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# HTTP-based Adaptive Streaming for Mobile Clients using Markov Decision Process

Ayub Bokani    **Mahbub Hassan**    Salil Kanhere

# Challenges of Decision Making in DASH

- Difficult decisions to make
- Frequent decision making - as much as 30 decisions per minute (2-sec chunks)
- All decisions are made by the mobile device (client)

Options for chunk #5

1. Quality 1, 2 KB
2. Quality 2, 4 KB
3. Quality 3, 5 KB
4. Quality 4, 7 KB
5. Quality 5, 9 KB



Server

non-deterministic network connection



Which option to select for chunk #5  
(minimize deadline miss, maximize quality, but minimize quality switching)

Just downloaded  
chunk #4 in Quality 3



# Markov Decision Process

- A tool for optimizing decision making when outcomes are partly random and partly under the control of the decision maker
- The process goes through a finite set of **states**
- At each state, the decision maker can choose a particular **action** from a given set
- State transitions are random, but the **transition probabilities** are different for different actions
- Each action is associated with a **reward** and sometimes a **penalty**
- **Revenue** (award minus penalty) is used to evaluate the outcome for a given action taken at a given state
- Solving an MDP problem means finding the best action for each state that will maximize the overall revenue
- Can be solved using dynamic programming
- Once solved, the actions corresponding to the states is called the **optimal strategy**



# DASHing with MDP

- Researchers have confirmed that use of MDP improves streaming quality
- However, MDP increases computation overhead of decision making at the mobile
- The overhead is an issue if a new MDP problem has to be solved for each chunk
- Our goal is to explore different approaches to reduce decision making overhead for MDP-based DASH



# Presentation Overview

- Our MDP model for DASH
- 3 approaches to reduce MDP overhead
- Simulation setup for evaluating these approaches
- Results
- Conclusion



# MDP Formulation of DASH (1)

- **States  $S(\rho, q)$** :  $q$  is quality level of the downloaded chunk and  $\rho$  is the time available before the chunk's playback deadline (*current buffer occupancy measured in time*)
- **Actions (decisions)**: quality level of the next chunk to be downloaded
- **Rewards**: higher reward for watching a chunk in higher quality
- **Penalties**: there is a penalty for missing a deadline as well as switching quality from previous chunk to the next

quality level ( $q$ )	1	2	3	4	5
$u(q)$	1	2	4	7	10

Rewards for quality levels

Penalty for missing a deadline =  
probability of deadline miss x D

Penalties for switching quality levels

$i \setminus j$	1	2	3	4	5
1	0	1	5	10	25
2	10	0	1	5	10
3	50	10	0	1	5
4	250	50	10	0	1
5	500	250	50	10	0



# MDP Formulation of DASH (2)

## The role of Bandwidth CDF

- For a given action (chunk size), state transition probabilities can be calculated if we know the **CDF of network bandwidth**
  - CDF allows calculation of the probability of a given buffer occupancy ( $\rho$ ) when the next chunk is downloaded
  - $\rho$ , together with the action (quality level decision), defines the next state  $S(\rho, q)$
- Transition probabilities will change if we have a different CDF
- Different CDFs will lead to different MDP strategies



# Approaches to reduce MDP overhead

- **k-MDP**
  - Uses only online data to learn and estimate bandwidth CDF (no offline network measurement)
  - Updates CDF with every sample (each chunk download gives one sample), but recomputes MDP only every  $k$  chunk downloads
  - Smaller  $k$  means more precise MDP, but higher overhead, and vice versa
- **s-MDP**
  - Solves MDP with offline network measurement (no online MDP computation)
  - One MDP computed for a large region and used at all locations during a given trip
  - Single MDP strategy to maintain for a large area, but CDF may not be accurate for some locations
- **x-MDP**
  - One MDP per road segment, but they are all solved offline (no online MDP computation)
  - More precise CDF estimation for every location visited during a given streaming session
  - Requires availability of more extensive bandwidth measurements (measurement needed for every road segments of a city-wide road network)

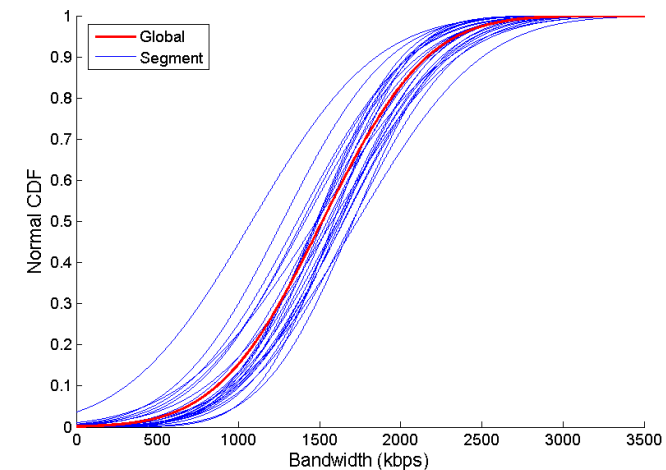
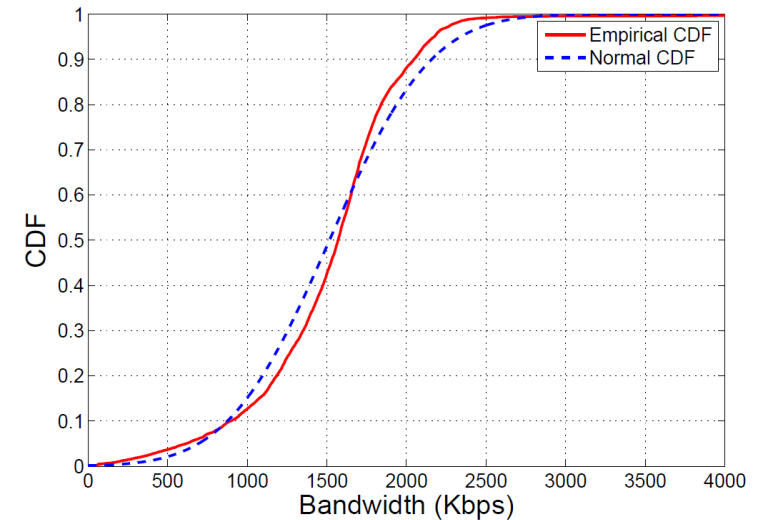




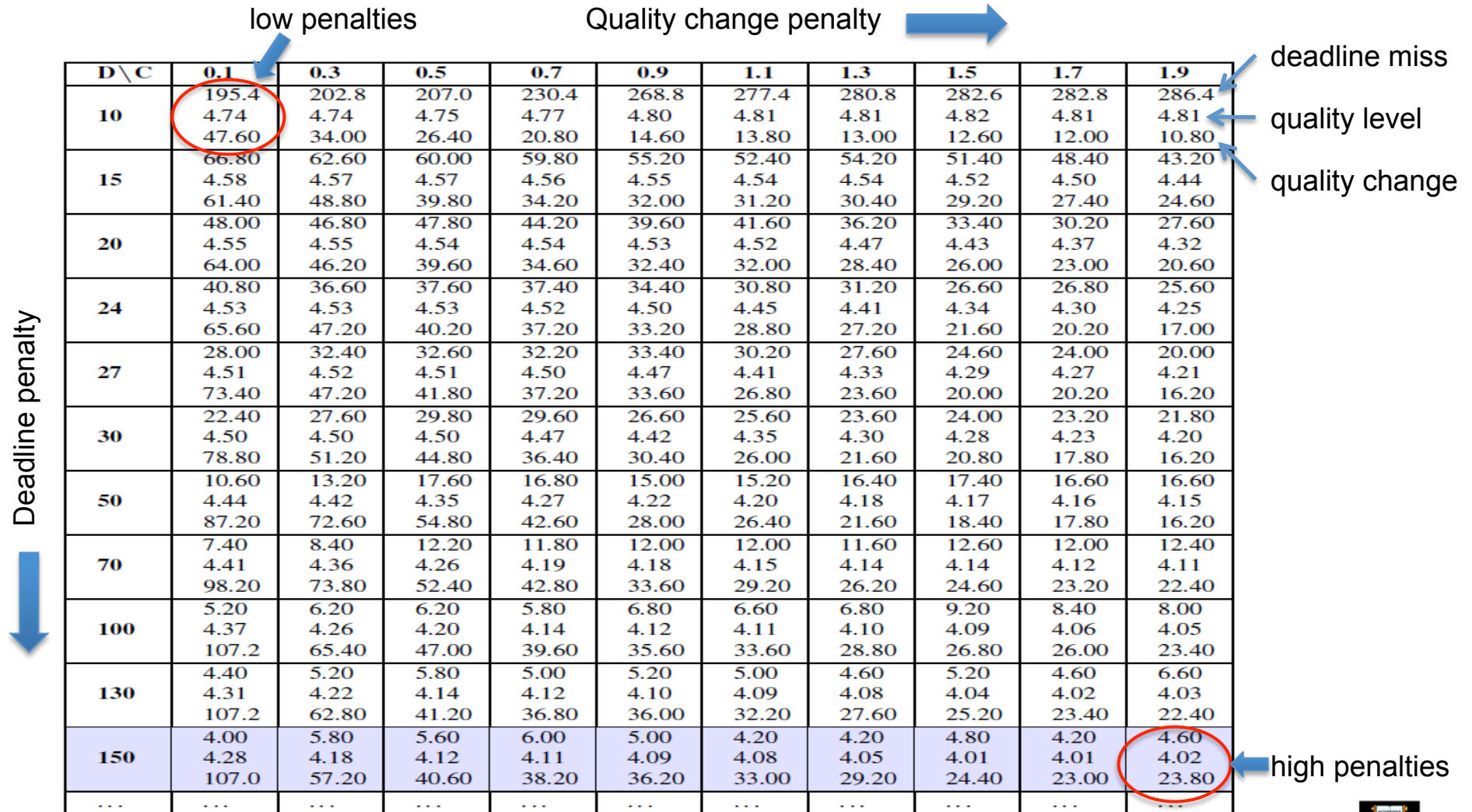
# Simulation Setup

- Big Buck Bunny encoded into 5 different bitrates (qualities) and segmented into 2-sec chunks
- Mobility and 3G bandwidth traces from actual driving in Sydney, Australia
  - 24 Km (22-30 minutes) repeated 70 times
  - Bandwidth measured at 10 sec interval
  - Useful data set to obtain CDF per road segment
  - Normal approx. appears close to empirical CDF (simplifies CDF storage and processing)

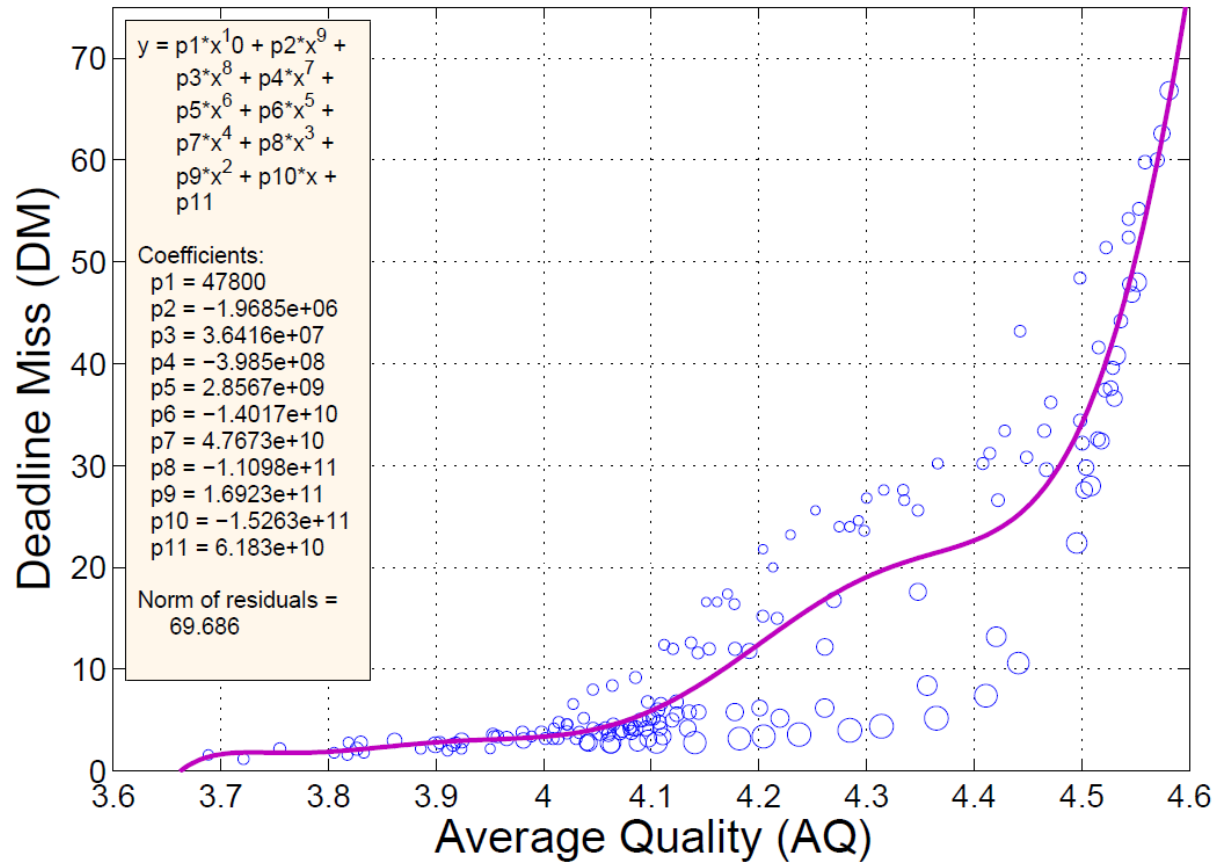
	time	latitude	longitude	bandwidth (Kbps)
1	1186549400	-33.919785	151.228913	1663.1440
2	1186549410	-33.919635	151.227787	1964.7330
3	1186549420	-33.91958	151.227322	2038.8659
4	1186549430	-33.91958	151.227322	2011.2631
5	1186549440	-33.91953	151.22692	1838.6578
6	1186549450	-33.91905	151.226322	1208.2767



# DASH performance as a function of MDP penalties



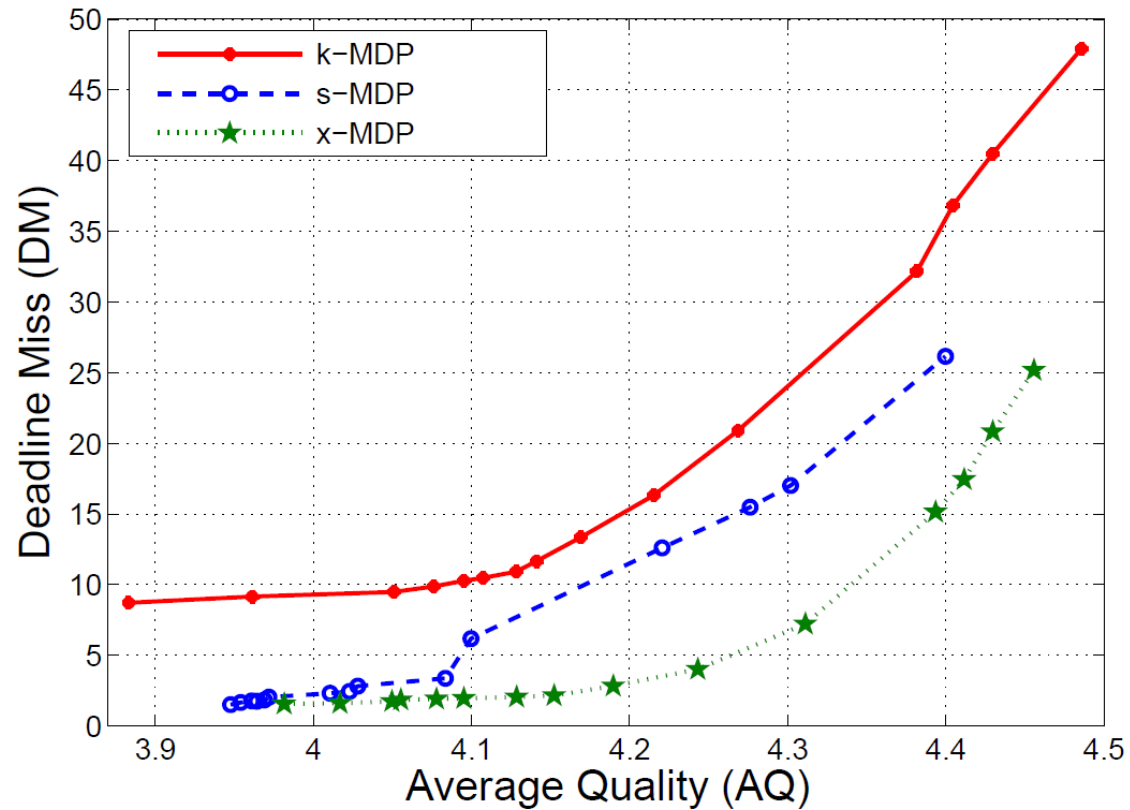
# MDP tuning allows trade off between quality and deadline miss



Performance of k-MDP with k=1



# Comparing k-MDP, s-MDP, and x-MDP



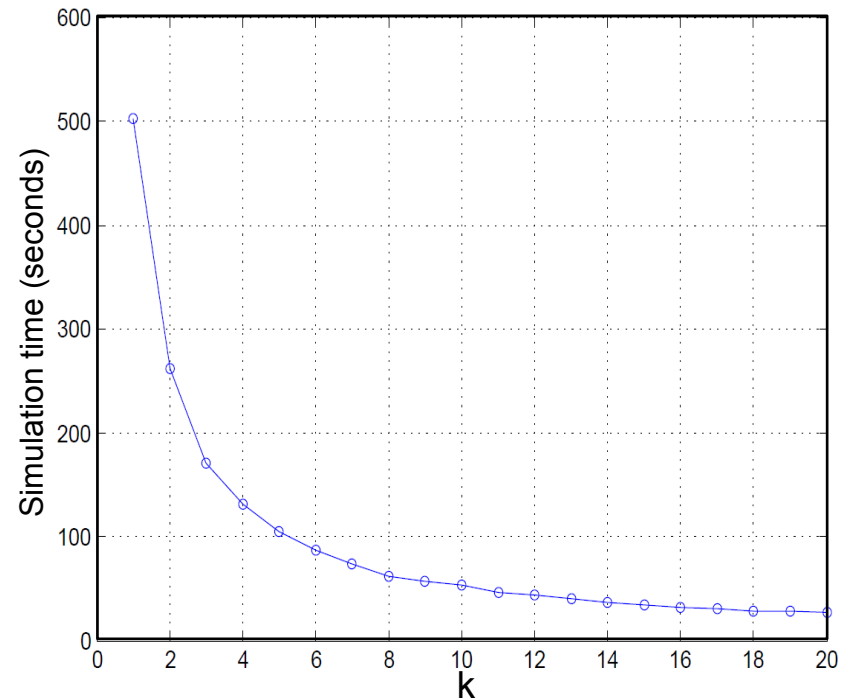
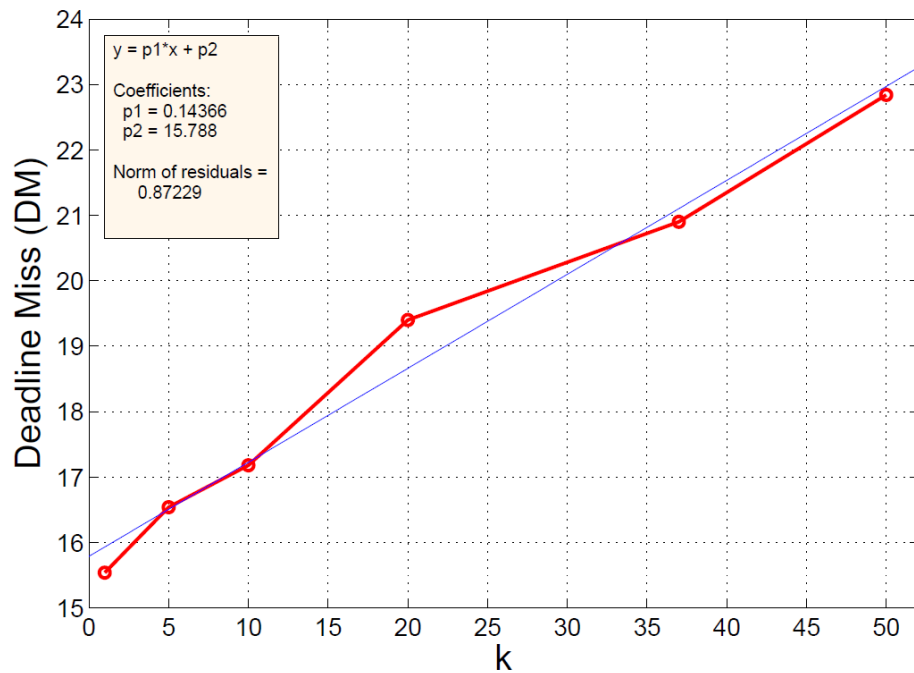
x = 1000 meter

k = 37 (about 37 chunks downloaded per 1000m on average)



# k-MDP's performance as a function of k

Deadline miss increases linearly with k



Online computation overhead is very high for  $k=1$ , but it can be reduced by a factor of  $k$  for  $k>1$



# Conclusion

- Although MDP has the potential to improve DASH performance, it increases decision making overhead in a mobile device
- We have explored 3 different implementations of MDP with a goal of reducing decision making overhead during a streaming session
- k-MDP reduces online optimization overhead of MDP by a factor of  $k$ 
  - but increases deadline miss linearly as a function of  $k$
  - it allows DASH decision making with no offline network measurements, but performs worse than offline optimization
- s-MDP and x-MDP can be used to solve MDP offline, eliminating the online computation overhead completely
  - but they require offline network measurement
- x-MDP outperforms s-MDP due to location-sensitivity of bandwidth distributions
  - but it requires more extensive offline network measurement and maintenance of a large number of MDP strategies for a given road network (one per road segment)

