

TRECVID 2016

AD-HOC VIDEO SEARCH TASK : OVERVIEW

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Ad-hoc Video Search Task Definition

- **Goal:** promote progress in content-based retrieval based on end user **ad-hoc queries** that include persons, objects, locations, activities and their combinations.
- **Task:** Given a test collection, a query, and a master shot boundary reference, return a ranked list of at most 1,000 shots (out of 335,944) which best satisfy the need.
- **New testing data:** 4,593 Internet Archive videos (IACC.3), 600 total hours with video durations between 6.5 min – 9.5 min.
- **Development data:** ~1400 hours of previous IACC data used between 2010-2015 with concept annotations.

Query Development

- Test videos were viewed by 10 human assessors hired by NIST
- 4 facet description of different scenes were used (if applicable):
 - **Who** : concrete objects and being (kind of persons, animals, things)
 - **What** : are the objects and/or beings doing ? (generic actions, conditions/state)
 - **Where** : locale, site, place, geographic, architectural
 - **When** : time of day, season
- In total assessors watched ~35% of the IACC.3 videos
- 90 Candidate queries chosen from human written descriptions to be used between 2016-2018.

TV2016 Query samples by complexity

- **Person + Action + Object + Location**

Find shots of a person playing guitar outdoors

Find shots of a man indoors looking at camera where a bookcase is behind him

Find shots of a person playing drums indoors

Find shots of a diver wearing diving suit and swimming under water

- **Person + Action + Location**

Find shots of the 43rd president George W. Bush sitting down talking with people indoors

Find shots of a choir or orchestra and conductor performing on stage

Find shots of one or more people walking or bicycling on a bridge during daytime

TV2016 Queries by complexity

- **Person + Action/state + Object**

Find shots of a person sitting down with a laptop visible

Find shots of a man with beard talking or singing into a microphone

Find shots of one or more people opening a door and exiting through it

Find shots of a person holding a knife

Find shots of a woman wearing glasses

Find shots of a person drinking from a cup, mug, bottle, or other container

Find shots of a person wearing a helmet

Find shots of a person lighting a candle

- **Person + Action**

Find shots of people shopping

Find shots of soldiers performing training or other military maneuvers

Find shots of a person jumping

Find shots of a man shake hands with a woman

TV2016 Queries by complexity

- **Person + Location**

Find shots of one or more people at train station platform

Find shots of two or more men at a beach scene

- **Person + Object**

Find shots of a policeman where a police car is visible

- **Object + Location**

Find shots of any type of fountains outdoors

- **Object**

Find shots of a sewing machine

Find shots of destroyed buildings

Find shots of palm trees

Training and run types

Four training data types:

- ✓ **A** – used only IACC training data (**4 runs**)
- ✓ **D** – used any other training data (**42 runs**)
- ✓ **E** – used only training data collected automatically using only the query text (**6 runs**)
- ✓ **F** – used only training data collected automatically using a query built manually from the given query text (**0 runs**)

Two run submission types:

- ✓ Manually-assisted (**M**) – Query built manually
- ✓ Fully automatic (**F**) – System uses official query directly

Evaluation

Each query assumed to be binary: absent or present for each master reference shot.

NIST sampled ranked pools and judged top results from all submissions.

Metrics: *inferred average precision per query.*

Compared runs in terms of **mean** *inferred average precision* across the 30 queries.

mean extended Inferred average precision (xinfAP)

2 pools were created for each query and sampled as:

- ✓ Top pool (ranks 1-200) sampled at 100%
- ✓ Bottom pool (ranks 201 - 1000) sampled at 11.1%
- ✓ % of sampled and judged clips from rank 201-1000 across all runs (min= 10.5%, max = 76%, mean = 35%)

30 queries
187,918 total judgments
7,448 total hits
4642 hits at ranks (1-100)
2080 hits at ranks (101-200)
726 hits at ranks (201-2000)

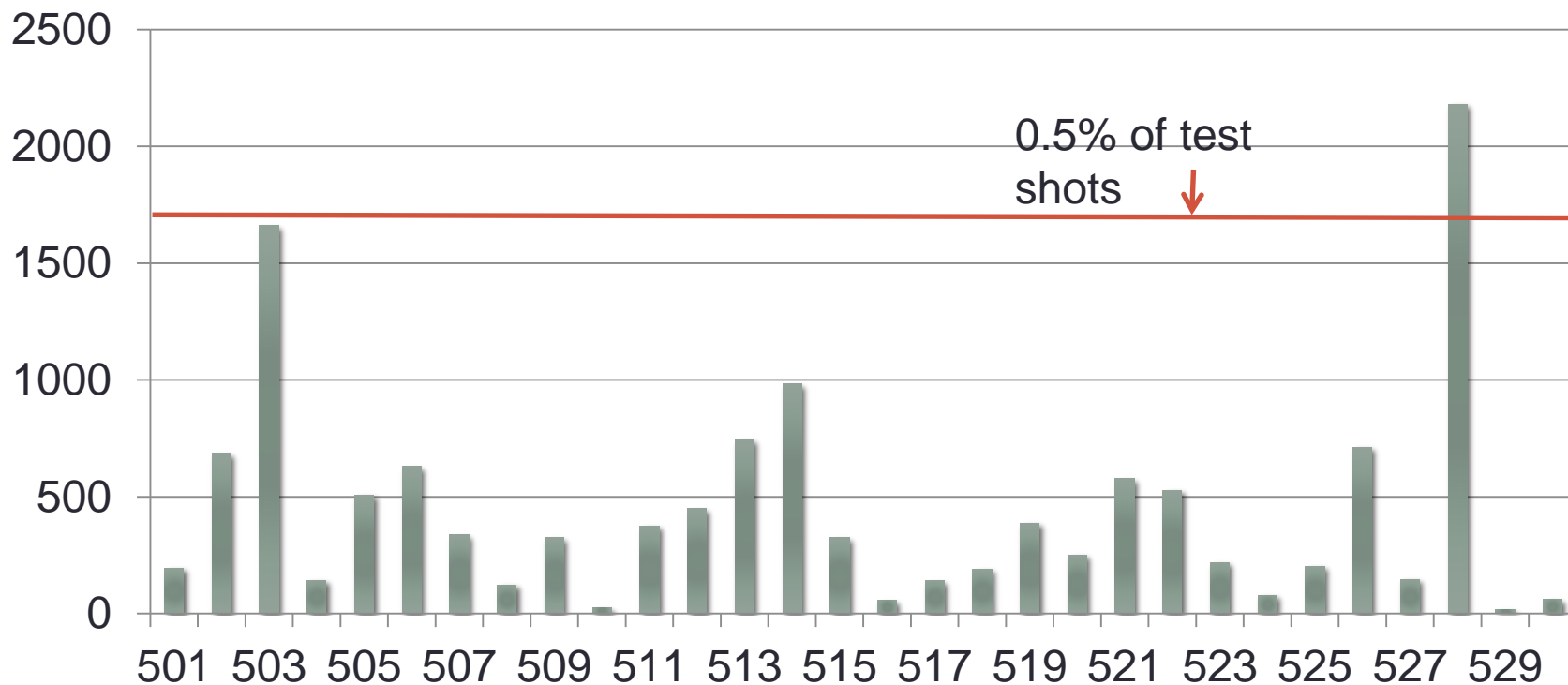
Judgment process: one assessor per query, watched complete shot while listening to the audio. infAP was calculated using the judged and unjudged pool by `sample_eval`

Finishers : 13 out of 29

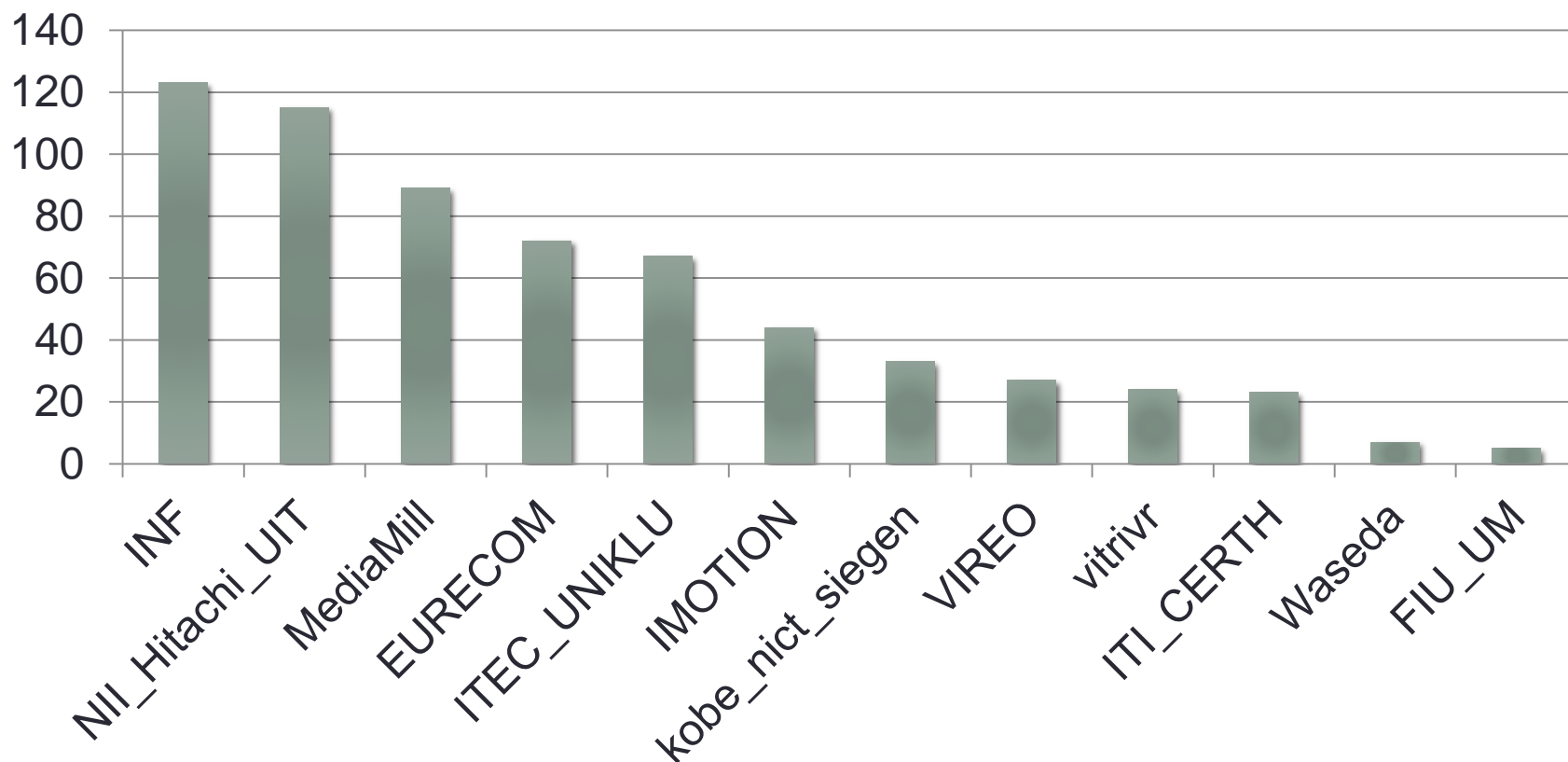
		M	F
INF	CMU; Beijing U. of Posts and Telecommunication; U. Autonoma de Madrid; Shandong U.; Xian JiaoTong U. Singapore	-	4
kobe_nict_siegen	Kobe U.; Japan National Institute of Information and Communications Technology, Japan U. of Siegen, Germany	3	-
UEC	Dept. of Informatics, The U. of Electro-Communications, Tokyo	2	-
ITI_CERTH	Inf. Tech. Inst., Centre for Research and Technology Hellas	4	4
ITEC_UNIKLU	Klagenfurt U.	-	3
NII_Hitachi_UIT	Natl. Inst. Of Info.; Hitachi Ltd; U. of Inf. Tech.(HCM-UIT)	-	4
IMOTION	U. of Basel, Switzerland; U. of Mons, Belgium; Koc U., Turkey	2	2
MediaMill	U. of Amsterdam Qualcomm	-	4
Vitrivr	U. of Basel	2	2
Waseda	Waseda U.	4	-
VIREO	City U. of Hong Kong	3	3
EURECOM	EURECOM	-	4
FIU_UM	Florida International U., U. of Miami	2	-

Inferred frequency of hits varies by query

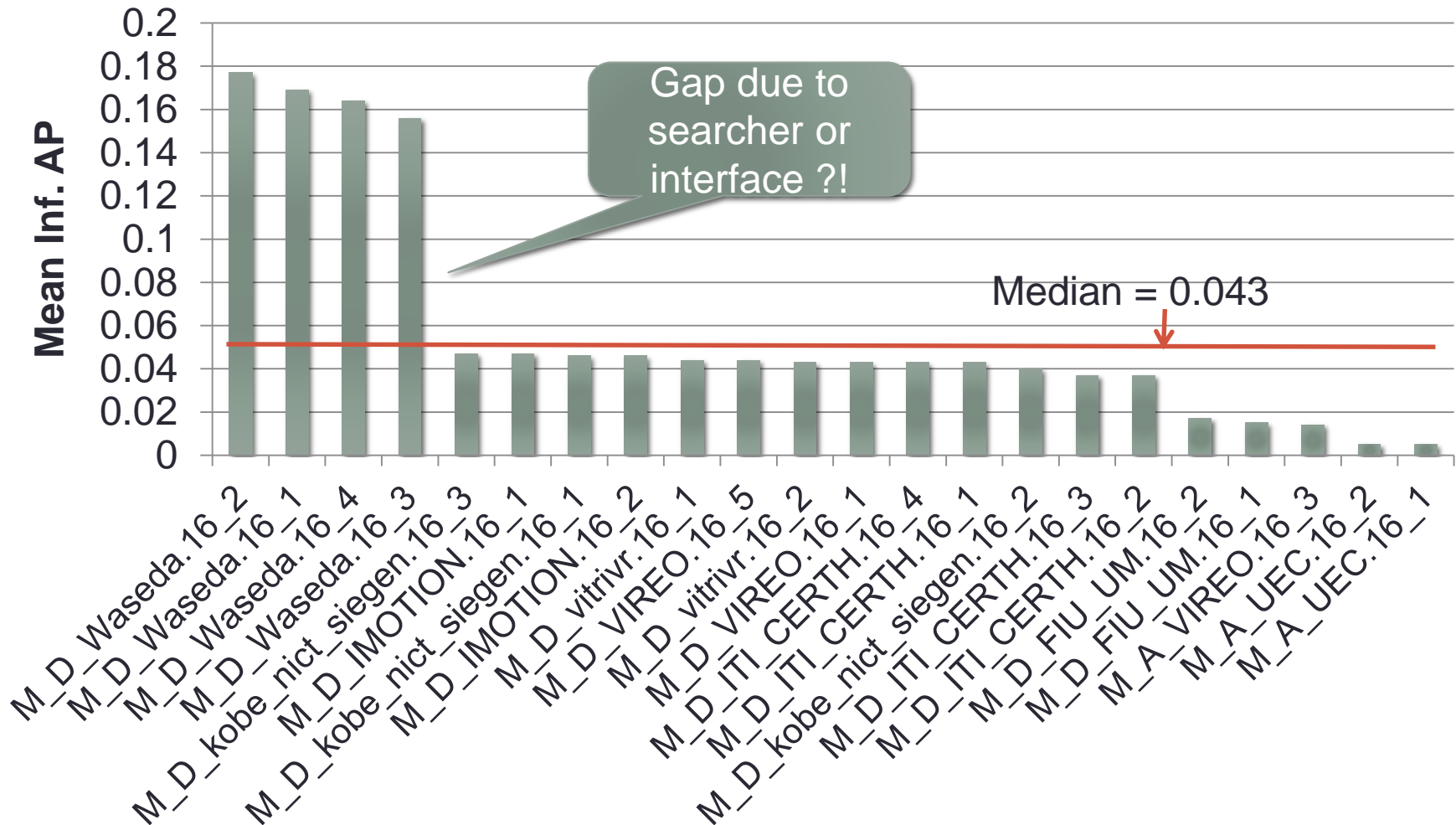
Inf. Hits / query



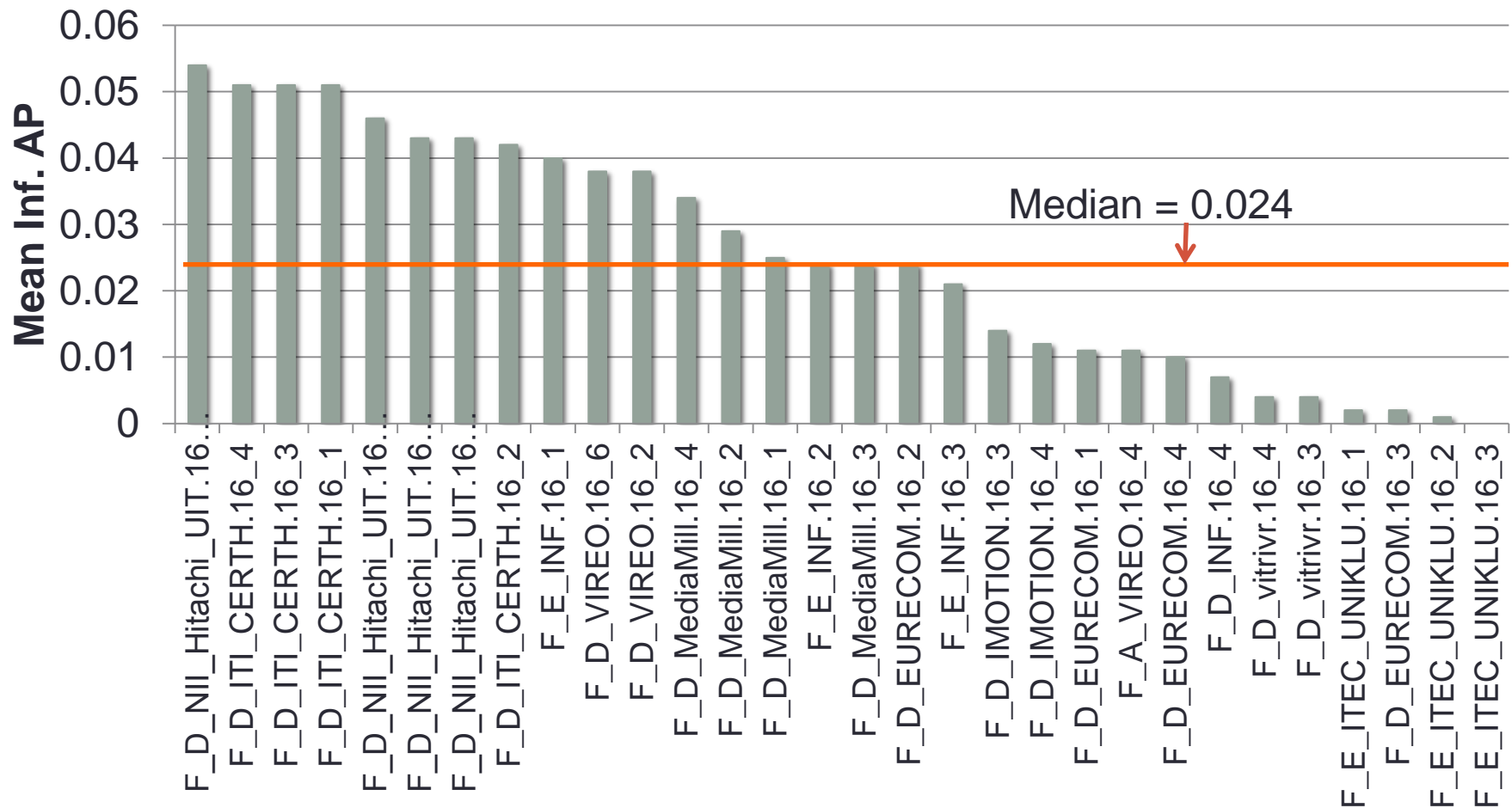
Total true shots contributed uniquely by team



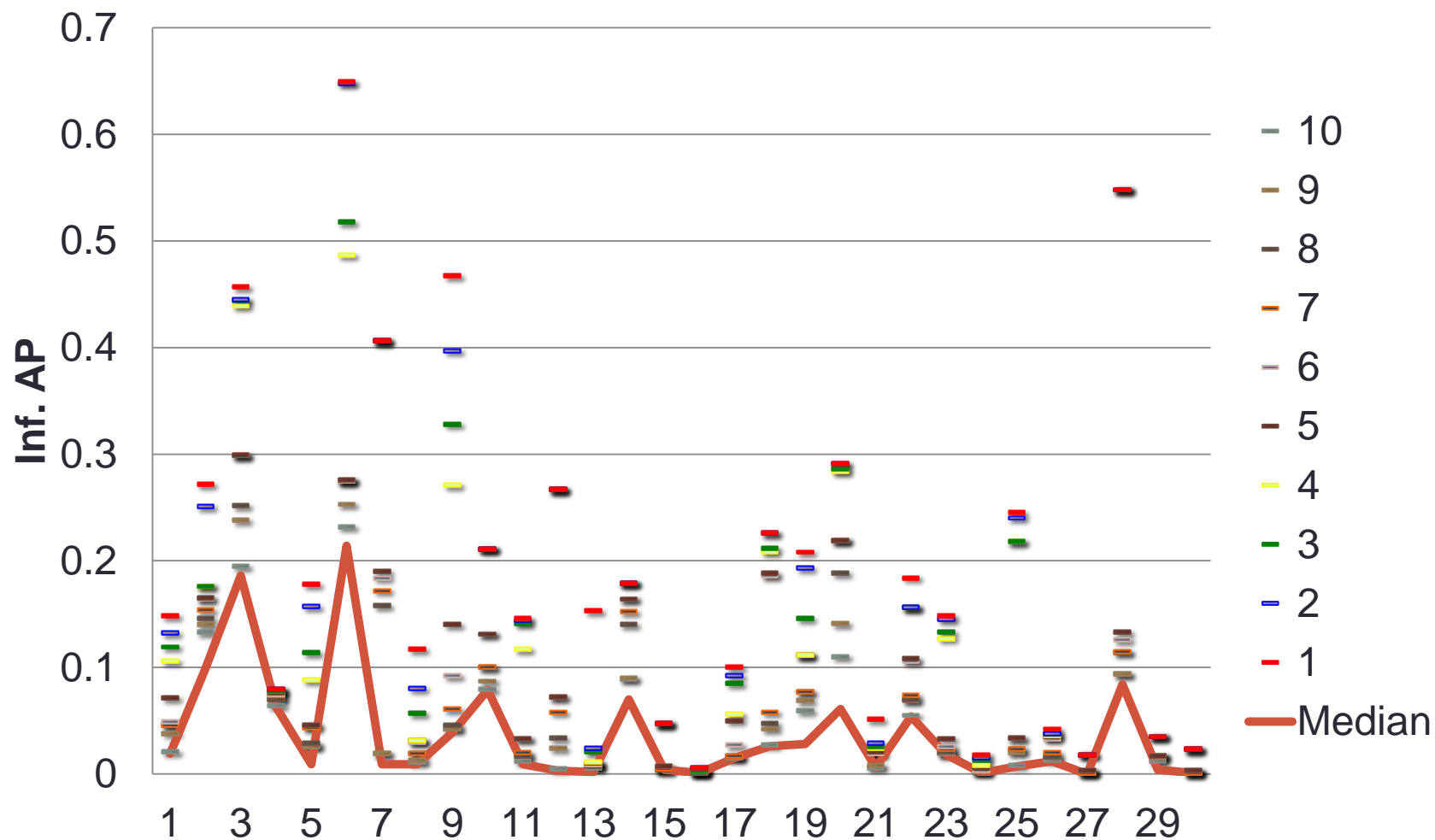
2016 run submissions scores (22 Manually-assisted runs)



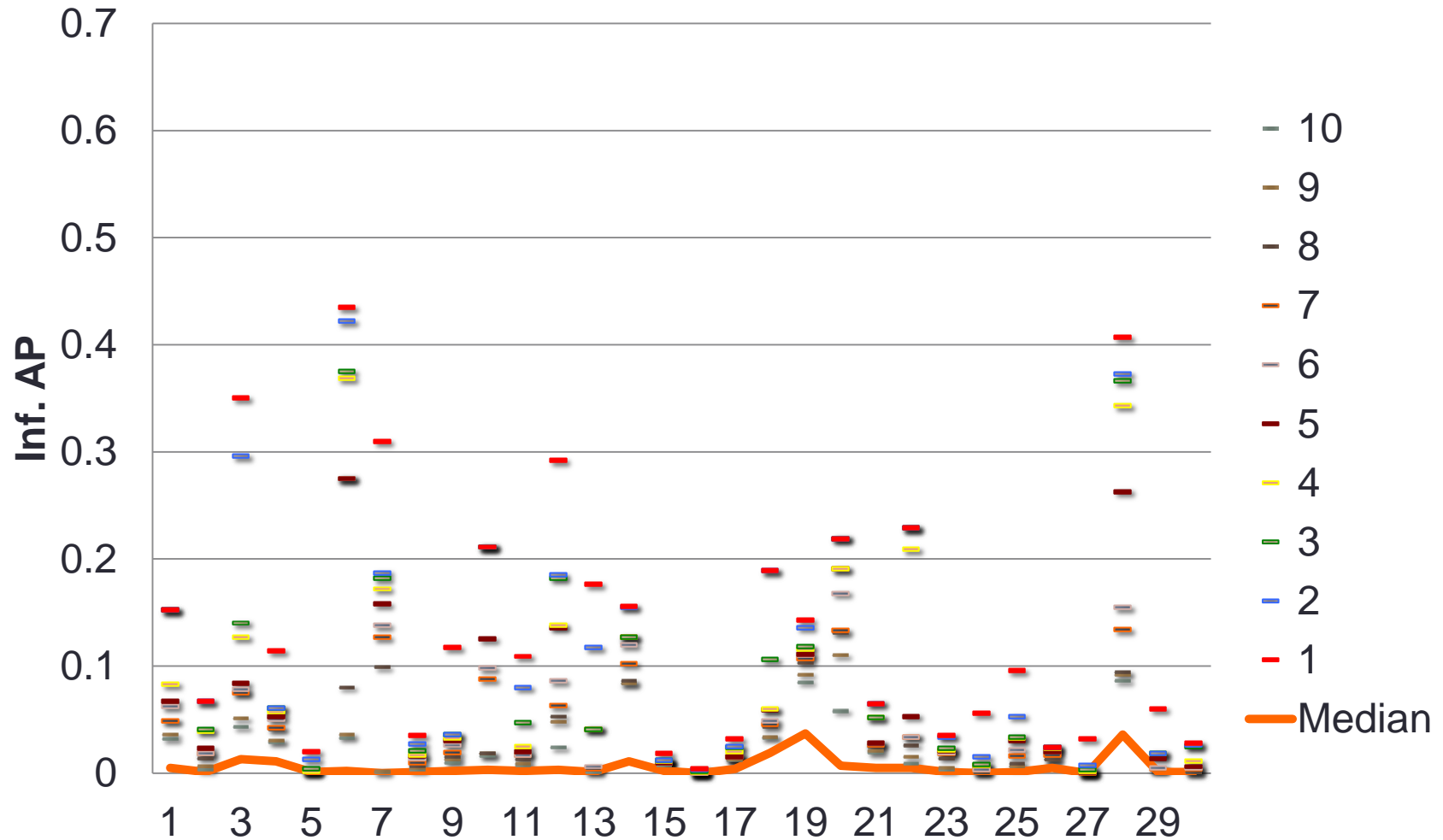
2016 run submissions scores (30 Fully automatic runs)



Top 10 infAP scores by query (Manually-assisted)



Top 10 infAP scores by query (Fully automatic)



Statistical significant differences among top 10 “M” runs (using randomization test, $p < 0.05$)

D_Waseda.16_2

- D_Waseda.16_3
 - D_kobe_nict_siegen.16_3
 - D_kobe_nict_siegen.16_1
 - D_IMOTION.16_1
 - D_IMOTION.16_2
 - D_vitrivr.16_1
 - D_VIREO.16_5
- D_Waseda.16_4
 - D_kobe_nict_siegen.16_3
 - D_kobe_nict_siegen.16_1
 - D_IMOTION.16_1
 - D_IMOTION.16_2
 - D_vitrivr.16_1
 - D_VIREO.16_5

D_Waseda.16_1

- D_Waseda.16_3
 - D_kobe_nict_siegen.16_3
 - D_kobe_nict_siegen.16_1
 - D_IMOTION.16_1
 - D_IMOTION.16_2
 - D_vitrivr.16_1
 - D_VIREO.16_5

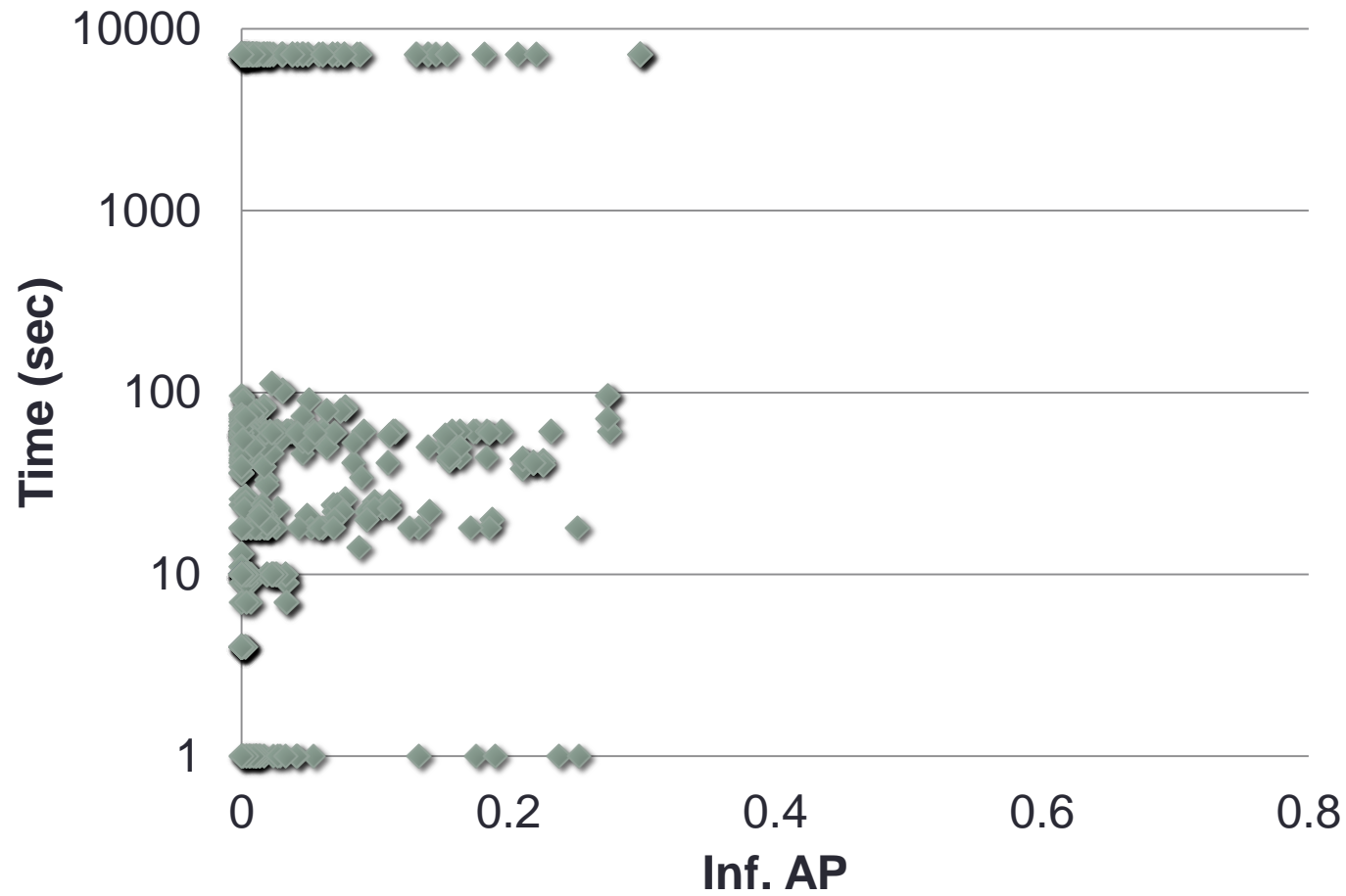
Run	Inf. AP score
D_Waseda.16_2	0.177 *
D_Waseda.16_1	0.169 *
D_Waseda.16_4	0.164 #
D_Waseda.16_3	0.156 #
D_kobe_nict_siegen.16_3	0.047 ^
D_IMOTION.16_1	0.047 ^
D_kobe_nict_siegen.16_1	0.046 ^
D_IMOTION.16_2	0.046 ^
D_vitrivr.16_1	0.044 ^
D_VIREO.16_5	0.044 ^

Statistical significant differences among top 10 “F” runs (using randomization test, $p < 0.05$)

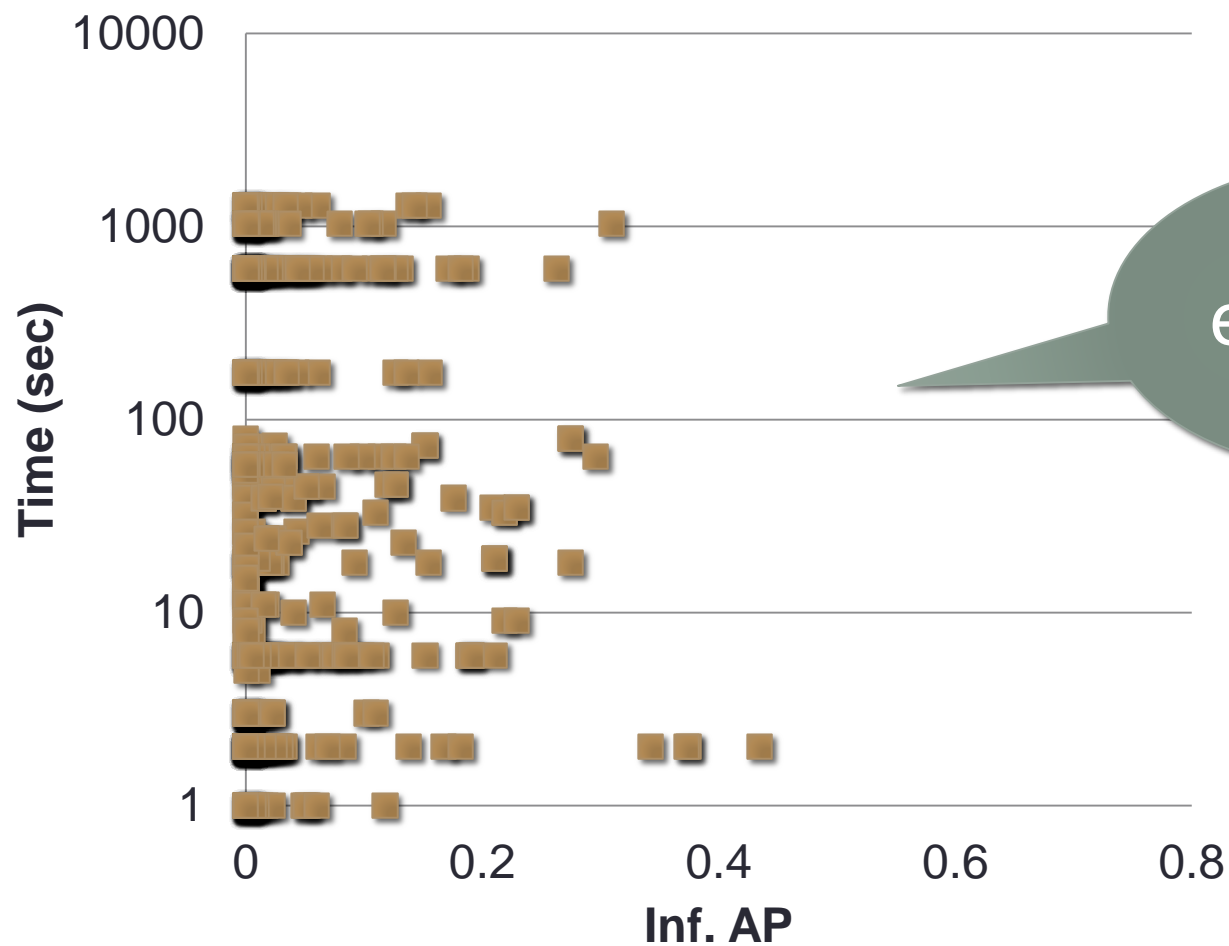
Run	Inf. AP score
D_NII_Hitachi_UIT.16_4	0.054
D_ITI_CERTH.16_4	0.051
D_ITI_CERTH.16_3	0.051
D_ITI_CERTH.16_1	0.051
D_NII_Hitachi_UIT.16_3	0.046
D_NII_Hitachi_UIT.16_2	0.043
D_NII_Hitachi_UIT.16_1	0.043
D_ITI_CERTH.16_2	0.042
E_INF.16_1	0.040
D_VIREO.16_6	0.038

No statistical
significant
differences among
the top 10 runs

Processing time vs Inf. AP ("M" runs)



Processing time vs Inf. AP ("F" runs)



Not fast enough?
!

2016 Observations / Questions

- Most teams relied on intensive visual concept indexing, leveraging on past SIN task and similar like ImageNet, Scenes ...
- Combined with manual or automatic query transformation
- Clever combination of concept scores (e.g. Waseda)
- Ad-hoc search is more difficult than simple concept-based tagging.
- Big gap between SIN best performance and AVS: maybe performance should be better compared with the “concept pair” task within SIN
- Manually-assisted runs performed better than fully-automatic.
- Most systems are not real-time (slower systems were not necessarily effective).
- Some systems reported 0 time!!!
- E and F runs are still rare compared to A and D
- Was the task/queries realistic enough?!
- Do we need to change/add/remove anything from the task in 2017 ?

Continued at MMM2017



6th Video Browser Showdown (VBS)

4-6 January, 2017 in Reykjavik, Iceland



- 10 Ad-Hoc Video Search (AVS) tasks, 5 of which are a random subset of the 30 AVS tasks of TRECVID 2016 and 5 will be chosen directly by human judges as a surprise. Each AVS task has several/many target shots that should be found.
- 10 Known-Item Search (KIS) tasks, which are selected completely random on site. Each KIS task has only one single 20-seconds long target segment
- Registration for the task is now closed

9:20 - 12:00 : Ad-hoc Video Search

- **9:20 - 9:40**, Task Overview
- **9:40 - 10:00**, NII_Hitachi UIT (National Institute of Informatics; Hitachi; U. of Inf. Tech.)
- **10:00 - 10:20**, ITI_CERTH (Centre for Research and Technology Hellas)
- **10:20 - 10:40**, **Break** with refreshments
- **10:40 - 11:00**, Waseda (Waseda University)
- **11:00 - 11:20**, kobe_nict_siegen (Kobe U.; Japan National Institute of Inf. and Communications Tech.;U. of Siegen)
- **11:20 - 11:40**, INF (Carnegie Mellon University, University of Technology Sydney, Renmin University of China, Shandong University)
- **11:40 - 12:00**, AVS discussion