ICCV DeeperAction Challenge 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?

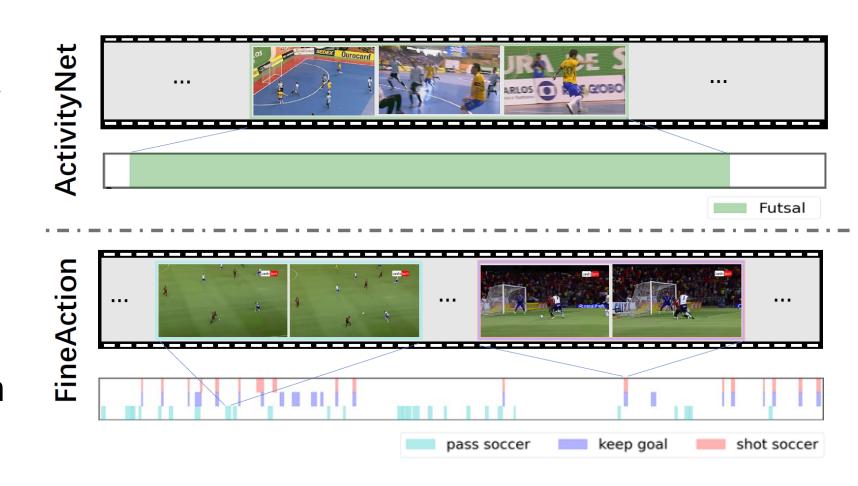


ActivityNet

- ☐ Sparse annotations
- □ Coarse-level action



- ☐ Dense annotations
- ☐ Fine-Grained action



Why to do?

THUMOS-14

- ☐ Small scale
- ☐ Specific domain



- ☐ Large scale
- □ Diversity domain



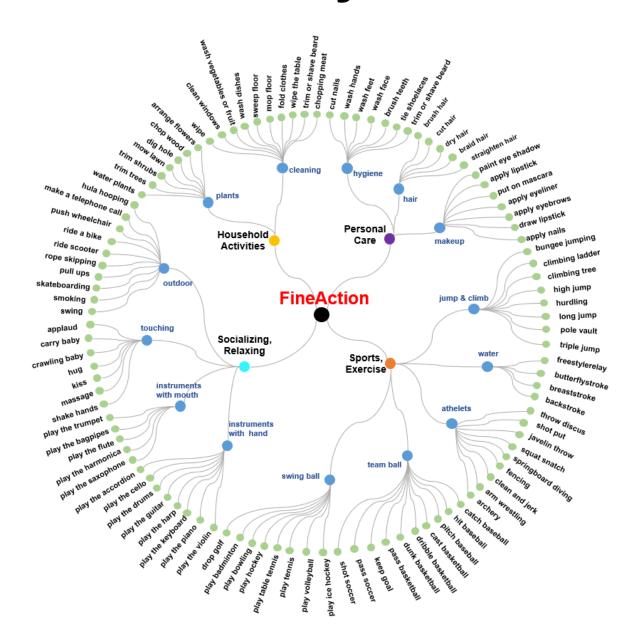


FineAction: New Taxonomy



FineAction

- □ 4 top-level categories
- ☐ 14 middle-level categories
- □ 106 bottom-level categories





FineAction: Video Collection



Video Select

- □ Exisiting video datasets
- YouTube 8M
- Kinetics 400
- FCVID
- □ Crawl from YouTube videos







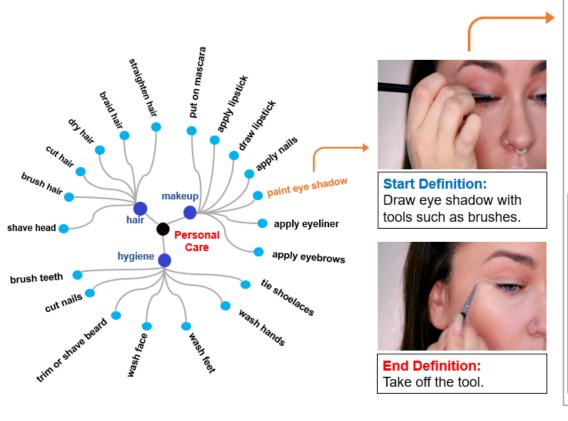


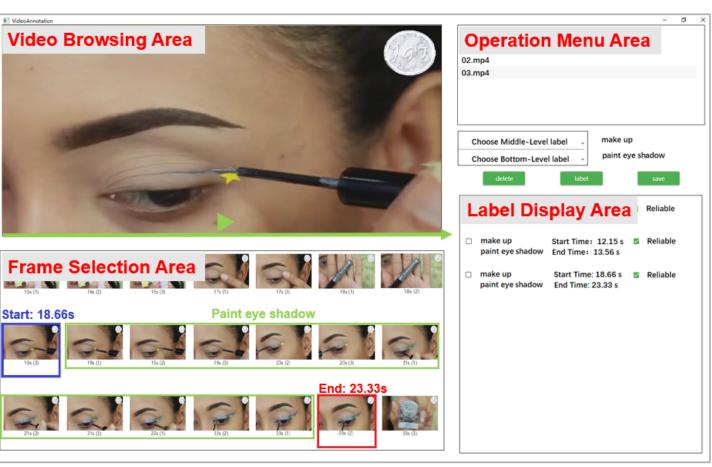


FineAction: Annotation



Framework of Annotation Process





(a) Category Taxonomy

(b) Annotation Guidance

(c) Annotation Tool



FineAction: Statistics



☐ Large-Scale & Fine-Grained

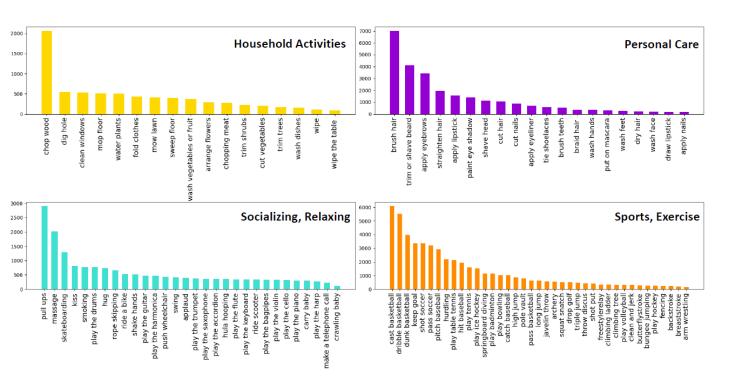
Database	Category	Video	Instance	Overlap	Duration	Action type
MPII Cooking	65	45	5,609	0.1%	11.1 m	kitchens
EPIC-Kitchens	4,025	700	89,979	28.1%	3.1 s	KICHCHS
FineGym V1.0	530	303	32,697	0.0%	1.7 s	enorte
THUMOS14	20	413	6,316	17.5%	4.3 s	sports
ActivityNet	200	19,994	23,064	0.0%	49.2 s	
HACS Segment	200	49,485	122,304	0.0%	33.2 s	daily events
FineAction (Ours)	106	16,732	103,324	11.5%	7.1 s	



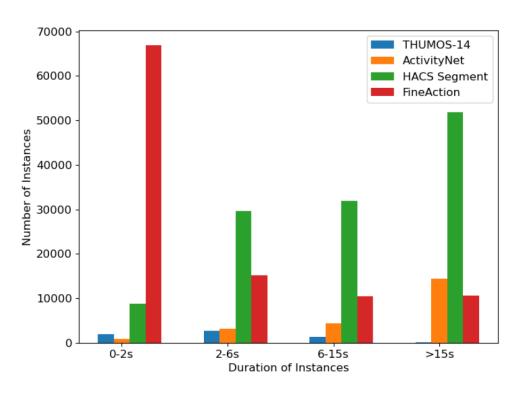
FineAction: Statistics



□ Instance distribution



☐ Instance duration





FineAction: Experiment



□ SOAT methods on FineAction

Mothod	Modelity	Act	tion Propos	sal Generati	on	Temporal Action Localization				
Method	Modality	AR@5	AR@10	AR@100	AUC	mAP@0.50	mAP@0.75	mAP@0.95	Avg.mAP	
	RGB	8.62	11.20	22.74	17.49	12.56	7.49	2.62	7.86	
BMN	Flow	9.85	12.72	24.18	18.94	14.49	8.92	3.19	9.23	
	RGB+Flow	9.99	12.84	24.34	19.19	14.44	8.92	3.12	9.25	
	RGB	6.82	9.01	21.26	15.48	8.57	5.01	1.93	5.31	
DBG	Flow	8.27	10.90	23.37	17.70	11.03	6.95	2.70	7.20	
	RGB+Flow	7.82	10.45	23.07	17.24	10.65	6.43	2.50	6.75	
G-TAD	RGB	7.96	10.45	20.86	16.06	10.88	6.52	2.19	6.87	
	Flow	8.87	11.60	22.01	17.09	12.58	8.18	2.56	8.26	
	RGB+Flow	9.02	11.83	22.60	17.65	13.74	8.83	3.06	9.06	

Dataset	AR@100	Classifier	mAP@0.5
THUMOS14	47.7	UNet	38.8
ActivityNet-1.3	75.0	TSN	50.1
HACS Segment	70.9	SlowFast	52.5
FineAction	24.3	TSN	14.4



FineAction: Experiment



□ Different feature length

Length	AR@5	AR@10	AR@100	AUC
100	9.99	12.84	24.34	19.19
150	10.56	13.99	26.89	21.06
200	10.69	14.41	28.57	22.16
250	11.11	14.81	29.56	22.94

□ Different duration instances

Instance Duration	0-2 s	2-5 s	5-10 s	>10 s
AR@100	7.46	32.58	51.24	71.73

☐ Top-3 worst & Top-3 best

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



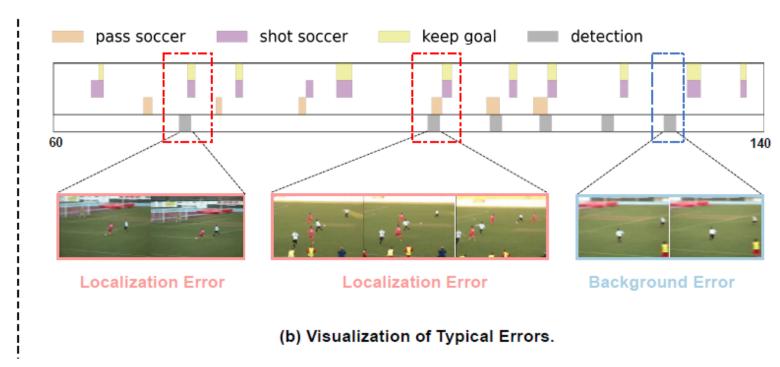
FineAction: Analysis



□ Error analysis

Background Err Localization Err Double Detection Err Confusion Err Wrong Label Err True Positive Removing Error Impact Error Breakdown (%) False Positive Profile Average-mAP_N Improvment (%) 2.5 1.0 0.30.3 0.5 0.4 16 26 36 36 36 46 66 66 96 96 106 Error Type **Top Predictions** (a) Error Analysis.

□ Error visualization



Part 2

FineAction Competition



Competition: FineAction Track



https://competitions.codalab.org/competitions/32363



ICCV DeeperAction Challenge - FineAction Track on Temporal Action Localization

Organized by yiliu

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

Jun 01, 2021-Sep 12, 2021 232 participants

- **□ Validation phase:** 2021.06.01-2021.09.01
- **☐ Testing phase:** 2021.09.01-2021.09.12



FineAction Track: Evaluation



☐ 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 tloU thresholds [0.5:0.05:0.95].

FineAction Track: Statistics



- **□** Valid Participants : 228
- □ Valid Teams: 18 + 14 (Company, University, Institute)







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FineAction Track: Results



□ Valid Submission: 95 (Val Phase) + 42 (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	Dahua_001	10	09/10/21	35.044	32.999	30.599	28.390	26.055	23.348	20.494	17.277	12.707	6.607	23.352 (1)
2	DeepakSridhar	1	09/03/21	20.244	18.854	17.581	16.055	14.305	12.697	11.043	9.059	6.774	3.615	13.023 (2)
3	foolwood	2	09/12/21	19.446	18.050	16.659	15.254	13.845	12.048	10.470	8.775	6.628	4.002	12.518 (3)





1st Place Winner



Learning Efficient Feature Representation for Temporal Action Localization

Chenglu Wu*, Xuefeng Yang*, Fuzhi Duan, Yanxun Yu, Yayun Wang, Jun Yin Zhejiang Dahua Technology Co., Ltd.
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2nd Place Winner



An Empirical Study of Feature Representation for Actionness based Temporal Action Localization

Qiang Wang and Rongliang Cheng Institute of Automation, Chinese Academy of Sciences.

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Over & Thank you!



Homepage: https://deeperaction.github.io/fineaction/