

The Visual Object Tracking Challenge Results VOT-RGBT 2019

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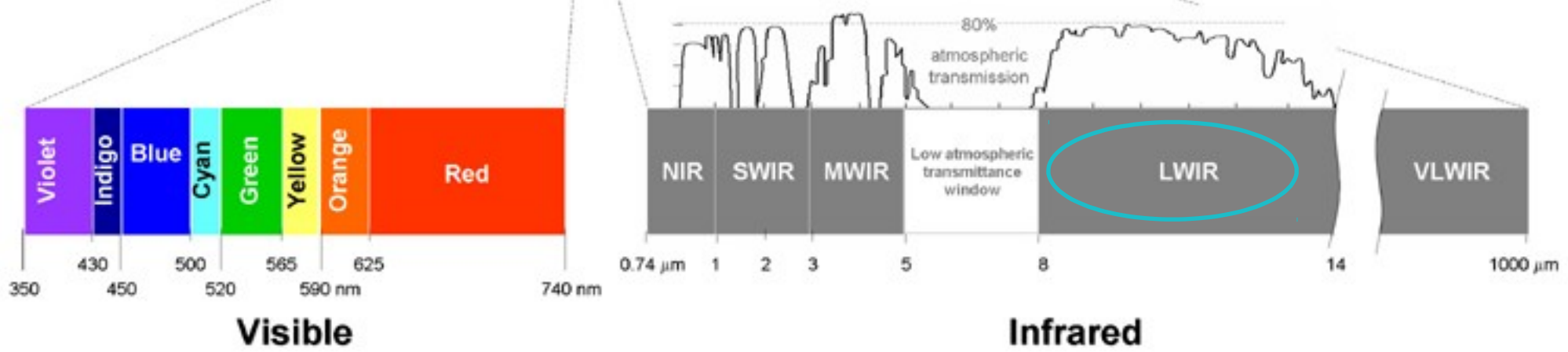
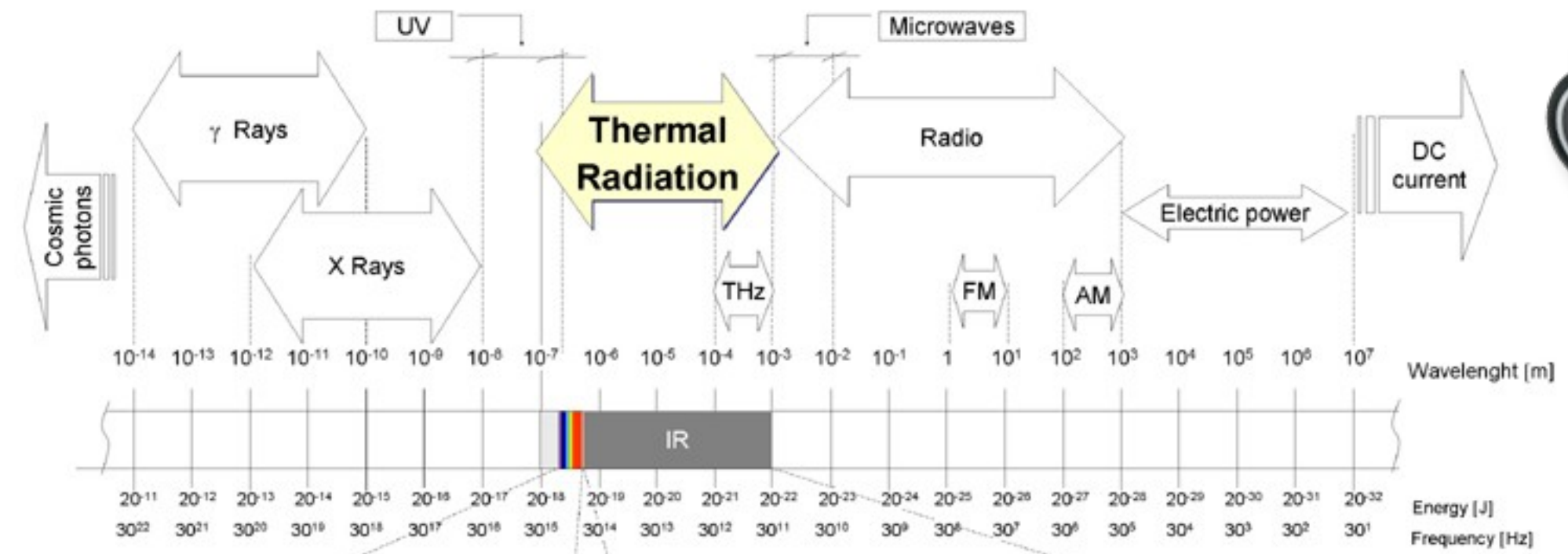


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Why adding Thermal Image Modality?



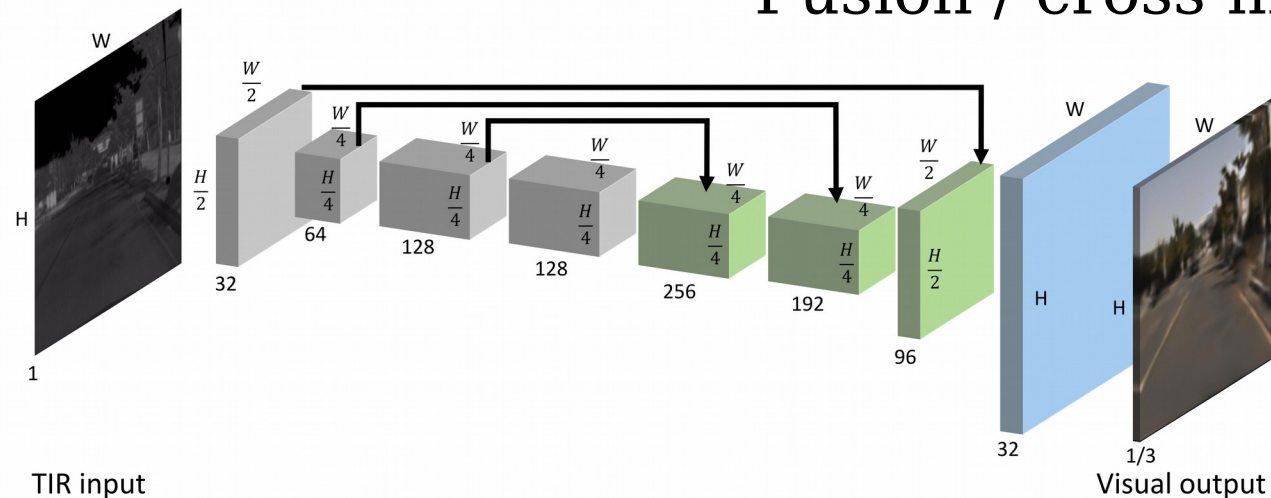


Applications of TIR

- Scientific research
- Security
- Fire monitoring
- Search and rescue
- Automotive safety
- Personal use
- ~~Military~~

Challenges

- Interpretation of TIR images
 - TIR2RGB
- Tracking: RGB and TIR
 - Calibration and registration
 - Understanding the similarities and complementarities (VOT-TIR)
 - Fusion / cross modality (VOT-RGBT)



Pre-VOT datasets for tracking in TIR

Name	Purpose	Resolution	#Bits	Stat/Mov
OSU Pedestrian [5]	Pedestrian detection and tracking.	360×240	8	Y/N
OSU Color-Thermal [6]	Pedestrian detection, tracking and thermal/visual fusion.	360×240	8	Y/N
Terravic Motion [7]	Detection and tracking	320×240	8	Y/N
LITIV [8]	Visible-infrared registration.	320×240	8	Y/N
ASL-TID [9]	Object (pedestrian, cat, horse) detection and tracking.	324×256	8/16	N/Y
BU-TIV [10]	Various visual analysis tasks. Single-object, multiple-object and multiple sensor tracking as well as motion patterns.	Up to 1024×1024	16	Y/N



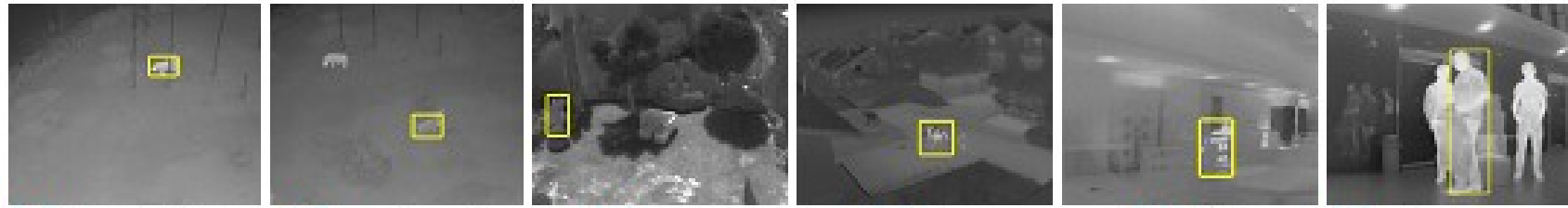
Why a separate challenge?

Tracking in TIR different from tracking in low resolution grayscale visual?

Many similarities but also interesting differences

- 16-bit
- Constant values if radiometric
- Less structure/edges/texture
- No shadows
- Noise: blooming, resolution, dead pixels

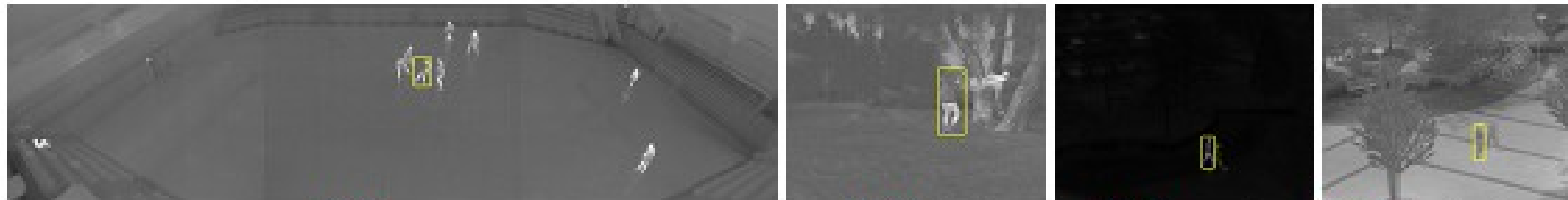
Towards VOT-TIR: Linköping Thermal InfraRed (LTIR) dataset



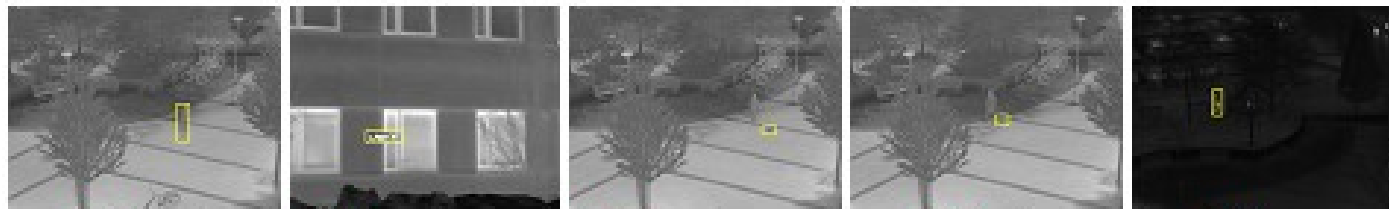
(1) rhino behind tree (2) running rhino (3) garden (4) horse (5) hiding (6) mixed distractors



(7) saturated (8) street (9) car (10) crouching (11) crowd



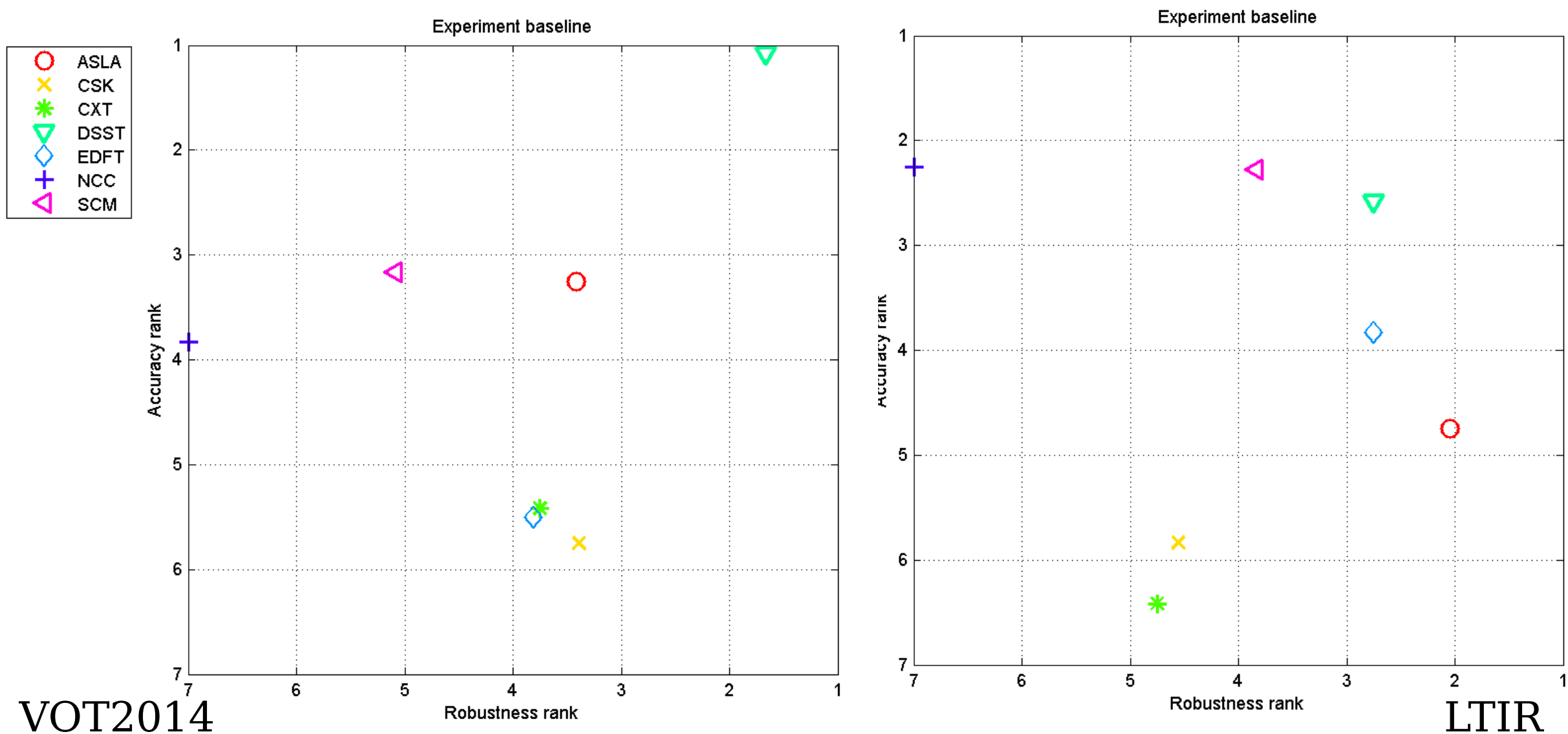
(12) soccer (13) birds (14) crossing (15) depthwise crossing



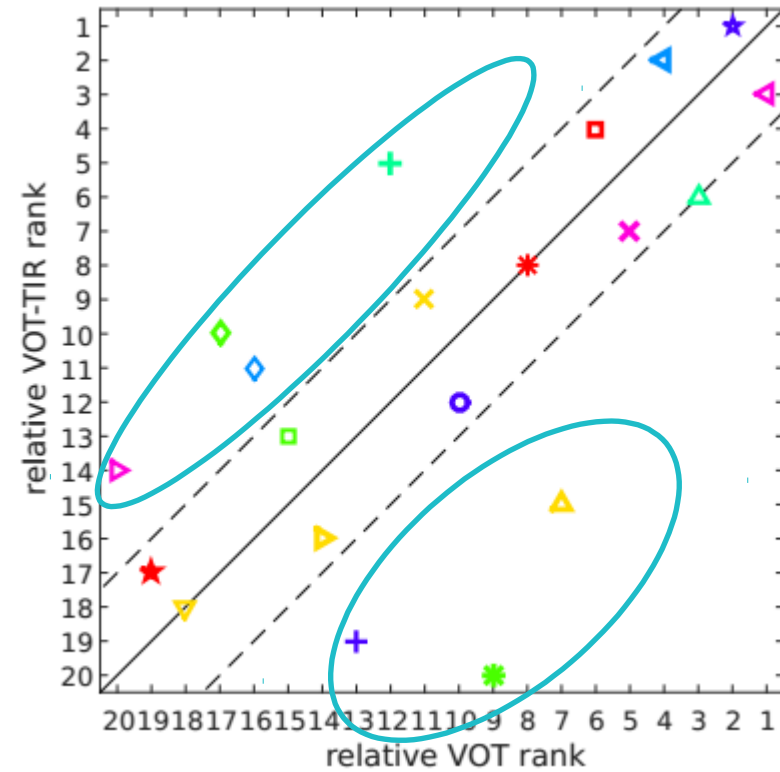
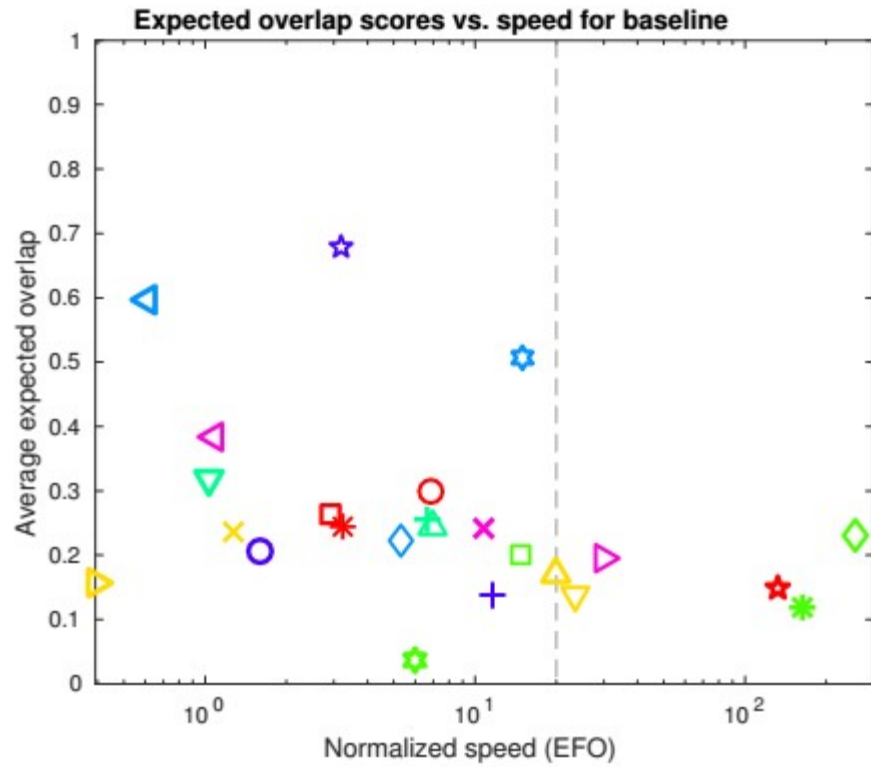
(16) jacket (17) quadrocopter (18) quadrocopter2 (19) selma (20) trees

A. Berg,
J. Ahlberg,
M. Felsberg,
*A Thermal
Object
Tracking
Benchmark.*
AVSS 2015.

Will it be different? Test against VOT2014



VOT2015 vs VOT-TIR2015



- | | | | | | | | |
|--------|---------|--------|-----------|--------|------------|------------|-----------|
| ○ ABCD | × AOG | * ASMS | ▽ CCFP | ◇ CMIL | + Dtracker | ◁ EBT | ★ FoT |
| ▷ GGT | □ KCFv2 | △ LDP | ☆ MCCT | ○ MKCF | × NSAMF | * OACF | ▽ PKLTF |
| ◇ sKCF | + SME | ▷ sPST | ☆ SRDCFir | ▷ STC | □ Struck | △ SumShift | * HotSpot |

Modifications of LTIR

- VOT-TIR2015 was already saturated
- Call for sequences - limited success (3 new sources, too easy)
- Easiest sequences have been removed: *Crossing*, *Horse*, and *Rhino behind tree*
- New, more difficult sequences have been added: *Bird*, *Boat1*, *Boat2*, *Car2*, *Dog*, *Excavator*, *Ragged*, and *Trees2*

Beihang
University



Properties

- 25 Sequences
- Average sequence length 740
- Annotations in accordance with VOT
 - Bounding-box
 - 11 global attributes (per-sequence)

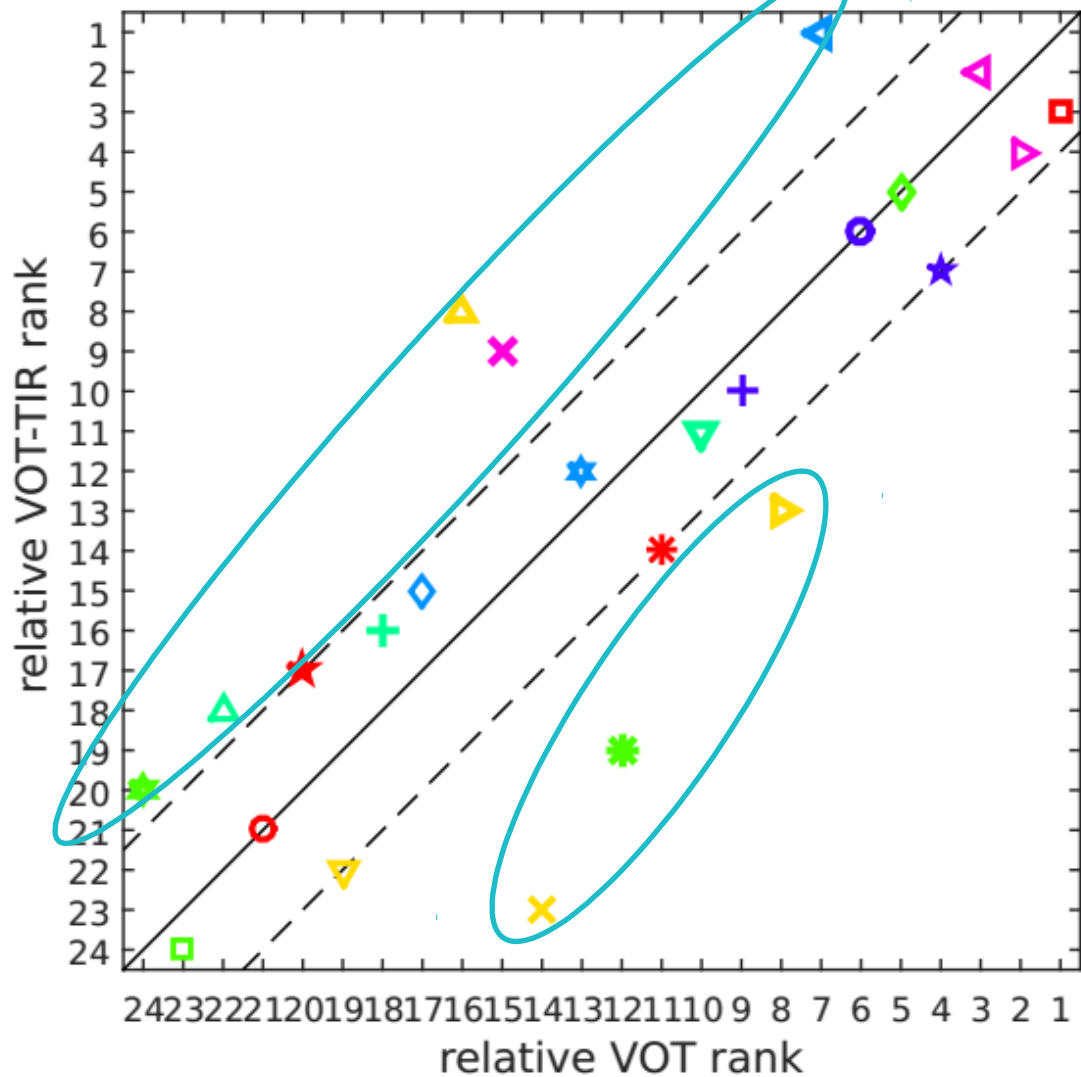
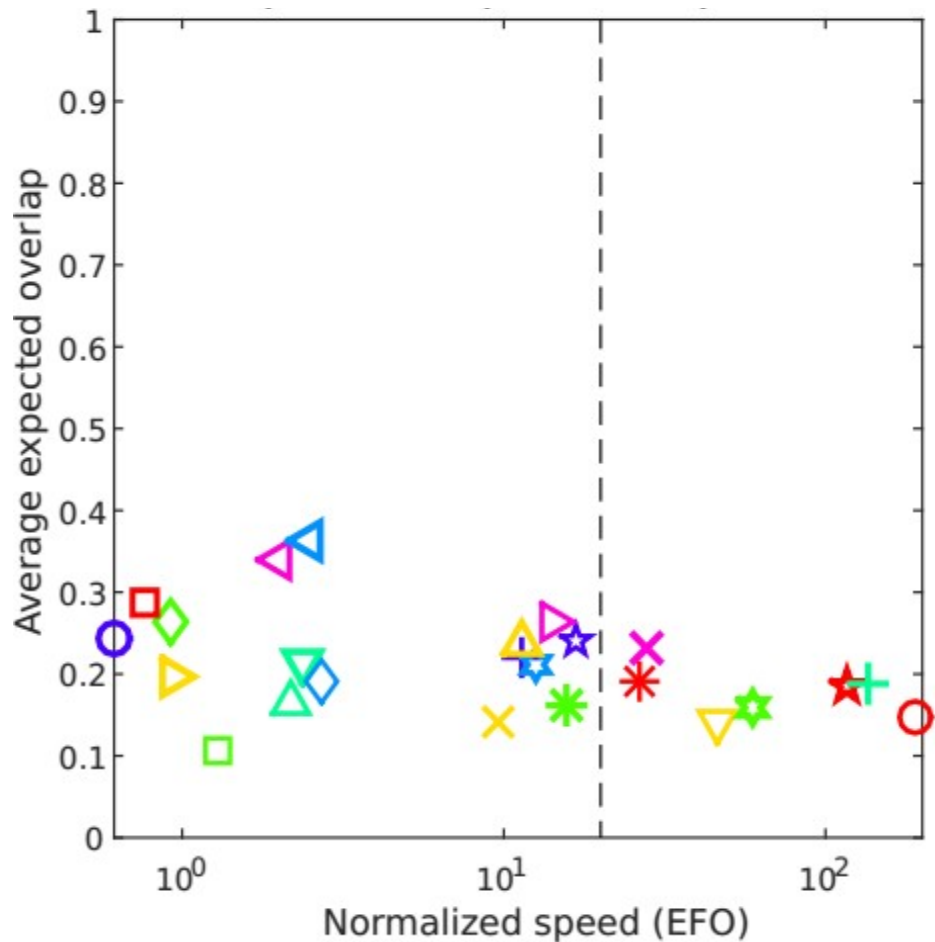
Blur, **dynamics change**, **temperature change**, object motion, size change, camera motion, background clutter, aspect ratio change, object deformation, scene complexity, neutral

- 6 local attributes (per-frame)

Occlusion, **dynamics change**, motion change, size change, camera motion, neutral



VOT2016 vs VOT-TIR2016



- | | | | | | |
|-----------|-----------|--------------|-------------|------------|--------|
| ○ BDF | × BST | * DAT | ▽ deepMKCF | ◇ DPCF | + DPT |
| △ EBT | ☆ FCT | ▽ GGTv2 | □ LoFT-Lite | △ LT-FLO | ☆ MAD |
| ○ MDNet-N | × MvCF | * NSAMF | ▽ PKLTF | ◇ SHCT | + sKCF |
| △ SRDCFir | ☆ STAPLE+ | ▽ Staple-TIR | □ TCNN | △ DSST2014 | * NCC |

RGBT-dataset

- RGBT234-dataset from: C. Li, X. Liang, Y. Lu, N. Zhao, and J. Tang. RGB-T object tracking: Benchmark and baseline. Pattern Recognition (96), 2019
- 234 sequences with an average length of 335 frames
- Same clustering in 11-dim attribute space, but now 60 sequences
- Local attribute illumination/dynamics change not used
- Original axis-aligned annotation has been replaced with new rotated bboxes

Issues

- Spatial accuracy (addressed by re-annotation)
- Synchronization (considered part of challenge)



Semi-automatic (re-)annotation

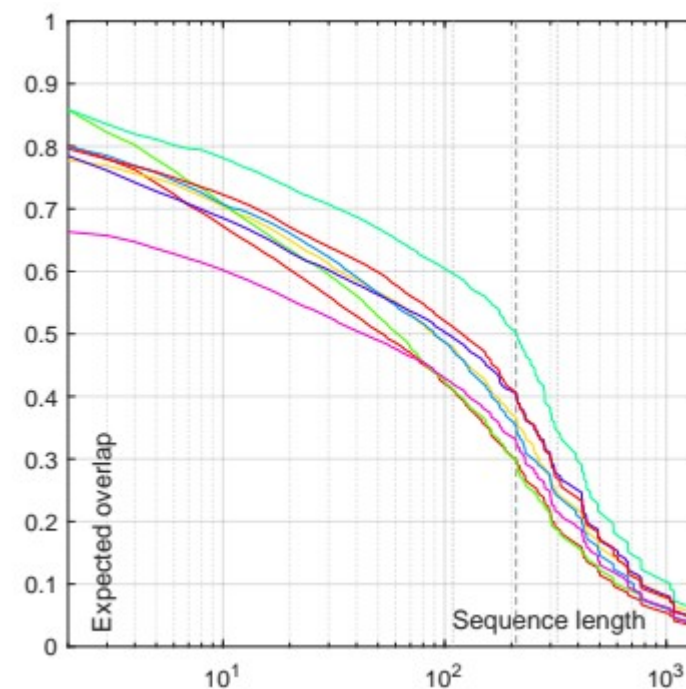
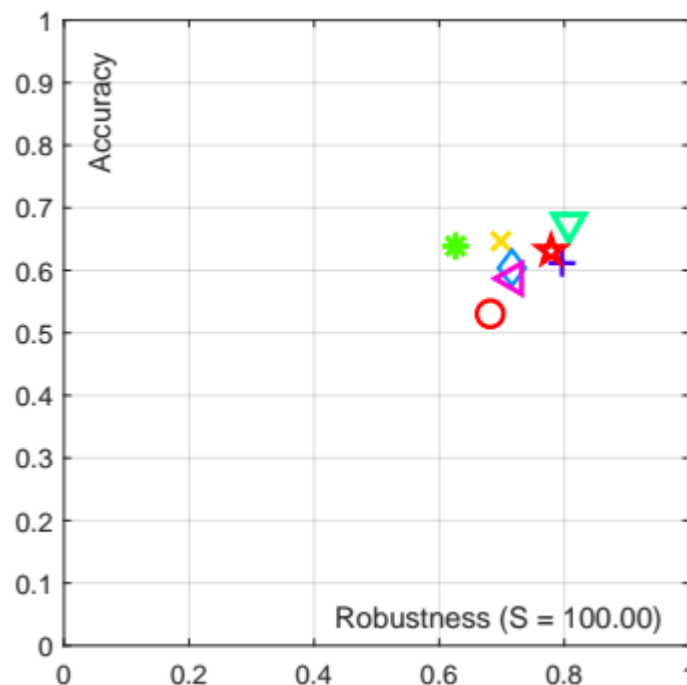
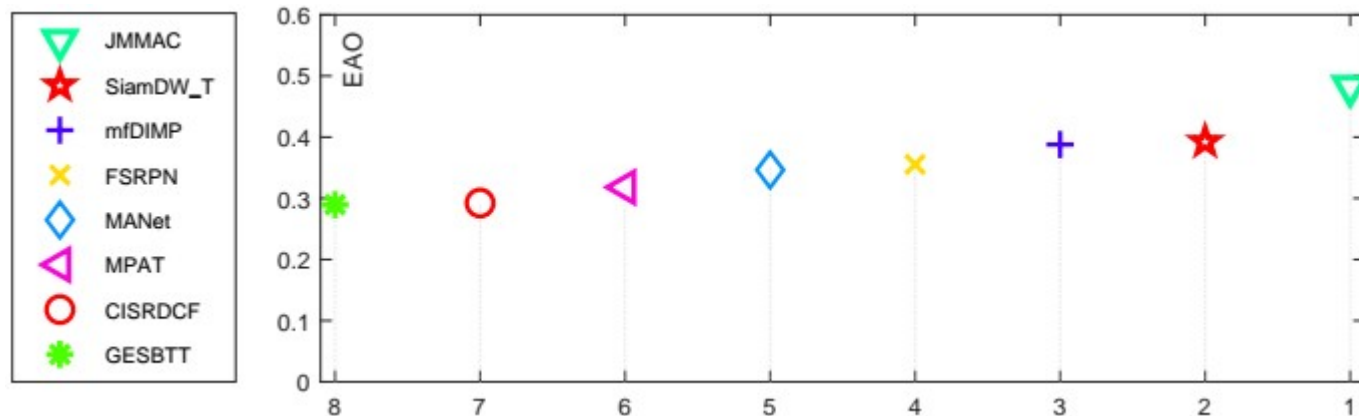
- Procedure described in paper #2:
A. Berg et al. *Semi-automatic annotation of objects in visual-thermal video*.
- Step 1: semi-automatic video segmentation based on: J. Johnander et al. *A generative appearance model for end-to-end video object segmentation*. In CVPR, 2019.
- Step 2: bounding box determination: T. Vojir and J. Matas. *Pixel-wise object segmentations for the VOT 2016 dataset*. Research Report CTU-CMP-2017-01.
- Synchronization issue: TIR is used as reference
- Spatial accuracy: EAO RGB-TIR 0.75
- Evaluation is performed in the same way as for VOT-ST 2019
- Top-ranked trackers on the public dataset run by the committee on the sequestered dataset
- Top-ranked tracker on the sequestered dataset is the winner

Submitted tracker

- 10 trackers in total, 8 unique submissions with code
 - 5 ST_1 , 3 ST_0
 - 7 uniform dynamic model, 1 random walk
 - 4 trackers based on discriminative correlation filters: CISRDCF, GESBTT, JMMAC, and mfDiMP
 - 4 trackers based on multiple CNNs: MANet, mfDiMP, MPAT, and SiamDW_T
 - 4 trackers make use of Siamese CNNs: FSRPN, mfDiMP, MPAT, and SiamDW_T
- 2 trackers apply a Kalman filter: GESBTT and JMMAC
- 1 tracker makes use of optical flow: GESBTT
- 1 tracker makes use of ransac: JMMAC
- 5 trackers use combinations of several features
- 6 trackers use CNN features
- 3 trackers use hand-crafted features
- 2 trackers use keypoints
- 2 trackers use grayscale features

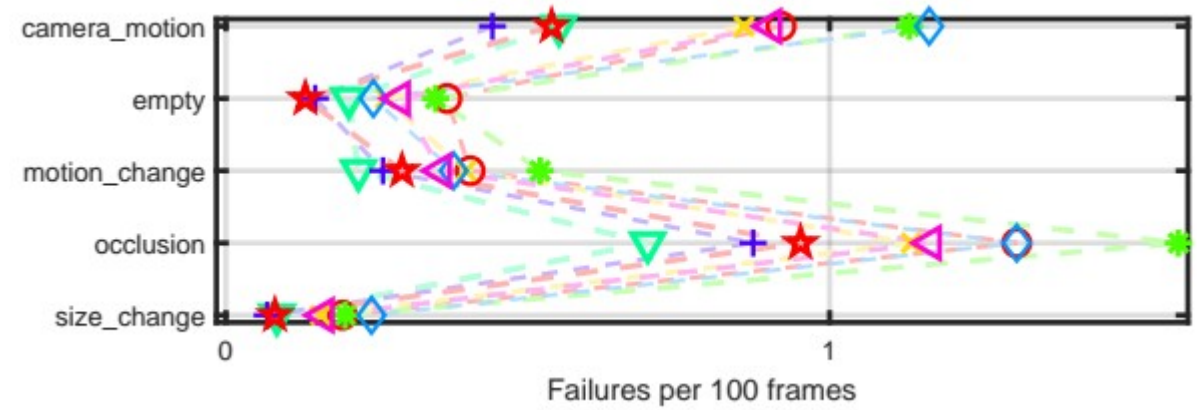
Results on public dataset

- All top-5 trackers use CNN features
- Respectively 3 out of these 5 trackers use
 - DCFs
 - Multiple CNNs
 - Siamese CNNs
 - JMMAC is working significantly better than the other two DCF-based trackers – RANSAC reason?



Further results

- EAO is stronger correlation to robustness than accuracy
- Robustness is most challenging for occlusion and camera motion
- Changed order for sequestered dataset



	Tracker	EAO	A	R
1.	mfDiMP	0.2347 ①	0.6133	0.3160 ①
2.	SiamDW_T	0.2143 ②	0.6515 ②	0.2714 ②
3.	MANet	0.2041 ③	0.5784	0.2592 ③
4.	JMMAC	0.2037	0.6337 ③	0.2441
5.	FSRPN	0.1873	0.6561 ①	0.1755

VOT-ST2019 Winners

Winners of the VOT-RGBT 2019 challenge:

mfDiMP by: L. Zhang, A. Gonzalez-Garcia, J. van de Weijer

“Multi-modal fusion for end-to-end RGB-T tracking”

(The talk up next!)



Summary

- CNN-features dominating
- The ranking changes on sequestered dataset
- Overall performance decreases on sequestered dataset
- Robustness most important
- Occlusion and camera motion largest challenges
- For the future:
 - Attract more participants
 - Measure the effect of spatial missalignment and synchronization errors?
 - Potential other changes in the evaluation system

Thanks

- The VOT2019 committee



- Everyone who participated or contributed

Matej Kristan¹, Jiří Matas², Aleš Leonardis³, Michael Felsberg⁴, Roman Pflugfelder^{5,6}, Joni-Kristian Kamarainen⁷, Luka Čehovin Zajc¹, Ondrej Drbohlav², Alan Lukežič¹, Amanda Berg^{4,8}, Abdelrahman Eldesokey⁴, Jani Käräyinen⁷, Gustavo Fernández⁵, Abel Gonzalez-Garcia¹⁸, Alireza Memarmoghdam⁵⁰, Andong Lu⁹, Anfeng He⁵², Anton Varfolomeiev³⁷, Antoni Chan¹⁷, Ardhendu Shekhar Tripathi²³, Arnold Smeulders⁴⁵, Bala Suraj Pedasingu²⁹, Bao Xin Chen⁵⁸, Baopeng Zhang¹², Baoyuan Wu⁴³, Bi Li²⁸, Bin He¹⁰, Bin Yan¹⁹, Bing Bai²⁰, Bing Li¹⁶, Bo Li⁴⁰, Byeong Hak Kim^{25,33}, Chao Ma⁴¹, Chen Fang³⁵, Chen Qian⁴⁰, Cheng Chen³⁸, Chenglong Li⁹, Chengquan Zhang¹⁰, Chi-Yi Tsai⁴², Chong Luo³⁴, Christian Micheloni⁵⁵, Chunhui Zhang¹⁶, Dacheng Tao⁵⁴, Deepak Gupta⁴⁵, Dejjia Song²⁸, Dong Wang¹⁹, Efstratios Gavves⁴⁵, Eunu Yi²⁵, Fahad Shahbaz Khan^{4,30}, Fangyi Zhang¹⁶, Fei Wang⁴⁰, Fei Zhao¹⁶, George De Ath⁴⁹, Goutam Bhat²³, Guangqi Chen⁴⁰, Guangting Wang⁵², Guoxuan Li⁴⁰, Hakan Cevikalp²¹, Hao Du³⁴, Haojie Zhao¹⁹, Hasan Saribas²², Ho Min Jung³³, Hongliang Bai¹¹, Hongyuan Yu^{16,34}, Houwen Peng³⁴, Huchuan Lu¹⁹, Hui Li³², Jiakun Li¹², Jianhua Li¹⁹, Jianlong Fu³⁴, Jie Chen⁵⁷, Jie Gao⁵⁷, Jie Zhao¹⁹, Jin Tang⁹, Jing Li²⁶, Jingjing Wu²⁷, Jingtuo Liu¹⁰, Jinqiao Wang¹⁶, Jinqing Qi¹⁹, Jinyue Zhang⁵⁷, John K. Tsotsos⁵⁸, Jong Hyuk Lee³³, Joost van de Weijer¹⁸, Josef Kittler⁵³, Jun Ha Lee³³, Junfei Zhuang¹³, Kangkai Zhang¹⁶, Kangkang Wang¹⁰, Kenan Dai¹⁹, Lei Chen⁴⁰, Lei Liu⁹, Leida Guo⁵⁹, Li Zhang⁵¹, Liang Wang¹⁶, Liangliang Wang²⁸, Lichao Zhang¹⁸, Lijun Wang¹⁹, Lijun Zhou⁴⁸, Linyu Zheng¹⁶, Litu Rout³⁹, Luc Van Gool²³, Luca Bertinetto²⁴, Martin Danelljan²³, Matteo Dunnhofer⁵⁵, Meng Ni¹⁹, Min Young Kim³³, Ming Tang¹⁶, Ming-Hsuan Yang⁴⁶, Naveen Paluru²⁹, Niki Martinel⁵⁵, Pengfei Xu²⁰, Pengfei Zhang⁵⁴, Pengkun Zheng³⁸, Pengyu Zhang¹⁹, Philip H.S. Torr⁵¹, Qi Zhang, Qiang Wang^{16,31}, Qing Guo⁴⁴, Radu Timofte²³, Rama Krishna Gorthi²⁹, Richard Everson⁴⁹, Ruize Han⁴⁴, Ruohan Zhang⁵⁷, Shan You⁴⁰, Shao-Chuan Zhao³², Shengwei Zhao¹⁶, Shihu Li¹⁰, Shikun Li¹⁶, Shiming Ge¹⁶, Shuai Bai¹³, Shuosun Guan⁵⁹, Tengfei Xing²⁰, Tianyang Xu³², Tianyu Yang¹⁷, Ting Zhang¹⁴, Tomáš Vojtíšek⁴⁷, Wei Feng⁴⁴, Weiming Hu¹⁶, Weizhao Wang³⁸, Wenjie Tang¹⁴, Wenjun Zeng³⁴, Wenyu Liu²⁸, Xi Chen⁶⁰, Xi Qiu⁵⁶, Xiang Bai²⁸, Xiao-Jun Wu³², Xiao-Jun Wu³², Xiaoyun Yang¹⁵, Xier Chen⁵⁷, Xin Li²⁶, Xing Sun⁵⁹, Xingyu Chen¹⁶, Xinmei Tian⁵², Xu Tang¹⁰, Xue-Feng Zhu³², Yan Huang¹⁶, Yanan Chen⁵⁷, Yanchao Lian⁵⁷, Yang Gu²⁰, Yang Liu³⁶, Yanjie Chen⁴⁰, Yi Zhang⁵⁹, Yinda Xu⁶⁰, Yingming Wang¹⁹, Yingping Li⁵⁷, Yu Zhou²⁸, Yuan Dong¹³, Yufei Xu⁵², Yunhua Zhang¹⁹, Yunkun Li³², Zeyu Wang, Zhao Luo¹⁶, Zhaoliang Zhang¹⁴, Zhen-Hua Feng⁵³, Zhenyu He²⁶, Zhichao Song²⁰, Zhihao Chen⁴⁴, Zhipeng Zhang¹⁶, Zhiron Wu³⁴, Zhiwei Xiong⁵², Zhongjian Huang⁵⁷, Zhu Teng¹², and Zihan Ni¹⁰

- VOT2019 sponsor:



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