

# An In-depth Study of Bandwidth Allocation across Media Sources in Video Conferencing

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# Video Conferencing Applications (VCAs)



Online  
Conferences



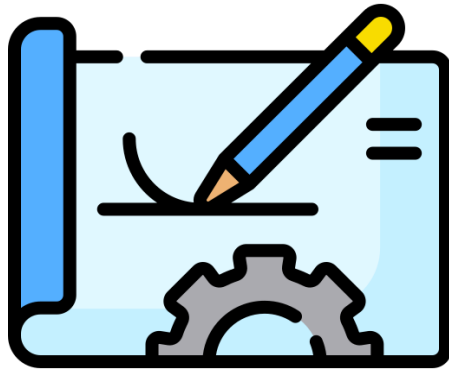
Online  
Education



Online  
Discussion

# Characterizing VCAs: Existing Work

*e.g., Xu et al. [IMC'12], Chang et al. [IMC'21], MacMillan et al. [IMC'21], Saini et al. [JAZ'23], etc.*



System Design



Quality-of-Experience (QoE)



Network Utilization



- Limited media sources: only audio and/or video stream(s)
- Lack of a quantitative QoE assessment

# Characterizing VCAs: Research Problems

- How do VCAs allocate bandwidth across diverse media sources?
  - When bandwidth is limited
- What are the QoE implications of different bandwidth allocation strategies?
- Can we develop a unified QoE model for modern VCAs?



# Our Contribution



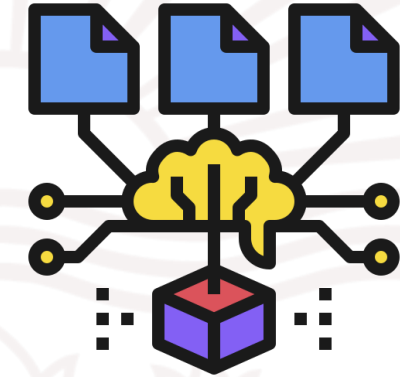
## A measurement study

- Media sources: audio, video, screen
- Bandwidth allocation strategy
- Zoom case study



## A large-scale user study

- 800 participants & 45,000+ ratings



## A generic QoE model

- Across VCAs, scenarios, media sources
- VCAs QoE evaluation



# Challenges

- Commercial VCAs are closed-source
  - How to efficiently acquire QoE metrics, e.g., data rate, resolution, latency, *etc.*
- User study design
  - How to determine a representative subset of bandwidth allocation samples
- QoE modeling across VCAs/scenarios/media sources
  - How to ensure the generalizability and robustness

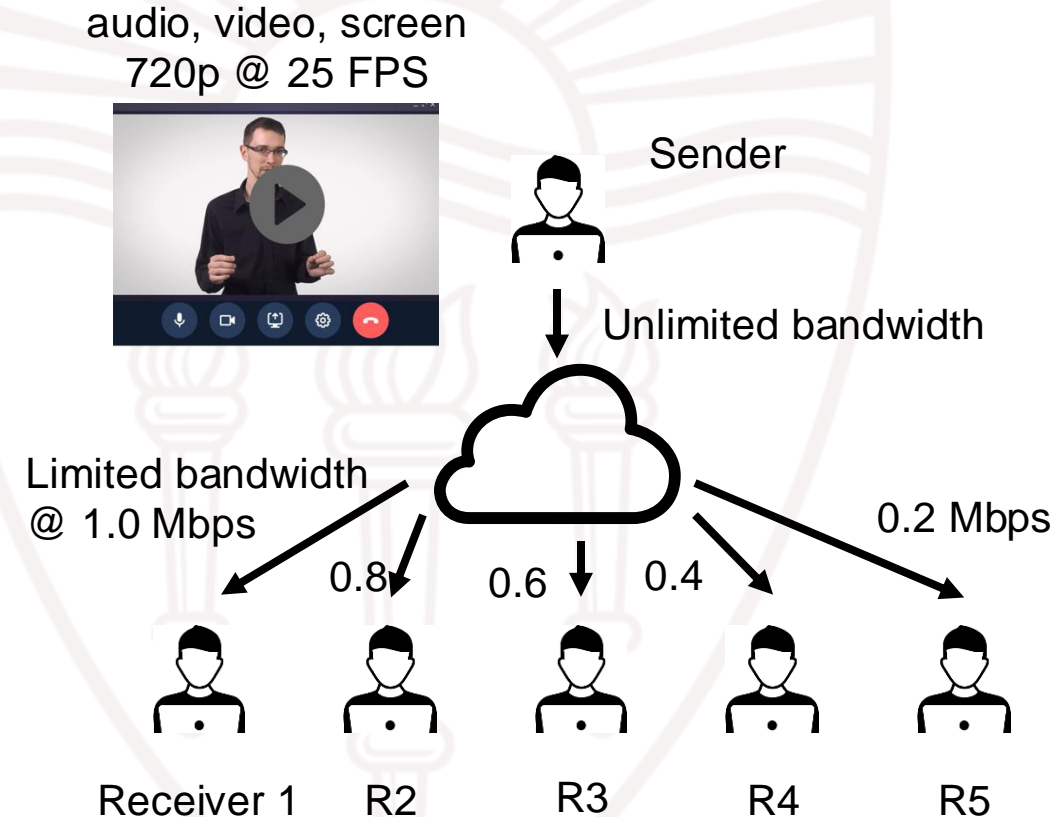
# Understanding Bandwidth Allocation Strategy

- Methodology
  - 4 {audio, video, screen} combination scenarios
  - 3 VCAs: Zoom, Webex, Google Meet
  - More details are in our paper

	Audio	Video	Screen
Scenario 1	✓	(Full-Screen)	
Scenario 2	✓		(Full-Screen)
Scenario 3	✓	(Thumbnail)	(Full-Screen)
Scenario 4	✓	(Half-Screen)	(Half-Screen)

## Key takeaway

- 3 VCAs share the **same prioritization** when allocating bandwidth: **audio > video > screen**



# Case Study: Zoom

- Video transmission

	Sender (Unlimited)	Receiver1 (Unlimited)	Receiver2 (750Kbps)	Receiver3 (500Kbps)	Receiver4 (250Kbps)	Receiver5 (150Kbps)
Data rate (Kbps)	1158±120	883±130	647±85	453±44	218±35	144±20
Framerate (FPS)	21±3	21±3	13±2	10±2	8±1	7±2
Resolution	360p	360p	360p	360p	180p	144p
SSIM		0.89	0.87	0.84	0.82	0.8
VMAF		91	73	66	41	22

Ensure a better framerate by sacrificing resolution and SSIM → Fluency

- Screen transmission

	Sender (Unlimited)	Receiver1 (Unlimited)	Receiver2 (750Kbps)	Receiver3 (500Kbps)	Receiver4 (250Kbps)	Receiver5 (150Kbps)
Data rate (Kbps)	1482±230	1439±230	547±150	326±85	168±40	118±20
Framerate (FPS)	10±2	10±2	4±1	2±1	1±1	<1
Resolution	720p	720p	720p	720p	720p	720p
SSIM		0.91	0.89	0.88	0.87	0.85
VMAF		90	86	82	75	64

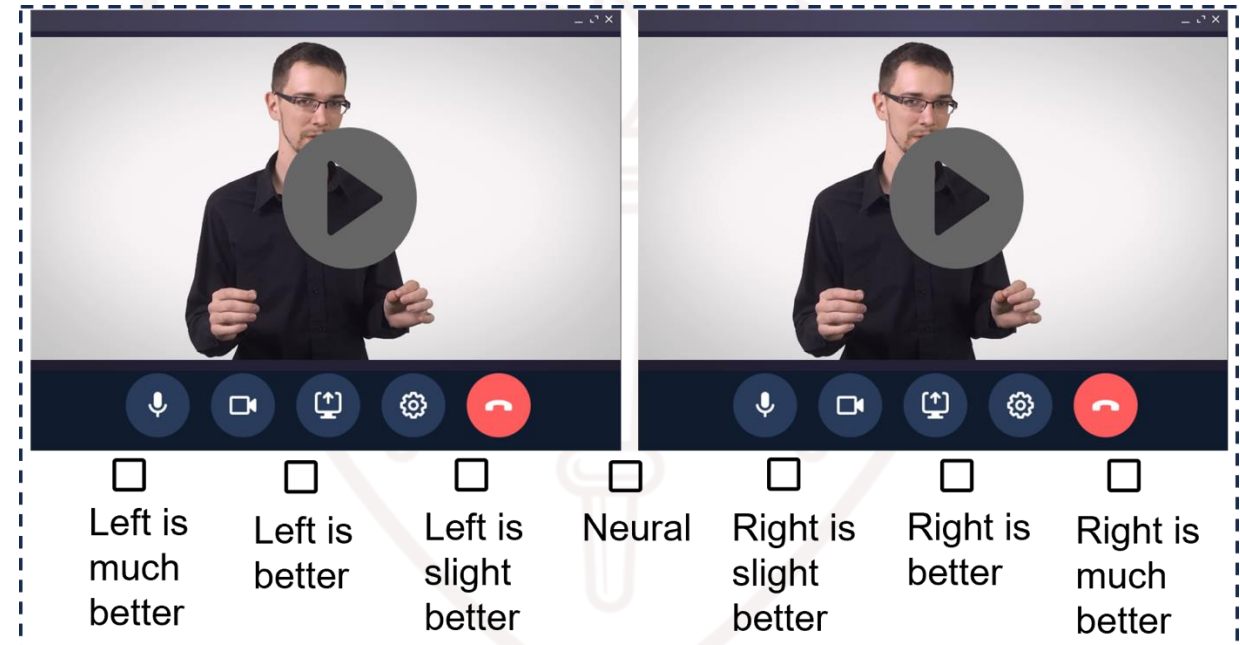
Ensure a high resolution by decreasing the framerate → Clarity



# QoE Modeling: User Study

- Setup
  - {27, 27, 243, 243} bitrate combination samples
    - For Scenario {1, 2, 3, 4}
  - Each clip > 15s
- 800 participants & 45,000+ ratings
  - Ratings: accumulated score
  - Translate to QoE values: PageRank
- More details are in the paper

	Audio	Video	Screen
<b>Scenario 1</b>	✓	(Full-Screen)	
<b>Scenario 2</b>	✓		(Full-Screen)
<b>Scenario 3</b>	✓	(Thumbnail)	(Full-Screen)
<b>Scenario 4</b>	✓	(Half-Screen)	(Half-Screen)



# QoE Modeling: Model Design

- General & scenario-specific QoE model
  - Parameters
    - General: Audio + Video + Screen + Bandwidth + Others
    - Scenario 1: Audio + Video + Bandwidth
    - Scenario 2: Audio + Screen + Bandwidth
    - Scenario 3 & 4: Audio + Video + Screen + Bandwidth

Category	Parameter
Audio	[audio bitrate]
Video	[video resolution, video framerate]
Screen	[screen resolution, screen framerate]
Bandwidth	[overall bitrate]
Others	[ratio of window size between video and screen]

**Table 7: Input parameters of each media source.**

# QoE Modeling: Model Evaluation

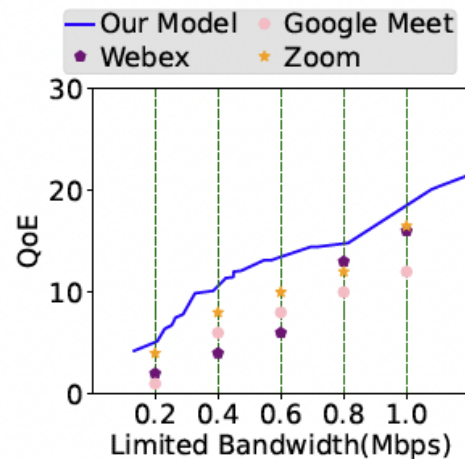
- Dataset: from user study
  - Training : Evaluation = 8 : 2
- Investigated ML models
  - Logistic Regression
  - Random Forest Regression (RF)
  - Gradient Boosting Decision Tree (GBDT)
  - Multi-layer Perceptron Regression (MLP)

- MLP has the best performance
- Our QoE modeling is **generic** across diverse scenarios/media sources

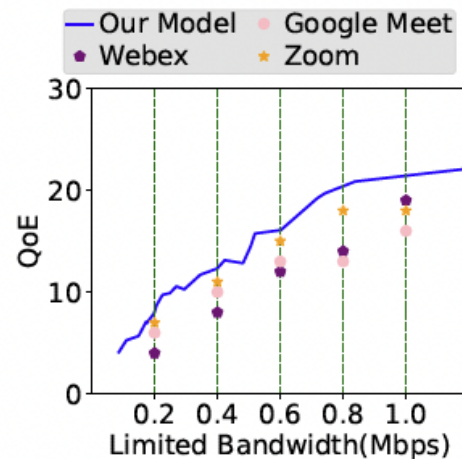
Scenario	Logistic			RF			GBDT			MLP		
	MAE	RMSE	Accuracy	MAE	RMSE	Accuracy	MAE	RMSE	Accuracy	MAE	RMSE	Accuracy
Scenario 1	0.12	0.92	82.13%	<u>0.12</u>	0.86	81.3%	0.15	0.93	82.1%	<u>0.12</u>	<u>0.83</u>	<u>84.61%</u>
Scenario 2	0.12	0.96	81.90%	0.13	0.87	82.15%	0.16	0.97	82.20%	<u>0.11</u>	<u>0.85</u>	<u>84.55%</u>
Scenario 3	0.15	2.81	78.79%	<u>0.13</u>	2.50	81.12%	0.16	2.87	81.04%	<u>0.13</u>	<u>2.19</u>	<u>84.37%</u>
Scenario 4	0.14	2.92	80.48%	<u>0.11</u>	2.28	82.06%	0.15	2.67	81.49%	0.12	<u>2.24</u>	<u>82.62%</u>
General	0.19	4.06	70.51%	0.09	<u>1.75</u>	81.63%	0.13	2.56	81.79%	<u>0.08</u>	1.78	<u>82.86%</u>

# QoE Evaluation of Zoom, Webex, Google Meet

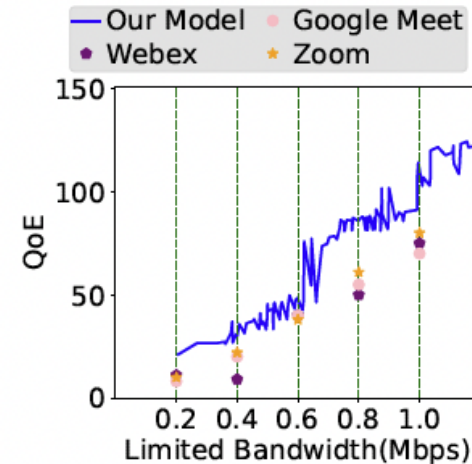
- Setup
  - MLP-based general QoE model
  - Bandwidth: {0.2, 0.4, 0.6, 0.8, 1.0} Mbps
  - Optimal QoE predicted by the model vs. VCA's QoE



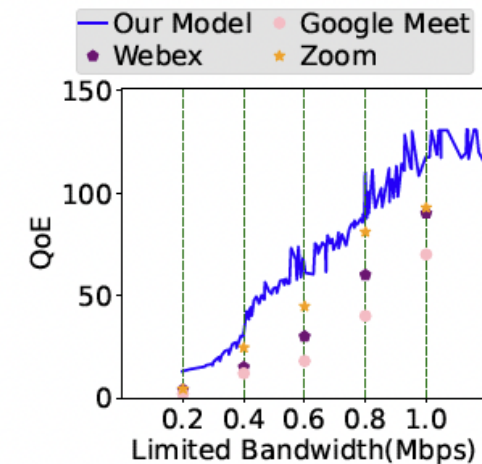
(a) Scenario 1



(b) Scenario 2



(c) Scenario 3



(d) Scenario 4

- **Zoom** consistently achieves higher QoE than Webex and Google Meet in most cases
- **None** of three VCAs achieve optimal QoE

# Summary

- Limitations of existing VCA measurement studies
- Our work
  - A measurement study of VCAs (Zoom, Webex, Google Meet)
    - Bandwidth allocation strategies across {audio, video, screen}
    - Zoom case study
  - A large-scale user study
    - 800 participants & 45,000+ ratings
  - A generic QoE model for VCAs
    - Across VCAs, scenarios, media sources
    - QoE evaluation of VCAs
- Q & A