

Resource-Efficient RGBD Aerial Tracking

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Overview

- **Background:** current drone-based research scenes focus on urban environments from a birds-eye view.
- **Motivation:** adding depth information can more effectively deal with target and background interference, while RGBD aerial tracking is still unexplored.
- **Contribution 1:** we **explore RGBD aerial tracking** in an overhead space, which can greatly enlarge the development of drone-based visual perception.
- **Contribution 2:** we **propose a large-scale benchmark** for RGBD aerial tracking, containing 1,000 drone-captured RGBD videos with dense annotations.
- **Contribution 3:** we **propose an efficient RGBD tracker** named EMT. Our tracker runs at over 100 fps on GPU, and 25 fps on the edge platform, benefiting from its efficient multimodal fusion and feature matching.

Motivation

- Drones have visibility and illumination limitations in color view. While RGBD aerial tracking is effective to tackle such failures.
- UAV tracking datasets record videos in urban environments from a birds-eye view, while overhead space (2 - 5 meters above the ground) is unexplored.
- New task brings challenges: complex real-world circumstances, limited onboard computational resources, real-time practical applications, and so on...

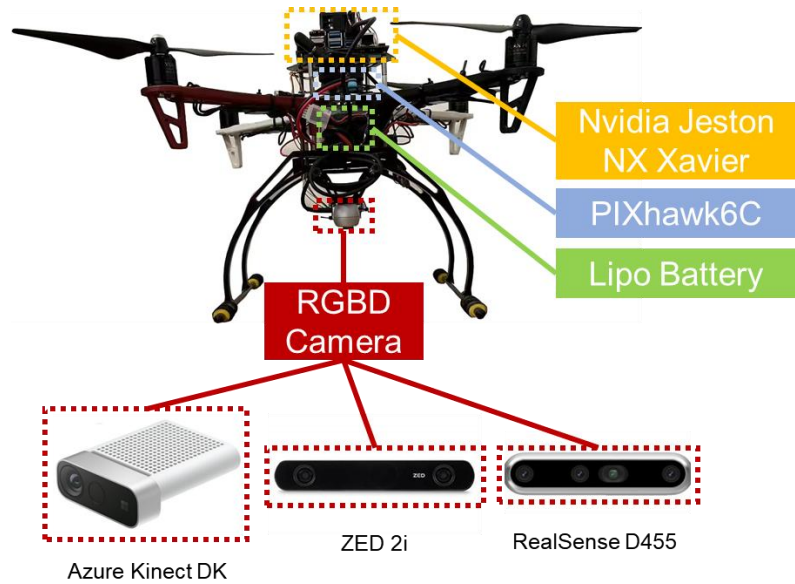
Therefore, datasets and solutions for RGBD aerial tracking are necessary.

Contributions

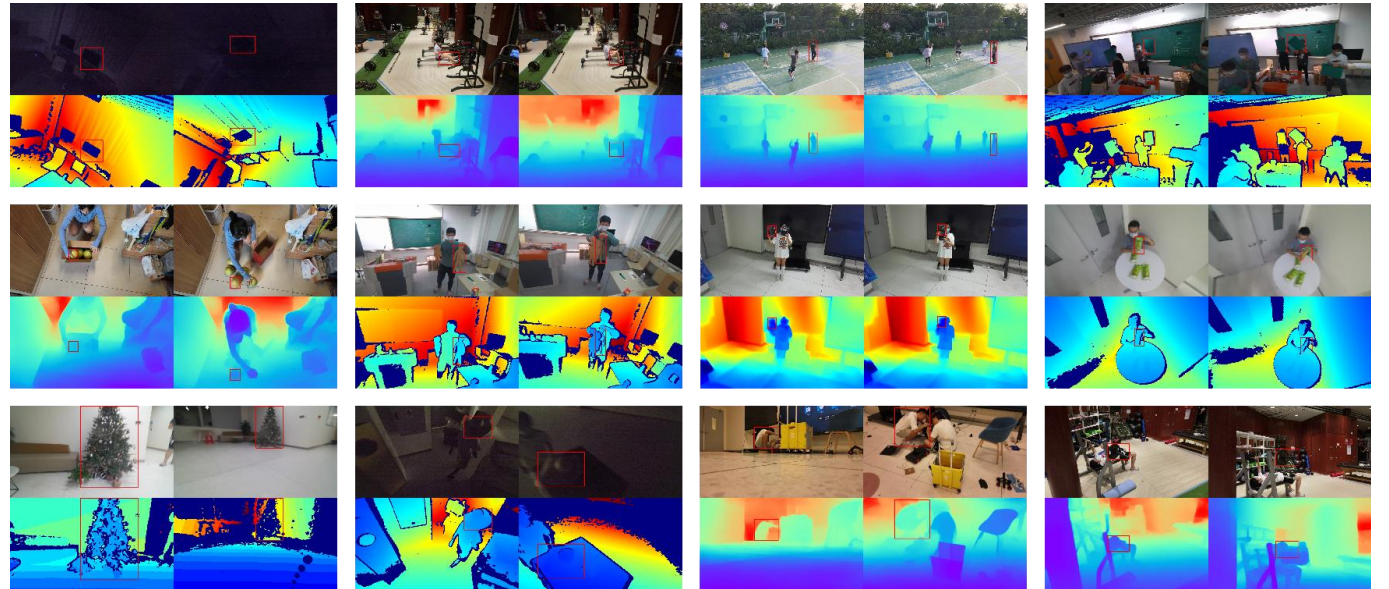
- **New Problem:** We propose a new task of RGBD aerial tracking for newly defined overhead space (2m - 5m).
- **New Benchmark:** We construct a large-scale high-diversity benchmark for RGBD aerial tracking. The advantage is that more categories can be considered than existing aerial tracking datasets.
- **New Baseline:** An efficient tracking baseline is proposed for RGBD aerial tracking, which is the first real-time tracker for efficient on-board multimodal tracking.

D²Cube Dataset

- Data Collection:



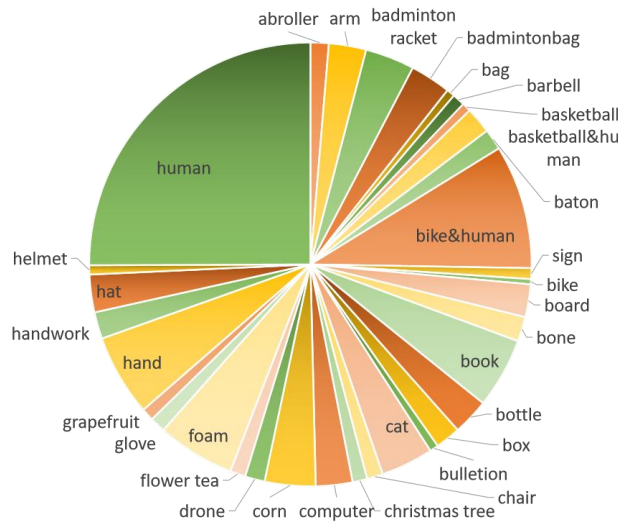
(a) Flight platform



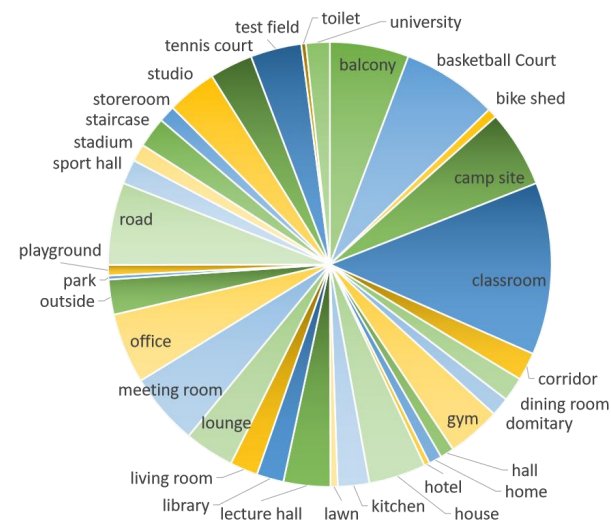
(b) Example video sequences

D²Cube Dataset

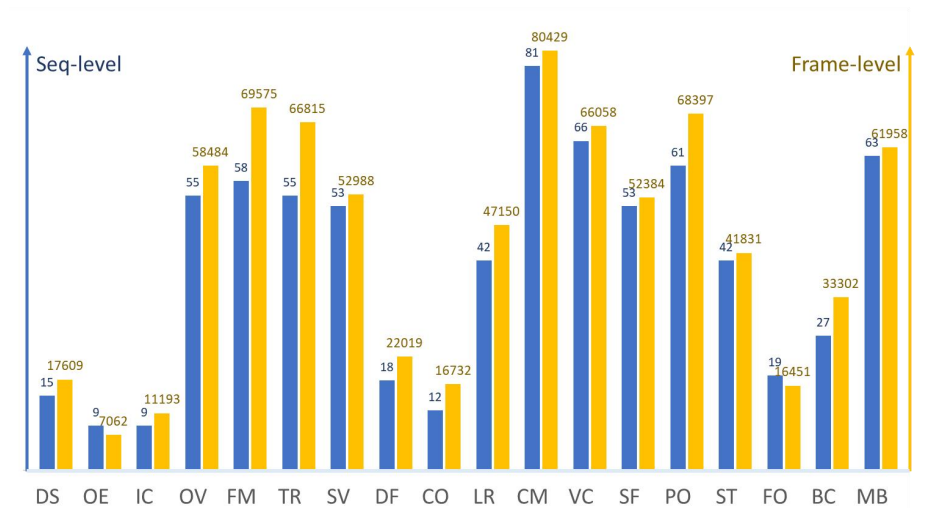
- Data Statistics:



(a) Data distribution of objects

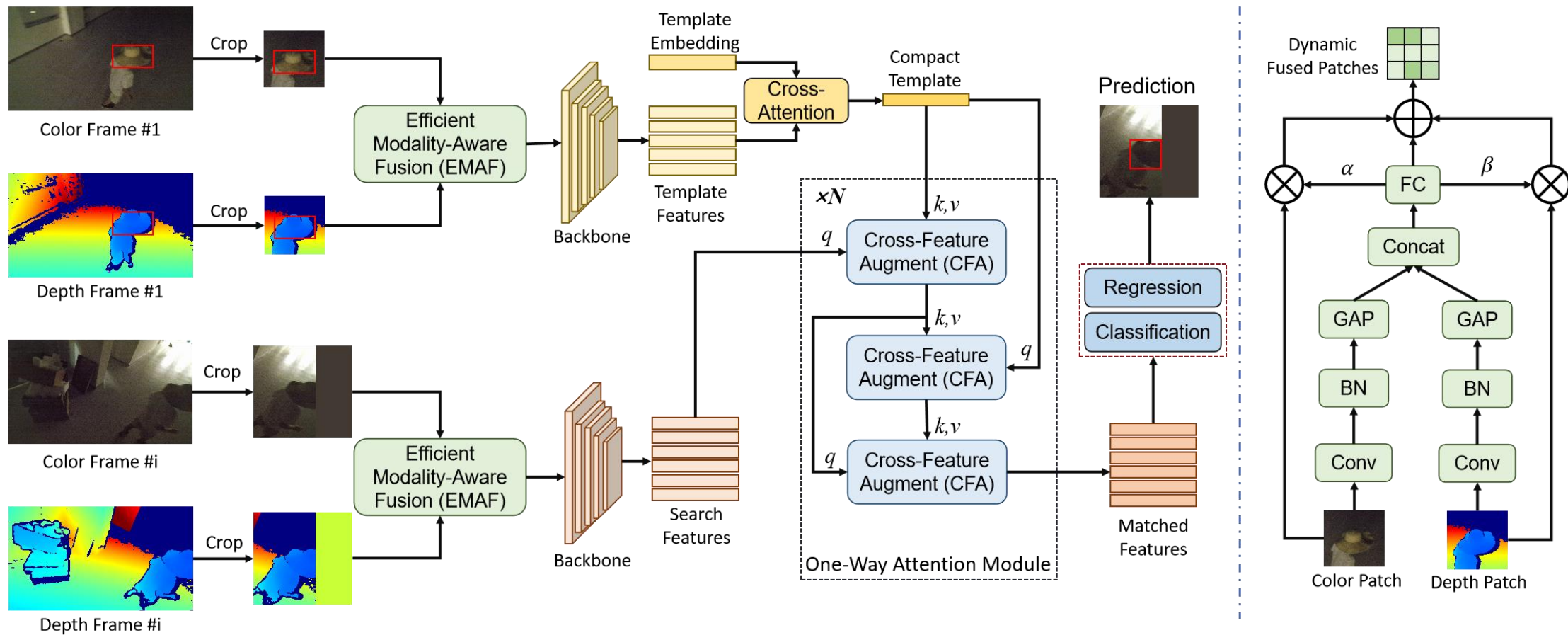


(b) Data distribution of scenarios



(c) Data distribution of attributes

Efficient Multimodal Tracker

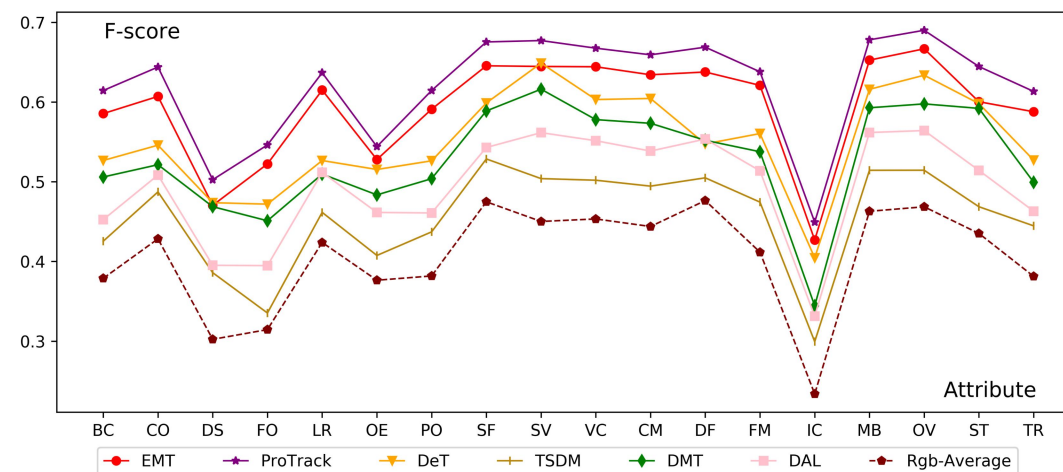


Experiments

- Experimental comparison:

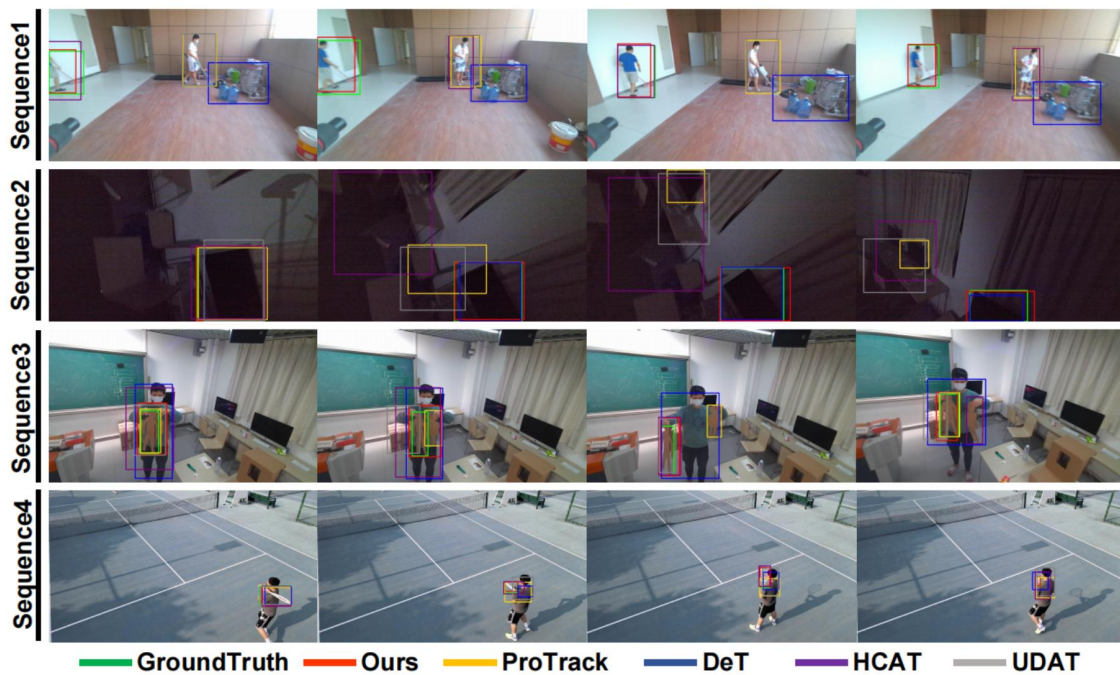
Method	Pr	Re	F-score	Speed
LightTrack [33]	0.500	0.531	0.515	119.5
HiFT [2]	0.404	0.430	0.417	66.9
TCTrack [4]	0.416	0.448	0.432	78.1
SiamAPN [9]	0.413	0.441	0.427	140.2
SiamAPN++ [3]	0.411	0.436	0.423	114.9
DaSiamRPN [44]	0.392	0.415	0.403	200.6
HCAT [5]	0.544	0.578	0.561	148.2
UDT+ [31]	0.387	0.412	0.399	50.4
SiamRPN++ [23]	0.459	0.488	0.473	83.3
UDAT-CAR [39]	0.462	0.492	0.476	33.9
EMT	0.653	0.609	0.630	120.3

Method	DAL [28]	TSDM [42]	DeT [34]	DMT [13]	ProTrack [37]	EMT
Pr	0.529	0.521	0.608	0.584	0.669	0.653
Re	0.565	0.492	0.587	0.569	0.644	0.609
F-score	0.547	0.506	0.597	0.576	0.656	0.630
MACs	15.78G	74.08G	30.57G	40.44G	82.58G	3.43G
Params	19.60M	114.59M	34.63M	38.97M	159.61M	10.05M
Speed	21.3	18.2	26.8	25.5	5.4	120.3

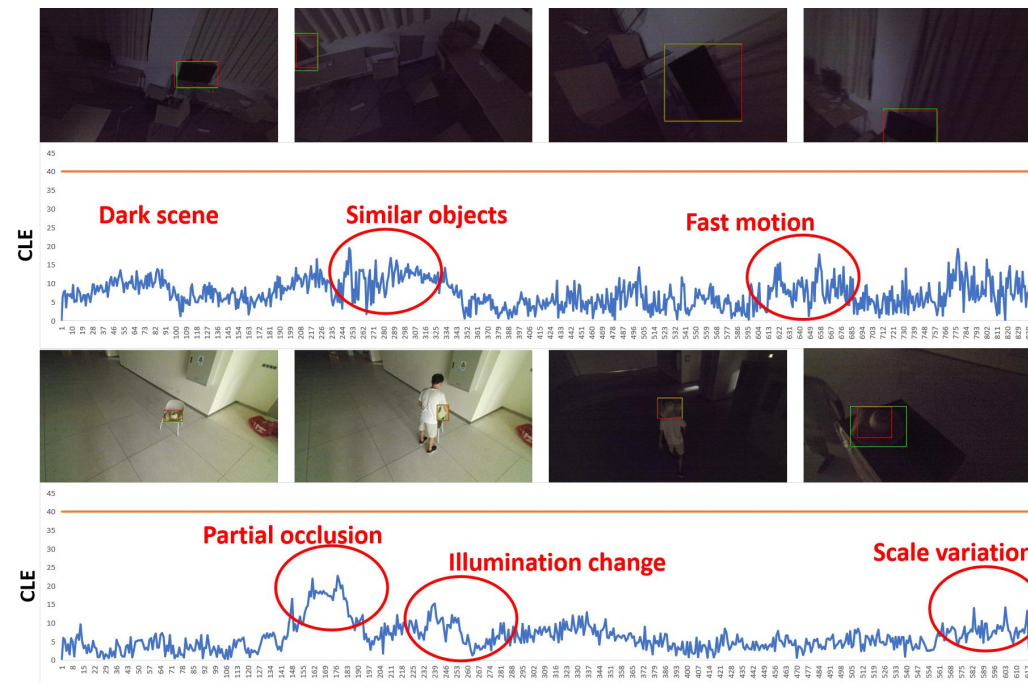


Experiments

➤ Qualitative results:



➤ On-board tests:





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Thanks for Watching!

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