MTMMC: A Large-Scale Real-World Multi-Modal Camera Tracking Benchmark

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Arxiv: https://arxiv.org/pdf/2403.20225

Project: https://sites.google.com/view/mtmmc



Multi-Target Multi-Camera Tracking

:tracking multiple objects simultaneously across different camera views.



Larger, Longer, and Diverse Tracking Benchmark

Dataset	# Cameras	# ID	# Frames	ames Camera Coverage Extra Modality		Resolution
PETS2009	8	30	1,200	outdoor	×	768 × 576
USC Campus	3	146	135,000	outdoor	×	852 × 480
Passageway	4	4	120,000	outdoor	×	320 × 240
NLPR MCT	≤ 5	≤ 235	355,500	in & outdoor	door 🗶 320	
CamNet	8	50	360,000	in & outdoor	×	640 × 480
WILDTRACK	7	N/A	66,626	outdoor	×	1920 × 1080
DukeMTMC	8	2,834	2,448,000	outdoor	×	1920 × 1080
MTA	6	2,840	2,007,360	simulated	×	1920 × 1080
MMPTRACK	≤ 6	≤ 140	2,979,900	indoor	×	640 × 320
MTMMC (Ours)	16	3,669	3,052,800	in & outdoor	(Thermal)	1920 × 1080

Multi-modal, Multi-view, Multi-object Videos



Cam 3

Cam 1



Cam 15





<image>







Cam 9

Cam 10

Cam 13

Cam 5

Cam 14

Cam 15

• Sites: <u>Campus</u>, Factory









Cam 4





Cam 5



Cam 6





Cam 8



Cam 12



Cam 16



Cam 13







Cam 14



Cam 15





Sites: Campus, <u>Factory</u>









Cam 4











Cam 11



Cam 12



Cam 16





Cam 9

Cam 13







Cam 14



Cam 15



• **Camera Topology:** <u>Indoor, Outdoor</u>, Multiple Floors, Overlapping Views



Indoor

Outdoor

• **Camera Topology:** Indoor, Outdoor, <u>Multiple Floors</u>, Overlapping Views















Camera Topology: Indoor, Outdoor, Multiple Floors, <u>Overlapping Views</u>

1st floor of Factory



Multi-modal Multi-camera Tracking

RGB and Thermal



RGBT camera with Coaxial Optical System

Thermal

RGB





1. Sub Tasks: Detection, Re-ID

more **challenging** and **generalizable**

Method	Train on	Eval on	mAP
Easter PCNN	COCO-Person	MOT17	29.8
raster Kenn	MTMMC-Person	MOT17	31.3
VOLOX	COCO-Person	MOT17	34.2
TOLOA	MTMMC-Person	MOT17	38.3

Person Detection

Method	Train on	Eval on	Rank 1	mAP
AGW	Market-1501	Market-1501	95.3	88.2
	MSMT17	MSMT17	78.3	55.6
	MTMMC-reID	MTMMC-reID	76.0	45.6
	MSMT17	Market-1501	64.3	34.2
	MTMMC-reID	Market-1501	66.5	35.4

Person Re-Identification

1. Sub Tasks: Multi-object tracking

more **challenging** and **generalizable**

Mathod		Eval on MTMMC					Eval on MOT17						
Method	MTMMC	MOT17	Misc	IDF1	MOTA	FP	FN	IDs	IDF1	MOTA	FP	FN	IDs
	✓			42.4	74.6	146678	859893	30767	48.0	40.9	2311	29084	329
JDE		\checkmark	cccpe	34.0	52.3	206112	1694301	27347	63.6	60.0	2927	18155	486
	\checkmark	\checkmark	cccpe	43.7	72.6	125770	964863	25725	70.5	65.7	2232	15759	469
QDTrack	✓			53.0	84.5	157529	475242	14542	55.3	43.6	10548	80197	449
		\checkmark		34.3	52.3	286382	1643818	21470	66.8	65.3	9324	45441	1383
	\checkmark	\checkmark		54.2	84.6	439646	439646	14106	70.0	68.6	6927	42903	1005
	✓			50.8	78.6	504642	353525	16972	55.0	45.3	17718	69870	903
CenterTrack		\checkmark		25.2	37.0	629624	1911628	40656	62.1	60.5	6678	55446	1710
Center Hack		\checkmark	CH_{pre}	27.1	45.7	518692	1662554	40746	63.7	66.2	7128	45939	1611
	\checkmark	\checkmark	CH_{pre}	51.6	80.9	415132	351162	16938	65.7	66.7	6138	46338	1407
	✓			64.8	89.7	112835	300354	7153	69.1	55.9	16896	54106	230
ByteTrack		\checkmark		40.2	56.8	506286	1283368	13585	76.8	75.0	4539	8693	224
Byte Hack		\checkmark	CH	56.9	77.7	267550	640084	7547	79.5	76.6	10128	27250	479
	\checkmark	\checkmark	CH	64.6	89.1	147385	289854	7184	78.7	76.9	8504	28302	517

2. Pre-Training: Real-world vs. Synthetic Data

Powerful pre-trained representations and synergy with synthetic data.

Method	Tra	Eval on MOT17					Method	Train on		Eval on MOT17					
	MTMMC	MOTSynth	IDF1	MOTA	FP	FN	IDs	Method	MTMMC	MOTSynth	IDF1	MOTA	FP	FN	IDs
	~		55.3	43.6	10548	80197	449		✓		68.6	66.6	9963	43074	957
QDTrack		\checkmark	54.1	43.1	11178	80178	615	QDTrack		\checkmark	70.8	68.7	9813	39882	921
	\checkmark	\checkmark	60.8	48.9	14724	67029	870		✓	\checkmark	72.0	70.2	8367	39135	750

(a) w/o finetune

(b) w/ finetune

3. Multi-modal Learning: Setups and Baselines

- **Modality Fusion**: Integrating thermal data either at input or feature level improves performance compared to using single modalities.
- **Modality Drop**: Training with both modalities but evaluating only on RGB shows effective feature transfer and model robustness.



3. Multi-modal Learning: Setups and Baselines

Enhancing Tracking Performance through Modality Fusion.

Method	Fusion	IDF1	MOTA	mAP	Method	w/o fi	ne-tune	w/ fine-tune		
RGB	×	53.0	84.5	92.8	Mediod	IDF1	MOTA	IDF1	MOTA	
T		55.0	70.2	2.0	RGB-Unimodal (baseline)	55.3	43.6	68.6	66.6	
T	×	44.5	19.2	89.9	Knowledge Distill.	55.1	43.2	70.5	68.0	
RGBT-I	Input	54.0	85.6	93.1	Multi-modal Recon.	57.9	46.2	68.3	67.6	
RGBT-F	Feature	53.9	86.0	93.5	Multi-modal Contrastive.	59.7	48.4	68.3	67.3	

(a) Modality Fusion in MTMMC

(b) Modality Drop in MTMMC \rightarrow MOT17

4. Multi-modal MTMC

Multi-Target Multi-Camera Tracking Results in MTMMC.

Method	IDF1	MOTA	FP	FN	IDs	Fusion	IDF1	ΜΟΤΑ	FP	FN	IDs
TrackTA	32.8	76.9	10604	18715	13364	RGBT-I	42.2	81.1	7823	14264	10803
H. Cluster	41.6	80.9	8012	14663	11072	RGBT-F	43.5	81.7	7301	13592	9916

(a) RGB-based MTMC

(b) Multi-modal MTMC

Conclusion

- □ The first multi-modal tracking benchmark
- The new multi-modal tracking setups
- The baselines (subtask, multi-modal tracking)