# Oboe: Auto-tuning Video ABR Algorithms to Network Conditions

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★: Co-primary authors











# Internet Video Streaming Today

- Internet video is delivered over:
  - Heterogeneous networks: WiFi, wired, 3G/4G LTE
  - Highly varying or challenging network conditions



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Quality of experience (QoE) issues are common place



Low quality



Rebuffering

# Internet Video Streaming Today

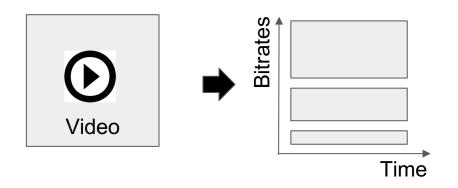
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  - Highly varying or challenging network conditions



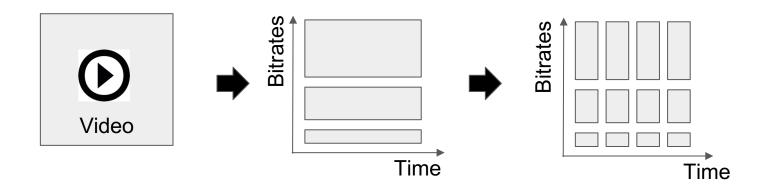
Quality of experience (QoE) issues are common place



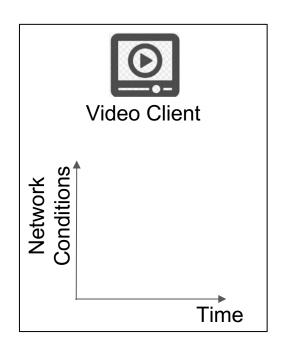


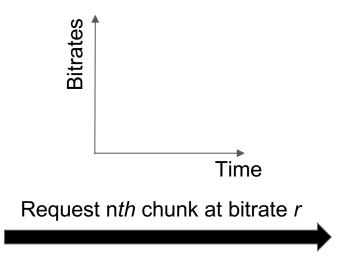


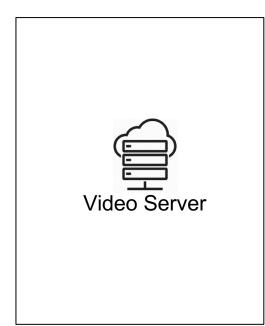
A video clip is encoded with multiple qualities (bitrates)

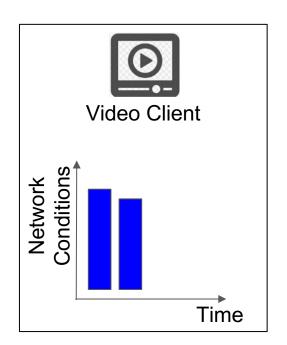


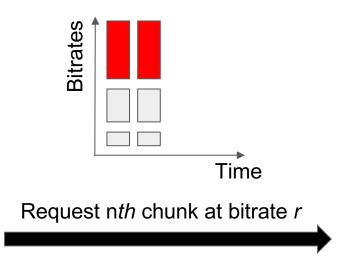
Each bitrate is split into chunks

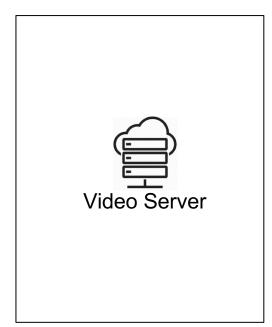


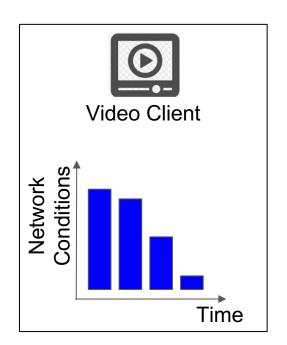


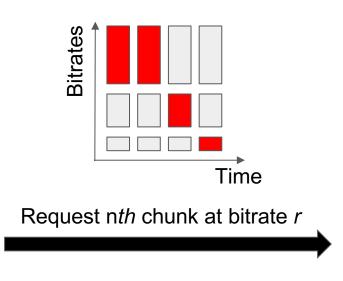


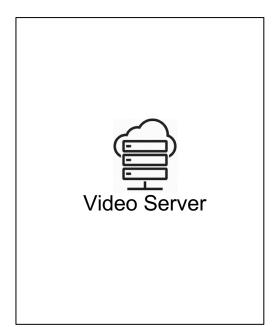


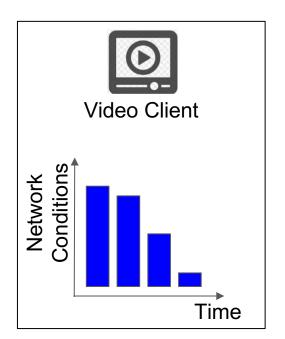


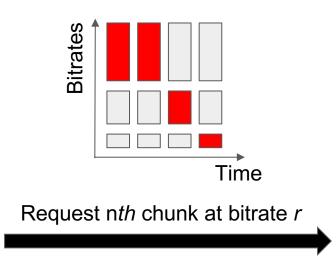


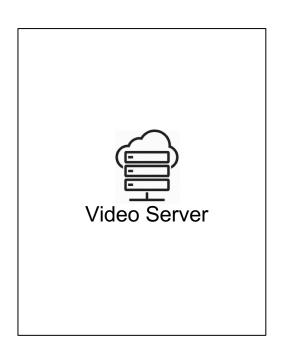




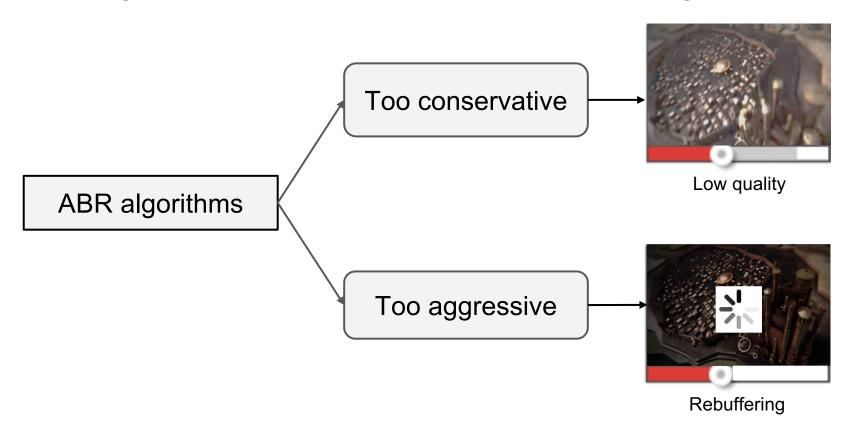




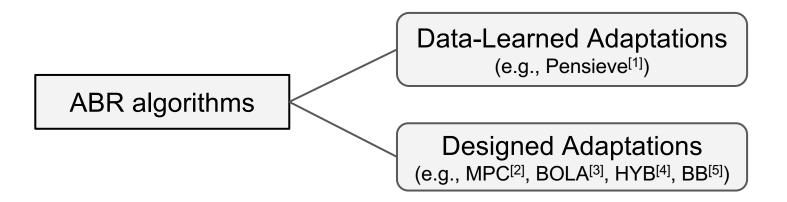




**Adaptive Bitrate Algorithms(ABR)** 



# Background: ABR algorithms



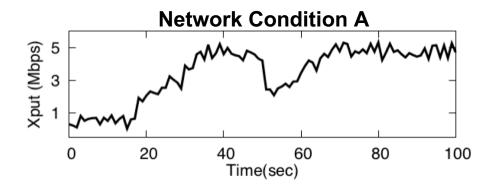
# Performance of Designed Adaptation based ABRs critically depends on configurable parameters

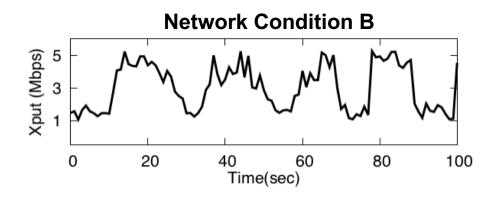
<sup>[1]</sup> Hongzi Mao, et al., SIGCOMM, 2017.

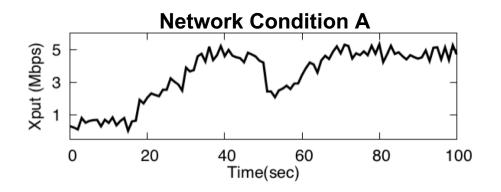
<sup>[2]</sup> Xiaoqi Yin, et al., SIGCOMM, 2015.

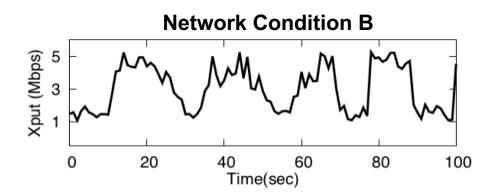
<sup>[3]</sup> Kevin Spiteri, et al., INFOCOM, 2016.

<sup>[4]</sup> An ABR algorithm that's widely used in industry.

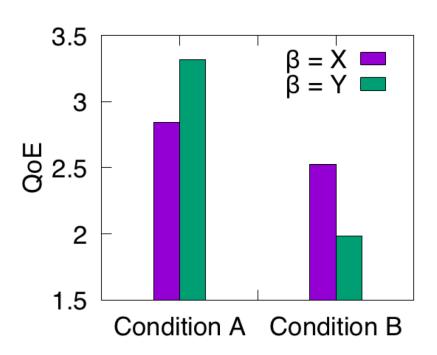




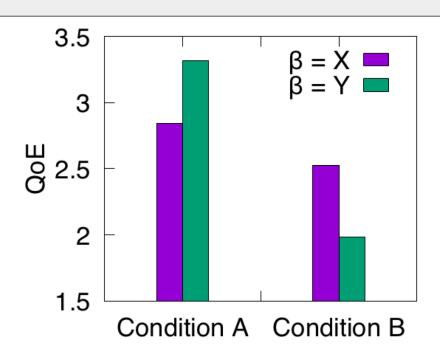




Widely deployed ABR algorithm with parameter  $\beta$ 



# Parameters of ABRs must be set in a manner sensitive to network conditions



# The problem with ABR algorithms

ABR algorithms	Parameter
MPC	Discount factor d
BOLA	Parameter $\gamma$
HYB	Safety margin $\beta$
BB	Reservoir r

ABR algorithms use **fixed parameter value** or **simple heuristic** 

# The problem with ABR algorithms

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ABR algorithms use **fixed parameter value** or **simple heuristic** 

#### Do not perform well across all network conditions

#### Goal of our work

Design a system to make ABR algorithms work better over a wide range of network conditions

# Key Challenges

How to model network conditions?

How to find the best parameter for a given condition?

How to adapt to changes in network conditions?

#### Contributions

How to model network conditions?

Leverage stationarity of network connections

How to find the best parameter for a given condition?

Pre-compute offline

How to adapt to changes in network conditions?

Detect change points online and adjust parameters

#### Contributions

How to model network conditions?

Leveraging stationarity of network connections

# Our system, Oboe improves state-of-art ABRs (MPC, BOLA and HYB) upto 38% and outperforms Pensieve by 24%

Online change point detection and adjusting ABR algorithms in online

# Key Challenges

How to model network conditions?

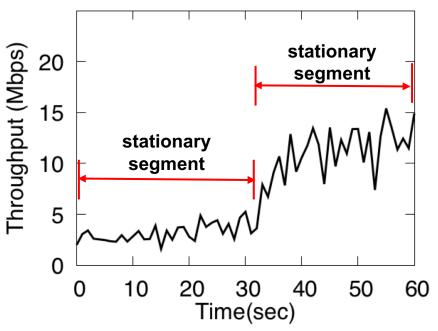
How to find the best parameter for a given condition?

Oboe Offline Stage

How to adapt to changes in network conditions?

Oboe Online Stage

# Modeling network conditions



TCP connection throughput can be modeled as a piecewise stationary<sup>[6-10]</sup> sequence of network states

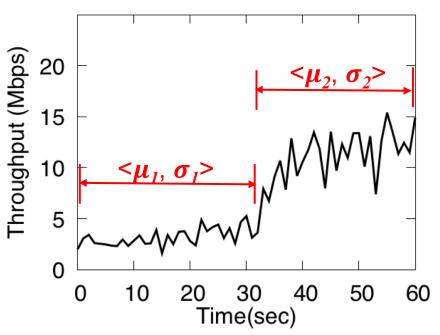
<sup>[6]</sup> Hari Balakrishnan, et. al. SIGMETRICS, 1997

<sup>[7]</sup> James Jobin, et. al. INFOCOM, 2004

<sup>[8]</sup> Dong Lu, et. al. ICDCS, 2005

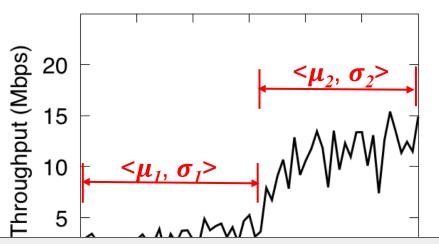
<sup>[9]</sup> Guillaume Urvoy-Keller. PAM, 2005.

# Modeling network conditions



Network state  $\mathbf{s} = \langle \mu_s, \sigma_s \rangle$  where  $\mu_s$  is the mean and  $\sigma_s$  is the standard deviation of throughput

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Network state  $\mathbf{s} = \langle \mu_s, \sigma_s \rangle$ where  $\mu_s$  is the mean and  $\sigma_s$  is the standard deviation of throughput

Key idea: Use the best parameter for each network state

# Key Challenges

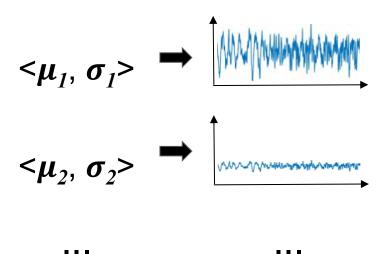
How to model network conditions?

How to find the best parameter for each network state?

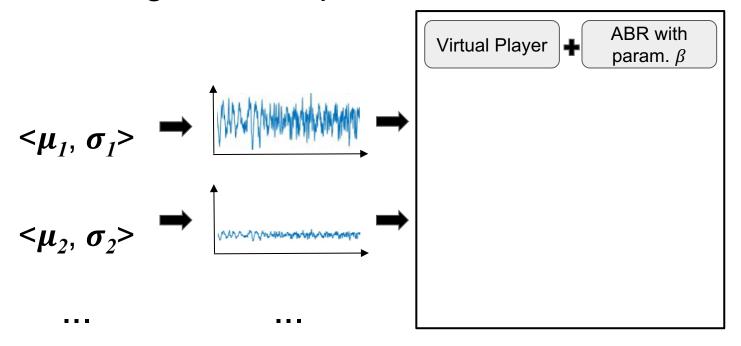
Oboe Offline Stage

How to adapt to changes in network state?

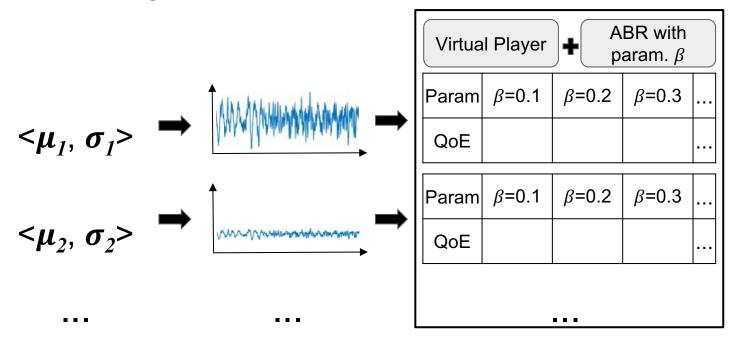
Oboe Online Stage



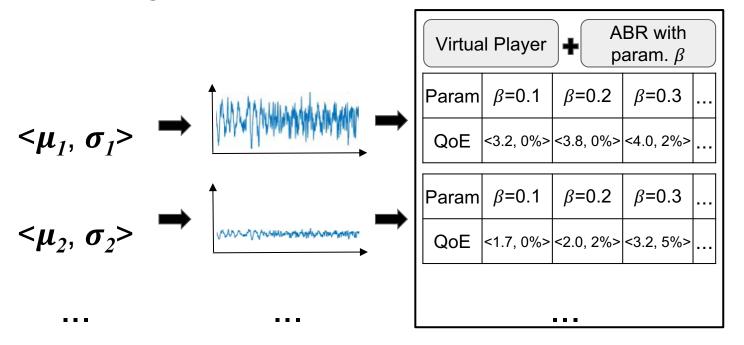
Generate synthetic stationary traces for each network state



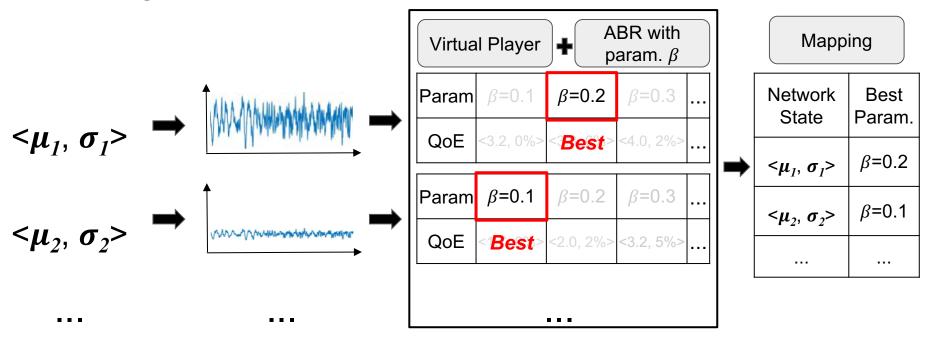
Explore parameter space for each state and get QoE vectors



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Explore parameter space for each state and get QoE vectors



Find the best parameter by vector dominance for each state

# Oboe Offline Stage: Design Questions

- Use real or synthetic traces?
- How to quantize network state space?
- How to reduce the cost of parameter space exploration?
- How to decouple ABR algorithms from Virtual Player?
- How to take publisher preferences into account?

# Key Challenges

How to model network conditions?

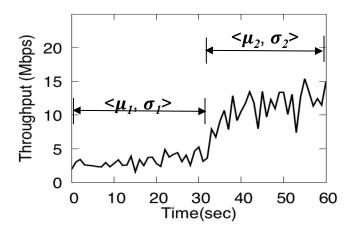
How to find the best parameter for each network state?

Oboe Offline Stage

How to adapt to changes in network state?

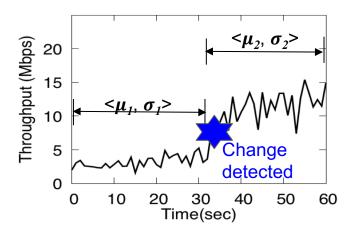
Oboe Online Stage

# Adapting to network state changes



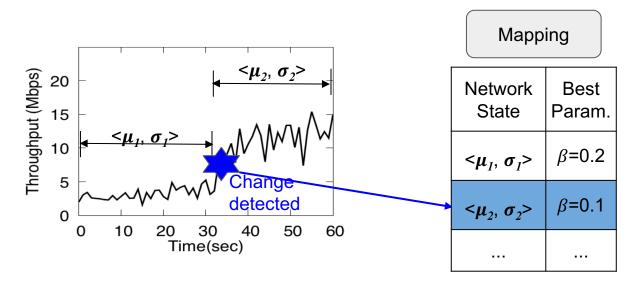
Online change point detection[11] algorithm identifies network throughput distribution changes in real time

# Adapting to network state changes: Online Step 1



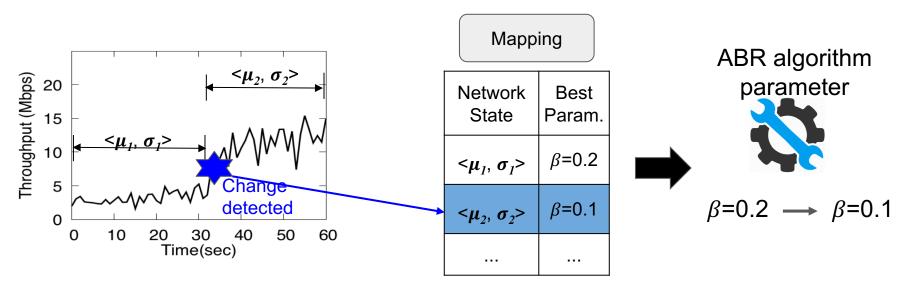
Online change point detection[11] algorithm identifies network throughput distribution changes in real time

# Adapting to network state changes: Online Step 2



Find the best parameter from the mapping for a new state

# Adapting to network state changes: Online Step 3



Reconfigure ABR algorithm parameter

## **Oboe Online Design Questions**

- Why not a simple moving average?
- How many throughput samples to detect changes?
- Are computational overheads acceptable in real time?

# Evaluation methodology

## Comparison

Existing ABRs vs. Existing ABRs + Oboe

#### **QoE Metrics**

- Average Bitrate
- Rebuffering Ratio
- Bitrate change magnitude
- QoE-lin (Linear combination of three metrics)

# Evaluation methodology

### TestBed Setup

#### **Chrome Browser Video Client**



Video player

- Dash.js
- Chrome DevTool API



#### **Apache Video Server**



Video

- 3 min long
- Encoded with 6 bitrates

#### **Dataset**

- 600 throughput traces from real users
- Real users used a desktop or a mobile
- Upto 6 Mbps

## **Various Evaluations**

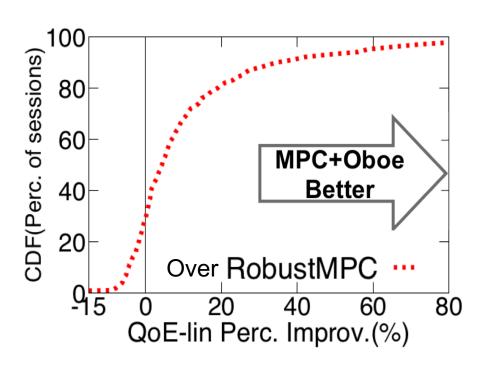
Method	Detail	
ABRs performance	MPC, BOLA, BB, HYB and Pensieve	
Public datasets	HSDPA <sup>[11]</sup> and FCC <sup>[12]</sup>	
Various settings	Live setting and different videos	
Alternative predictors	Ideal predictors on MPC	
Publisher specification	Different rebuffering tolerance	
Pilot Deployment	Partial deployment on AWS	

<sup>[11]</sup> Haakon Riiser, et. al. MMSys, 2013

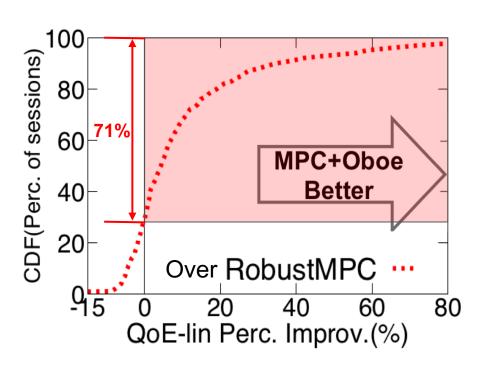
Solves an optimization problem to choose bitrates using predicted throughputs and player buffer occupancy

#### Discount factor d

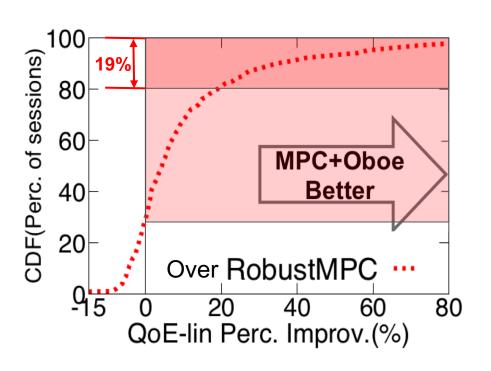
- Reduces a predicted throughput by d to compensate prediction errors
- Simple heuristic based on previous prediction errors



- Improves QoE-lin for 71% of sessions
- For 19% of the sessions,
  more than 20% benefit

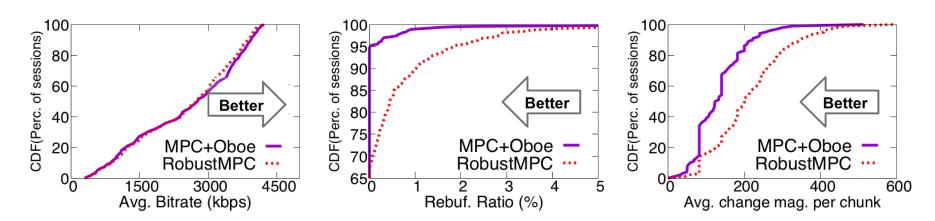


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#### Where does benefit come from over RobustMPC

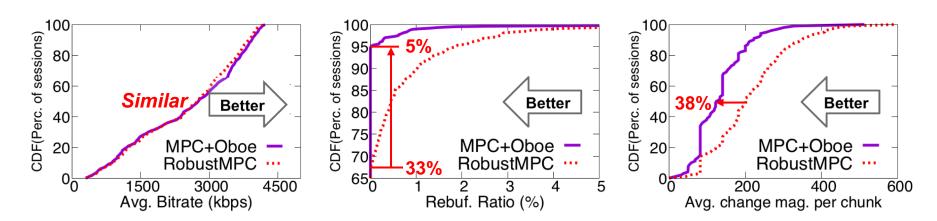


#### Similar average bitrates

Reduces the # of sessions with rebuffering from 33% to 5%

Improves the median per chunk change magnitude by 38%

#### Where does benefit come from over RobustMPC

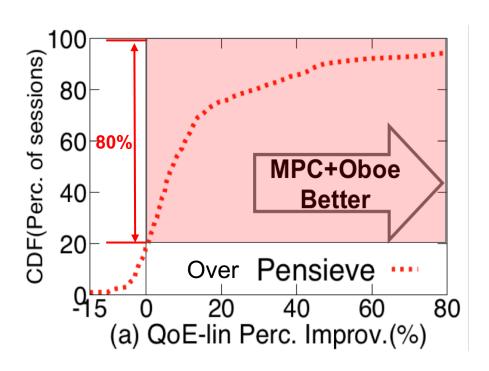


#### Similar average bitrates

Reduces the # of sessions with rebuffering from 33% to 5%

Improves the median per chunk change magnitude by 38%

#### MPC+Oboe vs Pensieve



- Improves QoE-lin for 80% of sessions
- 24% better in average
  QoE-lin

#### Where does benefit come from over Pensieve?

 Hypothesis: Pensieve performs better when it is trained with a constrained throughput range

Model	Trained	Tested	Result
Pen-Specialized	0 - 3 Mbps	0 2 Mbps	Better
Original Pensieve	0 - 6 Mbps	0 - 3 Mbps	

- Pensieve unable to specialize to network conditions
- Oboe specializes parameters for every network state

# Summary

# Oboe is a system to make ABR algorithms work better in a wide range of network conditions

by auto-tuning parameters to current network state

## Oboe is general

- Can be applied to many existing ABRs
- Improves existing ABRs upto 38% in QoE metrics
- Outperforms Pensieve by 24% in average QoE

## Live demo tomorrow in the demo session!

https://github.com/USC-NSL/Oboe

Thanks!