

ActEV19: Activities in Extended Video t (Summary Results)

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TRECVID 2019 Workshop

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Disclaimer

Certain commercial equipment, instruments, software, or materials are identified in this paper to specify the experimental procedure adequately. Such identification is not intended to imply recommendation or endorsement by NIST, nor necessarily the best available for the purpose.

<u>The views and conclusions contained herein are</u> <u>those of the authors</u> and should not be interpreted as necessarily representing the official policies or endorsements, either expressed or implied, of IARPA, NIST, or the U.S. Government.



Outline

- ActEV Overview
- TRECVID ActEV19 Evaluation
- ActEV19 Tasks and Measures
- ActEV19 Dataset
- ActEV19 Results and Analyses
- Next Steps



ActEV Overview





What is ActEV?



vehicle_turning_left-00



vehicle_turning_left-04



Closing-01



Closing-02



Closing-00



Loading-00



Loading-01



Loading-02





What is ActEV's Goal?

- To advance video analytics technology that can automatically <u>detect a target activity</u> and <u>identify and</u> <u>track objects</u> associated with the activity.
- A series of challenges are also designed for:
 - Activity detection in a multi-camera environment
 - Temporal (and spatio-temporal) localization of the activity for reasoning



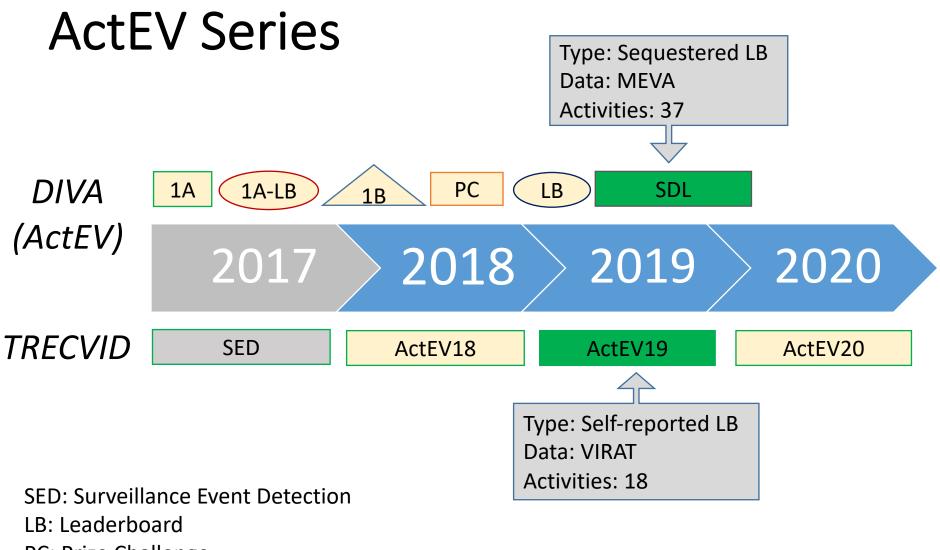
NIST, IARPA, and Kitware

- NIST developed the ActEV evaluation series to support the metrology needs of the Intelligence Advanced Research Projects Activity (IARPA) Deep Intermodal Video Analytics (DIVA) Program
- The ActEV's datasets were collected and annotated by Kitware, Inc.









PC: Prize Challenge

SDL: Sequestered Data Leaderboard

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TRECVID ActEV19 Evaluation





Evaluation Framework

- Target applications
 - <u>Retrospective analysis of archives</u> (e.g., forensic analytics)
 - Real-time analysis of live video streams (e.g., alerting and monitoring)
- Evaluation Type
 - Self-reported (& take-home) evalulation
 - TRECVID ActEV19
 - Independent (& sequestered) evalulation
 - DIVA ActEV SDL



ActEV19 Tasks and Measures





Evaluation Tasks (AD)

- "Activity" definition for this evaluation
 - One or more people performing a specified movement, or interacting with an object or group of objects (including driving)
- Activity Detection (AD) task
 - Given a target activity, a system automatically 1) detects its presence and then temporally localizes all instances of the activity in video sequences
 - The temporal overlap must fall within a minimal requirement
 - The system output includes:
 - Start and end frames indicating the temporal location of the target activity
 - A presence confidence score that indicates how likely the activity occurred



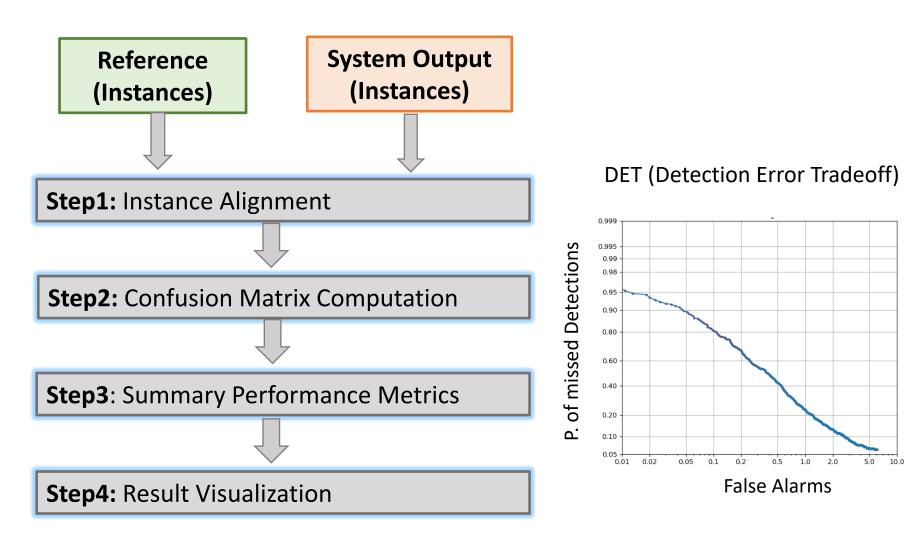


Past Evaluation Tasks (AOD and AODT)

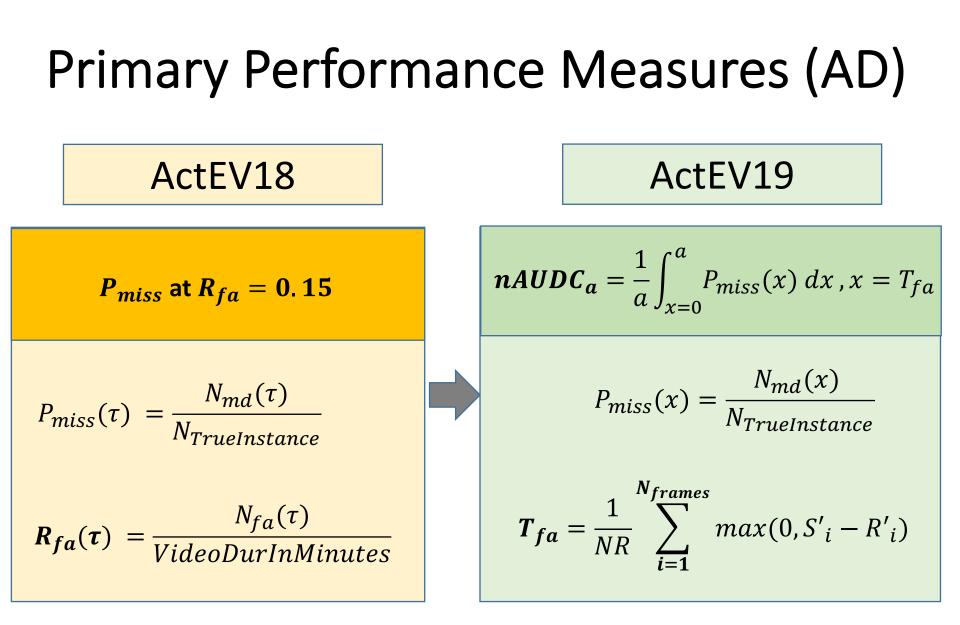
- Activity and Object Detection (AOD)
 - A system not only 1) detects/localizes the target activity, but also 2) detects the presence of required objects and spatially localizes the objects that are associated with the activity
- Activity Object Detection/Tracking (AODT)
 - A system 1) correctly detects/localizes the target activity, 2) correctly detects/localizes the required objects in that activity, and 3) correctly tracks those objects over time.
- The AOD and AODT tasks are <u>NOT addressed</u> in ActEV19 evaluations



Performance Metric Calculation







nAUDC (normalized partial Area Under the DET Curve)



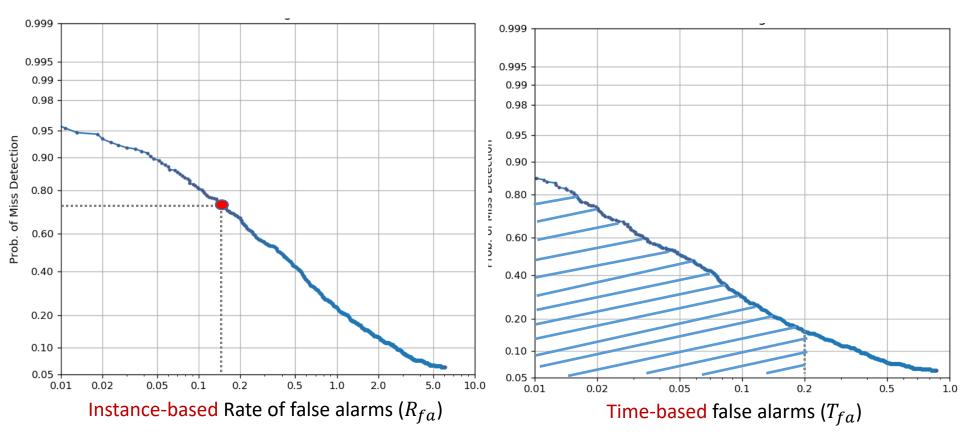
Performance Measures (AD)

ActEV18



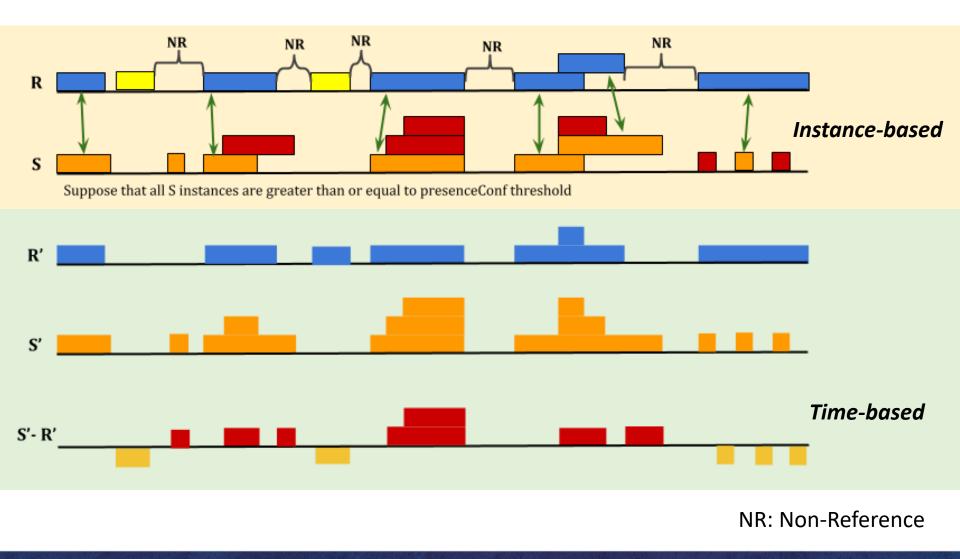
ActEV19

 $nAUDC_a$, a = 0.2





Instance vs Time based False Alarms



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ActEV19 Dataset





Activities and Number of Instances

	ActEV	18 (V1)	ActEV19 (V1V2)				
Activity Type	Train	Validation	Train	Validation			
Closing	126	132	126	132			
Closing_trunk	31	21	31	21			
Entering	70	71	70	71			
Exiting	72	65	72	65			
Loading	38	37	38	37			
Open_Trunk	35	22	35	22			
Opening	125	127	125	127			
Transport_HeavyCarry	45	31	45	31			
Unloading	44	32	44	32			
Vehicle_turning_left	152	133	152	133			
Vehicle_turning_right	165	137	165	137			
Vehicle_u_turn	13	8	13	8			
Interacts	88	101	x	х			
Pull	21	22	21	22			
Riding	21	22	21	22			
Talking	67	41	67	41			
Activity_carrying	364	237	364	237			
Specialized_talking_phone	16	17	16	17			
Specialized_texting_phone	20	5	20	5			

Due to ongoing evaluations, the test sets are not included in the table



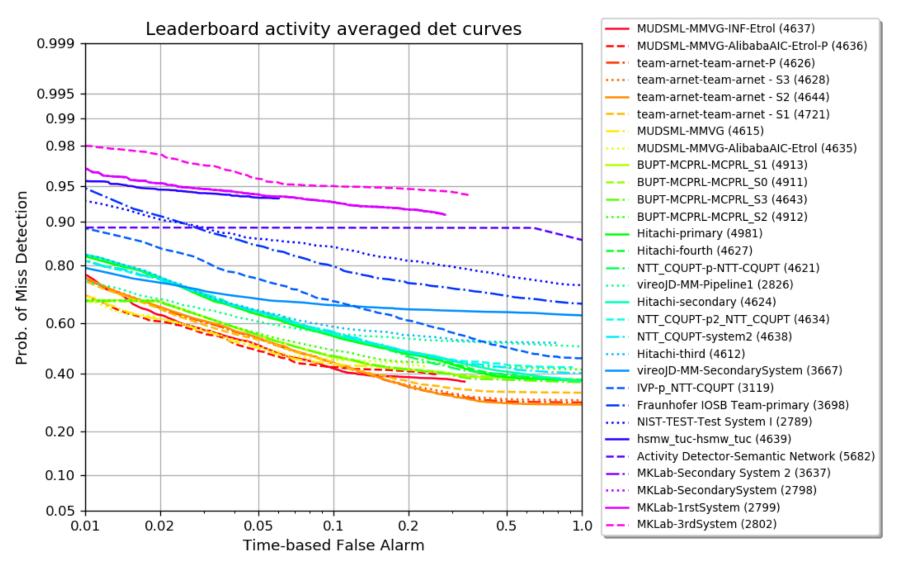


ActEV19 Results and Analyses





As of 11/13/2019



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ActEV19 Participants

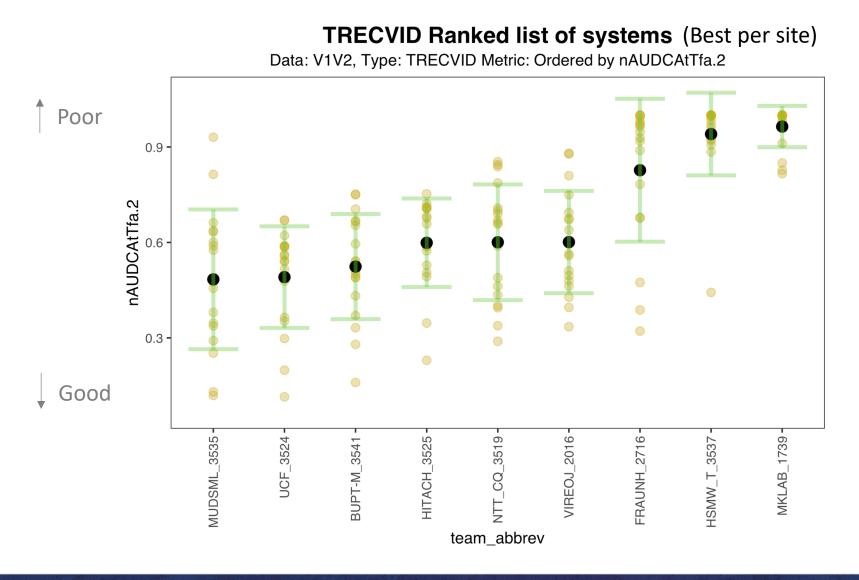
 256 submissions (as of 11/1/2019) from 9 teams from 6 countries (best system result per site)

Team	Organization	nAUDC
BUPT-MCPRL	Beijing University of Posts and Telecommunications, China	0.524
Fraunhofer IOSB	Fraunhofer Institute, Germany	0.827
HSMW_TUC	University of Applied Sciences Mittweida and Chemnitz University of Technology, Germany	0.941
MKLab (ITI_CERTH)	Information Technologies Institute, Greece	0.964
MUDSML	Monash University, Australia and Carnegie Mellon University, USA	0.484
NII_Hitachi_UIT	National Institute of Informatics, Japan Hitachi, Ltd., Japan University of Information Technology, Vietnam	0.599
NTT_CQUPT	NTT company & Chongqing University of Posts and Telecommunications, China	0.601
UCF	University of Central Florida, USA	0.491
vireoJD-MM	City University of Hong Kong and JD AI Research, China	0.601





Performance Ranking (AD)

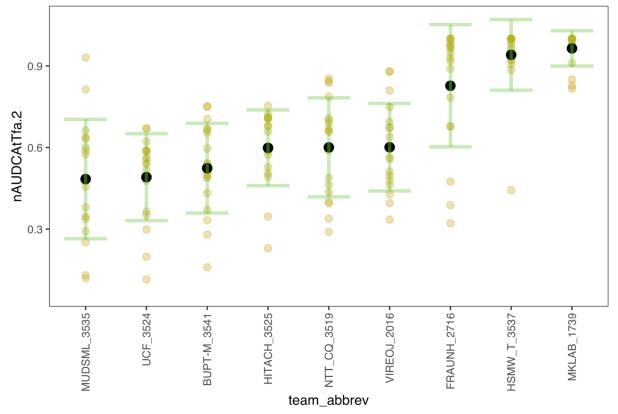


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TRECVID Ranked list of systems

Data: V1V2, Type: TRECVID Metric: Ordered by nAUDCAtTfa.2

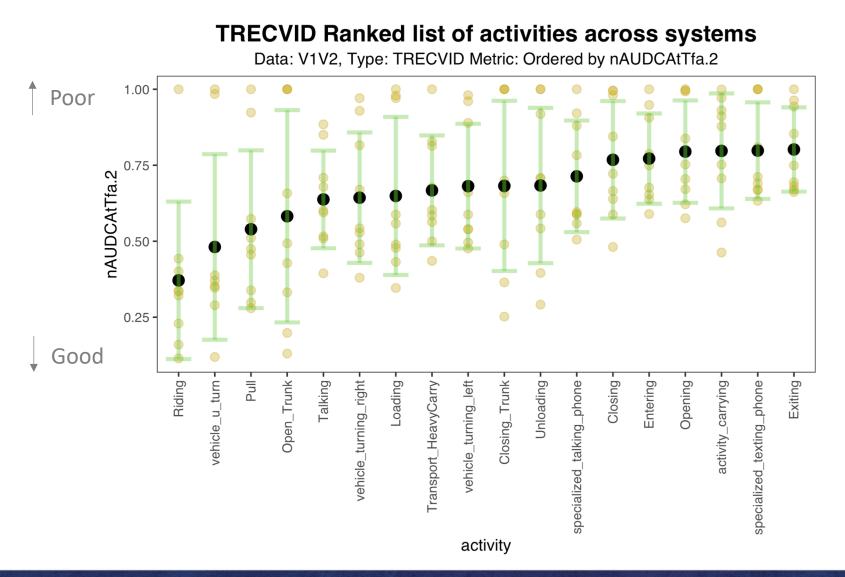


Observation

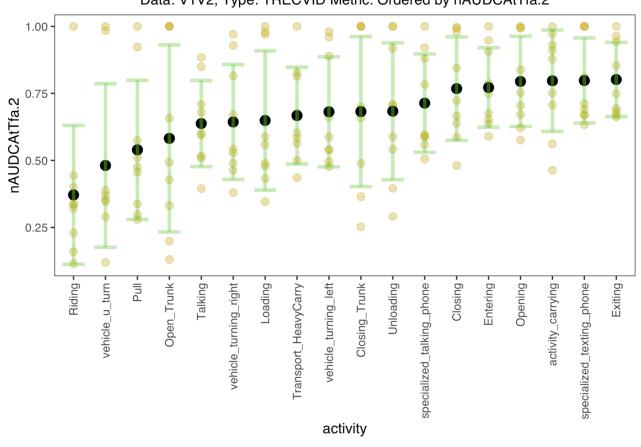
- Highest performance on activity detection:
 - MUDSML (nAUDC: 48.4%) followed by UCF (nAUDC: 49.1%)
- A large variance of the 18 activities across systems



Activity Ranking (AD)







TRECVID Ranked list of activities across systems

Data: V1V2, Type: TRECVID Metric: Ordered by nAUDCAtTfa.2

Observation

- Given the dataset and the 18 activities, "Riding" is the easiest to detect while "Exiting" is the hardest across the 9 systems
- "Open_Truck" and "Closing_Truck" have lager variance across systems



Which activities are easier or more difficult to detect?

TRECVID Summary of Activitian Difficulty

		IKE				•					-
		D	ata: v	172, 1	ype:	REC	/ID Me	etric: n	AUDC	Atita	.2
	Riding -	3.2	1.0	1.0	1.0	1.0	13.5	5.0	4.0	1.0	1.0
	vehicle_u_turn -	4.6	4.0	2.0	2.0	9.0	13.5	1.0	1.0	4.0	5.0
	Pull -	5.5	2.0	3.0	6.0	5.0	13.5	8.0	2.0	3.0	7.0
	vehicle_turning_right -	6.8	7.0	9.0	5.0	8.0	1.0	7.0	11.0	9.0	4.0
	Talking -	7.4	9.0	5.0	15.0	2.0	3.0	13.0	3.0	7.0	10.0
	Open_Trunk -	7.5	3.0	17.0	3.0	15.0	13.5	2.0	9.0	2.0	3.0
	vehicle_turning_left -	7.9	10.0	7.0	8.0	7.0	6.0	9.0	10.0	8.0	6.0
>	Loading -	7.9	5.0	12.0	7.0	15.0	5.0	6.0	7.0	6.0	8.0
activity	Transport_HeavyCarry -	9.0	8.0	4.0	9.0	15.0	2.0	17.0	5.0	12.0	9.0
ä	specialized_talking_phone -	9.7	12.0	6.0	4.0	4.0	13.5	12.0	8.0	10.0	18.0
	Unloading -	10.5	11.0	8.0	13.0	15.0	13.5	4.0	14.0	14.0	2.0
	Closing_Trunk -	10.6	6.0	17.0	10.0	15.0	13.5	3.0	13.0	5.0	13.0
	Entering -	12.3	13.0	10.0	11.0	3.0	13.5	15.0	15.0	15.0	15.0
	activity_carrying -	12.4	18.0	13.0	14.0	11.0	4.0	18.0	6.0	11.0	17.0
	Closing -	12.6	14.0	14.0	17.0	10.0	7.0	10.0	17.0	13.0	11.0
	Exiting -	13.8	17.0	11.0	12.0	6.0	13.5	16.0	18.0	17.0	14.0
	Opening -	14.1	16.0	15.0	18.0	15.0	8.0	11.0	16.0	16.0	12.0
	specialized_texting_phone -	15.2	15.0	17.0	16.0	15.0	13.5	14.0	12.0	18.0	16.0
		AVG	BUPT-M_3541-	FRAUNH_2716 -	HITACH_3525 -	HSMW_T_3537 -	MKLAB_1739 -	MUDSML_3535 -	NTT_CQ_3519 -	UCF_3524 -	VIREOJ_2016 -
					t	team_	abbrev	/			

- X-axis: team names and and average activity ranking (AVG)

- Y-axis:18 activities -Numbers in the matrix: the ranking of 18 activities per system

15

10

The activity class was characterized by systems and baseline performance

Observation: the Riding, vehicle_u_turn, and Pull activities are easier to detect compared to the rest of the other activities

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Comparison of ActEV18 and ActEV19 Results

	ActE	V18	ActEV19				
Team	Self(12)	LB (19)	LB (18)				
	PR.15↓	PR.15↓	PR.15↓	nAUDC			
UMD	0.618	x	x	x			
SeuGraph	0.624	x	x	x			
Team_Vision	0.710	0.709	x	x			
UCF	0.759	0.733	0.680	0.491			
STR-DIVA Team	0.827	x	x	x			
JHUDIVATeam	0.887	x	x	x			
MUDSML (INF)	0.896	0.844	0.789	0.484			
SRI	0.927	x	x	x			
VANT	0.940	0.882	x	x			
HSMW_TUC	0.961	x	0.951	0.941			
BUPT-MCPRL	0.990	0.749	0.736	0.524			
USF Bulls	0.991	0.934	x	x			
MKLab (ITI_CERTH)	0.999	x	0.968	0.964			
UTS-CETC	x	0.925	x	x			
NII_Hitachi_UIT	x	0.925	0.819	0.599			
Fraunhofer IOSB	x	x	0.849	0.827			
NTT_CQUPT	x	x	0.878	0.601			
vireoJD-MM	x	x	0.714	0.601			

T: TRECVID, D: DIVA, Self: Self-reported eval, LB: Leaderboard eval PR.15: μP_{miss} at $R_{FA} = 0.15$

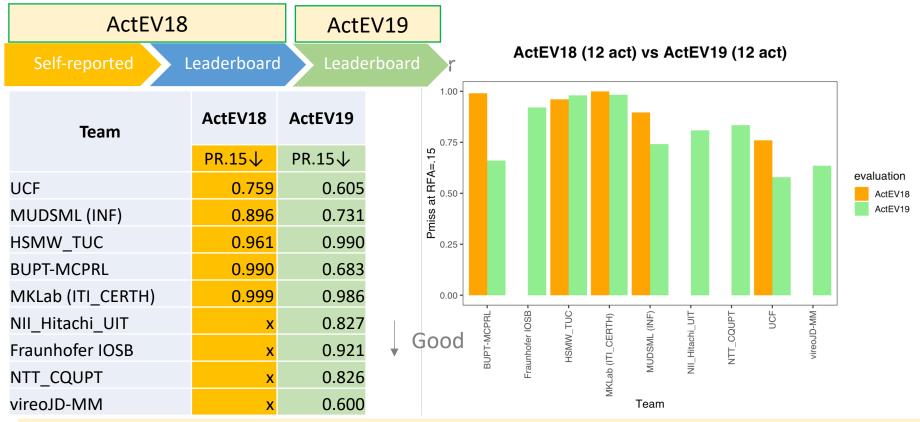


Comparison of ActEV18 vs ActEV19 (Leaderboard only)

	ActEV18	ActEV19]	PR . 1	15: <i>P</i>	miss	at R_{f}	_a =. 1	l 5 (<i>Ac</i>	tEV1	8 <i>sco</i>	ring	prot	ocol)
Dataset	VIRAT V1	VIRAT V1V2	↑ Po	oor		Act	EV18 (19 ac	et) vs A	CtEV	19 (18	3 act)		
# Activities	19	18		1.00				_]
Metric	PR.15	PR.15	1						_					
	ActEV1	8 ActEV19		0.75 -								_	_	
Team	LB (19)	LB (18)	۲ ۲	CT.=ATRA=.020-										ovelvetie
	PR.15	∕ PR.15↓		1 0.50 -										evaluatio
UCF	0.7	<mark>33</mark> 0.680	0 0	ISS a										ActEV
MUDSML (INF)	0.8	<mark>44</mark> 0.789	2 0	E 0.25 -										
HSMW_TUC		x 0.951		0.20										
BUPT-MCPRL	0.7	<mark>49</mark> 0.736												
MKLab (ITI_CERTH)		x 0.968		0.00 -										
NII_Hitachi_UIT	0.9	<mark>25</mark> 0.819	🗼 G	iood	CPRL	IOSB	TUC	RTH)	(INF)	TIU_ir	CQUPT	UCF	D-MM	
Fraunhofer IOSB		x 0.849			BUPT-MCPRL	Fraunhofer IOSB	HSMW_TUC	MKLab (ITI_CERTH)	MUDSML (INF)	NII_Hitachi_UIT			vireoJD-MM	
NTT_CQUPT		x 0.878			BU	Fraur	I	-ab (IT	MUI	IN	Z		-	
vireoJD-MM		x 0.714						MKL	Teers					
									Team					

Observation: System performance improved from last year for leaderboard eval. For example, reduced ~12% relative error rate NII_Hitachi_UIT, ~7% for and UCF and MUDSML

Comparison of ActEV18 vs ActEV19 (12 Activities only)



Observation:

- System performance on 12 activities improved largely from ActEV18 to ActEV19

 Reduced 31% relative error rate for BUPT-MCPRL, 21% for UCF, and 18% for MUDSML



Summary

- New performance measure to be more relevant to the user cases
- 256 submissions out of 9 teams
- Given the test set and the 18 activities, "Riding" is the easiest while "Exiting" is the hardest across the 9 systems
- Large system improvements this year from last year



Next Steps





Next Steps

 WACV HADCV'20 (Human Activity Detection in multi-camera, Continuous, long-duration Video) workshop (*paper submission deadline: Dec 15, 2019)* the details at <u>https://wacv20.wacv.net</u>



- Resources: <u>https://actev.nist.gov</u> (click "Resources")
 - Datasets (training data)
 - Baseline algorithms
 - Annotation Tools
- TRECVID ActEV20 plan
 - ActEV Task Discussion Session (including new M1 data release)



Sequestered Data Leaderboard (SDL)

- Anyone can *submit their system to NIST*, which will then run the system on *sequestered data* (MEVA), post the results to the leaderboard
- Visit ongoing ActEV SDL Evaluation at <u>https://actev.nist.gov/sdl</u>
- MEVA data (<u>https://mevadata.org/</u>)
 - 37 activities (72 video hours) : Indoor and outdoor scenes, night and day, crowds and individuals, EO (Electro-Optical) and IR (Infrared) sensors
 - New M1 data release



9:00 - 9:30 am	Activities in Extended Video (ActEv) Task Overview
9:30 - 10:00 am	BUPT-MCPRL at TRECVID 2019: ActEv
	BUPT_MCPRL Team - Beijing University of Posts and Telecommunications
10:00 - 10:30 am	AI Surveillance System for Spatial-Temporal Activity Detection in Surveillance Scenarios
	MUDSML + INF Teams - Monash University; Carnegie Mellon University
10:30 - 11:00 am	Our ActEv approach with object detection and custom tracking algorithm
	HSMW_TUC Team - University of Applied Sciences Mittweida
11:00 - 11:30 am	Break with refreshments
11:30 - 12:00 am	Real-time activity detection in surveillance videos
	UCF Team - University of Central Florida
12:00 - 12:30 pm	Event Detection with Specialized Object Detectors
	Hitachi Team
12:30 - 2:00	Lunch
2:00 - 2:30 pm	Traffic Danger Recognition with Surveillance Cameras Without Training Data, Invited Talk: Lijun Yu, Carnegie Mellon University
2:30 - 2:50 pm	ActEv Task Discussion





Questions?

https://actev.nist.gov/

Contact: actev-nist@nist.gov





TRECVID ActEV19 Feedback and ActEV20 Discussion



Sequestered Data Leaderboard (SDL)

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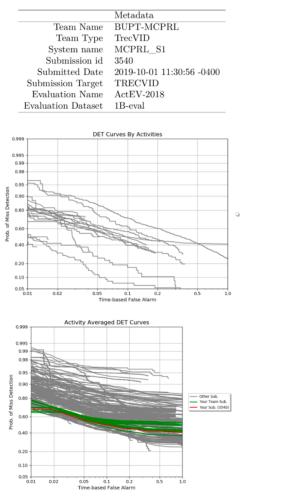


2019 ActEV feedback and 2020 plans

- What do the teams think about the ActEV task ?
- Any feedback on the new Scoring Metric compared to the 2018 Metric?
- Any feedback on the data repo to download data (VIRAT, MEVA, ..) ?
- Any feedback on the scoring server and different documents?
- Besides the ActEV leaderboard, we have added the ActEV reports (report on next slide), any feedback?
- Current Plan is to continue the ActEV task with the VIRAT dataset with more activities (28 or more activities)



Current TRECVID ActEV reports (at the end of the evaluation)



BUPT-MCPRL-SYS-00293-20191001-113056-4859

