



# A Control-Theoretic Approach for Dynamic Adaptive Video Streaming over HTTP

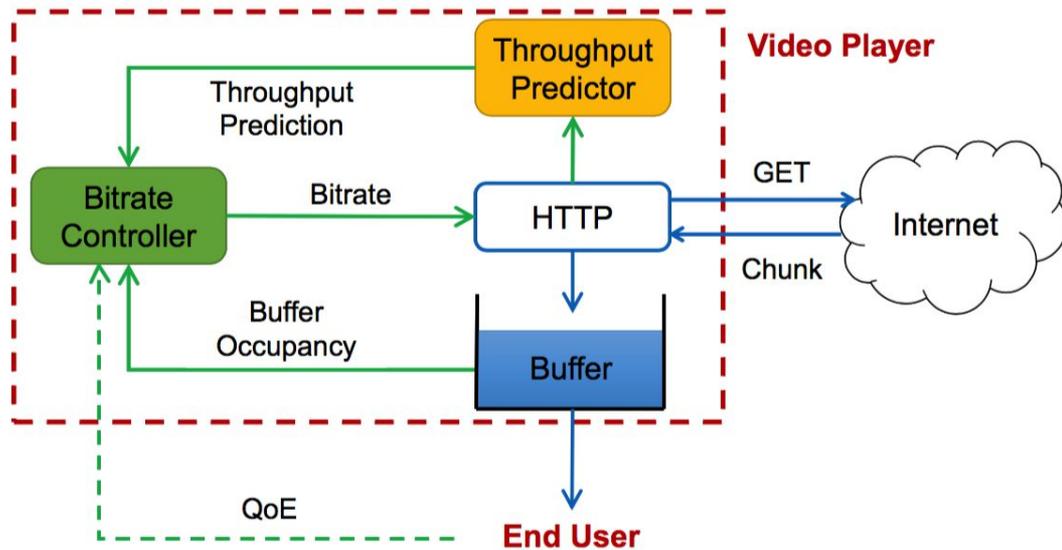
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# + Adaptive Video Player



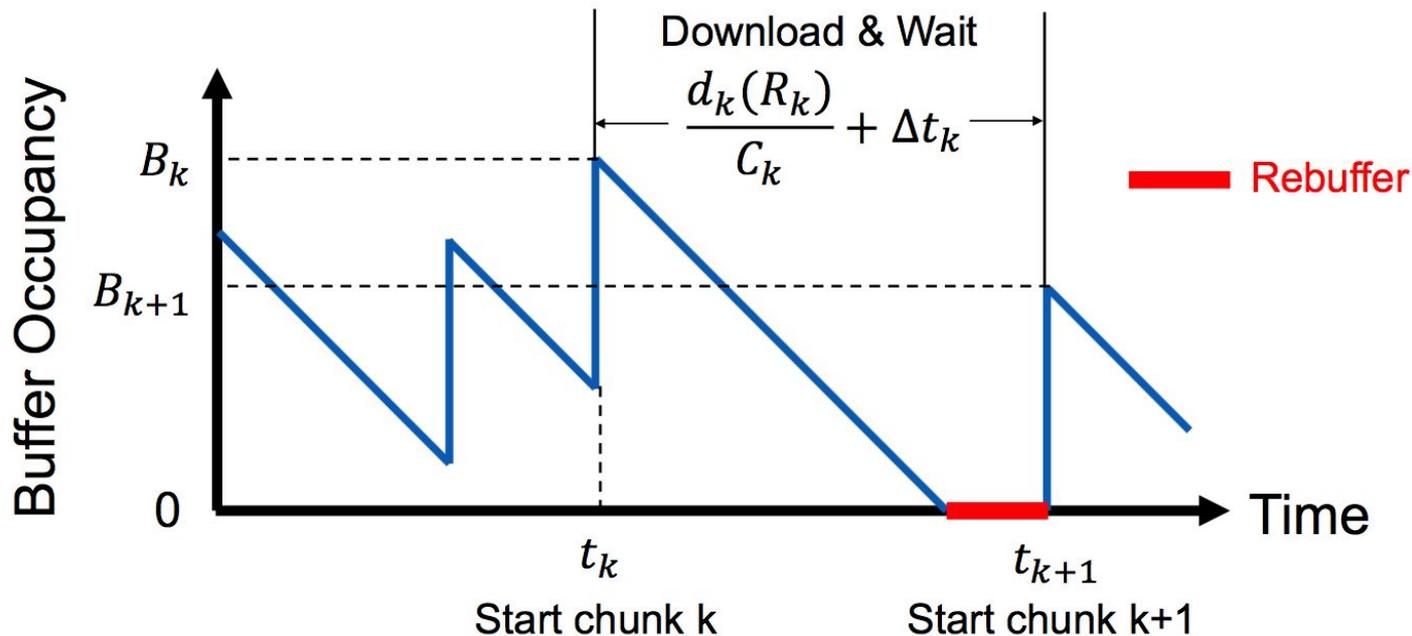
- In the HTTP-based delivery model that predominates today, videos are typically chunked and encoded at different bitrate levels.
- The goal of an adaptive video player is to choose the bitrate level for future chunks to deliver the highest possible QoE
- Bitrate adaptation logic in the client-side video player becomes critical to optimize user experience

# + Video Streaming Model



**Figure 1: Abstract model of DASH players**

# + Buffer Dynamics



**Figure 2: Illustration of buffer dynamics**

## + Background: HTTP-based Adaptive Streaming



- Many conflicting goals:
  - minimizing rebuffering events where the playback buffer is empty and cannot render the video
  - delivering as high a playback bitrate as possible within the throughput constraints
  - minimizing startup delay so that the user does not quit while waiting for the video to load
  - keeping the playback as smooth as possible by avoiding frequent or large bitrate jumps

# + A client-side solution



- Why client-side:
  - Most immediately deployable alternative comparing to
    - network support
    - server-side support
    - lower-layer transport protocols
  - The client is often in the best position to quickly detect performance issues and respond to dynamics.

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# QoE Maximization Problem

- the QoE of video segment 1 through K by a weighted sum of the aforementioned components:

$$\begin{aligned}
 QoE_1^K = & \sum_{k=1}^K q(R_k) - \lambda \sum_{k=1}^{K-1} |q(R_{k+1}) - q(R_k)| \\
 & - \mu \sum_{k=1}^K \left( \frac{d_k(R_k)}{C_k} - B_k \right)_+ - \mu_s T_s \quad (5)
 \end{aligned}$$

Average video quality (points to  $q(R_k)$ )  
 Average quality variations (points to  $|q(R_{k+1}) - q(R_k)|$ )  
 Total rebuffer time (points to  $\mu \sum_{k=1}^K \left( \frac{d_k(R_k)}{C_k} - B_k \right)_+$ )  
 Startup Delay (points to  $\mu_s T_s$ )

# + QoE Maximization Problem

$$\max_{R_1, \dots, R_K, T_s} QoE_1^K \quad (6)$$

$$s.t. \quad t_{k+1} = t_k + \frac{d_k(R_k)}{C_k} + \Delta t_k, \quad (7)$$

$$C_k = \frac{1}{t_{k+1} - t_k - \Delta t_k} \int_{t_k}^{t_{k+1} - \Delta t_k} C_t dt, \quad (8)$$

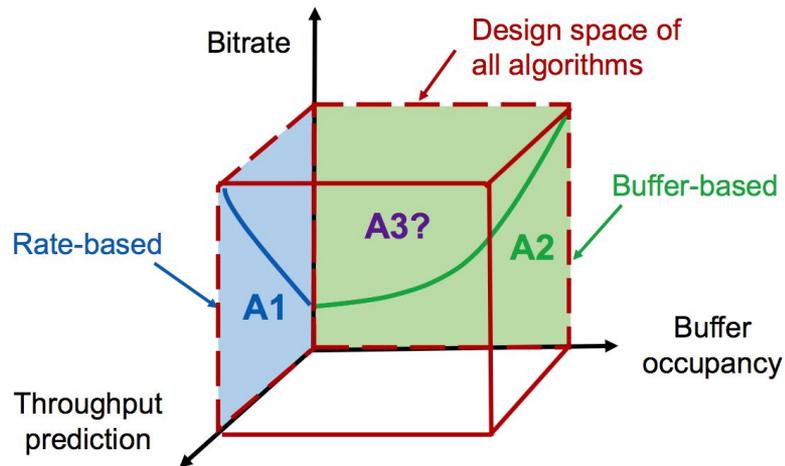
$$B_{k+1} = \left( \left( B_k - \frac{d_k(R_k)}{C_k} \right)_+ + L - \Delta t_k \right)_+, \quad (9)$$

$$B_1 = T_s, \quad B_k \in [0, B_{max}] \quad (10)$$

$$R_k \in \mathcal{R}, \quad \forall k = 1, \dots, K. \quad (11)$$

**Figure 3: Formulation for QoE maximization ( $QOE\_MAX_1^K$ ) subject to buffer and throughput dynamics.**

# + QoE Bitrate Adaption



**Figure 4: Design space of algorithms for the video adaptation problem: Most current approaches choose the bitrate as a function of only one variable; e.g., A1 is rate-based (RB) while A2 is buffer-based (BB).**

# + Model Predictive Control



- Why MPC?
  - PID control is designed to work in continuous time and state space and using it in a highly discrete system such as ours may result in performance degradation or instability
  - with MPC (MDP) we could consider formulating the throughput and buffer state transition as Markov processes

# + Basic MPC Algorithm

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**Algorithm 1** Video adaptation workflow using MPC

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- 1: Initialize
  - 2: **for**  $k = 1$  to  $K$  **do**
  - 3:     **if** player is in startup phase **then**
  - 4:          $\hat{C}_{[t_k, t_{k+N}]} = \text{ThroughputPred}(C_{[t_1, t_k]})$
  - 5:          $[R_k, T_s] = f_{mpc}^{st} \left( R_{k-1}, B_k, \hat{C}_{[t_k, t_{k+N}]} \right)$
  - 6:         Start playback after  $T_s$  seconds
  - 7:     **else if** playback has started **then**
  - 8:          $\hat{C}_{[t_k, t_{k+N}]} = \text{ThroughputPred}(C_{[t_1, t_k]})$
  - 9:          $R_k = f_{mpc} \left( R_{k-1}, B_k, \hat{C}_{[t_k, t_{k+N}]} \right)$
  - 10:     **end if**
  - 11:     Download chunk  $k$  with bitrate  $R_k$ , wait till finished
  - 12: **end for**
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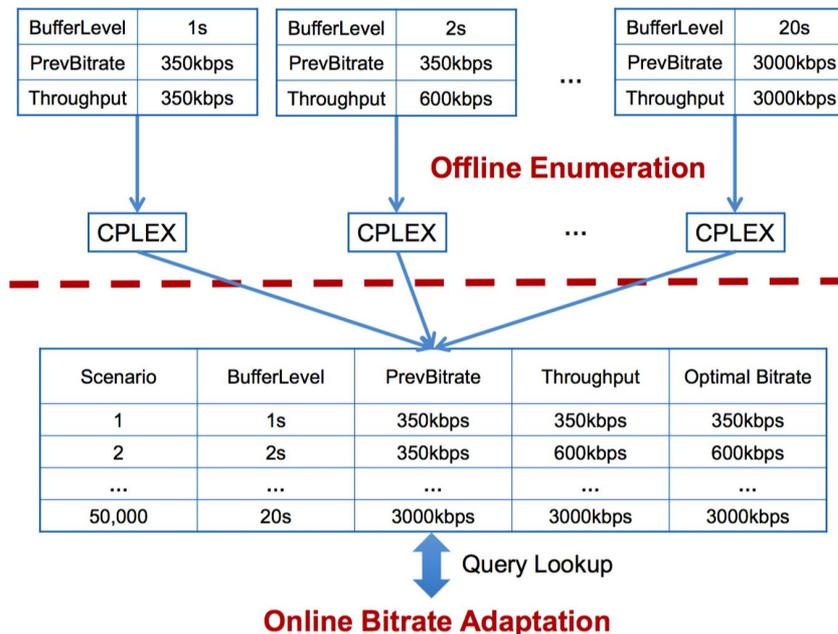
# + Robust MPC



- The basic MPC algorithm assumes the existence of an accurate throughput predictor.
- Robust MPC essentially optimizes the worst-case QoE assuming that the actual throughput can take any value in a range in contrast to a point estimate
- Robust MPC makes conservative estimation

# + FastMPC

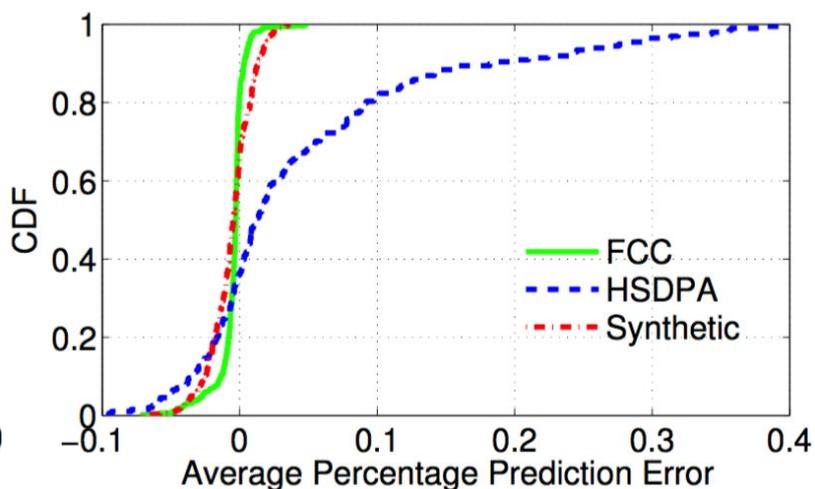
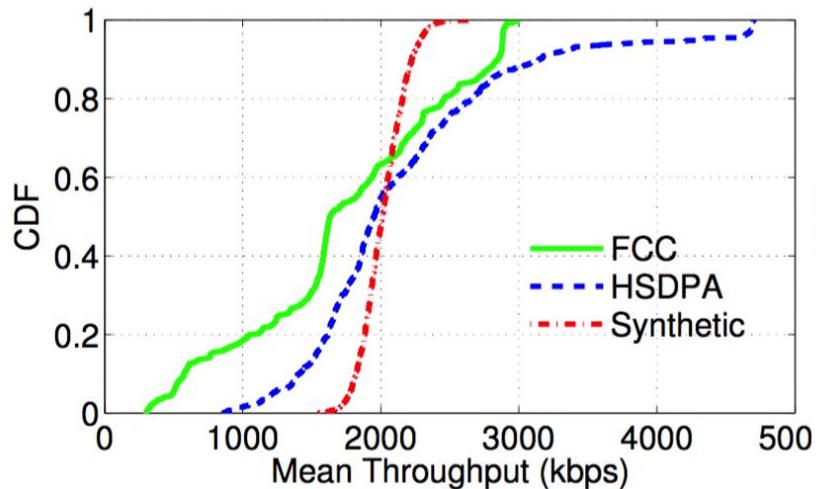
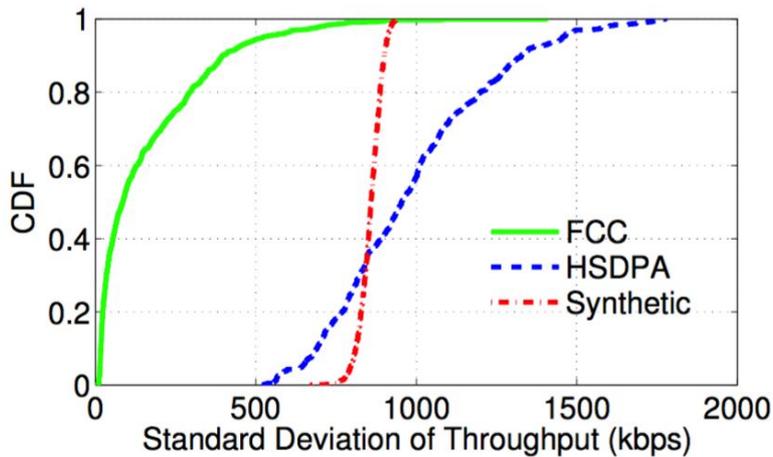
- Why FastMPC?
  - MPC has large computational overhead
  - Video player cannot be bundled with the solver
- Solution: Table lookup



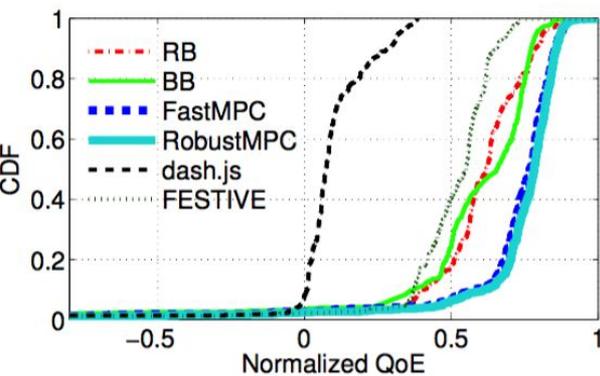
**Figure 5: “FastMPC” idea: We enumerate possible scenarios and create a table indexing the optimal decision for each scenario.**

# + Evaluation

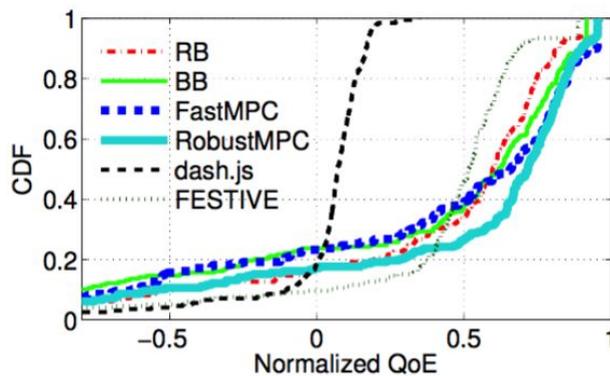
- FCC
- HSDPA
- Synthetic



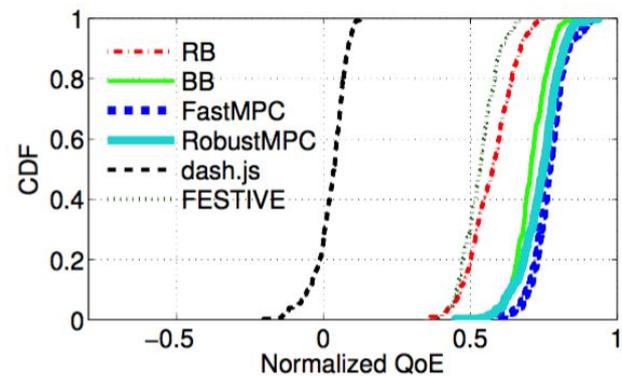
# + Real Player Evaluation



(a) FCC dataset



(b) HSDPA dataset

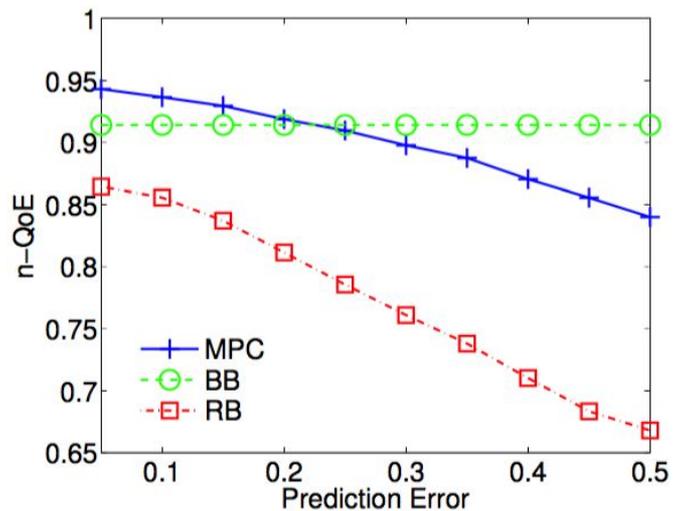


(c) Synthetic dataset

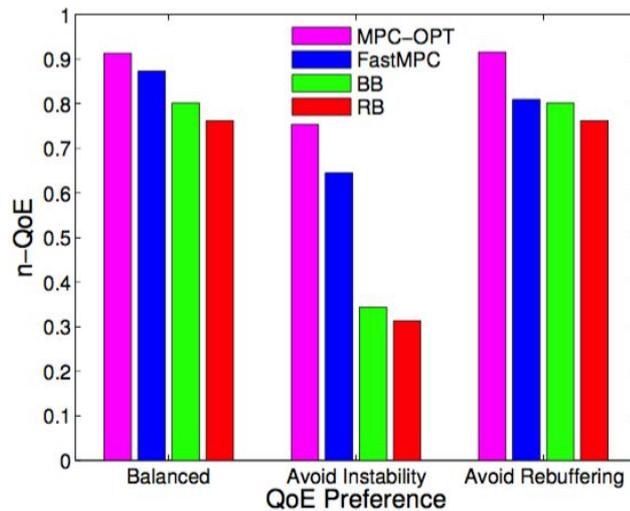
**Figure 8: Real experiment results with different throughput traces**

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# Sensitivity Analysis



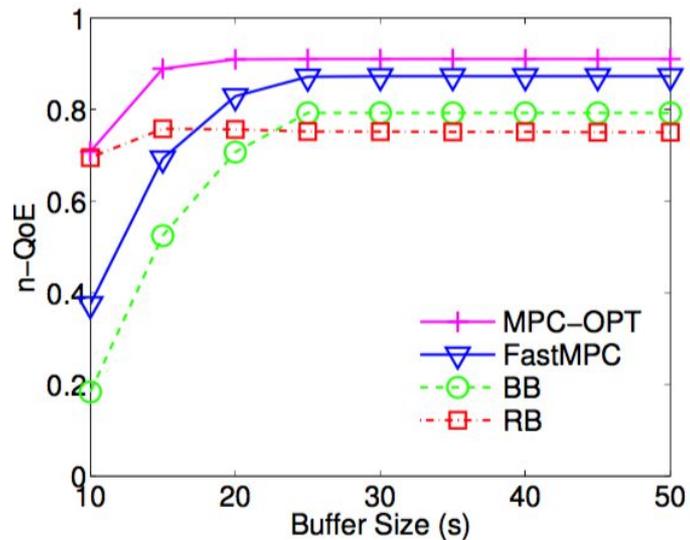
(a) Prediction error



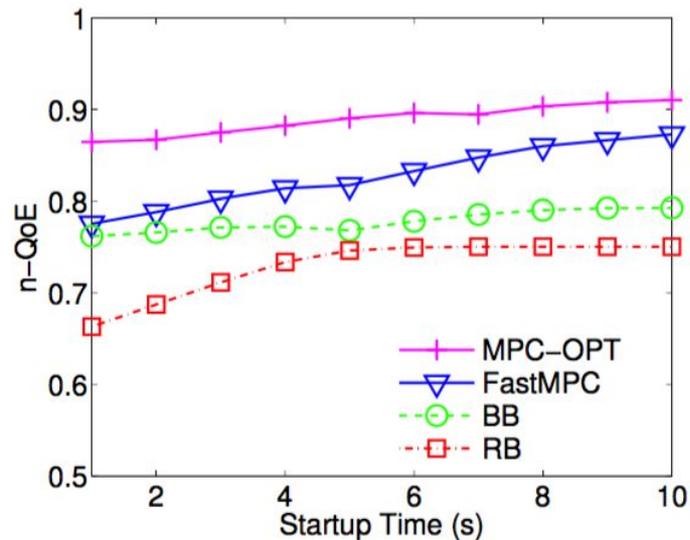
(b) QoE preferences

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# Sensitivity Analysis



(c) Buffer size



(d) Startup time

# + Takeaways and questions



Goal: developing a first-principles approach via control theory to develop a general framework to reason about classes of algorithms.

Questions to think about:

- What are some of the possible issues with MPC, comparing to BB and RB?
- What are the assumptions made by the paper that may not be realistic in practice?