

EECE5698

Networked XR Systems

The slides are borrowed from Dr. Timmerer [AAU/Bitmovin]'s tutorial at Sigcomm'17

Lecture Outline for Today

- Video Streaming
- Adaptive Bitrate Algorithms

Explosion in Video Applications



Importance of Internet Media



- **Real-time entertainment:** Streaming video and audio; **>70% of Internet traffic** at peak periods
- **Popular services**
 - YouTube (17.53%), Netflix (35.15%), Amazon Video (4.26%), Hulu (2.68%); all delivered **over-the-top (OTT)**
- **Forecast: Visual Networking Index (VNI) 2016-2021 (Sep'17)**
 - IP video traffic will be **82% of all consumer Internet traffic by 2021** (up from 73% in 2016); will **grow threefold** from 2016 to 2021
 - Live Internet video will account for **13% of Internet video traffic by 2021**; will **grow 15-fold** from 2016 to 2021
- More people now subscribe to Netflix (50.85M) than cable TV (48.61M) in the US (Q1 2017)

Beginning of Internet Video

IPTV

Managed delivery

Emphasis on quality

Mostly linear TV

Always a paid service

IP Video

Best-effort delivery

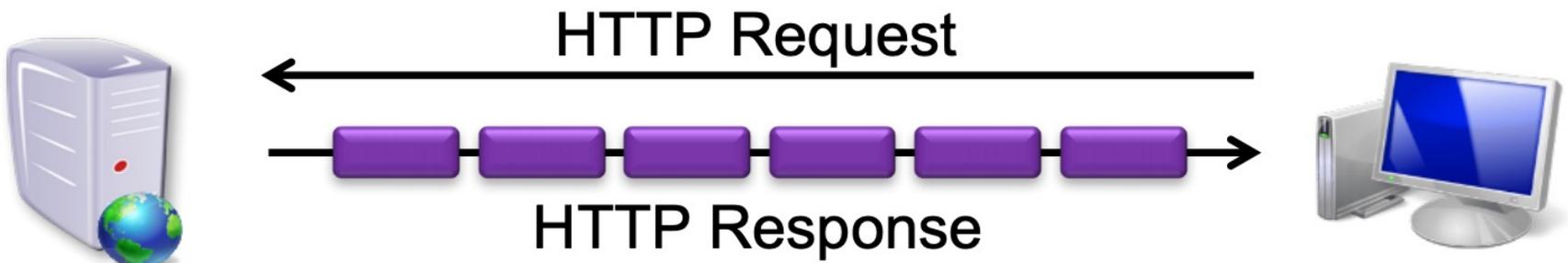
Quality not guaranteed

Mostly on demand

Paid or ad-based service

Progressive Download

One request, One response



What is Streaming?

Streaming is transmission of a continuous content from a server to a client and its simultaneous consumption by the client

Two Main Characteristics

1. Client consumption rate may be limited by real-time constraints as opposed to just bandwidth availability
2. Server transmission rate (loosely or tightly) matches to client consumption rate

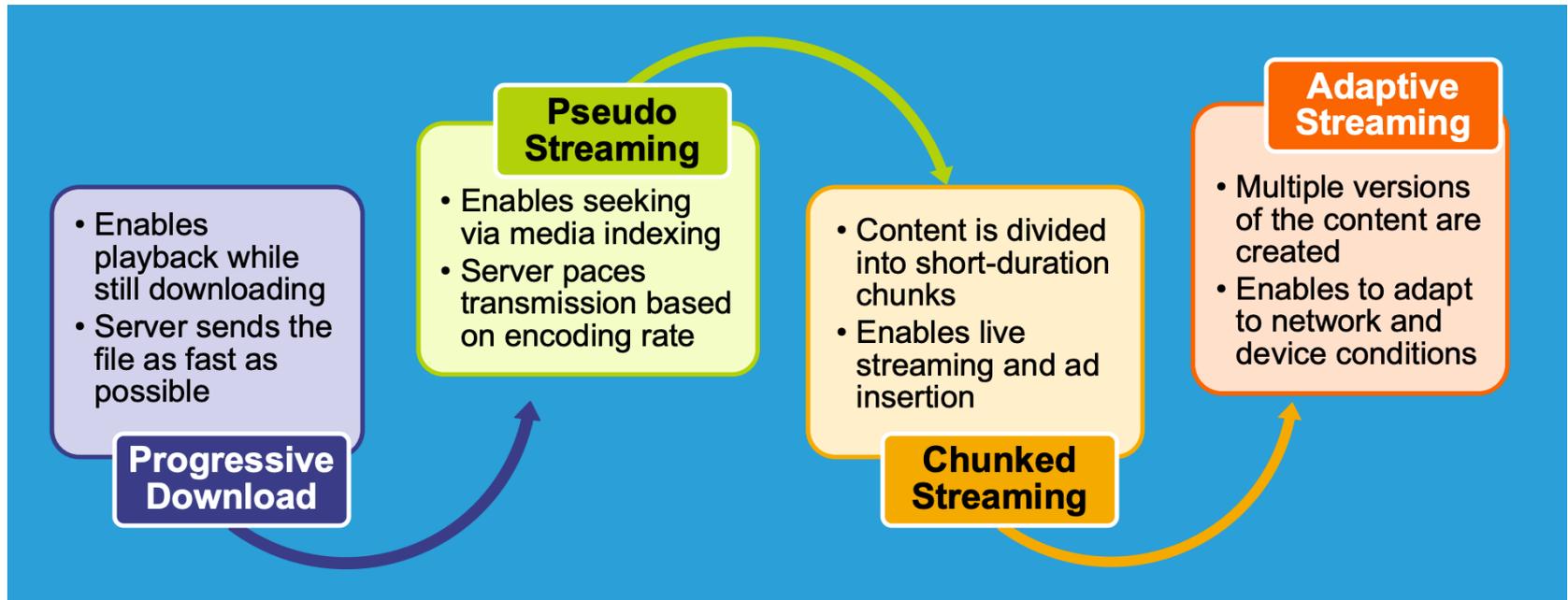
Some common Streaming Issues

Stalls, Slow Start-Up, Plug-In and DRM Issues

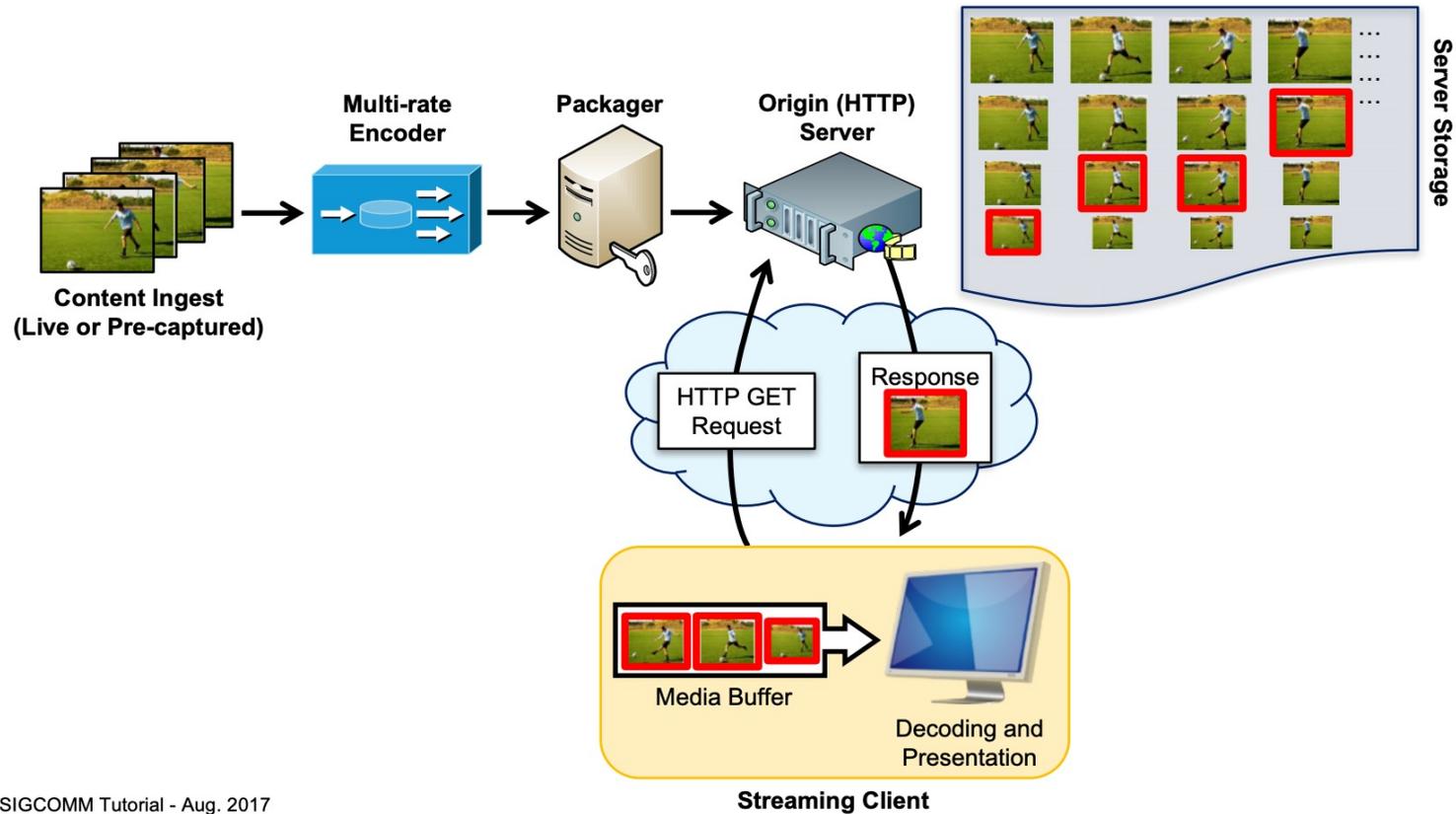
- Unsupported/wrong
 - protocol
 - plug-in
 - codec
 - format
 - DRM
- Slow start-up
- Poor quality, quality variation
- Frequent freezes/glitches
- Lack of seeking



Evolution of Video Streaming

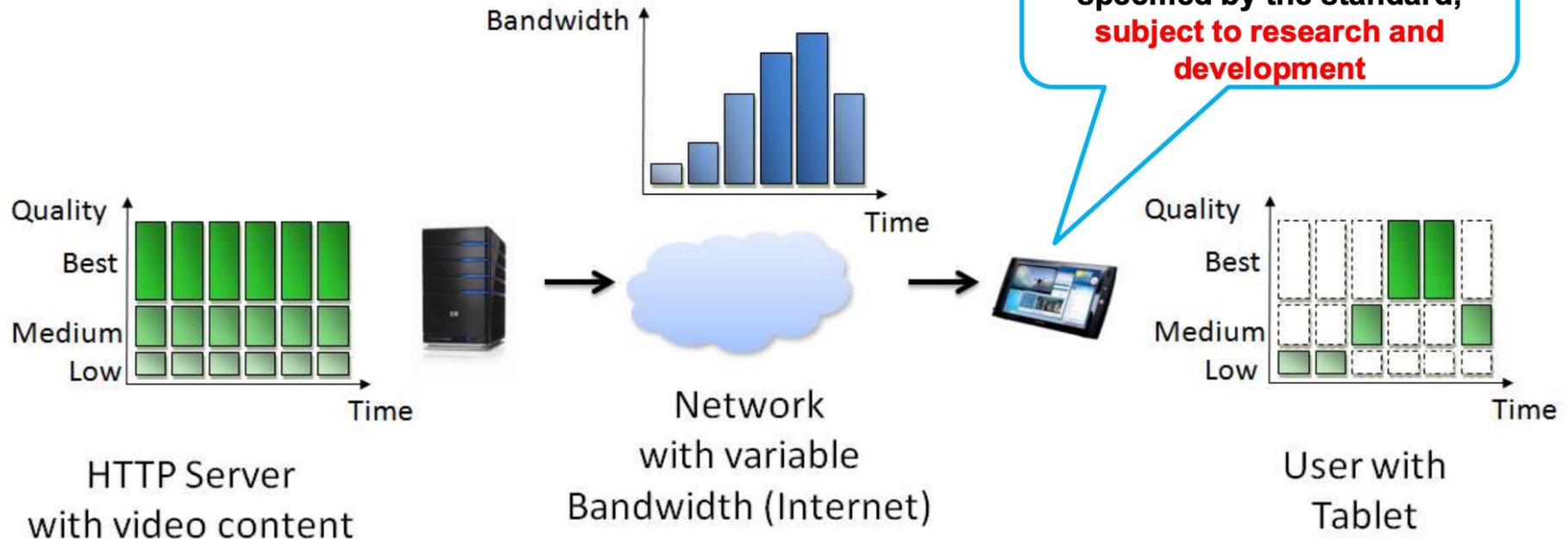


Adaptive Streaming



Adaptive Streaming

- In a nutshell...



Example Representations

Vancouver 2010

	Encoding Bitrate	Resolution
Rep. #1	3.45 Mbps	1280 x 720
Rep. #2	1.95 Mbps	848 x 480
Rep. #3	1.25 Mbps	640 x 360
Rep. #4	900 Kbps	512 x 288
Rep. #5	600 Kbps	400 x 224
Rep. #6	400 Kbps	312 x 176

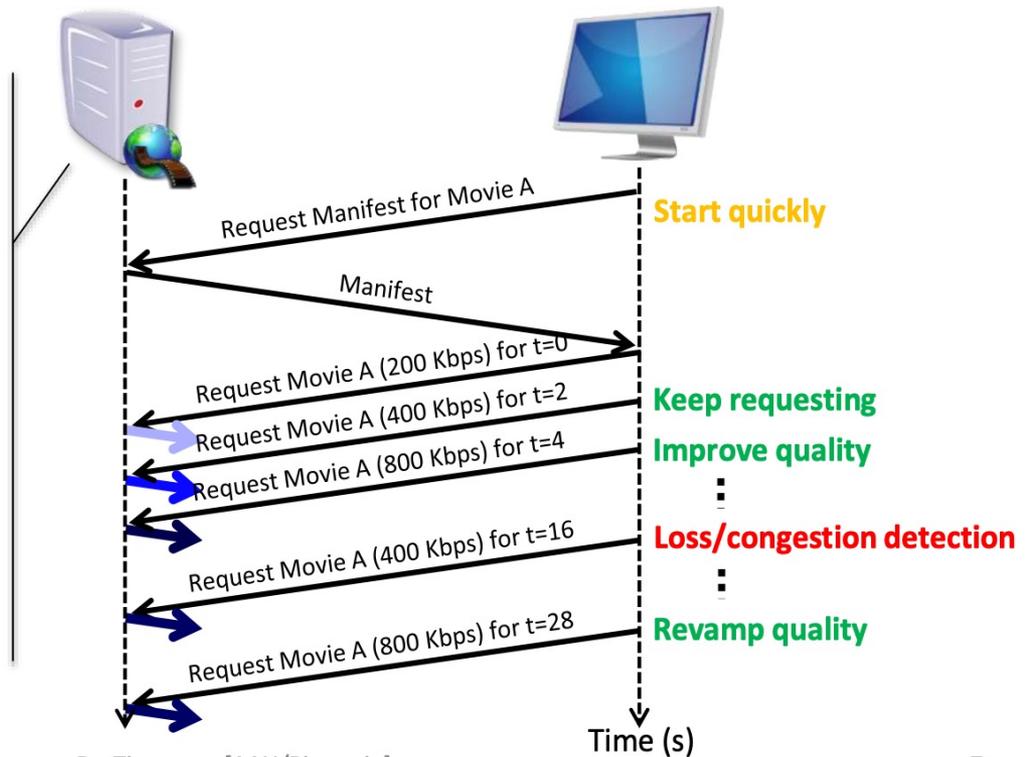
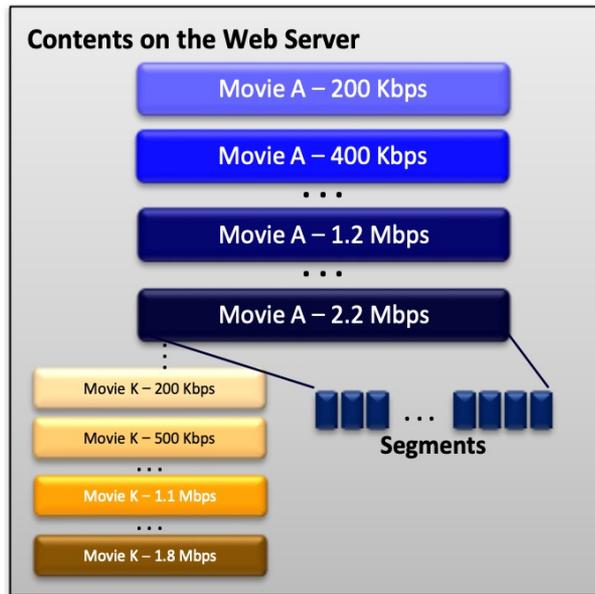
Sochi 2014

	Encoding Bitrate	Resolution
Rep. #1	3.45 Mbps	1280 x 720
Rep. #2	2.2 Mbps	960 x 540
Rep. #3	1.4 Mbps	960 x 540
Rep. #4	900 Kbps	512 x 288
Rep. #5	600 Kbps	512 x 288
Rep. #6	400 Kbps	340 x 192
Rep. #7	200 Kbps	340 x 192

FIFA 2014

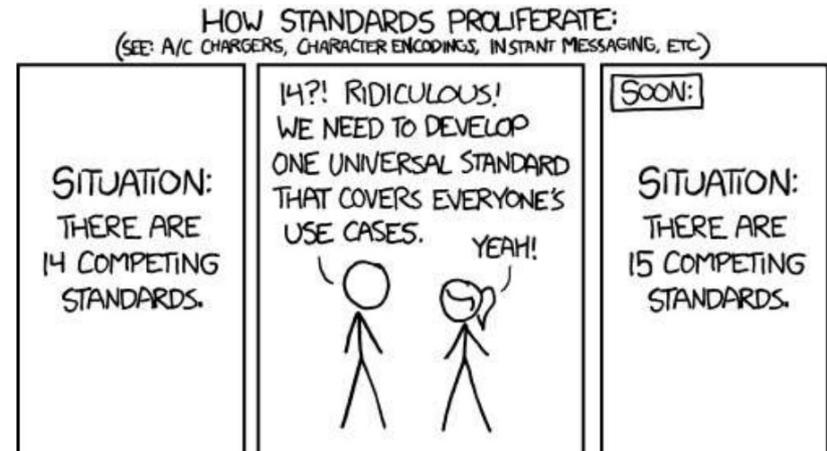
	Encoding Bitrate	Resolution
Rep. #1	3.45 Mbps	1280 x 720
Rep. #2	2.2 Mbps	1024 x 576
Rep. #3	1.4 Mbps	768 x 432
Rep. #4	950 Kbps	640 x 360
Rep. #5	600 Kbps	512 x 288
Rep. #6	400 Kbps	384 x 216
Rep. #7	250 Kbps	384 x 216
Rep. #8	150 Kbps	256 x 144

Representation switching



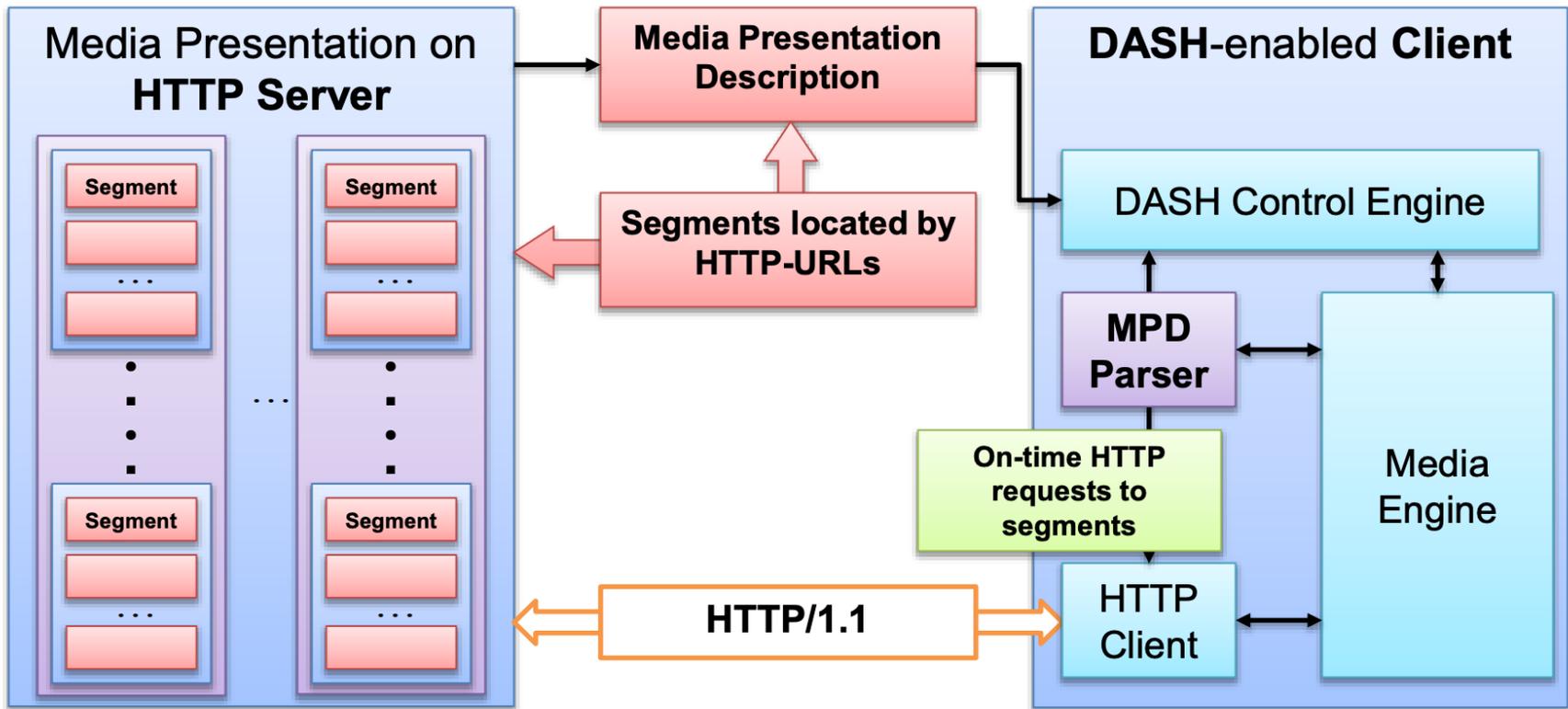
Adaptive Streaming Standards

- **Adobe**
 - **HTTP Dynamic Streaming (HDS)**
 - Switched to DASH
- **Apple**
 - **HTTP Live Streaming (HLS)**
 - Required for iOS
- **Microsoft**
 - **Smooth Streaming**
 - Switched to DASH, almost..
- **MPEG Dynamic Adaptive Streaming over HTTP (DASH)**
 - **Supported by Netflix, YouTube, Bitmovin, etc.**
- **MPEG Common Media Application Format (MPEG-A Part 19)**
 - **The new kid on the block** – support for “fragmented mp4 in HLS”
 - DASH/HLS convergence at segment level – some open issues with encryption format

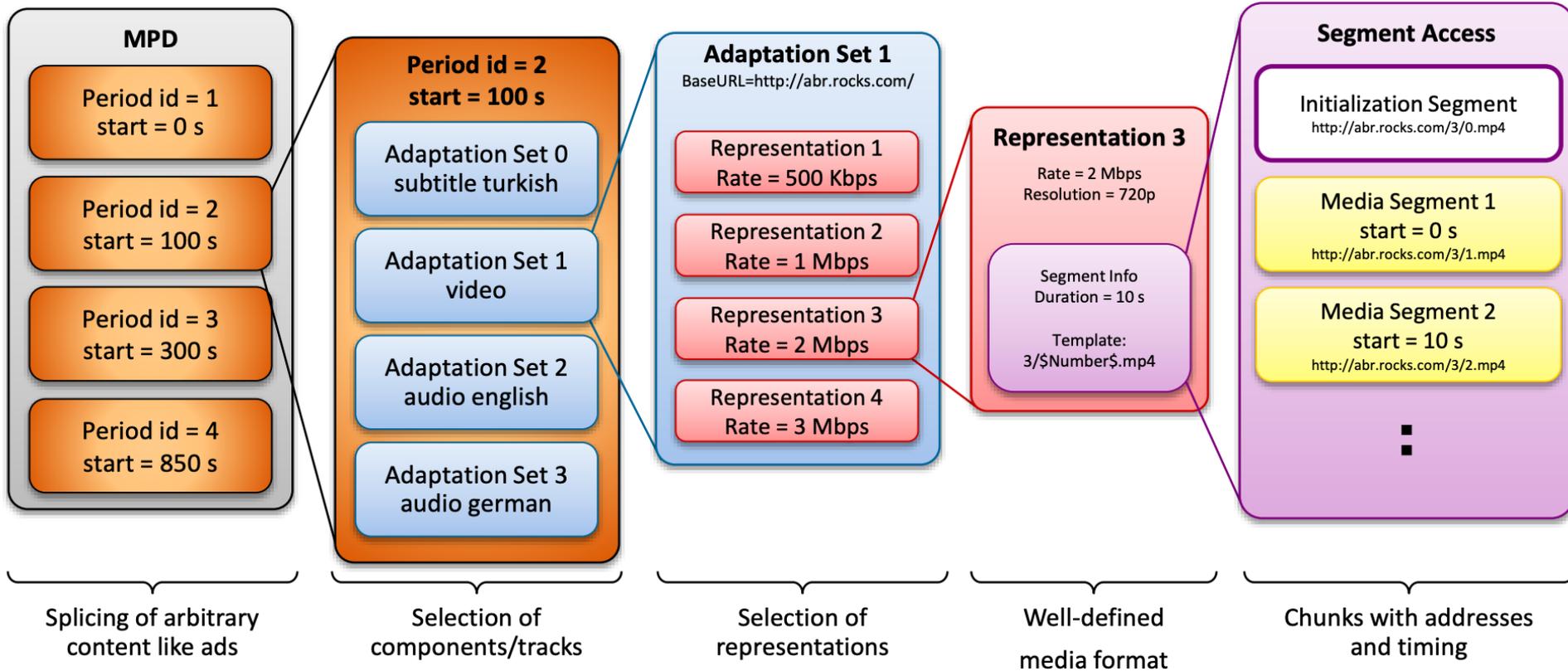


Source: <http://xkcd.com/927/>

DASH Internals

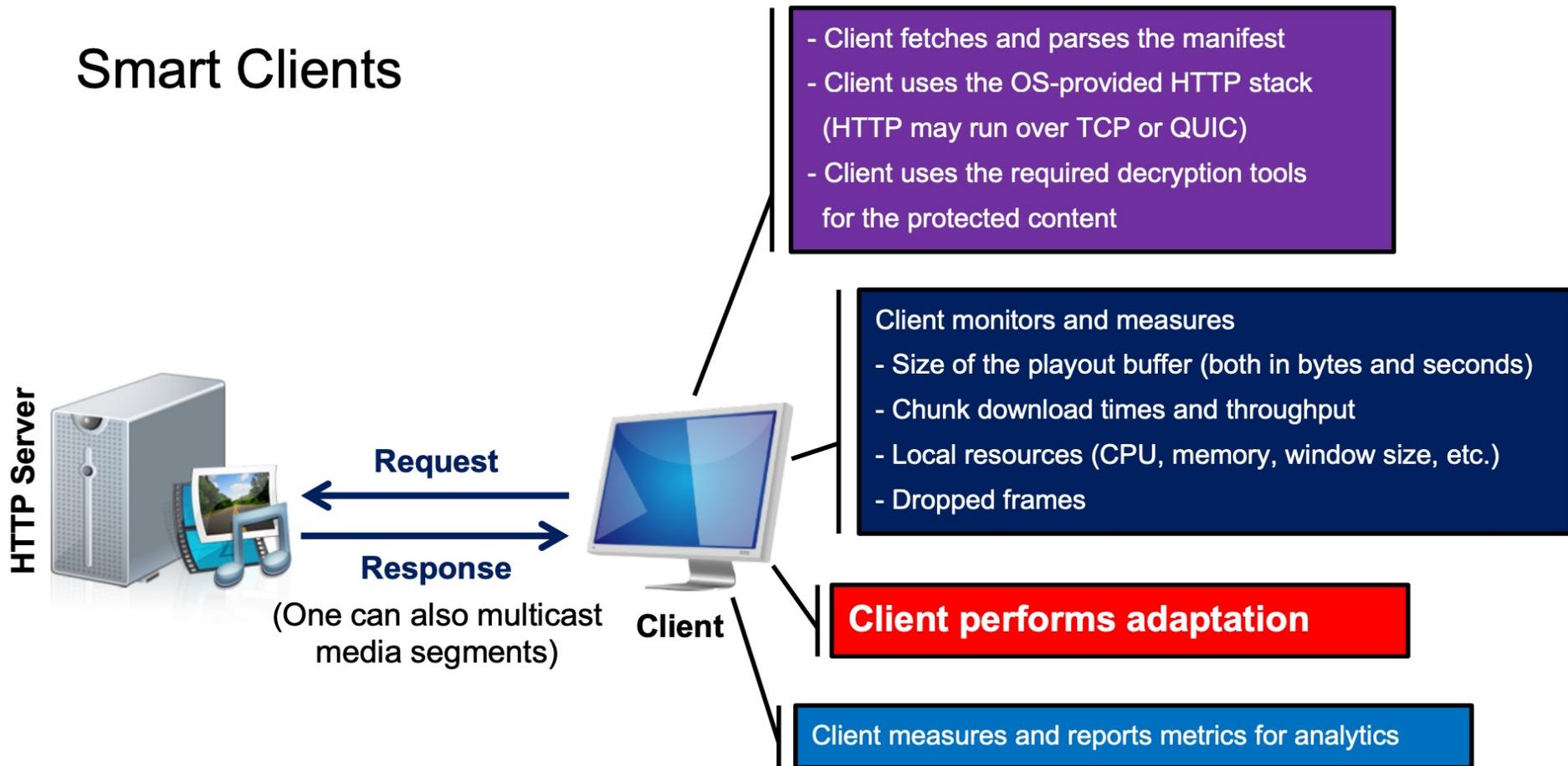


DASH Internals – Manifest File



A Typical Client Does

Smart Clients



Video Streaming - Quality of Experience

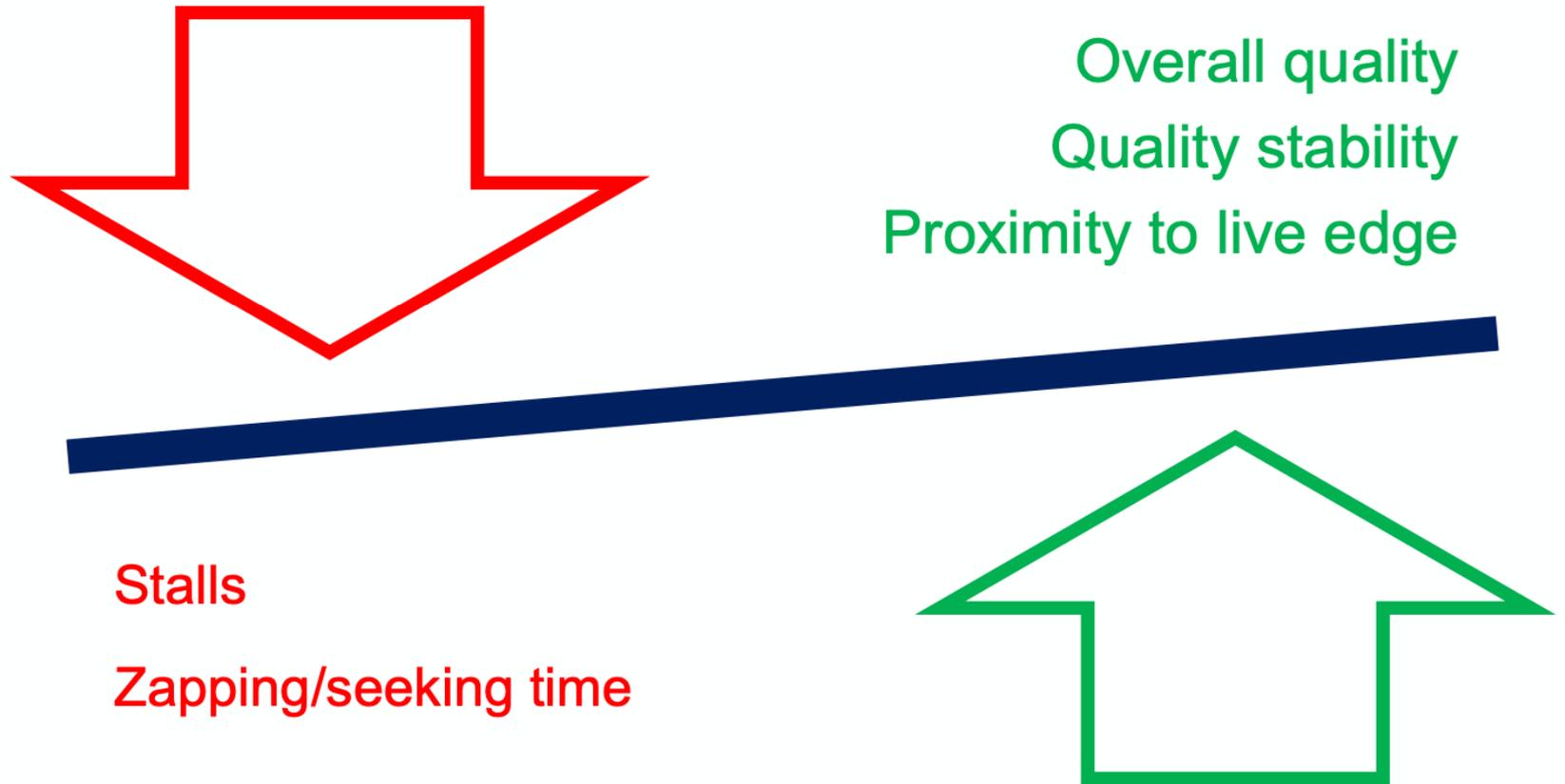
- **Objective**

- Initial or startup **delay** (low)
- Buffer underrun / **stalls** (zero)
- Quality **switches** (low)
- **Media throughput** (high)
- [Other media-related configuration: encoding, representations, segment length, ...]

- **Subjective**

- **Mean Opinion Score** (MOS) – various scales
- Various **methodologies** (e.g., DSCQS, DSIS, ACR, PC, ...)

Adaptive Bitrate (ABR) Algorithm



Bitrate Selection is the key

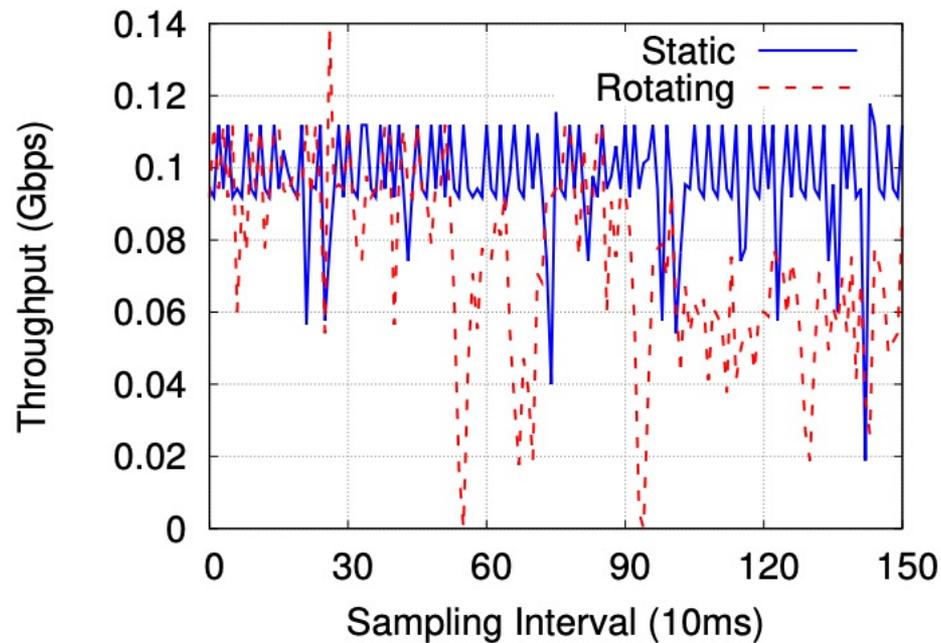
ABR algorithms

- The bitrate adaptation logic is typically controlled by the client
 - ◆ Bitrate adaptation heuristics based on:
 1. Available bandwidth
 2. Playback buffer size
 3. Chunk scheduling
 4. Hybrid-based
- Interesting algorithms
 - ◆ Li et al (2014), Liu et al (2011)
 - ◆ Huang et al (2014), Mueller et al (2015)
 - ◆ Jiang et al (2012), Chen et al (2013)
 - ◆ Yin et al (2015), Li et al. (2014), De Cicco et al. (2013), Miller et al. (2012), Zhou et al. (2012)

Classes of ABR algorithms

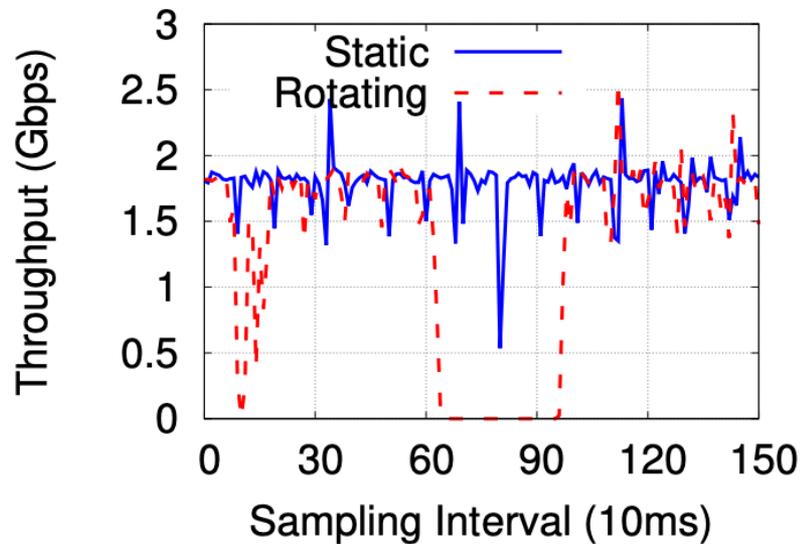
- Bandwidth prediction
 - Simple statistics
 - ML
- ABR Algorithms
 - Buffer based (e.g., BBA, BOLA)
 - Throughput based (e.g., Festive)
 - Hybrid (e.g., MPC, Pensieve)
- Identifying root causes
 - Client, Server, CDN, Routers

Some Wireless Network Conditions

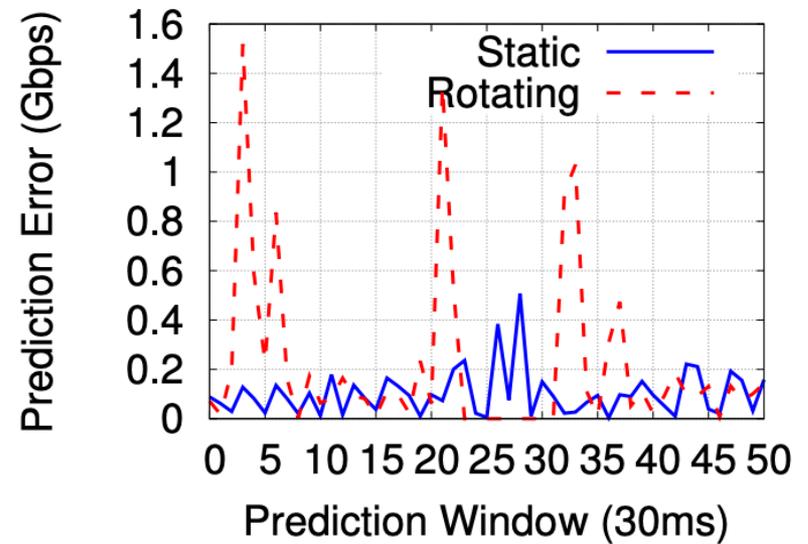


Reasons for fluctuation?

Some Wireless Network Conditions



(a) Throughput Variation



(b) Prediction Error

Increased bandwidth is not enough: reliability matters.

ABR Algorithms

- Key performance metrics
 - Quality (Q), Rebuffering (R), Quality switches (S)
- A general objective
 - Quality of Experience (QoE)

$$\mathbf{QoE = w1.Q - w2.R - w3.S}$$

Maximize QoE

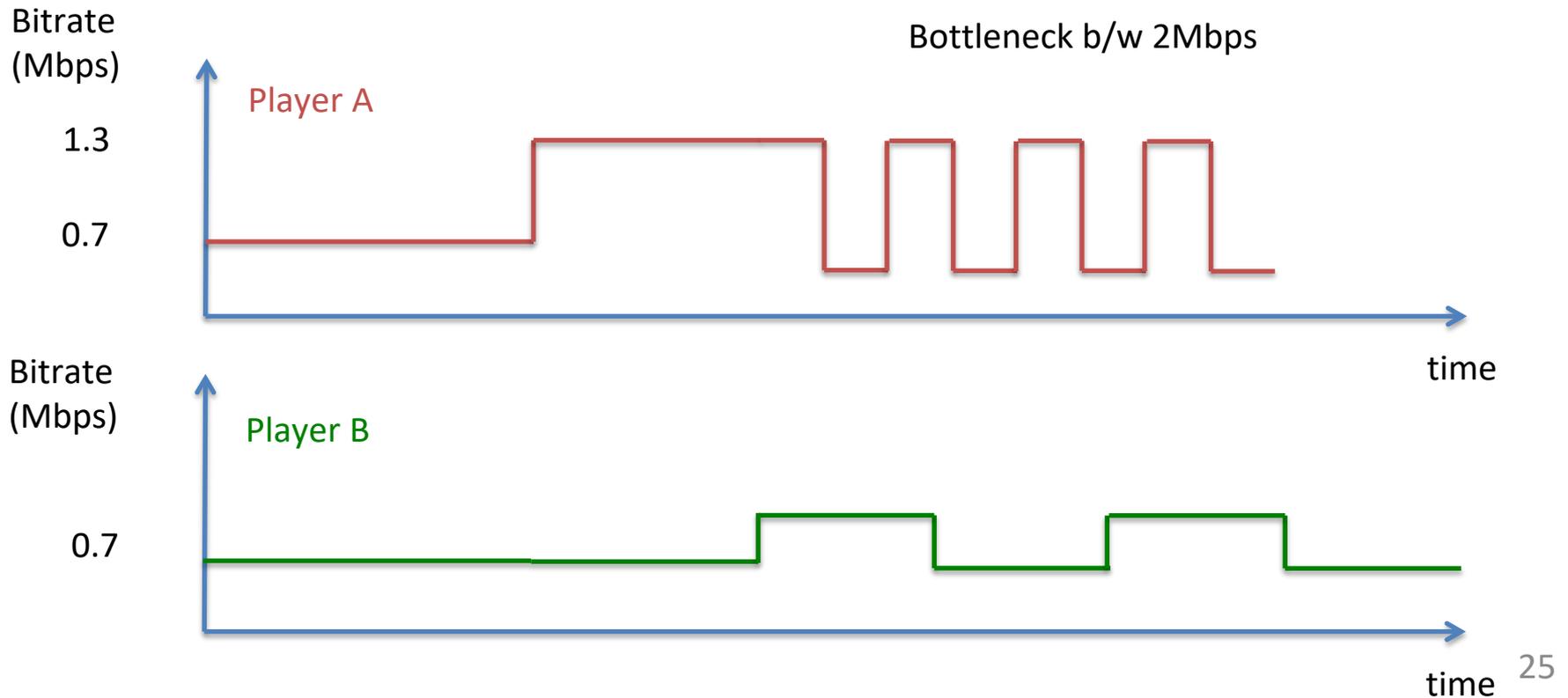
s.t network constraints

Throughput Based ABR (FESTIVE)

Inefficiency: Fraction of bandwidth not being used or overused

Unfairness: Discrepancy of bitrates used by multiple players

Instability: The frequency and magnitude of recent switches



Throughput Based ABR (FESTIVE)

Many players use this to keep fixed video buffer

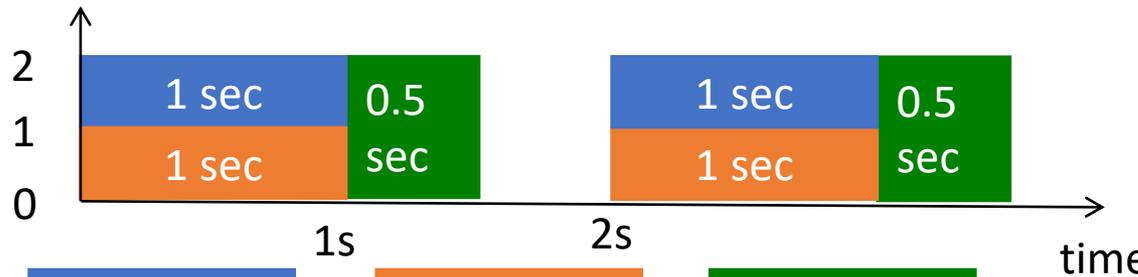
e.g., if chunk duration = 2 sec, chunk requests at $T=0,2,4,\dots$ sec

Example setup: Total bandwidth: 2Mbps

Bitrate 0.5 Mbps, 2 sec chunks

Chunk size: 0.5 Mbps x 2 sec = 1.0Mb

b/w (Mbps)



Throughput: 2 Mbps

Throughput: 1 Mbps

Throughput: 1 Mbps

time

Player A,
 $T=0,2,4,\dots$

Player B
 $T=0,2,4,\dots$

Player C
 $T=1,3,5,\dots$

Unfair! Start time impacts observed throughput

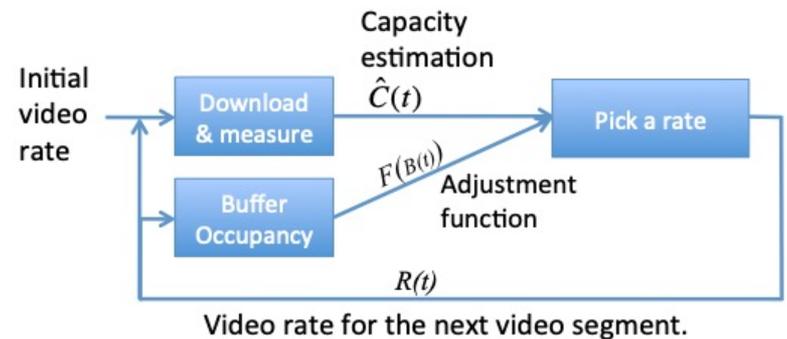
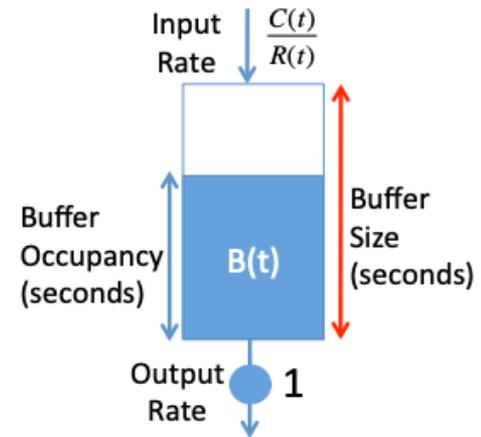
NOT a TCP problem!

Buffer Based ABR (BBA)

- Adaptive Bit-Rate (ABR) selection algorithms
 - Adjust the bit-rate based on estimated capacity of the network
 - Estimating capacity is difficult b/c of variability
 - Playback buffer occupancy can be used as a heuristic

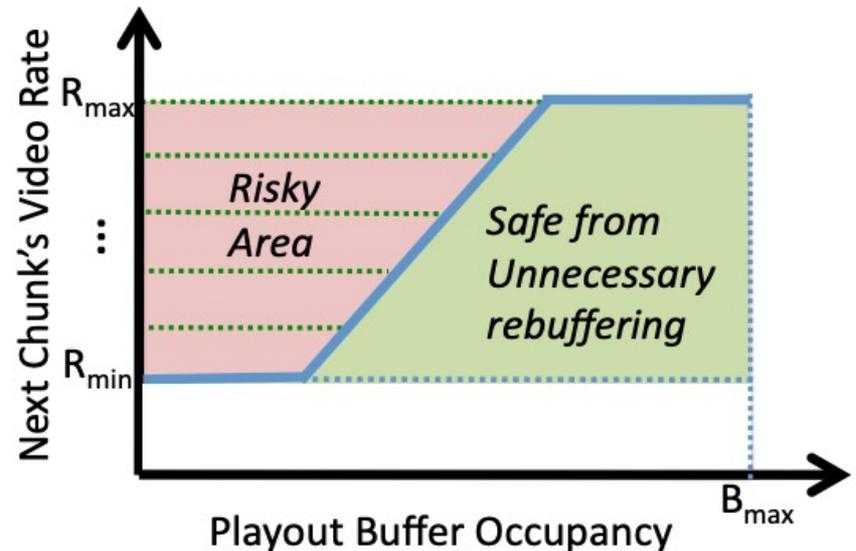
Buffer Based ABR (BBA)

- Picks Video Rate $R(t)$, with capacity $C(t)$
- Capacity overestimation leads to buffer depletion quickly
- Difficult to pick the right adjustment function when $C(t)$ is highly variable in practice

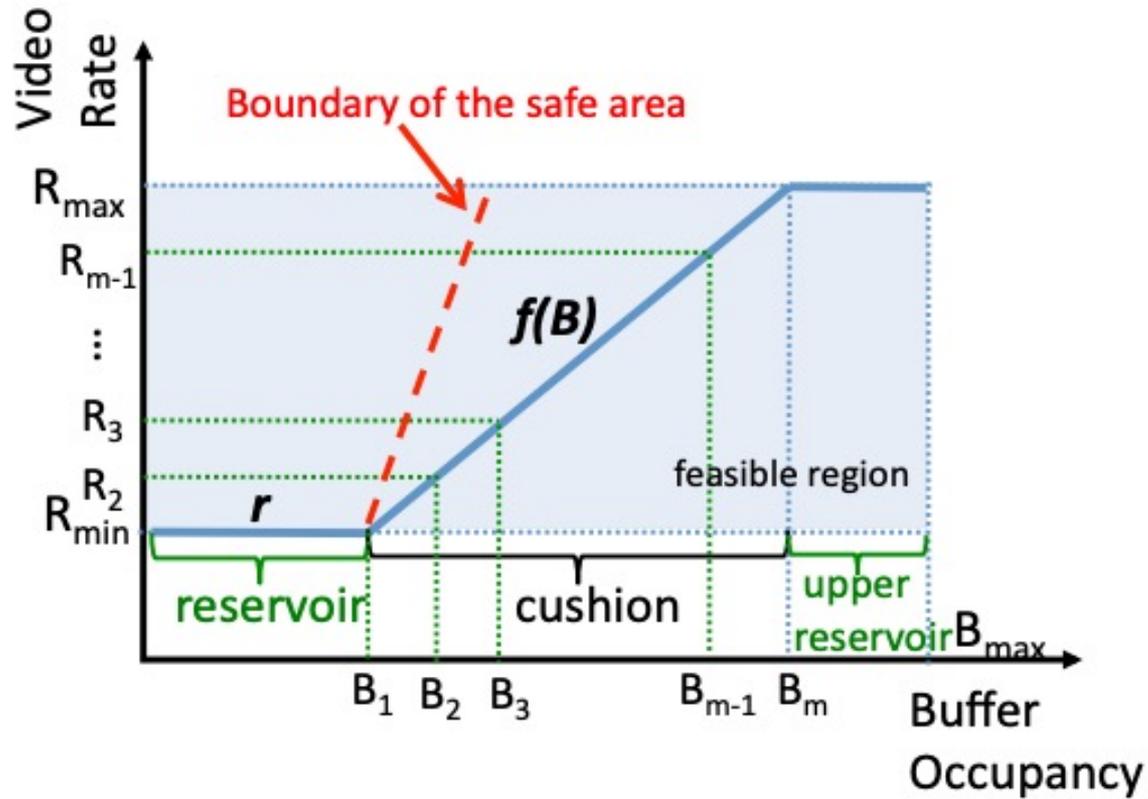


Buffer Based ABR (BBA)

- Pure Buffer based approach
- A function of the current buffer occupancy, $B(t)$
- A function that produces a video rate between R_{\min} and R_{\max} given the current buffer occupancy.



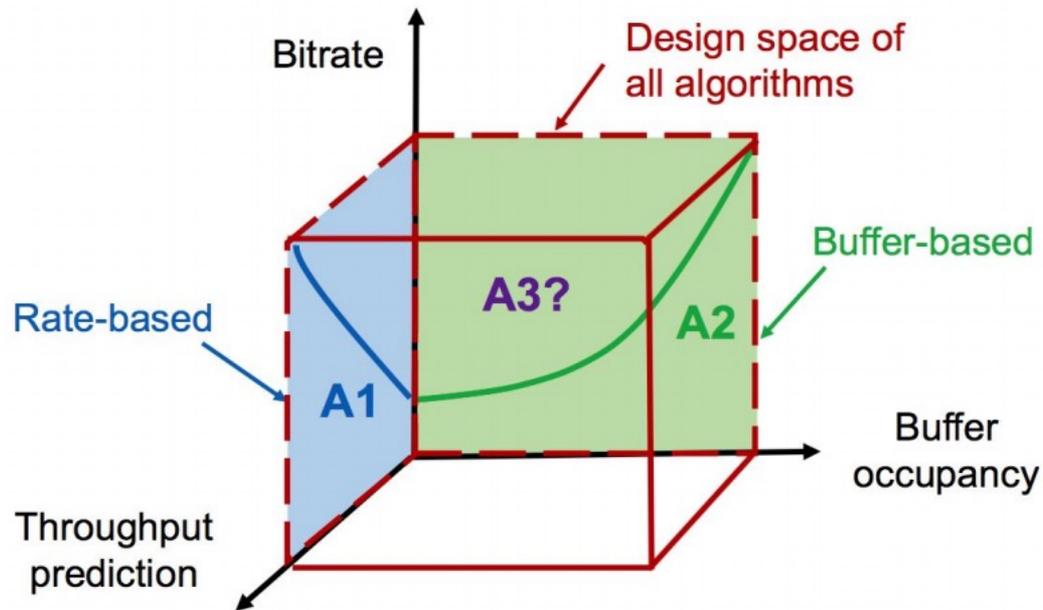
Buffer Based ABR (BBA)



Buffer Based ABR (BBA2)

- BBA-1 still suffers from poor video rate while the buffer is still filling up initially
 - The algorithm is too conservative in startup
 - Chunk map is useful for steady-state
 - Capacity estimate is still useful for startup
 - Estimate based on the throughput of the last chunk

Throughput and Buffer Based ABR (MPC)



Model Predictive Control - MPC

Throughput and Buffer Based ABR (MPC)

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

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

The diagram illustrates the components of the QoE equation (5). Four callout boxes, each with an orange border and a pointer to a specific term in the equation, are present:

- Average video quality**: Points to the first term, $\sum_{k=1}^K q(R_k)$.
- Average quality variations**: Points to the second term, $-\lambda \sum_{k=1}^{K-1} |q(R_{k+1}) - q(R_k)|$.
- Total rebuffer time**: Points to the third term, $-\mu \sum_{k=1}^K \left(\frac{d_k(R_k)}{C_k} - B_k \right)_+$.
- Startup Delay**: Points to the fourth term, $-\mu_s T_s$.

Throughput and Buffer Based ABR (basicMPC)

Algorithm 1 Video adaptation workflow using MPC

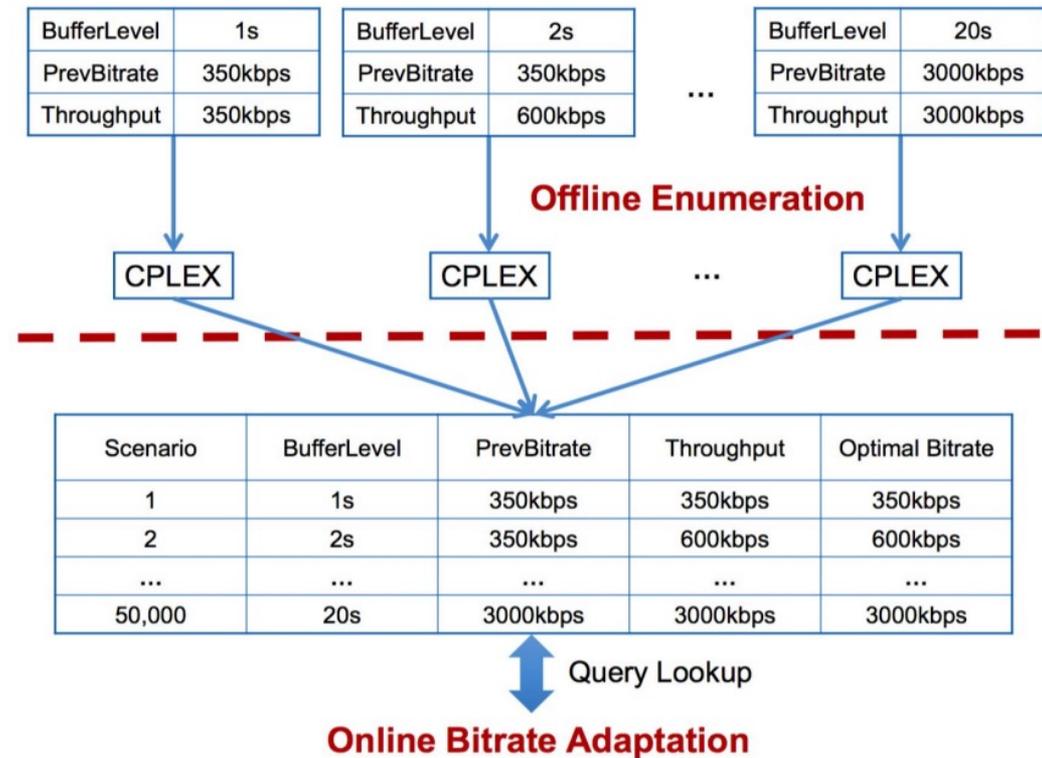
```
1: Initialize
2: for  $k = 1$  to  $K$  do
3:   if player is in startup phase then
4:      $\hat{C}_{[t_k, t_{k+N}]} = \text{ThroughputPred}(C_{[t_1, t_k]})$ 
5:      $[R_k, T_s] = f_{mpc}^{st} \left( R_{k-1}, B_k, \hat{C}_{[t_k, t_{k+N}]} \right)$ 
6:     Start playback after  $T_s$  seconds
7:   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
```

Throughput and Buffer Based ABR (robustMPC)

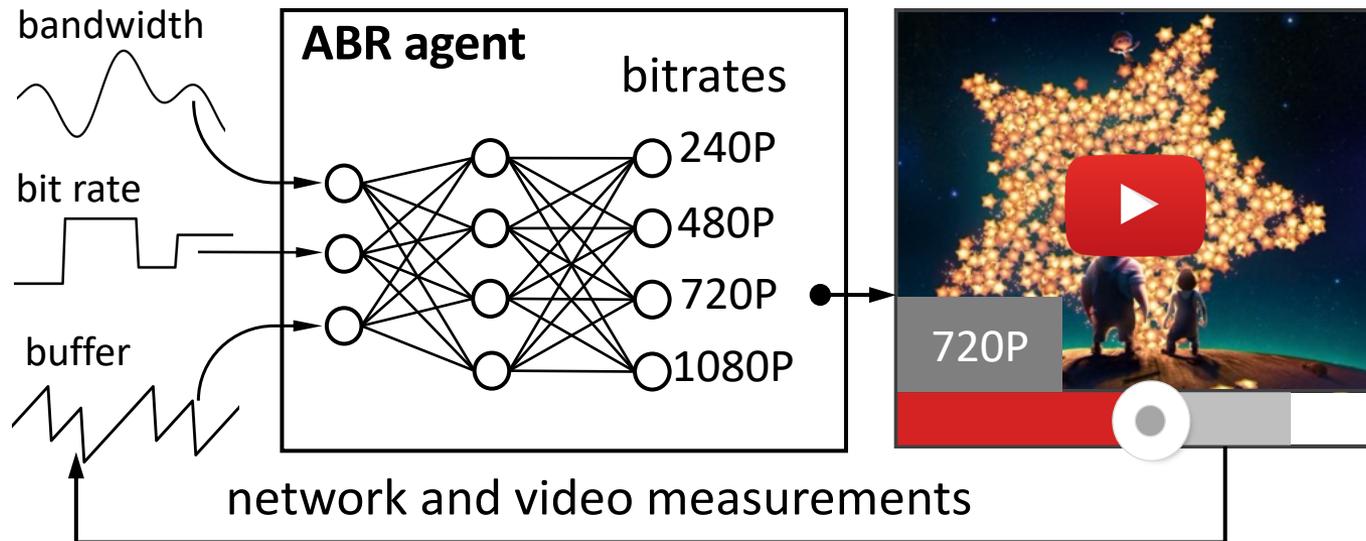
- The basic MPC algorithm assumes the existence of an accurate throughput predictor.
- Robust MPC essentially optimizes the worst-case QoE assuming that the actual throughput can take any value in a range in contrast to a point estimate
- Robust MPC makes conservative estimation

Throughput and Buffer Based ABR (fastMPC)

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



Learning Based ABR (Pensieve)

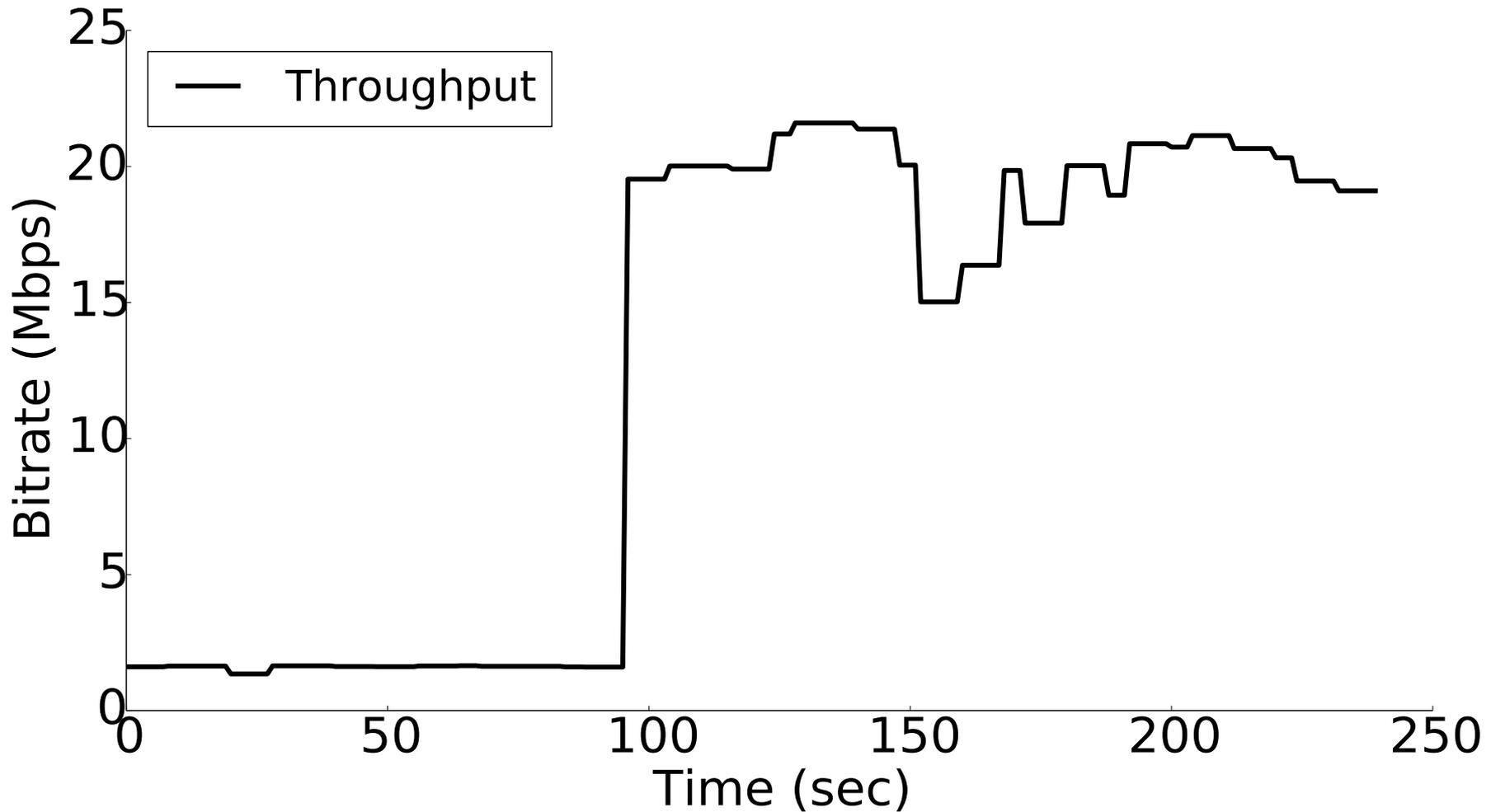


Pensieve **learns** ABR algorithm **automatically** through experience

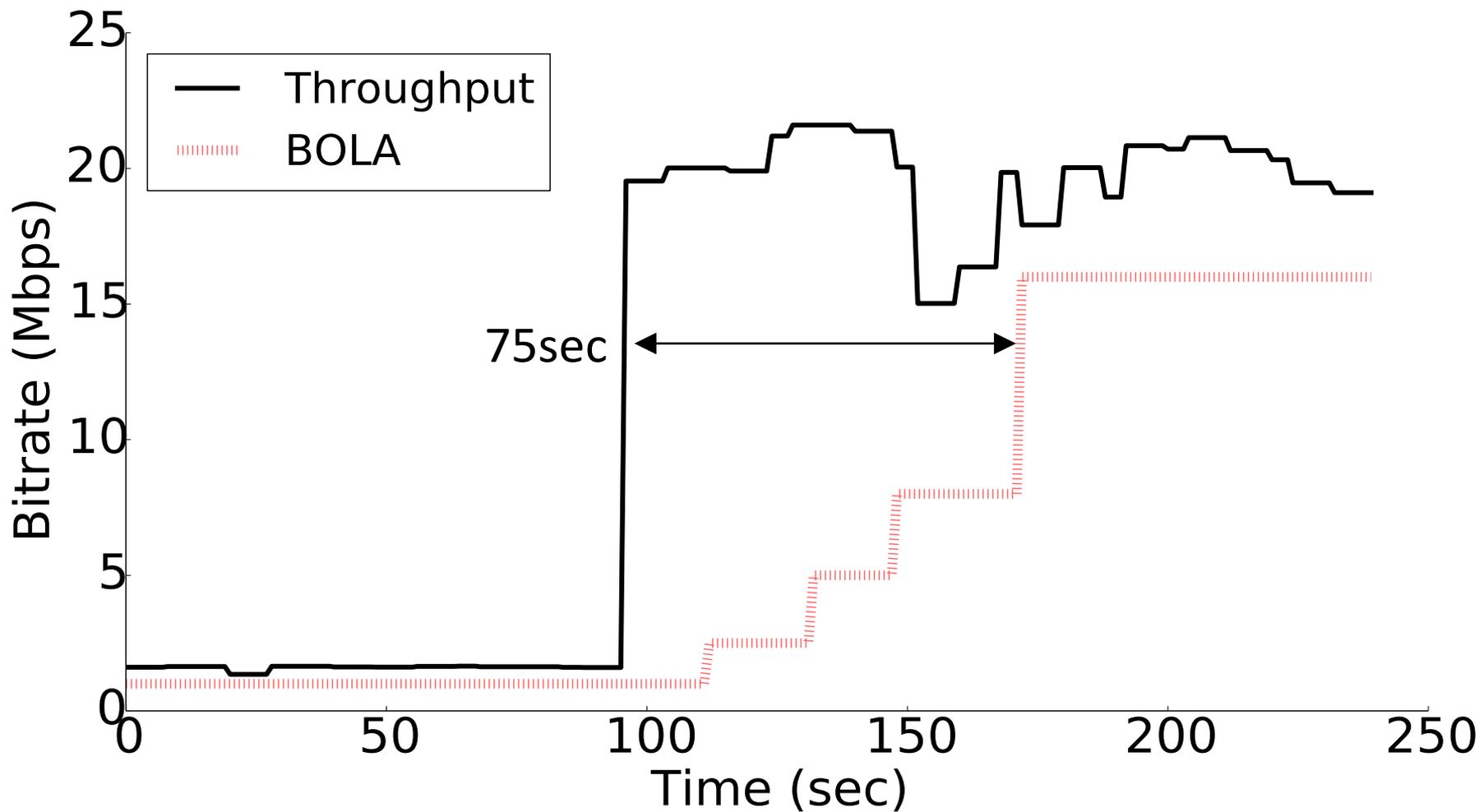
Fast Switching

- Responsiveness to higher network throughput by replacing segments already in the buffer
- Decide whether to download a new segment or a replacement segment
- Determine which segment to replace
- Works with both throughput based and buffer-based ABR algorithms
 - Any problems with buffer based?

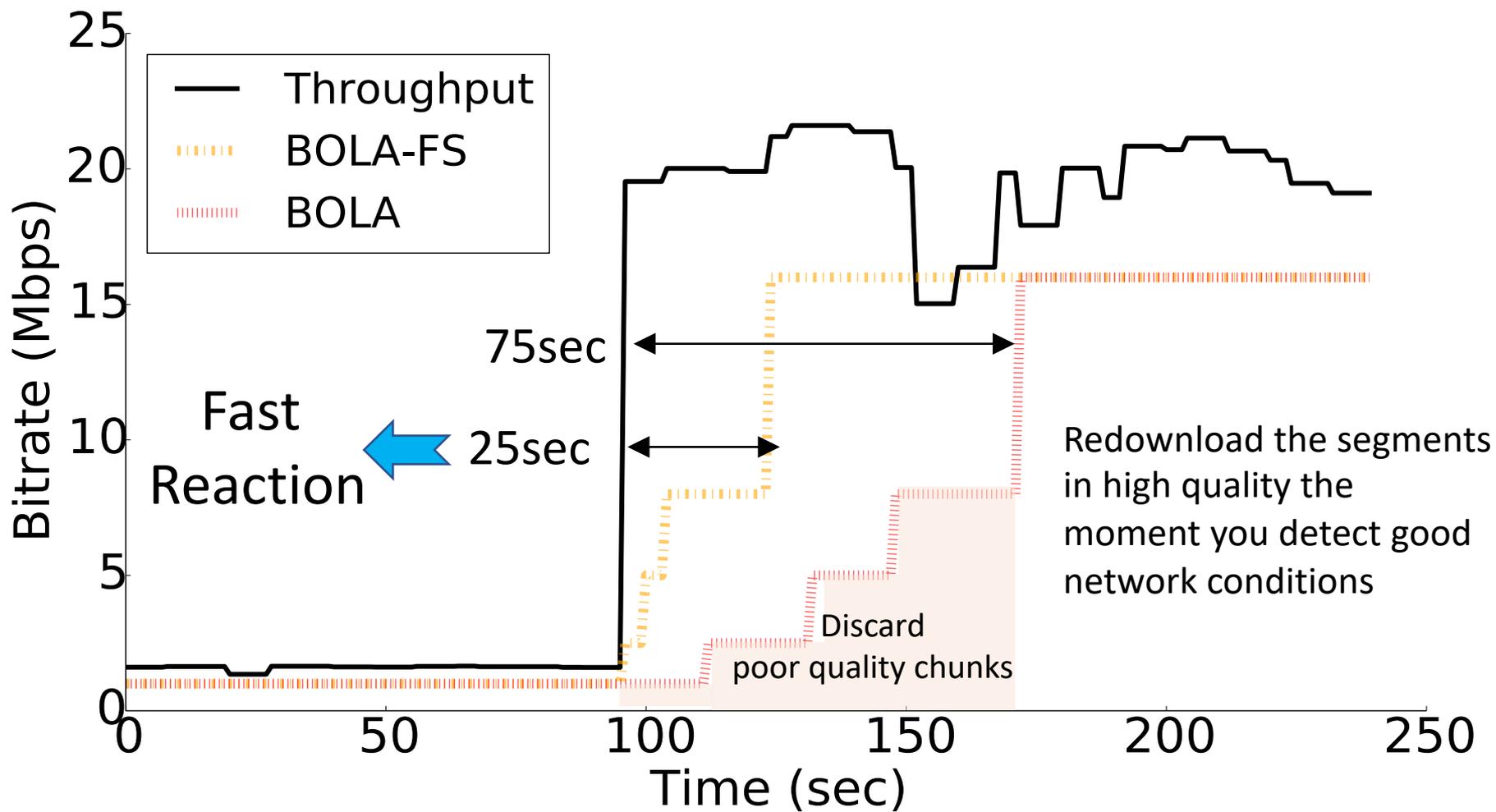
Fast Switching



Fast Switching



Fast Switching



Summary of the Lecture

- ABR for advanced video applications
- ABR Algorithms
 - Buffer based
 - Throughput based
 - Hybrid
 - Learned
- Bandwidth prediction
- Fast switching for high throughput network