

DeeperAction Workshop @ ECCV2022

FineAction Track on Temporal Action Localization

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Outline

Part 1: Dataset Introduction

Part 2: FineAction Competition

Part 1



Dataset Introduction

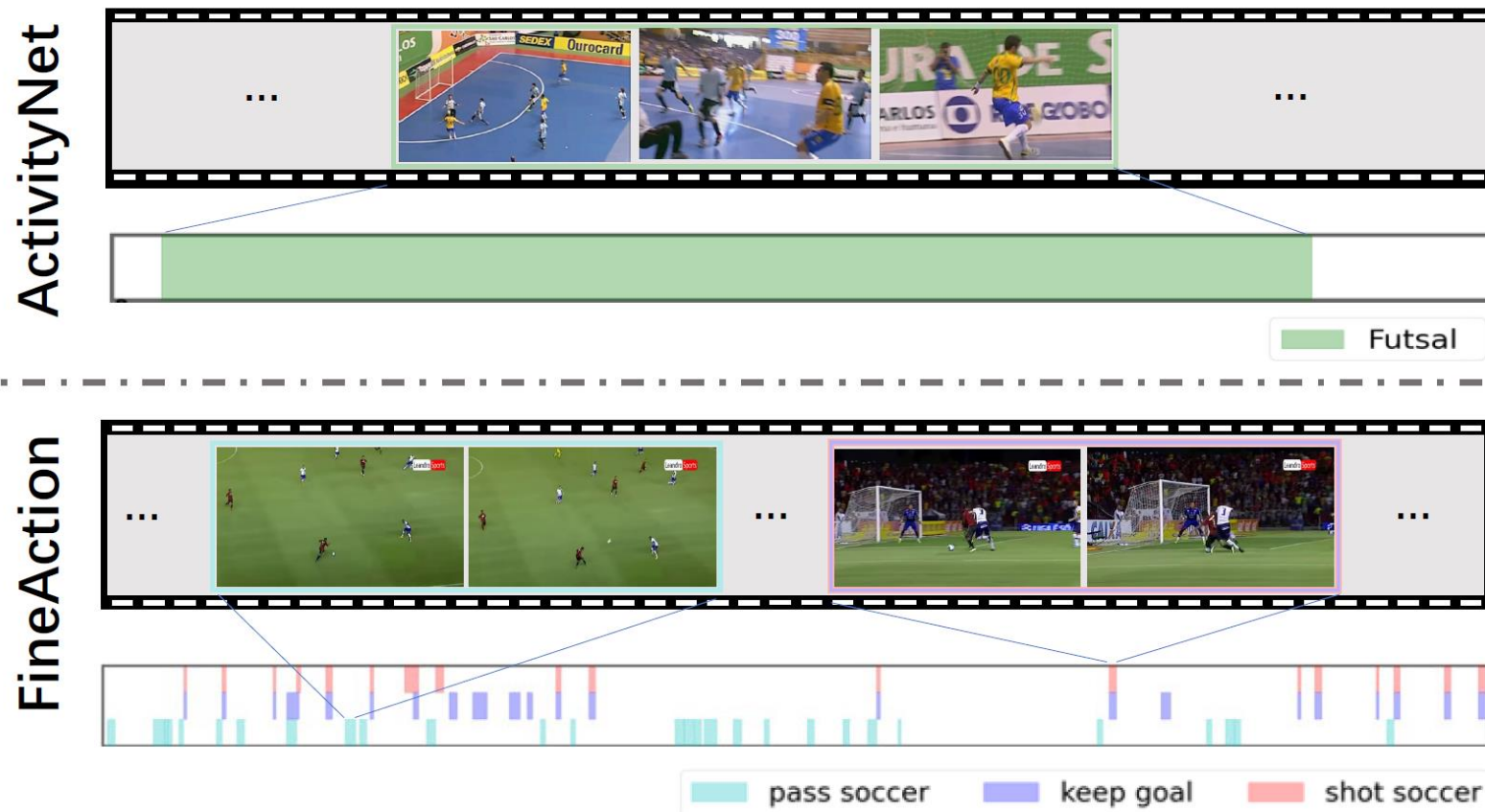
Why to do?

ActivityNet

- ❑ *Sparse annotations*
- ❑ *Coarse-level action*



- ❑ **Dense annotations**
- ❑ **Fine-Grained action**



THUMOS-14

- ❑ *Small scale*
- ❑ *Specific domain*

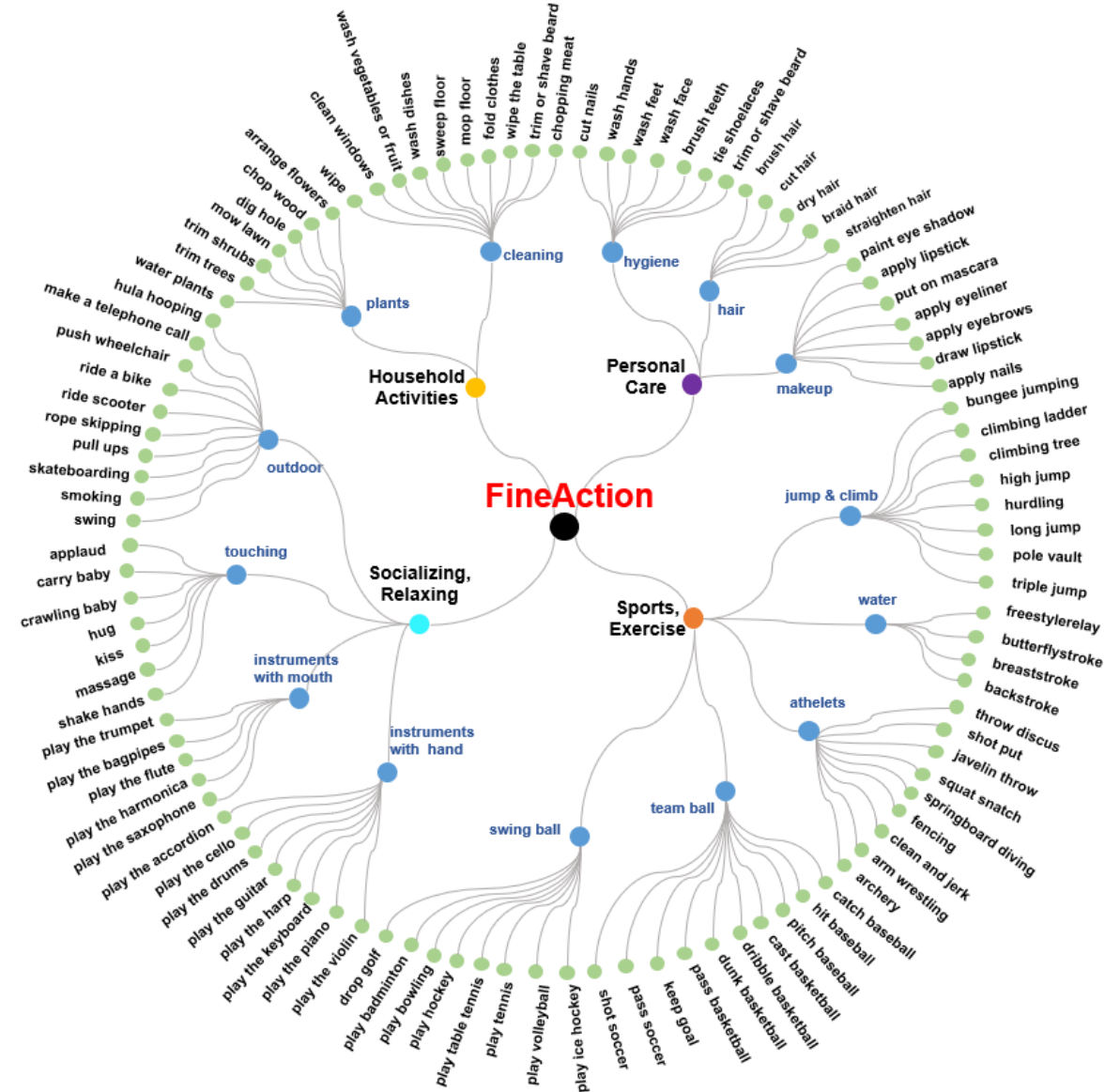


- ❑ **Large scale**
- ❑ **Diversity domain**



FineAction

- ❑ 4 top-level categories
- ❑ 14 middle-level categories
- ❑ 106 bottom-level categories

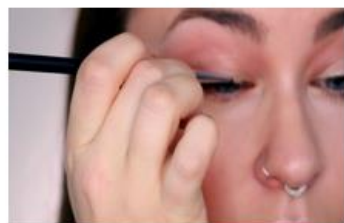
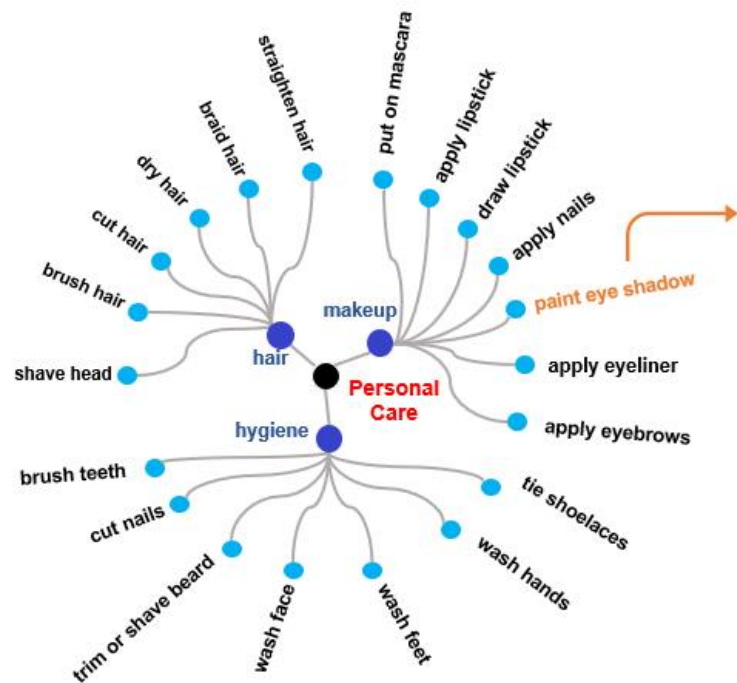


Video Selection

- ❑ Existing video datasets
 - *YouTube 8M*
 - *Kinetics 400*
 - *FCVID*
- ❑ Crawl from YouTube videos



Framework of Annotation Process



Start Definition:
Draw eye shadow with tools such as brushes.



End Definition:
Take off the tool.

Video Browsing Area

Frame Selection Area

Start: 18.66s Paint eye shadow End: 23.33s

Operation Menu Area

02.mp4
03.mp4

Choose Middle-Level label: make up
Choose Bottom-Level label: paint eye shadow

delete label save

Label Display Area Reliable

<input type="checkbox"/>	make up	Start Time: 12.15 s	<input checked="" type="checkbox"/> Reliable
<input type="checkbox"/>	paint eye shadow	End Time: 13.56 s	
<input type="checkbox"/>	make up	Start Time: 18.66 s	<input checked="" type="checkbox"/> Reliable
<input type="checkbox"/>	paint eye shadow	End Time: 23.33 s	

(a) Category Taxonomy

(b) Annotation Guidance

(c) Annotation Tool

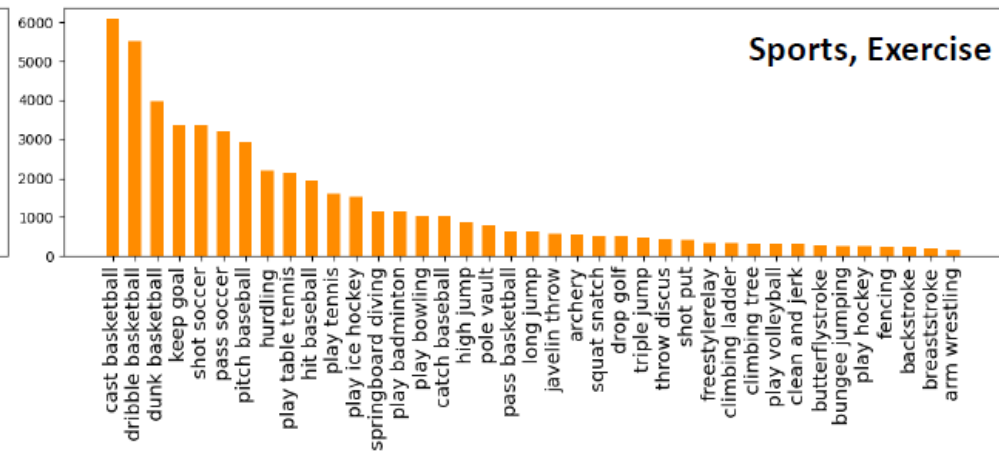
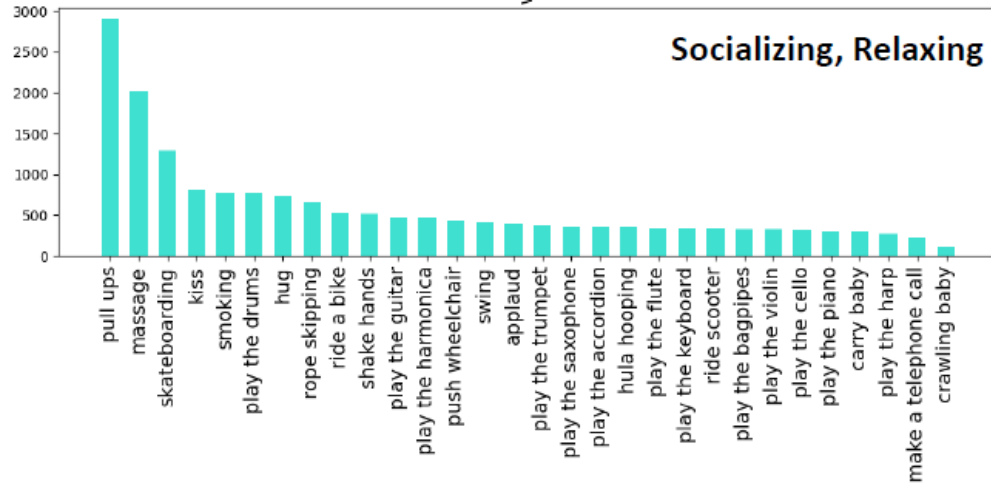
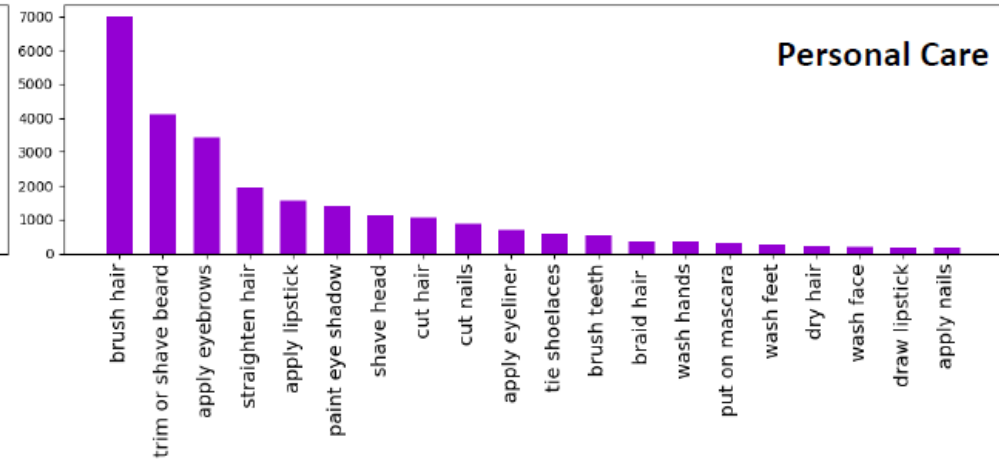
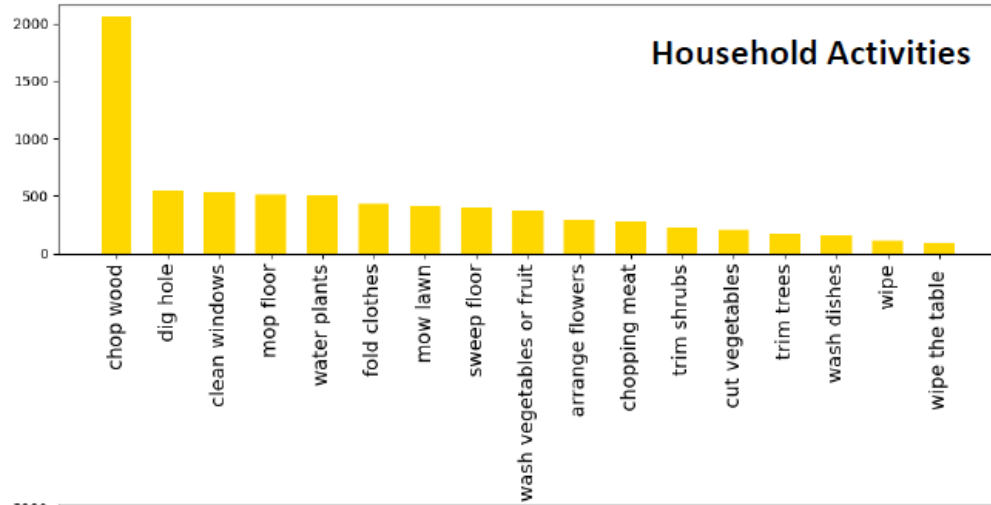
□ Large-Scale & Fine-Grained

Database	Category	M-L	Video	Instance	Overlap	Duration	Action type	Main task
MPII Cooking [24]	65	✓	45	5,609	0.1%	11.1 m	kitchens	Action Classification
EPIC-Kitchens [26]	4,025	✓	700	89,979	28.1%	3.1 s	kitchens	Action Classification
FineGym V1.0 [23]	530	✓	303	32,697	0.0%	1.7 s	sports	Action Classification
THUMOS14 [14]	20	×	413	6,316	17.5%	4.3 s	sports	Temporal Action Localization
ActivityNet [15]	200	×	19,994	23,064	0.0%	49.2 s	daily events	Temporal Action Localization
HACS Segment [16]	200	×	49,485	122,304	0.0%	33.2 s	daily events	Temporal Action Localization
FineAction	106	✓	16,732	103,324	11.5%	7.1 s	daily events	Temporal Action Localization

□ Instance duration

Database	0-2 s	2-6 s	6-15 s	>15 s	Ins / Vid
THUMOS14 [14]	2,029	2,753	1,437	99	15.29
ActivityNet [15]	900	3,253	4,426	14,485	1.15
HACS Segment [16]	8,874	29,644	31,982	51,804	2.47
FineAction	66,890	15,253	10,523	10,586	6.17

Instance distribution

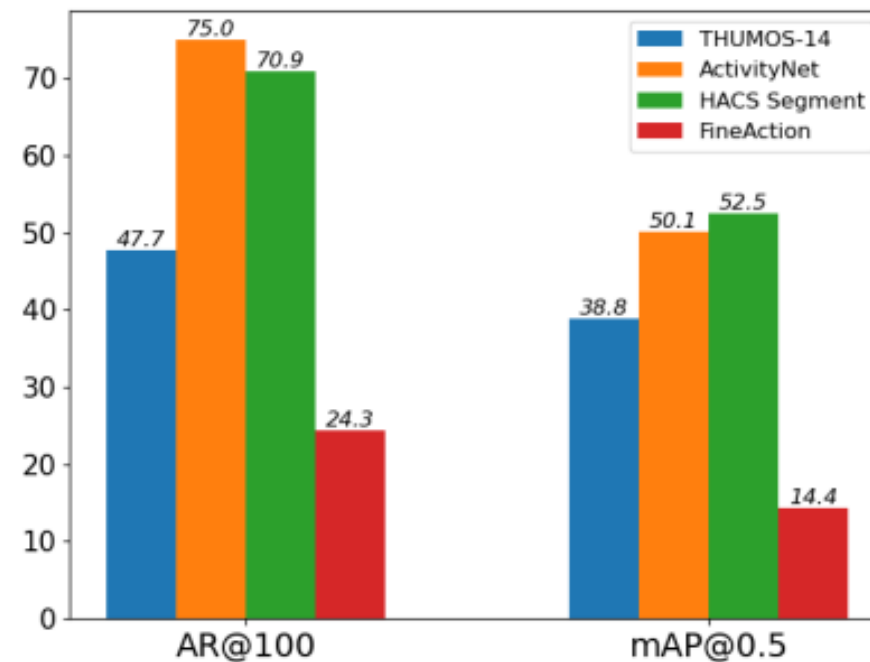


SOAT methods on FineAction

Method	Modality	Action Proposal Generation			
		AR@5	AR@10	AR@100	AUC
BMN [11]	RGB	8.62	11.20	22.74	17.49
	Flow	9.85	12.72	24.18	18.94
	RGB+Flow	9.99	12.84	24.34	19.19
DBG [13]	RGB	6.82	9.01	21.26	15.48
	Flow	8.27	10.90	23.37	17.70
	RGB+Flow	7.82	10.45	23.07	17.24
G-TAD [12]	RGB	7.96	10.45	20.86	16.06
	Flow	8.87	11.60	22.01	17.09
	RGB+Flow	9.02	11.83	23.17	17.65

Method	Modality	Temporal Action Localization			
		mAP@0.50	mAP@0.75	mAP@0.95	Avg.mAP
BMN [11]	RGB	12.56	7.49	2.62	7.86
	Flow	14.49	8.92	3.19	9.23
	RGB+Flow	14.44	8.92	3.12	9.25
DBG [13]	RGB	8.57	5.01	1.93	5.31
	Flow	11.03	6.95	2.70	7.20
	RGB+Flow	10.65	6.43	2.50	6.75
G-TAD [12]	RGB	10.88	6.52	2.19	6.87
	Flow	12.58	8.18	2.56	8.26
	RGB+Flow	13.74	8.83	3.06	9.06

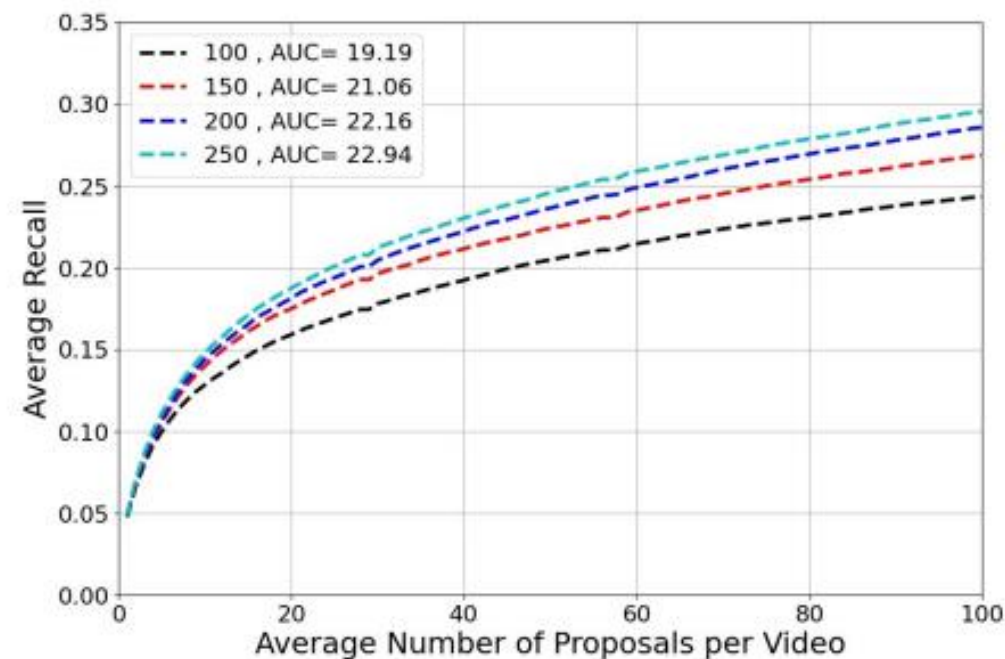
Benchmark comparison



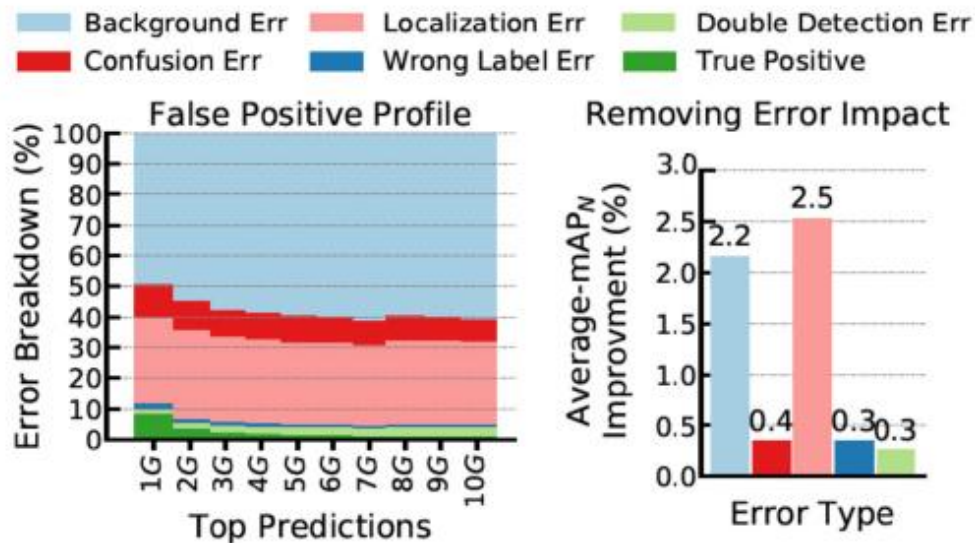
□ Cross-dataset evaluation

Dataset	Parameters Initial	AUC(%)
ActivityNet [15]	Scrath	67.29
FineAction	Scrath	19.19
ActivityNet [15]	FineAction	63.17 (↓ 4.12)
FineAction	ActivityNet [15]	18.42 (↓ 0.77)

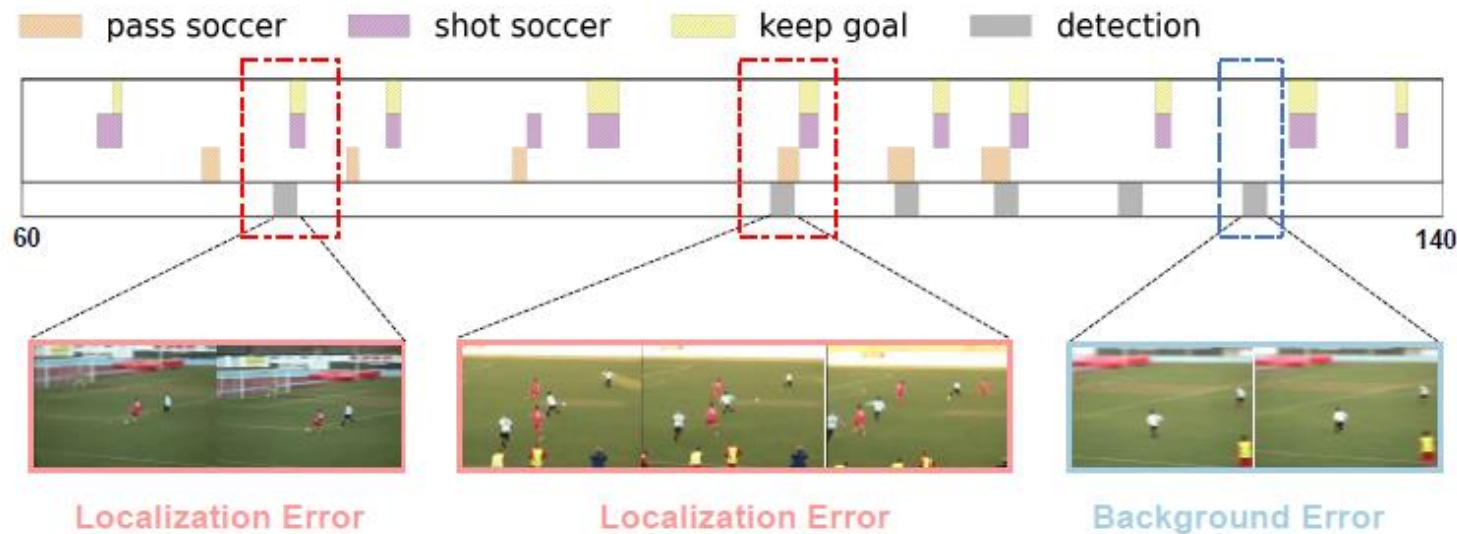
□ Different sequence length



□ Error analysis & visualization

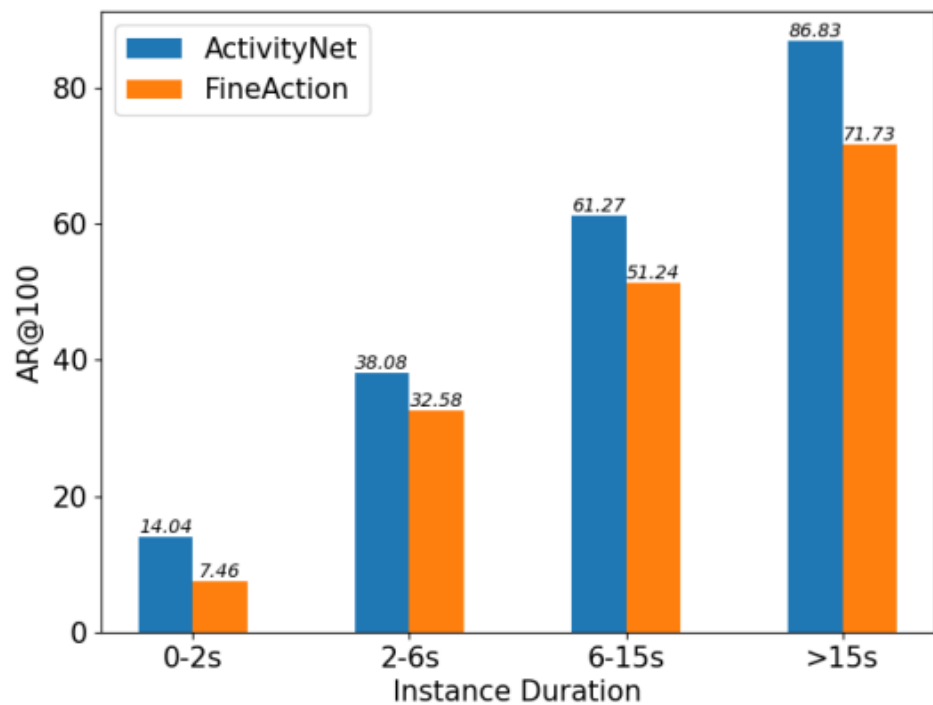


(a) Error Analysis.

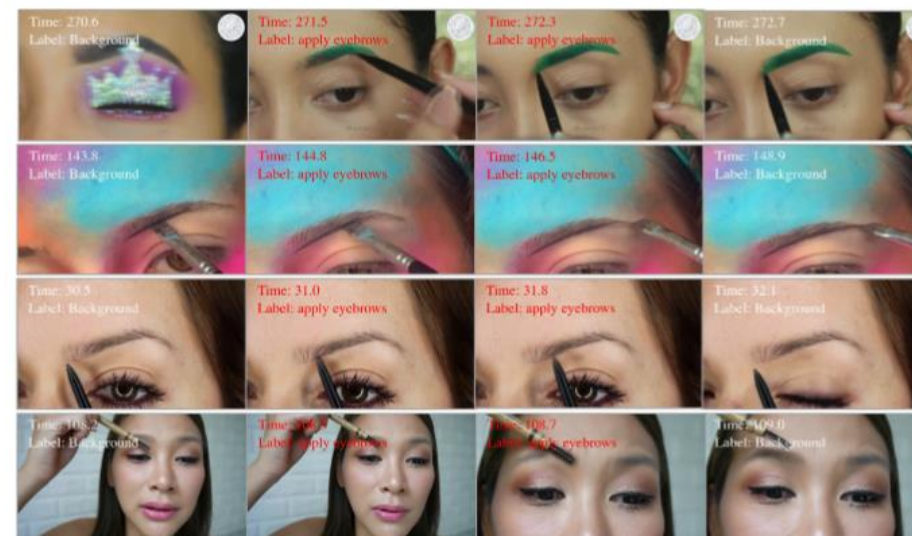


(b) Visualization of Typical Errors.

Challenging categories



FineAction	AR@100	Avg. Duration	Num
straighten hair	1.6	1.17 s	1934
apply eyebrows	1.7	1.13 s	3428
dig hole	2.2	0.83 s	544
freestyle relay	88.4	53.12 s	345
breaststroke	94.5	53.59 s	211
play the harp	96.2	69.21 s	268



Part 2



**FineAction
Competition**

<https://codalab.lisn.upsaclay.fr/competitions/4386>

Deeper
Action

ECCV DeeperAction Challenge - FineAction Track on
Temporal Action Localization

Organized by richard61

The challenge is Track 1 at ECCV DeeperAction Challenge. This track is
to detect and recognise all action instances within ...

May 01, 2022-Aug 31, 2022

140 participants

- ❑ **Validation phase:** 2022.05.01-2022.08.15
- ❑ **Testing phase:** 2022.08.16-2022.08.31

□ Goal

The challenge requires detecting and recognizing all-action instances within an untrimmed video.

□ Metric

Mean Average Precision (mAP) is a conventional evaluation metric, where Average Precision (AP) is calculated for each action category with tIoU thresholds [0.5:0.05:0.95].

- ❑ Valid Participants : 137
- ❑ Valid Teams: 8+9 (Company, University, Institute)



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Agency for
Science, Technology
and Research

SINGAPORE

❑ **Valid Submission: 58 (Val Phase) + 84 (Test Phase)**

Test Set (Mean Average Precision - mAP)

#	User	Entries	Date of Last Entry	mAP@0.50 ▲	mAP@0.55 ▲	mAP@0.60 ▲	mAP@0.65 ▲	mAP@0.70 ▲	mAP@0.75 ▲	mAP@0.80 ▲	mAP@0.85 ▲	mAP@0.90 ▲	mAP@0.95 ▲	Avg.mAP ▲
1	yptang	24	08/31/22	35.95 (1)	33.48 (1)	30.64 (1)	27.89 (1)	24.92 (1)	21.78 (1)	18.25 (1)	14.25 (1)	9.00 (1)	3.52 (3)	21.97 (1)
2	linxt	12	08/25/22	28.55 (2)	26.14 (2)	23.78 (2)	21.35 (2)	18.67 (2)	15.91 (2)	12.85 (3)	9.54 (4)	5.93 (4)	2.16 (4)	16.49 (2)
3	Strangelove	14	08/31/22	25.50 (3)	23.58 (3)	21.66 (3)	19.46 (3)	17.59 (3)	15.70 (3)	13.74 (2)	11.35 (2)	8.36 (2)	4.25 (2)	16.12 (3)

A One-Stage Method for FineAction Localization from Multiple Views

Yepeng Tang^{1,2†}, Weining Wang³, Chunjie Zhang^{1,2*}, Jie Jiang^{3,4},
Weitao Yuan^{3,4}, Sihao Chen^{3,4}, Jing Liu^{3,4}, Yao Zhao^{1,2}

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Technical Report for FineAction 2022 - Temporal Action Localization

Yaokun Zhong , Xiaotong Lin
Sun Yat-sen University, Guangzhou, China



中山大學
SUN YAT-SEN UNIVERSITY

Technical Report for FineAction 2022 - Temporal Action Localization

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Over & Thank you !

Homepage: <https://deeperaction.github.io/datasets/fineaction>

GitHub: <https://github.com/Richard-61/FineAction>

IEEE TRANSACTIONS ON IMAGE PROCESSING

FineAction: A Fine-Grained Video Dataset for Temporal Action Localization

Yi Liu, Limin Wang *Member, IEEE*, Yali Wang, Xiao Ma, Yu Qiao *Senior Member, IEEE*