# Subjective Analysis and Objective Characterization of Adaptive Bitrate Videos

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## **Motivation: Evaluating HAS QoE**

 Increasing number of services based on HTTP Adaptive Streaming (HAS)

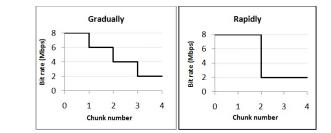


High interest on identifing the factors influencing on Quality of Experince (QoE) of HAS

# Challenges

### Multiple parameters can influence

- Switching frequency
- Switching amplitud
- Content
  - Genre
  - Objective characterization





### • Goal: Evaluation (improvment) of whole HAS session

- Analyzing the effect of each parameters
  - Testing certain patterns of quality switching
  - NOT the video quality in each time instant

### Challenges

- Adaptation event can last up to several seconds
- Lack of appropriate testing method
  - Mostly for assessment of short sequences (~10 sec)
    - Absolute Category Rating (ACR)
       PVS1
       ~10 sec
  - Recency, hystersis effect and decreasing user attention
    - Single Stimulus Continuous Quality Evaluation (SSCQE)

Test Sequence ≤ 5min

PVS2

Continuous vote

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- Hard to design experiments on event level
  - Test session should not last longer than 1.5 hrs (Rec. ITU-T P.913)
    - Avoiding user fatigue and boredom

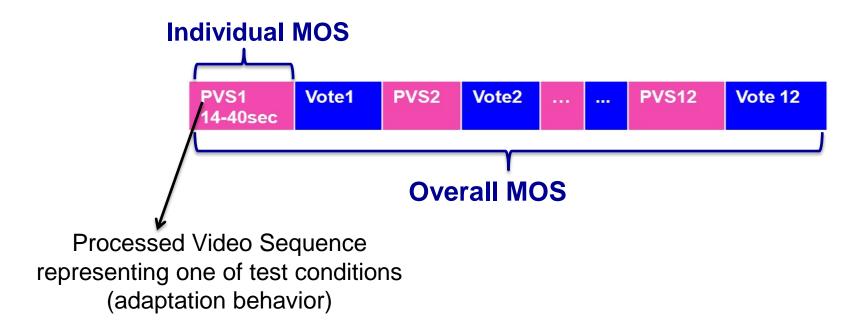
## **Objectives of this study**

- 1. Toward improving **subjective testing methodology** for adaptive streaming
  - Subjective Experiment
- 2. Objective characterization of adapted videos for understanding subjective ratings
  - Objective characterization

## **Subjective Experiment**

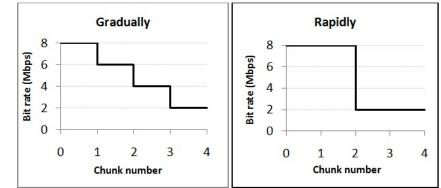
Toward improving subjective testing methodology:

- Impact of evaluation methodology on test subjects evaluation
- Relationship between MOS of individual adaptation events and overall MOS of whole sequence



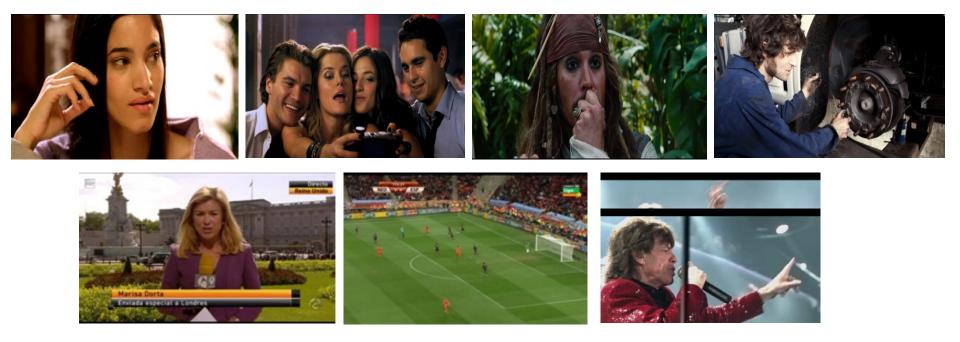
### **Adaptation study parameters**

- Switching behavior
  - Period (chunk length): 2 sec and 10 sec
  - Amplitude (difference between consequtive quality levels) gradual vs. rapid switching
  - o Decreasing and increasing
- Adaptation dimension
  - o Video quality (QP)
  - o H.264/AVC



- o 4 streams: 5Mbps, 3Mbps, 1Mbps, 600kbps
- o 1280x720/25fps
- MOS based on > 60 test subjects
  - o 132 adaptation events with variable length video

# **Test video sequences**



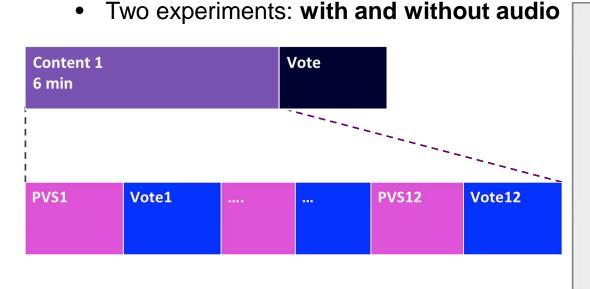
- 7 commercial content
  - Movie, Documentary, Sport, News, Music
- 6 min; originally 1080p Bluray video, 24/50 fps
- Different spatial and temporal characteristics

# **Evaluation methodology**

Cross-lab experiment (same PVS and rating question in both labs)

(1) **UPM:** Continuous presenation of 6min video including subsequent PVS

Evaluation of <u>individual PVS</u> and <u>overall quality</u> of each video





2 Acreo: Evaluation of overall quality of 6 min video

Content 1	Vote
6 min	

# **Experimental Setup**

- Environment and equipment in both labs
  - Satisfying Rec. ITU-R BT. 500-11
  - 46" Hyundai S465D display
  - Controlled lightning system
  - Viewing distance: 4H
- Observers' number (after screening)
  - UPM-Audio: 21 (6 female & 15 male)
  - UPM-NonAudio: 22 (5 female & 17 male)
  - Acreo: 30 (10 female & 20 male)
- Test presentation
  - One subject per session
  - Training





# **Objective Characteriation**

**<u>Aim</u>**: Identify the PVS characteristics influential on QoE

### Approach:

- No-Reference objective characterization tools
  - Blockiness
  - Blur
  - Brightness

- Noise
- Bitrate
- Contrast
- Full-Reference tool
  - Video Quality Model with Variable Frame Delays (VQM-VFD)
- Temporal pooling technique
  - Avrage
  - Av. low 10%

– Av. High 10%

- Av. first 2 sec
  Av. first 2 sec
- Av. first 2 sec
- Av. first & last 2 sec
  - Weighted Average

Spatial Activity
Temporal Activity

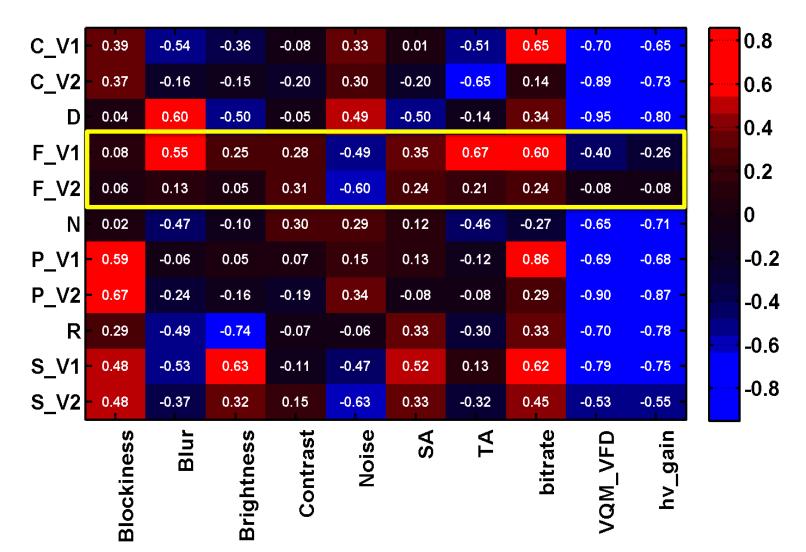
- Standard Deviation
- Minkowski
- Ninassi

### **SROCC Result: NR measurement and MOS**

	Blockiness	Blur	Brightness	Contrast	Noise	SA	ΤA	bitrate		-
Weighted Average	0.19	0.05	-0.23	0.31	-0.15	-0.06	-0.29	0.42		-0.4
verage, first & last 2s	0.07	-0.03	-0.22	-0.07	-0.15	0.22	0.04	0.36		-0.3
Standard Deviation	-0.46	-0.00	0.07	-0.08	-0.15	0.01	0.22	-0.14		-0.2
Ninassi	0.15	0.01	-0.21	0.29	-0.15	-0.10	-0.30	0.37		-0.1
Minkowski	-0.04	-0.04	-0.20	0.11	-0.19	-0.10	-0.13	-0.01		0
Average, last 2s	0.24	-0.10	-0.01	0.06	-0.08	0.29	0.07	0.37		
Average, first 2s	0.01	0.13	-0.25	-0.01	-0.10	0.17	0.10	0.17		0.1
Average High 10%	-0.02	0.05	-0.16	0.35	-0.17	-0.02	0.01	0.06		0.2
Average Low 10%	0.46	0.03	-0.19	0.24	-0.18	0.02	-0.09	0.23	<b>-</b>	0.3
Average	0.18	0.07	-0.24	0.32	-0.15	-0.07	-0.30	0.33	<b>.</b>	0.4

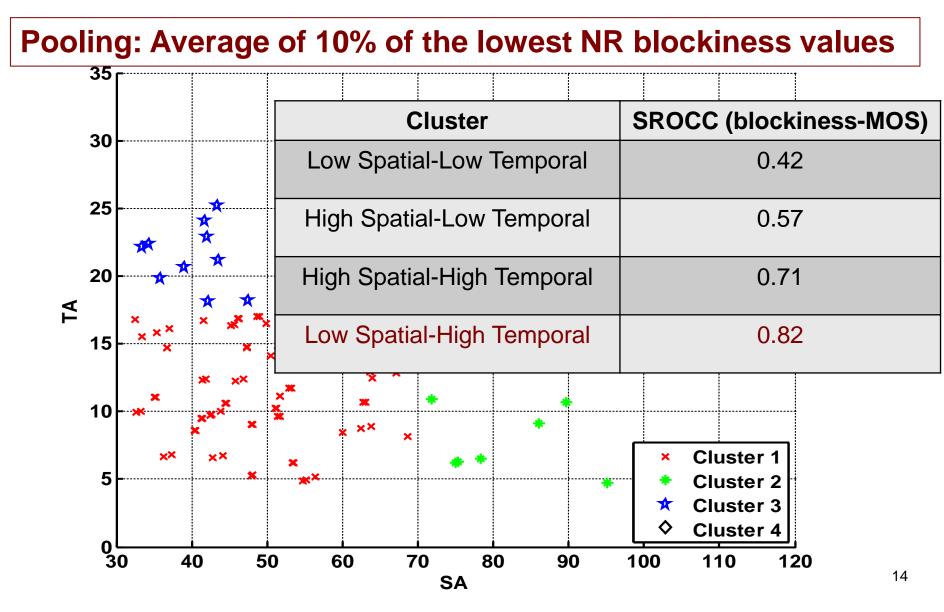
### SROCC Result: NR and FR measurement for each content

#### **Pooling approach for NR measurement: Average**

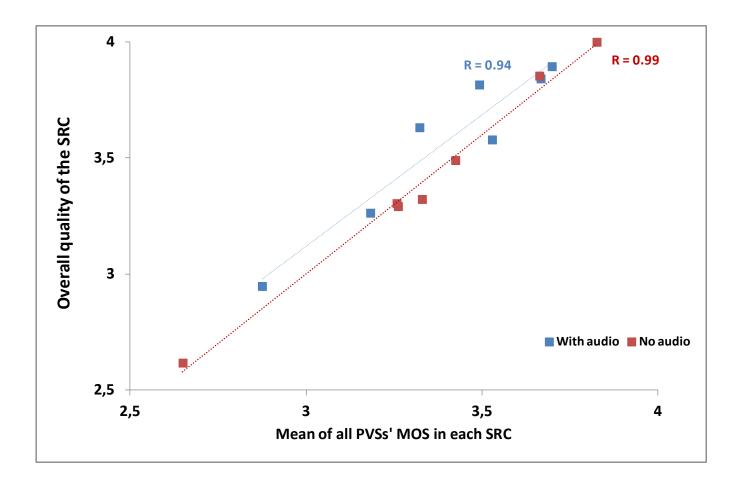


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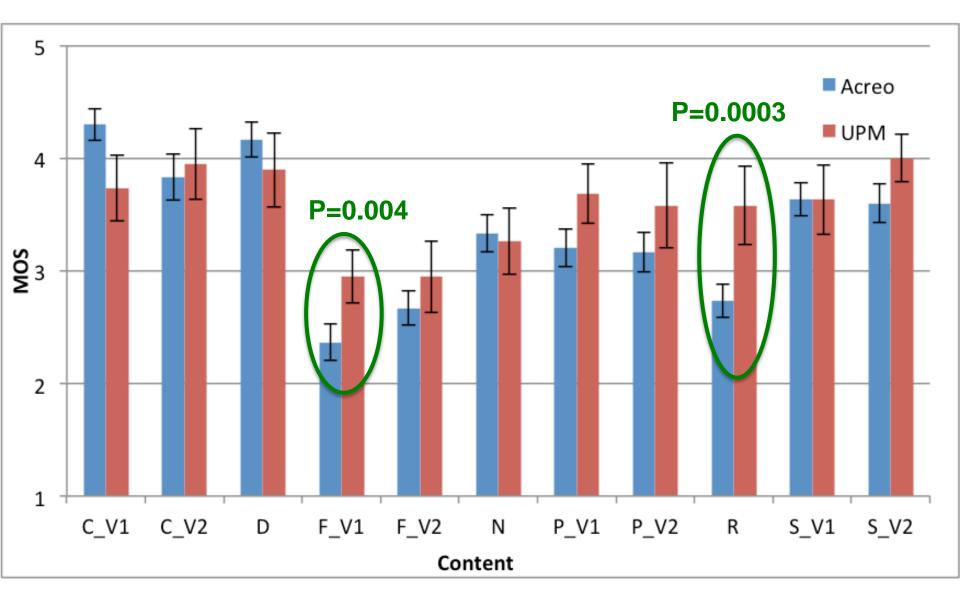
### **Clustering based on Spatial and Temporal Activity**



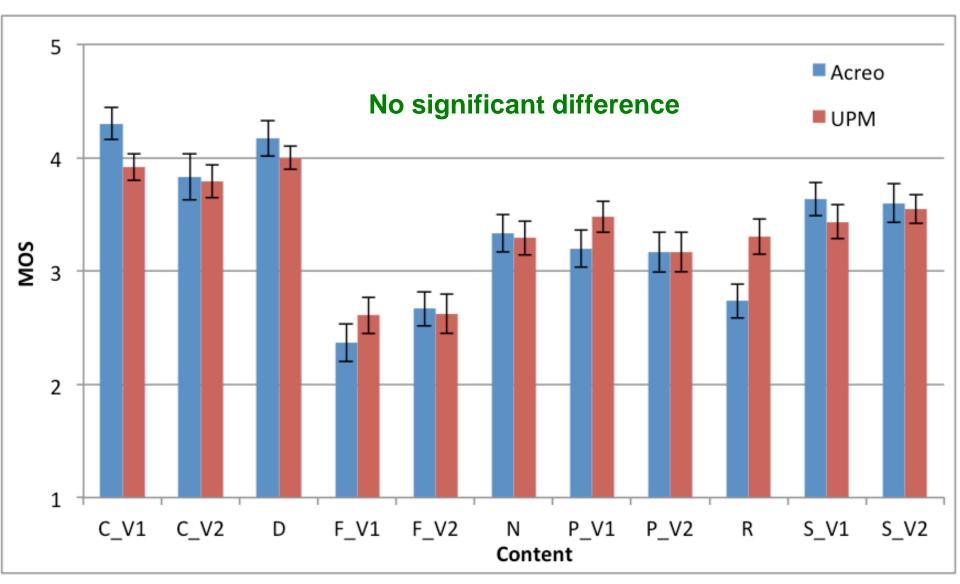
### **Earlier Subjective results**



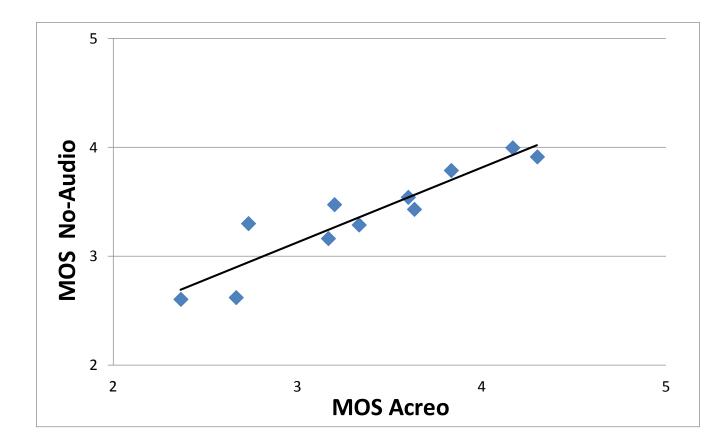
### Impact of evaluation methodology: Acreo vs. UPM-Audio



### Impact of evaluation methodology: Acreo vs. UPM-NoAudio

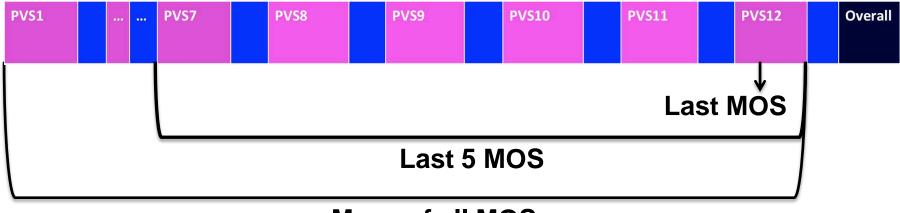


### Estimating overall QoE from individual MOS



# Estimating overall QoE from individual MOS?

Pearson Correlation	Overall	Last 5 MOS	Last MOS	Mean of All MOS
Acreo-Audio	0.79	0.66	0.54	0.81
Acreo-NoAudio	0.93	0.70	0.71	0.90



#### Mean of all MOS

### Discussion

On subjective evaluation methodology

- Effect of rating individual events on given overall scores (UPM experiments)
- Influence of language and culture
  - Big difference in Spanish language videos (Acreo vs. UPM)
  - High similarity of Acreo and No-Audio
    - More focus on visual quality
- Mean MOS of individual event: a good correlation with overall quality
  - Higher correlation in Audio-NoAudio

### Discussion

On objective characterization of adaptive streaming videos

- A selection of tools can be used
- Depends on the content the performance of the tools can be different
  - Football content due to the characteristics of the original video



- Mean of the MOS of individual events: a good predictor of the overall MOS for the full length 6 minute video
- Low performance of objective characterization (tools) in our dataset
- Clustering videos based on Spatial and Temporal characteristics: increasing the performance
- Best predictor: Full-Reference blockiness

Future works

- Repeating experiment with different content (English language)
- Incorporating other characteristics on perceived quality
  - e.g. scene changes

# Thank you

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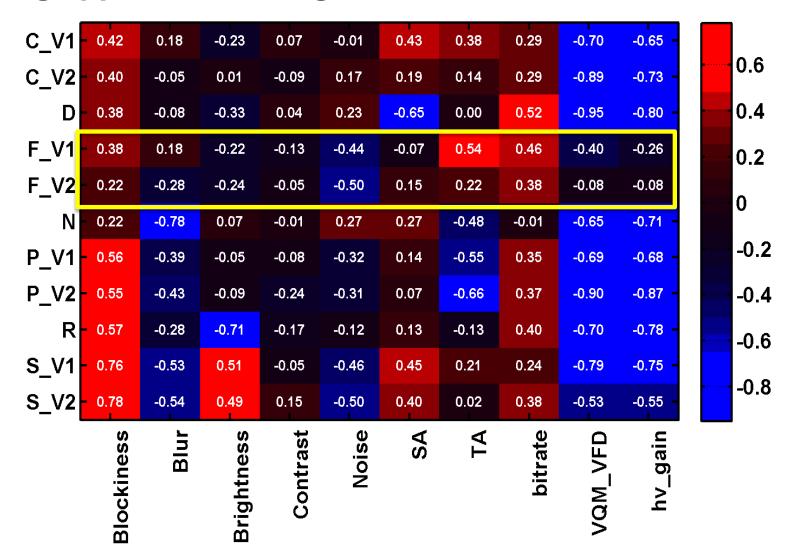


# **Adaptation scenarios**

Status	Possible Client	Code			
	Gradually	2 sec chunk	IGR2		
Increasing quality	(600-1-3-5)	10 sec chunk	IGR10		
	Rapidly	2 sec chunk	IRP2		
	(600-5)	10 sec chunk	IRP10		
Decreasing quality	Gradually	2 sec chunk	DGR2		
	(5-3-1-600)	10 sec chunk	DGR10		
	Rapidly	2 sec chunk	DRP2		
	(5-600)	10 sec chunk	DRP10		
	No degradation- The whole segment at 5Mbps				
Constant quality	No degradation- The whole	N3			
	No degradation- The whole	N1			
	No degradation- The whole	N600			

# SROCC Results between NR and FR measurements for each content

#### Pooling approach: Average of 10% of the lowest values



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