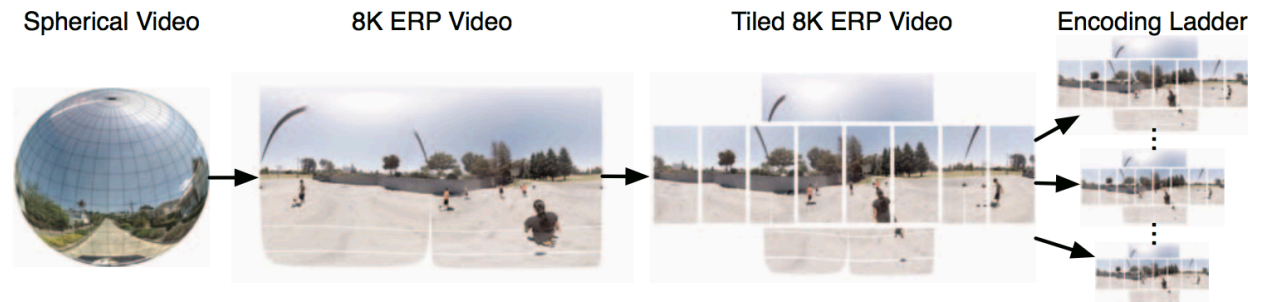


Estimation of Optimal Encoding Ladder for Tiled 360° VR Video in Adaptive Streaming Systems

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Introduction

- 360° video streaming is significantly challenging owing to its resource-intensive encoding and storage requirements
- Adaptive streaming: each 360° video is divided into a set of tiles that includes different bitrate levels of the tiled video
- Representation sets for the video content forms the **encoding ladder**



Cost-Optimal Encoding Ladders

- Tiles affects coding efficiency
 - Tiles have a different level of contribution for the overall 360° video viewing quality
- new encoding ladder configurations are required
- ⇒ focus on the configuration of cost-optimal encoding ladders in adaptive streaming systems
- considering both the provider's and client's perspective
 - develop an encoding ladder estimation method for tiled 360° video streaming

System Models: Distortion Modeling

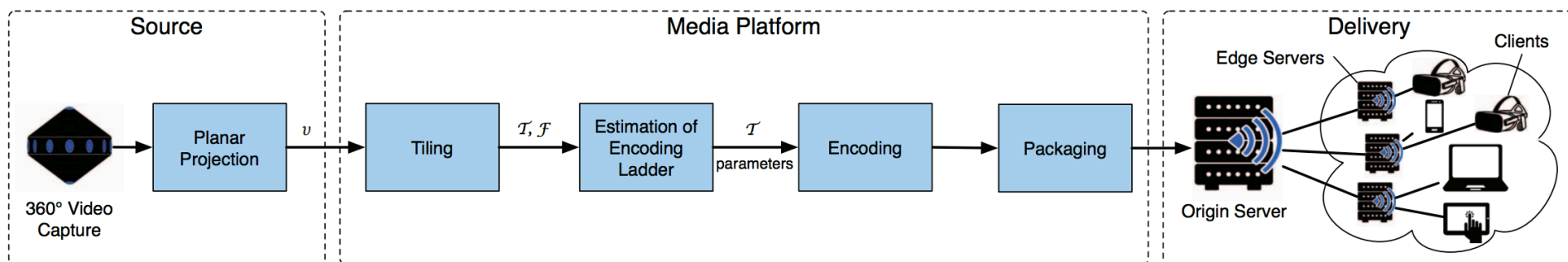
- Distortion function

$$FT_{ogB} = k_{og} Z_B^{\Omega_{og}} + \Phi_{og}$$

- Noise power for the i -th representation of the j -th tile

$$d_{ij} = \frac{\sum_{x \in W} \sum_{y \in H} ((t_j(x, y) - \tilde{t}_{ij}(x, y))^2 q_j(x, y))}{\sum_{x \in W} \sum_{y \in H} q_j(x, y)}$$

$$q_j(x, y) = \cos \frac{(y + 0.5 - H/2)\pi}{H}$$



System Models: Cost Modeling

- Encoding cost $c_{ij}^e = \begin{cases} \mu_e, & r_{ij} \leq 720p \\ 2\mu_e, & 720p < r_{ij} \leq 1080p \\ 4\mu_e, & 1080p < r_{ij} \leq 4K \\ 8\mu_e, & 4K < r_{ij} \leq 8K \end{cases}$
- Storage cost $c_{ij}^s = \mu_s b s_{ij}$

Resolution \mathcal{G}		g_1						g_2						g_3					
Model		Distortion			Data size			Distortion			Data size			Distortion			Data size		
		k	Ω	Φ	k	Ω	Φ	k	Ω	Φ	k	Ω	Φ	k	Ω	Φ	k	Ω	Φ
Content type \mathcal{O}	o_1	1809	-0.6959	5.649	0.7613	0.9901	52.54	4002	-0.7558	2.723	0.8005	0.9859	52.25	1829	-0.5587	-3.266	0.8264	0.9846	214.9
	o_2	220.1	-0.3583	6.447	0.6467	1.003	29.36	191.9	-0.2763	-5.728	0.6078	1.009	71.15	480.6	-0.3643	-5.728	0.5654	1.015	269
	o_3	820.4	-0.4702	6.2	0.6631	1.001	10.69	643	-0.3825	-2.625	0.6691	1	17.46	616.9	-0.2837	-23.78	0.5943	1.012	203.8

Problem Formulation

- Constraints

- Bandwidth: B_{\min} and B_{\max}
- Computational and storage costs: C_{\max} and S_{\max}
- Encoding rate: minimum step size τ

Profiles:	p_1	p_2	p_3	p_4
B^{\min} (Mbps)	1	3	15	25
B^{\max} (Mbps)	4	20	30	40
Λ	0.25	0.25	0.25	0.25

$$\mathcal{L}^* : \operatorname{argmin}_{\mathcal{L}} \sum_{i \in \mathcal{L}} \sum_{p \in \mathcal{P}} (\gamma c_i + (1 - \gamma) d_i) a_{ip}$$

$$c_i = \sum_{j \in \mathcal{T}} (c_{ij}^e + c_{ij}^s) \quad c_i \in \mathcal{P}$$

$$d_i = \sum_{j \in \mathcal{T}} d_{ij}$$

$$B_p^{\min} \leq b_i a_{ip} \leq B_p^{\max} \quad \forall i \in \mathcal{L} \text{ and } \forall p \in \mathcal{P},$$

$$\sum_{i \in \mathcal{L}} a_{ip} = \lfloor \frac{M \Lambda_p}{\sum_{p \in \mathcal{P}} \Lambda_p} \rfloor \quad \forall p \in \mathcal{P},$$

$$\sum_{p \in \mathcal{P}} a_{ip} \leq 1 \quad \forall i \in \mathcal{L},$$

$$\sum_{i \in \mathcal{L}} \sum_{p \in \mathcal{P}} s_i a_{ip} \leq S^{\max},$$

$$\sum_{i \in \mathcal{L}} \sum_{p \in \mathcal{P}} c_i a_{ip} \leq C^{\max},$$

$$\frac{b_i a_{ip}}{b_n^*} \geq \tau, \quad \forall i \in \mathcal{L}, \forall n \in \mathcal{L}^* \text{ and } \forall p \in \mathcal{P}.$$

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Experimental Setup

- N=10
- H.264/AVC
- Quality: WS-MSE and WS-PSNR
- Baselines: one-size-fits-all encoding ladders
- ILP using Pyomo

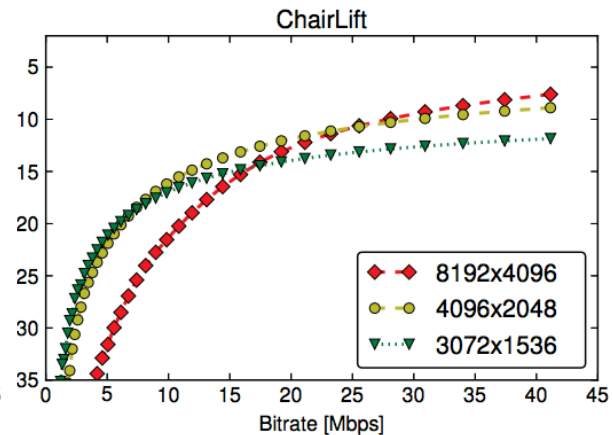
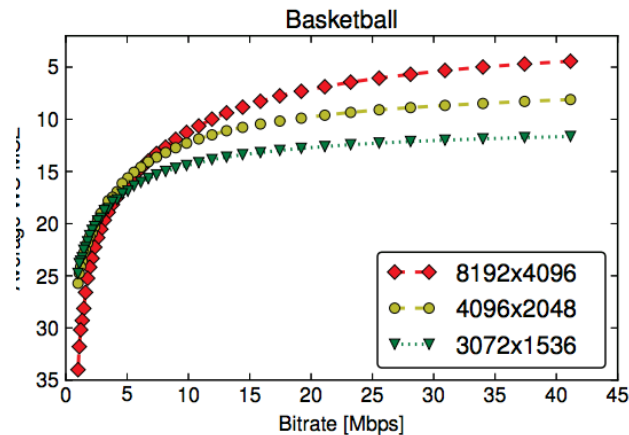
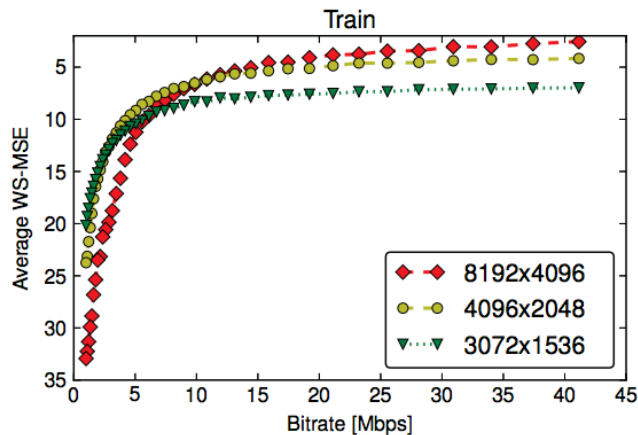
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Sequence	f_{spa}	f_{tmp}	\mathcal{O}
<i>Train</i>	0.977	0.065	
<i>Stitched_left_Dancing360_8K</i>	0.884	0.110	\mathcal{O}_1
<i>Basketball</i>	0.843	0.090	
<i>KiteFlite</i>	0.861	0.090	\mathcal{O}_2
<i>ChairLift</i>	0.789	0.212	
<i>SkateboardInLot</i>	0.827	0.521	\mathcal{O}_3

Apple [15]		Axinom [21]		Netflix [16]	
Z (Mbps)	$W \times H$	Z (Mbps)	$W \times H$	Z (Mbps)	$W \times H$
45	8192 × 4096	45	8192 × 4096	43	8192 × 4096
30	8192 × 4096	30	8192 × 4096	30	4096 × 2048
20	4096 × 2048	21	4096 × 2048	23.5	4096 × 2048
11	3072 × 1536	12	3072 × 1536	17.5	3072 × 1536

RD Performance Gain

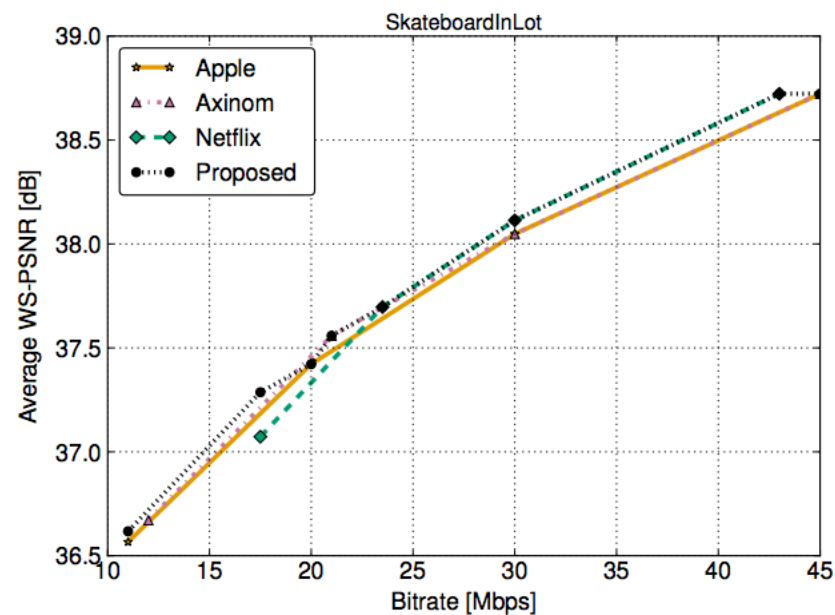
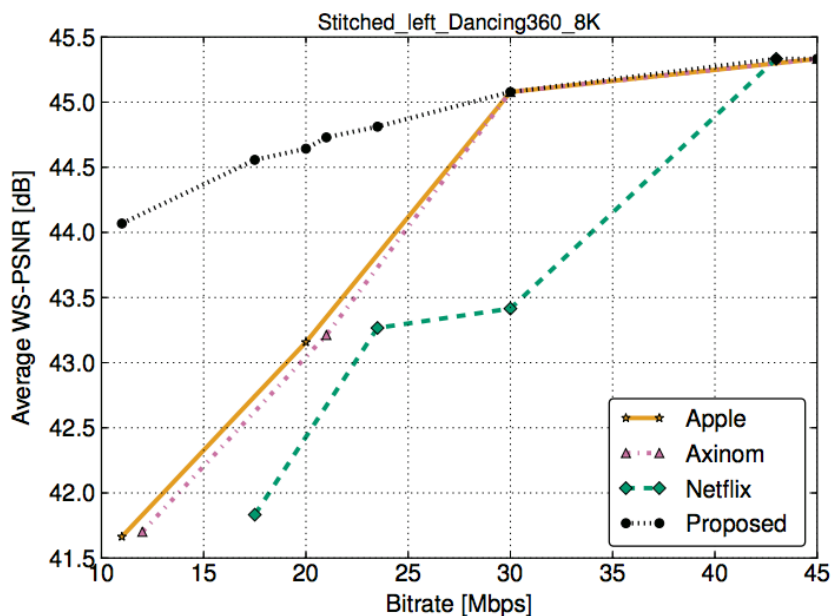
- High diversity in video content characteristics
- One-size-fits-all schemes cannot provide cost-optimal and high-quality streaming performances



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• $\gamma = 0$ $\mathcal{L}^* : \operatorname{argmin}_{\mathcal{L}} \sum_{i \in \mathcal{L}} \sum_{p \in \mathcal{P}} (\gamma c_i + (1 - \gamma) d_i) a_{ip}$



BD Rate (%)

Sequence v	Streaming vendor		
	Apple	Axinom	Netflix
<i>Stitched_left_Dancing360_8K</i>	-5.557	-5.885	-69.253
<i>KiteFlite</i>	-13.876	-14.436	-69.178
<i>SkateboardInLot</i>	-1.673	-1.701	-1.155

Experiment Scenario

- Constraints of S_{\max} and C_{\max} are 8000, $\tau = 1.2$, and $M = 12$

Sequence v	γ	Representation i											
		1	2	3	4	5	6	7	8	9	10	11	12
<i>Stitched_left_Dancing360_8K</i>	0.0	($g_1,1.47$)	($g_1,1.78$)	($g_1,2.15$)	($g_1,3.8$)	($g_1,4.6$)	($g_1,5.6$)	($g_2,10.84$)	($g_2,13.11$)	($g_2,15.87$)	($g_2,28.11$)	($g_3,34.01$)	($g_3,41.15$)
	0.1	($g_2,1.34$)	($g_2,1.61$)	($g_2,1.95$)	($g_2,2.60$)	($g_3,3.14$)	($g_3,3.80$)	($g_3,6.12$)	($g_3,7.40$)	($g_3,8.96$)	($g_3,17.45$)	($g_3,21.12$)	($g_3,25.55$)
	0.5	($g_2,1.00$)	($g_2,1.21$)	($g_2,1.47$)	($g_2,2.36$)	($g_3,2.86$)	($g_3,3.46$)	($g_3,6.12$)	($g_3,7.40$)	($g_3,8.96$)	($g_3,17.45$)	($g_3,21.12$)	($g_3,25.55$)
<i>KiteFlite</i>	0.0	($g_1,1.47$)	($g_1,1.78$)	($g_2,2.15$)	($g_2,3.80$)	($g_2,4.60$)	($g_3,5.56$)	($g_3,10.84$)	($g_3,13.11$)	($g_3,15.87$)	($g_3,28.11$)	($g_3,34.01$)	($g_3,41.15$)
	0.1	($g_1,1.47$)	($g_1,1.78$)	($g_2,2.15$)	($g_2,3.80$)	($g_2,4.60$)	($g_3,5.56$)	($g_3,6.73$)	($g_3,8.14$)	($g_3,9.85$)	($g_3,17.45$)	($g_3,21.12$)	($g_3,25.55$)
	0.5	($g_1,1.00$)	($g_1,1.21$)	($g_1,1.47$)	($g_2,2.36$)	($g_2,2.86$)	($g_2,3.46$)	($g_3,6.12$)	($g_3,7.40$)	($g_3,8.96$)	($g_3,17.45$)	($g_3,21.12$)	($g_3,25.55$)
<i>SkateboardInLot</i>	0.0	($g_1,1.47$)	($g_1,1.78$)	($g_1,2.15$)	($g_1,3.80$)	($g_1,4.60$)	($g_1,5.56$)	($g_2,10.84$)	($g_2,13.11$)	($g_2,15.87$)	($g_2,28.11$)	($g_3,34.01$)	($g_3,41.15$)
	0.1	($g_1,1.47$)	($g_1,1.78$)	($g_1,2.15$)	($g_1,2.86$)	($g_1,3.46$)	($g_1,4.18$)	($g_1,6.12$)	($g_1,7.40$)	($g_1,8.96$)	($g_1,17.45$)	($g_2,21.12$)	($g_2,25.55$)
	0.5	($g_1,1.21$)	($g_1,1.47$)	($g_1,1.78$)	($g_1,2.36$)	($g_1,2.86$)	($g_1,3.46$)	($g_1,6.12$)	($g_1,7.40$)	($g_1,8.96$)	($g_2,17.45$)	($g_2,21.12$)	($g_2,25.55$)

Sequence v	Δcost (%)		$\Delta\text{distortion}$ (%)	
	$\gamma = 0.1$	$\gamma = 0.5$	$\gamma = 0.1$	$\gamma = 0.5$
<i>Stitched_left_Dancing360_8K</i>	37.463	39.683	-13.628	-42.914
<i>KiteFlite</i>	33.165	39.206	-9.564	-25.326
<i>SkateboardInLot</i>	37.214	38.884	-8.977	-15.26

Conclusion

- A novel encoding ladder estimation method for tiled 360° video streaming systems, considering both the provider's and client's perspectives
- The developed system included classification of the content type, distortion modeling, cost modeling, and problem formulation
- Achieved significant bitrate savings compared to the one-size-fits-all encoding ladders
- Automatically find cost-optimal encoding ladders using several practical constraint