.930	0.798	0.672	0.356	0.490	0.212
Group2	0.899	0.350	0.114	0.114	0.114	0.114	0.143	0.114	0.114	0.114
Group3	0.399	0.130	0.362	0.363	0.366	0.367	0.365	0.352	0.349	0.365
Person2	0.881	1.000	1.000	1.000	1.000	1.000	1.000	0.881	1.000	0.996
Person4	0.298	0.266	0.937	0.877	0.457	0.294	0.450	0.299	0.298	0.299
Person5	0.742	1.000	0.980	1.000	1.000	1.000	0.709	0.683	0.660	0.508
Person7	0.885	0.321	0.330	0.352	0.357	0.196	0.505	0.203	0.186	0.143
Person14	0.976	0.801	0.063	0.064	0.063	0.064	0.064	0.064	0.064	0.064
Person17	0.994	0.406	0.605	0.999	0.993	0.996	0.894	0.994	0.605	0.605
Person19	0.138	0.233	0.297	0.303	0.221	0.301	0.201	0.138	0.173	0.156
Person20	0.236	0.474	0.376	0.228	0.173	0.480	0.217	0.236	0.445	0.317
Uav1	0.443	0.411	0.115	0.154	0.319	0.107	0.361	0.122	0.108	0.110
Average	0.624	0.514	0.507	0.500	0.482	0.457	0.437	0.422	0.368	0.311

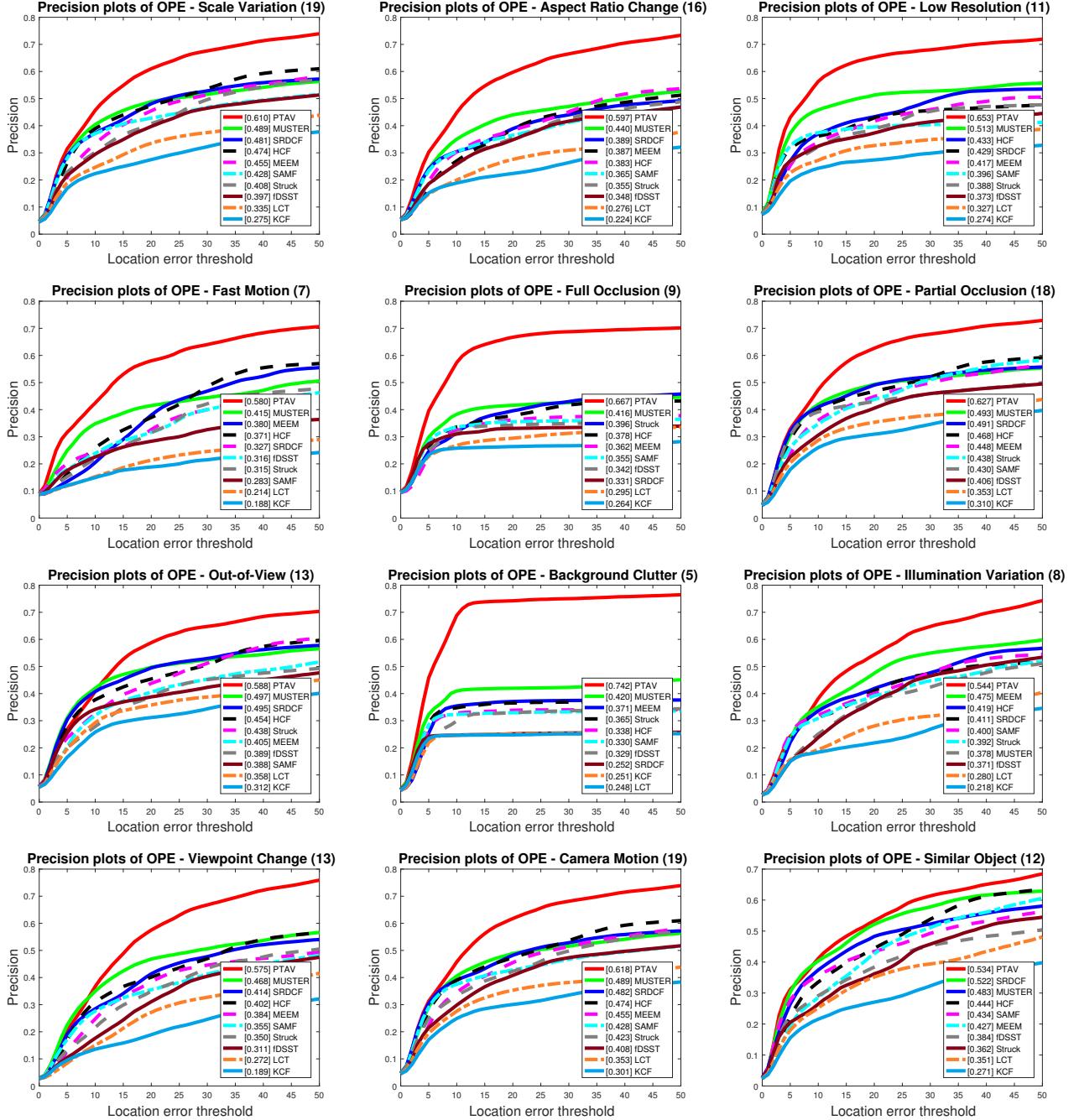


Figure 6. Distance precision plots on UAV20L [12] for 12 attributes, which are scale variation (SV), aspect ratio change (ARC), low resolution (LR), fast motion (FM), full occlusion (FOC), partial occlusion (POC), out-of-view (OV), background clutter (BC), illumination variation (IV), viewpoint change (VC), camera motion (CM) and similar object (SOB). The distance precision (DP) score is shown in the legend. Our PTAV ranks top 1 on all 12 attributes.

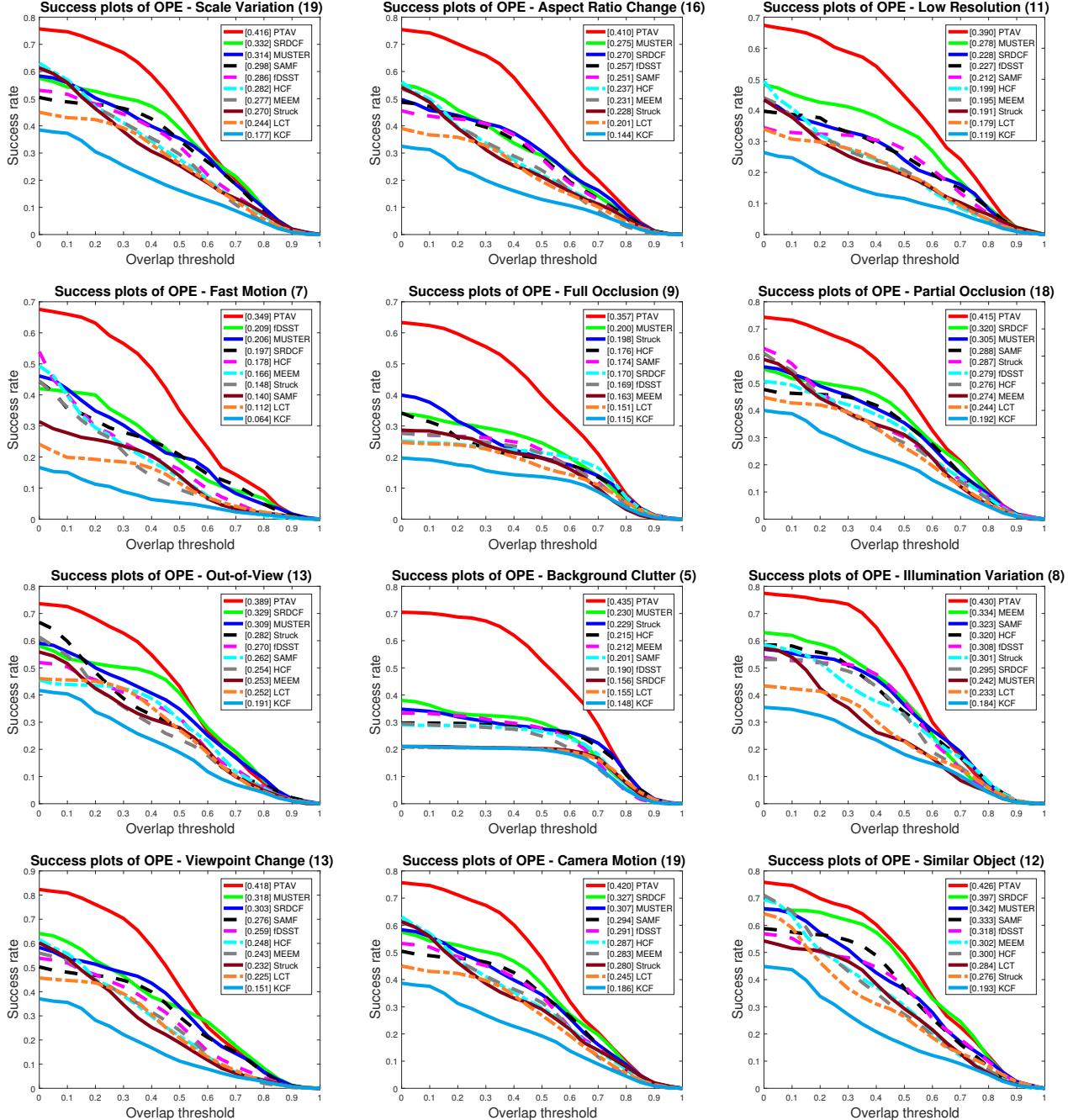


Figure 7. Overlap success plots on UAV20L [12] for 12 attributes, which are scale variation (SV), aspect ratio change (ARC), low resolution (LR), fast motion (FM), full occlusion (FOC), partial occlusion (POC), out-of-view (OV), background clutter (BC), illumination variation (IV), viewpoint change (VC), camera motion (CM) and similar object (SOB). The title of each attribute plot contains the name of the attribute and the number of videos associated with it. The overlap success (OS) score is shown in the legend. Our PTAV ranks top 1 on all 12 attributes.