



Visual and Textual Encoder Assembly for Ad-hoc Video Search

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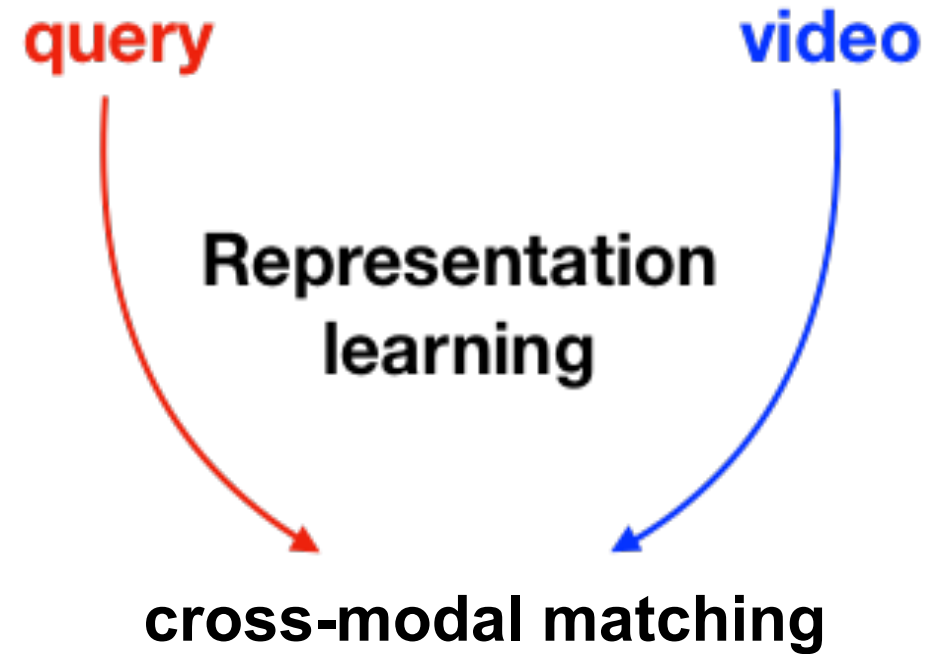
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<https://ruc-aimc-lab.github.io/>

7 December 2021

Ad-hoc Video Search (AVS)

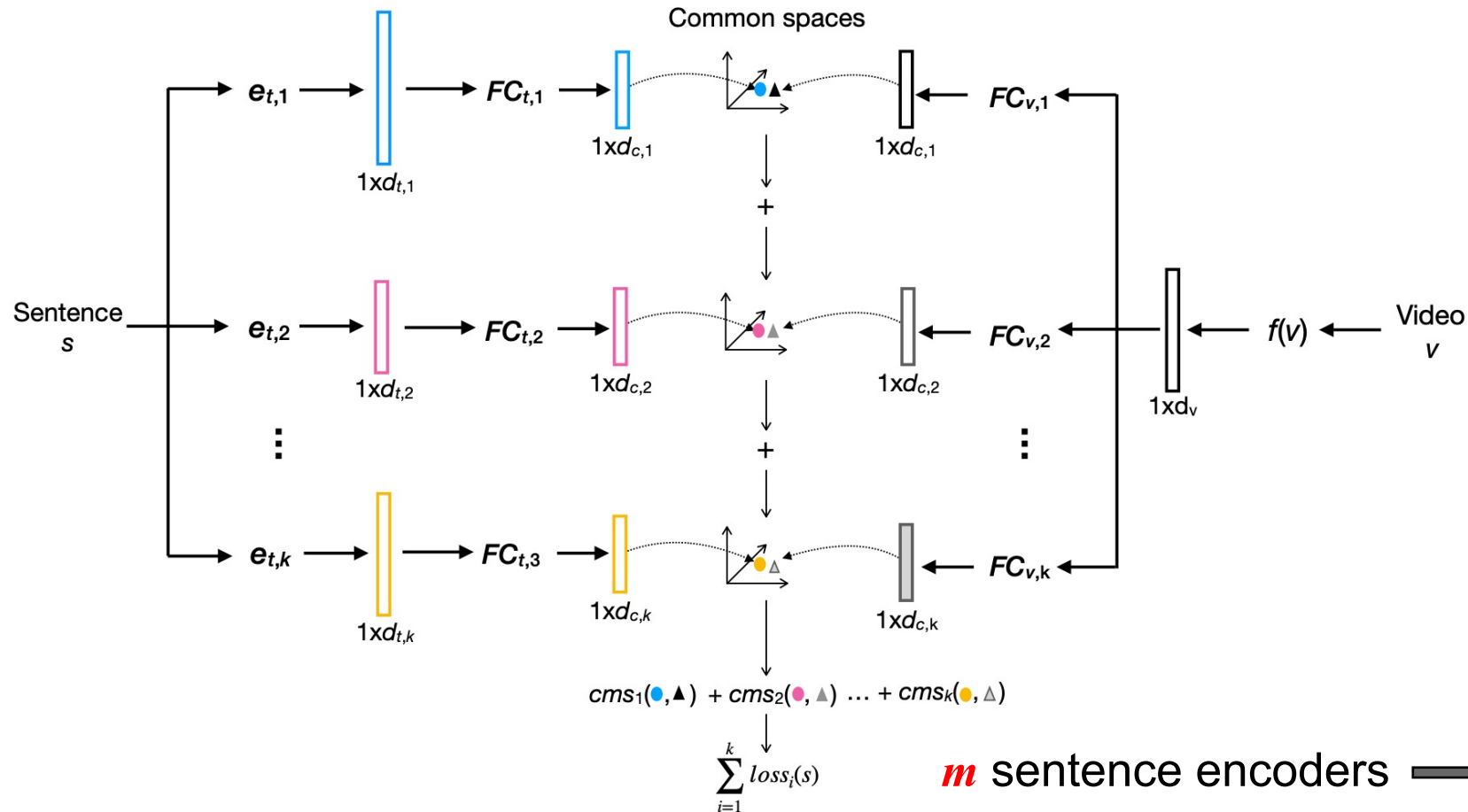
- Search for **unlabeled** videos by **natural-language** queries



Our Solution

Sentence Encoder Assembly (SEA)

- It supports text-video matching in multiple encoder-specific common spaces.



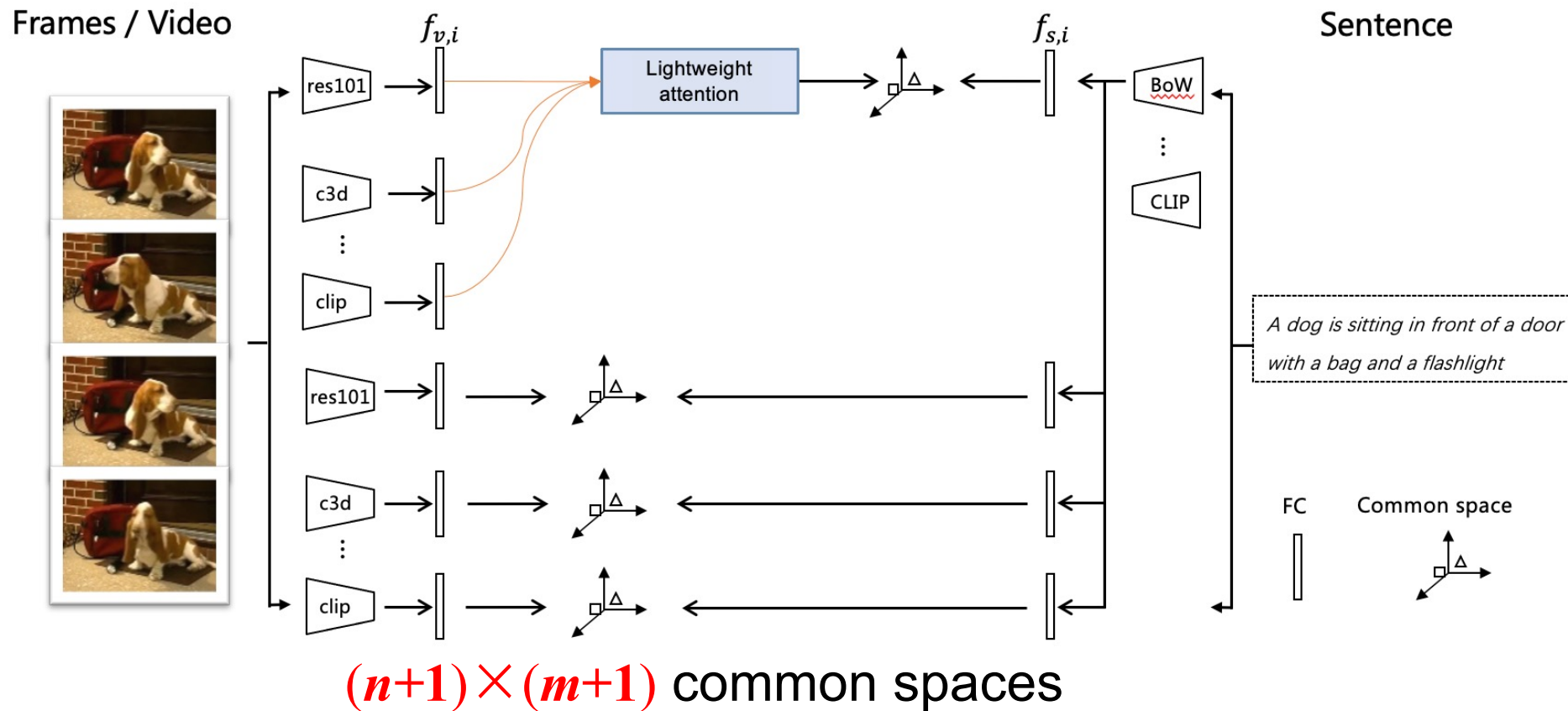
[1] Li et al., SEA: Sentence Encoder Assembly for Video Retrieval by Textual Queries, TMM 2020

[2] Li et al., Renmin University of China at TRECVID 2020: Sentence Encoder Assembly for Ad-hoc Video Search, TRECVID 2020

Our Solution

Multiple Encoder Assembly (MEA)

- It supports text-video matching in multiple feature-specific common spaces.



Choice of Visual Encoders

Seven 2D / 3D features :

2D feature (frame-level)

- rx101
- re152
- wsl
- clip

3D feature (segment-level)

- c3d
- ircsn
- tf

Rank	Feature	TV19	TV20	MEAN
1	<i>c3d</i>	0.036	0.098	0.067
2	<i>tf</i>	0.102	0.142	0.122
3	<i>ircsn</i>	0.088	0.193	0.14
4	<i>clip</i>	0.155	0.180	0.168
5	<i>wsl</i>	0.135	0.210	0.173
6	<i>rx101-wsl</i>	0.148	0.208	0.178
7	<i>clip-wsl</i>	0.148	0.208	0.178
8	<i>rx101-re152-wsl</i>	0.159	0.208	0.184
9	<i>rx101-re152</i>	0.167	0.201	0.184
10	<i>rx101-clip</i>	0.196	0.223	0.210
11	<i>rx101-re152-wsl-clip</i>	0.182	0.238	0.210
12	<i>rx101-wsl-clip</i>	0.183	0.239	0.211
13	<i>rx101-re152-clip</i>	0.199	0.233	0.216

Model: SEA(bow, w2v)

Choice of Textual Encoders

Three sentence encoders:

- Bag-of-Words (bow)
- Word2Vec (w2v)
- CLIP (ViT-B/32)^[1]

Feature	TV19	TV20	MEAN
Model: SEA(bow, w2v)			
<i>rx101-wsl-clip</i>	0.183	0.239	0.211
<i>rx101-re152-clip</i>	0.199	0.233	0.216
Model: SEA(bow, w2v, clip)			
<i>rx101-re152-clip</i>	0.199	0.249	0.224
<i>rx101-wsl-clip</i>	0.204	0.262	0.233

[1] Radford et al., Learning Transferable Visual Models from Natural Language Supervision, ICML 2021

Choice of (Pre-)Training Data

Five datasets :

- MSCOCO
- GCC
- MSR-VTT
- TGIF
- VATEX

Feature	TV19	TV20	MEAN
Pre-training on MS-COCO:			
<i>rx101-wsl-clip-c3d</i>	0.203	0.339	0.271
<i>rx101-wsl-clip-tf</i>	0.205	0.313	0.259
<i>rx101-wsl-clip-ircsn</i>	0.196	0.347	0.272
Pre-training on GCC:			
<i>rx101-wsl-clip-ircsn</i>	0.203	0.335	0.269
Pre-training on GCC and MS-COCO:			
<i>rx101-wsl-clip-ircsn</i>	0.209	0.357	0.283

Training data: MSR-VTT, TGIF and VATEX

Submissions(fully automatic track)

We submitted the following 4 runs:

MEA, SEA, CLIP and their combinations

Run id	Model	Feature	(Pre-)Training Data
Run1	Late fusion of Run3, Run4, CLIP	<i>rx101</i>	MSCOCO, GCC
Run2	MEA-reRank	<i>wsl</i>	MSR-VTT
Run3	MEA(bow, w2v,clip)	<i>clip</i>	TGIF
Run4	SEA(bow, w2v,clip)	<i>ircsn</i>	VATEX
-	CLIP (ViT/B32)	-	-

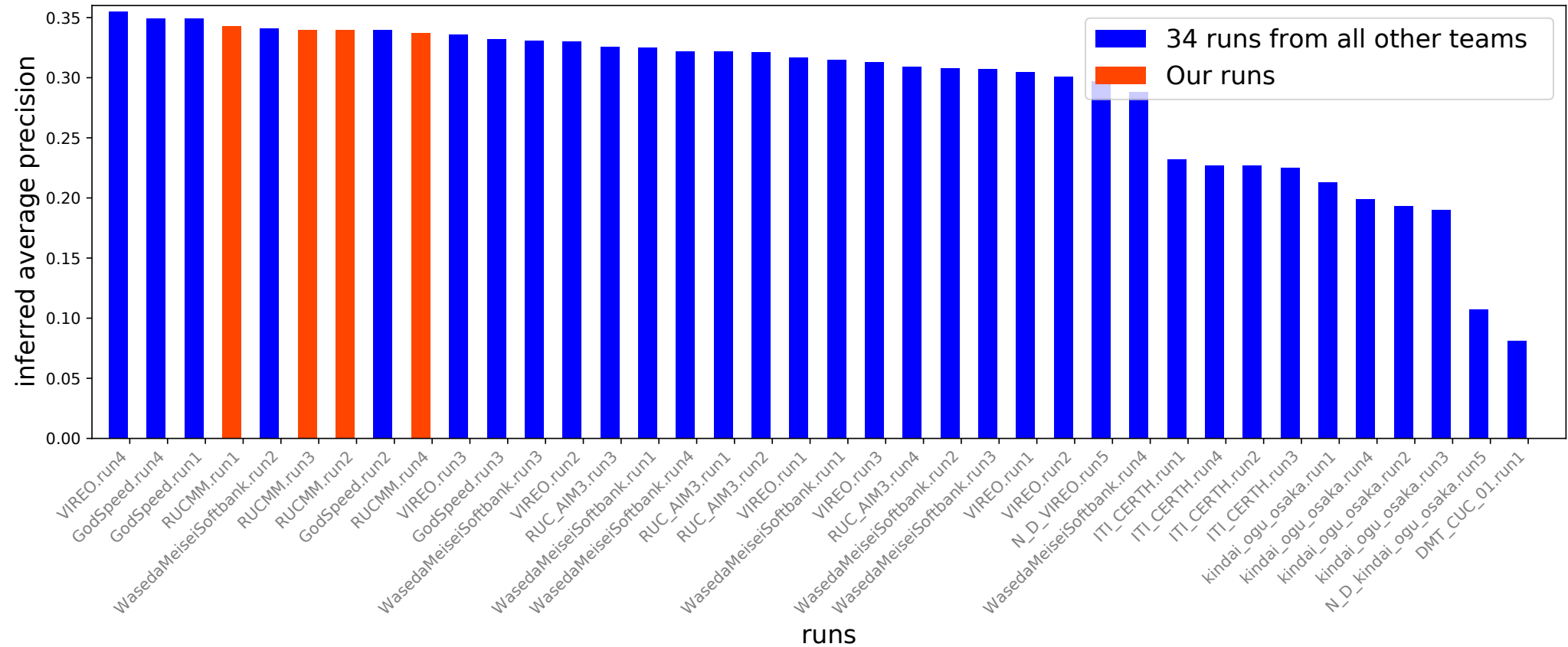
Performance on TV16-21 AVS Task

The performance of CLIP is lower, but it can bring better performance when doing late fusion.

Run id	Model	TV16	TV17	TV18	TV19	TV20	TV21	MEAN
Run1	Late fusion of Run3, Run4, CLIP	0.246	0.320	0.161	0.227	0.366	0.343	0.277
Run2	MEA-rerank	0.224	0.342	0.168	0.224	0.361	0.340	0.293
Run3	MEA	0.223	0.342	0.167	0.223	0.361	0.340	0.292
Run4	SEA(bow, w2v,clip)	0.232	0.255	0.135	0.213	0.358	0.337	0.239
-	CLIP	0.173	0.208	0.087	0.136	0.161	0.194	0.160
-	Late fusion of Run3, Run4	0.235	0.300	0.156	0.225	0.365	0.339	0.270

All fully automatic AVS submissions

Our submissions ranked the 3rd



Results of individual topics

More **blue** is better

More **red** is worse

Can be divided into 2 types:

- Easy topics: all **blue**
- Hard topics: all **red**

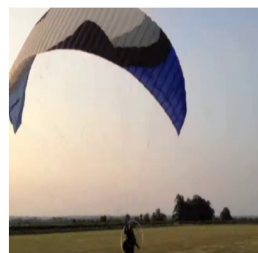
topic id	Run1	Run2	Run3	Run4
661	0.395	0.388	0.384	0.423
662	0.792	0.811	0.824	0.657
663	0.504	0.485	0.486	0.533
664	0.269	0.260	0.261	0.260
665	0.127	0.120	0.120	0.134
666	0.207	0.214	0.212	0.186
667	0.110	0.092	0.093	0.128
668	0.287	0.262	0.260	0.299
669	0.171	0.163	0.165	0.184
670	0.514	0.506	0.505	0.516
671	0.166	0.166	0.166	0.143
672	0.013	0.020	0.019	0.009
673	0.056	0.068	0.067	0.035
674	0.879	0.906	0.904	0.864
675	0.511	0.506	0.506	0.511
676	0.176	0.162	0.163	0.193
677	0.833	0.835	0.836	0.830
678	0.336	0.337	0.337	0.296
679	0.040	0.042	0.043	0.031
680	0.478	0.460	0.460	0.517

Per-topic analysis on TV21

topic id	Run1	Run2	Run3	Run4
661	0.395	0.388	0.384	0.423
662	0.792	0.811	0.824	0.657
663	0.504	0.485	0.486	0.533
664	0.269	0.260	0.261	0.260
665	0.127	0.120	0.120	0.134
666	0.207	0.214	0.212	0.186
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676	0.176	0.162	0.163	0.193
677	0.833	0.835	0.836	0.830
678	0.336	0.337	0.337	0.296
679	0.040	0.042	0.043	0.031
680	0.478	0.460	0.460	0.517

Easy query

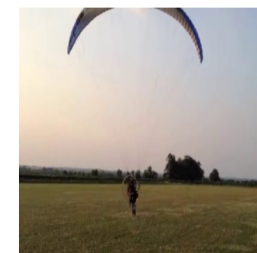
674 a **parachutist** descending towards a field on the ground in the **daytime**



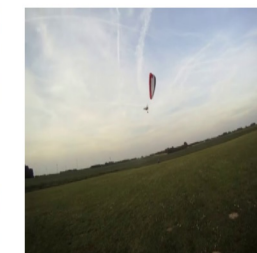
shot06988_28_62



shot06988_32_0



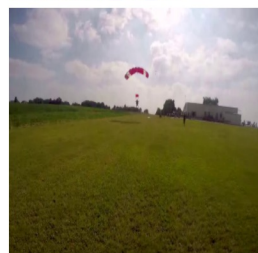
shot06988_31_62



shot05942_70_375



shot00615_100_0



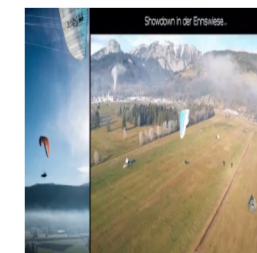
shot07270_124_108



shot07270_129_24



shot05675_1_11



shot06043_26_725



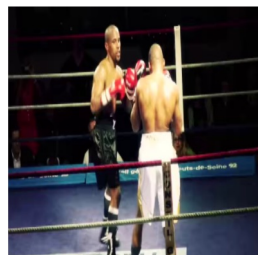
shot06988_46_125

Per-topic analysis on TV21

topic id	Run1	Run2	Run3	Run4
661	0.395	0.388	0.384	0.423
662	0.792	0.811	0.824	0.657
663	0.504	0.485	0.486	0.533
664	0.269	0.260	0.261	0.260
665	0.127	0.120	0.120	0.134
666	0.207	0.214	0.212	0.186
667	0.110	0.092	0.093	0.128
668	0.287	0.262	0.260	0.299
669	0.171	0.163	0.165	0.184
670	0.514	0.506	0.505	0.516
671	0.166	0.166	0.166	0.143
672	0.013	0.020	0.019	0.009
673	0.056	0.068	0.067	0.035
674	0.879	0.906	0.904	0.864
675	0.511	0.506	0.506	0.511
676	0.176	0.162	0.163	0.193
677	0.833	0.835	0.836	0.830
678	0.336	0.337	0.337	0.296
679	0.040	0.042	0.043	0.031
680	0.478	0.460	0.460	0.517

Easy query

677 two boxers in a ring



shot03605_272_0



shot03605_403_12



shot03605_265_37



shot03605_263_37



shot03605_266_12



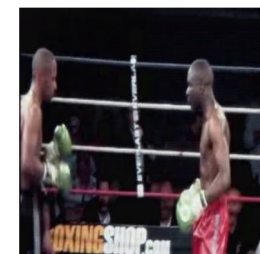
shot03605_330_25



shot03605_259_0



shot03605_412_12



shot03605_387_50



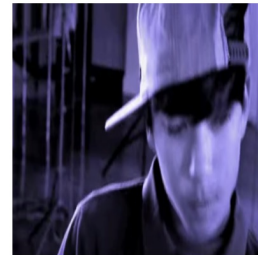
shot03605_276_25

Per-topic analysis on TV21

topic id	Run1	Run2	Run3	Run4
661	0.395	0.388	0.384	0.423
662	0.792	0.811	0.824	0.657
663	0.504	0.485	0.486	0.533
664	0.269	0.260	0.261	0.260
665	0.127	0.120	0.120	0.134
666	0.207	0.214	0.212	0.186
667	0.110	0.092	0.093	0.128
668	0.287	0.262	0.260	0.299
669	0.171	0.163	0.165	0.184
670	0.514	0.506	0.505	0.516
671	0.166	0.166	0.166	0.143
672	0.013	0.020	0.019	0.009
673	0.056	0.088	0.087	0.035
674	0.879	0.906	0.904	0.864
675	0.511	0.506	0.506	0.511
676	0.176	0.162	0.163	0.193
677	0.833	0.835	0.836	0.830
678	0.336	0.337	0.337	0.296
679	0.040	0.042	0.043	0.031
680	0.478	0.460	0.460	0.517

Hard query

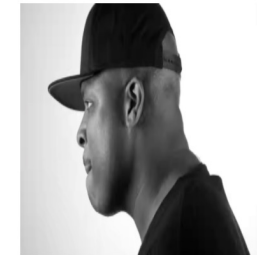
672 a person wearing a **cap backwards**



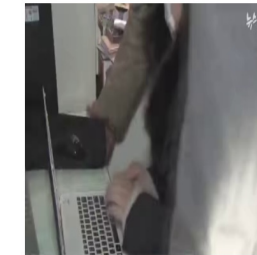
shot02563_158_57



shot06067_138_37



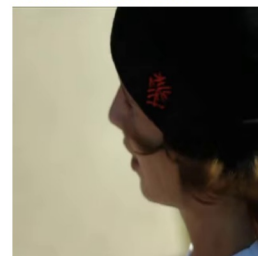
shot00375_31_11



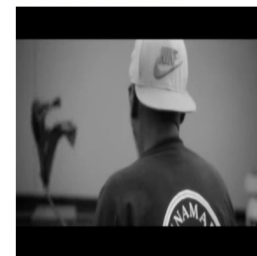
shot00306_180_0



shot03893_37_23



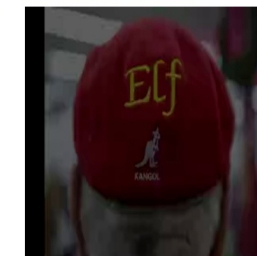
shot00799_6_0



shot01323_45_46



shot02563_152_23



shot07472_5_0



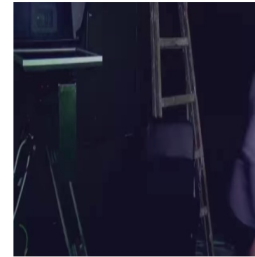
shot00375_253_0

Per-topic analysis on TV21

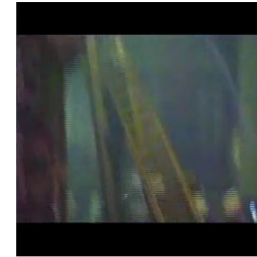
topic id	Run1	Run2	Run3	Run4
661	0.395	0.388	0.384	0.423
662	0.792	0.811	0.824	0.657
663	0.504	0.485	0.486	0.533
664	0.269	0.260	0.261	0.260
665	0.127	0.120	0.120	0.134
666	0.207	0.214	0.212	0.186
667	0.110	0.092	0.093	0.128
668	0.287	0.262	0.260	0.299
669	0.171	0.163	0.165	0.184
670	0.514	0.506	0.505	0.516
671	0.166	0.166	0.166	0.143
672	0.013	0.020	0.019	0.009
673	0.056	0.068	0.067	0.035
674	0.879	0.906	0.904	0.864
675	0.511	0.506	0.506	0.511
676	0.176	0.162	0.163	0.193
677	0.833	0.835	0.836	0.830
678	0.336	0.337	0.337	0.296
679	0.040	0.042	0.043	0.031
680	0.478	0.460	0.460	0.517

Hard query

679 a ladder with **less than 6 steps**



shot01890_300_0



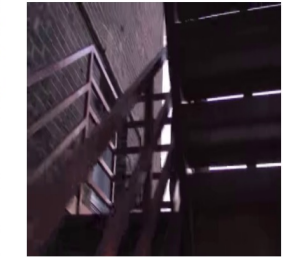
shot01303_172_0



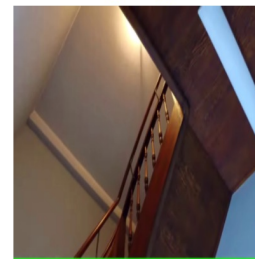
shot04153_76_0



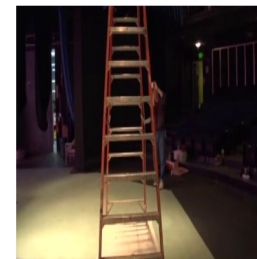
shot04153_77_37



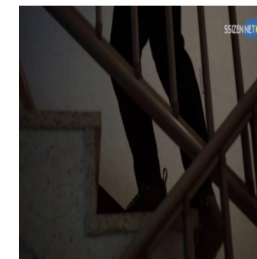
shot02132_95_14



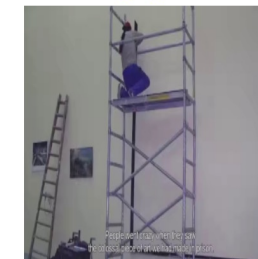
shot06200_45_50



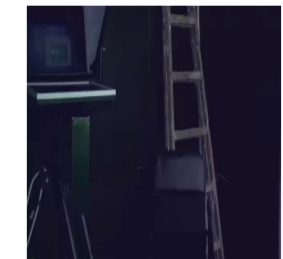
shot06079_245_80



shot07429_66_130



shot02608_106_12



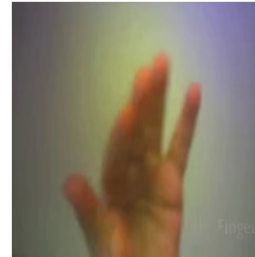
shot01890_273_0

Per-topic analysis on TV21

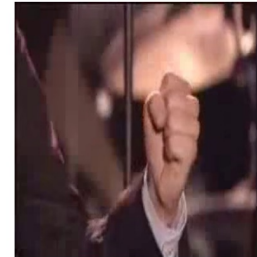
topic id	Run1	Run2	Run3	Run4
661	0.395	0.388	0.384	0.423
662	0.792	0.811	0.824	0.657
663	0.504	0.485	0.486	0.533
664	0.269	0.260	0.261	0.260
665	0.127	0.120	0.120	0.134
666	0.207	0.214	0.212	0.186
667	0.110	0.092	0.093	0.128
668	0.287	0.262	0.260	0.299
669	0.171	0.163	0.165	0.184
670	0.514	0.506	0.505	0.516
671	0.166	0.166	0.166	0.143
672	0.013	0.020	0.019	0.009
673	0.056	0.068	0.067	0.035
674	0.879	0.906	0.904	0.864
675	0.511	0.506	0.506	0.511
676	0.176	0.162	0.163	0.193
677	0.833	0.835	0.836	0.830
678	0.336	0.337	0.337	0.296
679	0.040	0.042	0.043	0.031
680	0.478	0.460	0.460	0.517

Hard query

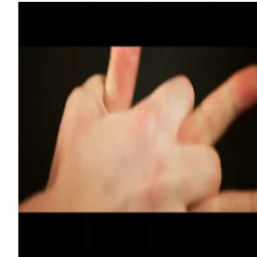
673 a man pointing with his finger



shot05271_82_12



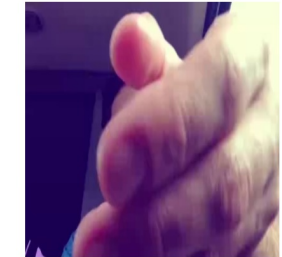
shot06476_90_0



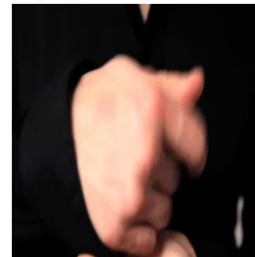
shot03237_174_0



shot07250_434_0



shot03918_505_0



shot02050_25_0



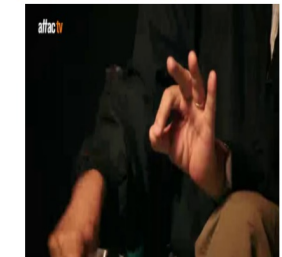
shot00123_209_0



shot00163_2_12



shot02305_68_72

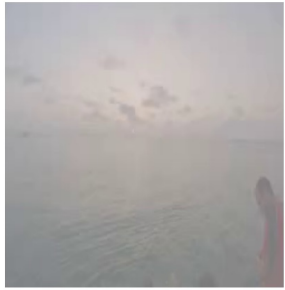


shot00169_147_12

Are we making progress?

Per-topic analysis on TV20

655 one or more persons **standing in a body of water**



shot07119_264_0



shot03670_19_0



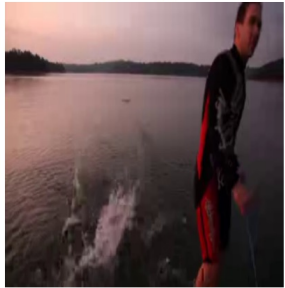
shot04386_80_0



shot05625_54_0



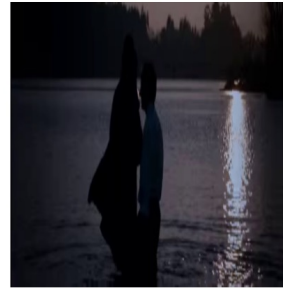
shot04342_137_12



shot02623_6_15



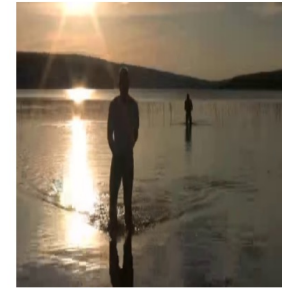
shot06853_69_0



shot04155_393_12



shot00119_72_0



shot00823_192_25

Topic id	RUCMM@TV20	RUCMM@TV21
641	0.281	0.316
642	0.522	0.640
643	0.092	0.228
644	0.662	0.861
645	0.166	0.089
646	0.155	0.254
647	0.436	0.364
648	0.099	0.063
649	0.469	0.835
650	0.079	0.085
651	0.072	0.610
652	0.130	0.228
653	0.231	0.416
654	0.341	0.365
655	0.053	0.064
656	0.629	0.730
657	0.084	0.106
658	0.055	0.062
659	0.342	0.426
660	0.476	0.574

Easy queries become easier, hard queries remain hard.

Are we making progress?

Per-topic analysis on TV20

658 two or more people **under a tree**



shot03193_249_37



shot03193_238_175



shot03193_242_275



shot06805_52_25



shot02499_169_25



shot06677_100_60



shot03484_8_29



shot03193_243_62



shot04413_19_72



shot05693_160_0

Topic id	RUCMM@TV20	RUCMM@TV21
641	0.281	0.316
642	0.522	0.640
643	0.092	0.228
644	0.662	0.861
645	0.166	0.089
646	0.155	0.254
647	0.436	0.364
648	0.099	0.063
649	0.469	0.835
650	0.079	0.085
651	0.072	0.610
652	0.130	0.228
653	0.231	0.416
654	0.341	0.365
655	0.053	0.064
656	0.629	0.730
657	0.084	0.106
658	0.055	0.062
659	0.342	0.426
660	0.476	0.574

Easy queries become easier, hard queries remain hard.

Conclusions

To boost AVS performance

- Cross modal Matching Model
- Video/text feature
- (Pre-)Training data

Understanding **fine-grained** queries is still hard

- Attributes: number
- Positions: in the water, under a tree