



ITUKALEIDOSCOPE

NANJING 2017

Challenges for a data-driven society

Machine Learning Approach for Quality Adaptation of Streaming Video Through 4G Wireless Network over HTTP

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Outline

- Motivation
- System Model and Development
- Machine Learning Approach
- Algorithm Development
- Experimental Results
- Conclusion

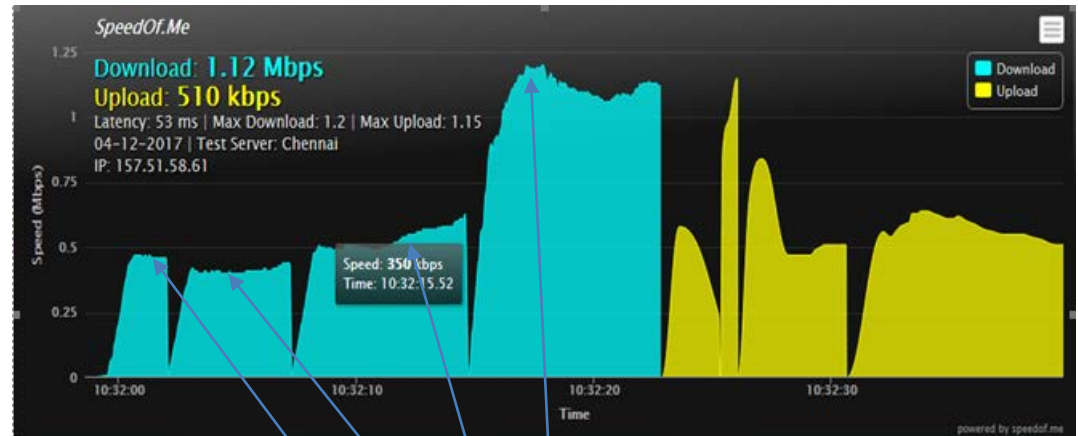


Motivation

- **Bit Rate Fluctuation in Internet over 4G Wireless**
 - Internet service based on Best-effort model
 - Received signal strength fluctuation in wireless channel

Solutions?

- Change in underlying hardware and network
- Migration to 5G
- Content adaptation on-the-top of HTTP



Trigger actions in those trends to adapt contents



System Model

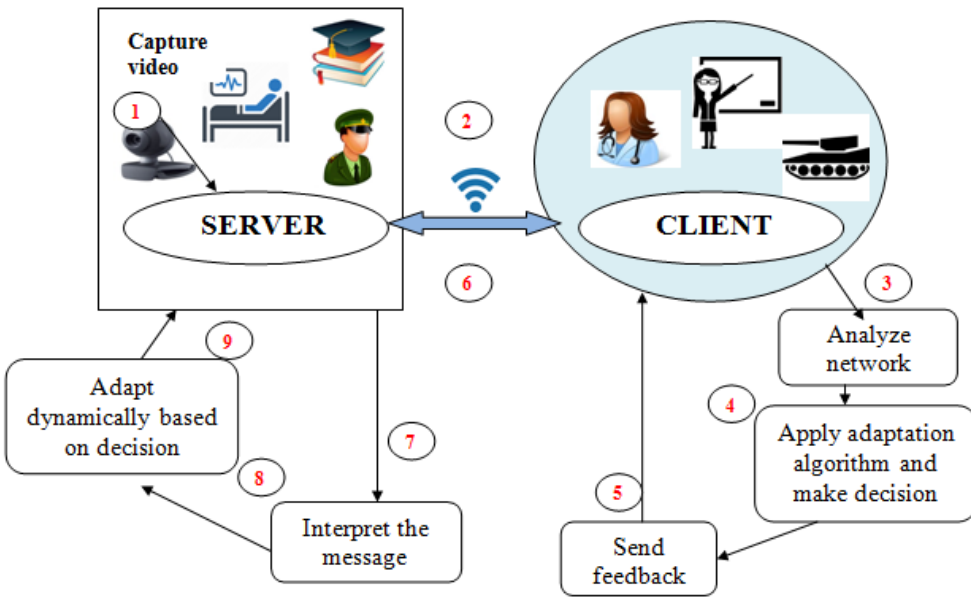


Fig.1. The Client Server Model

Client Process

Server Process

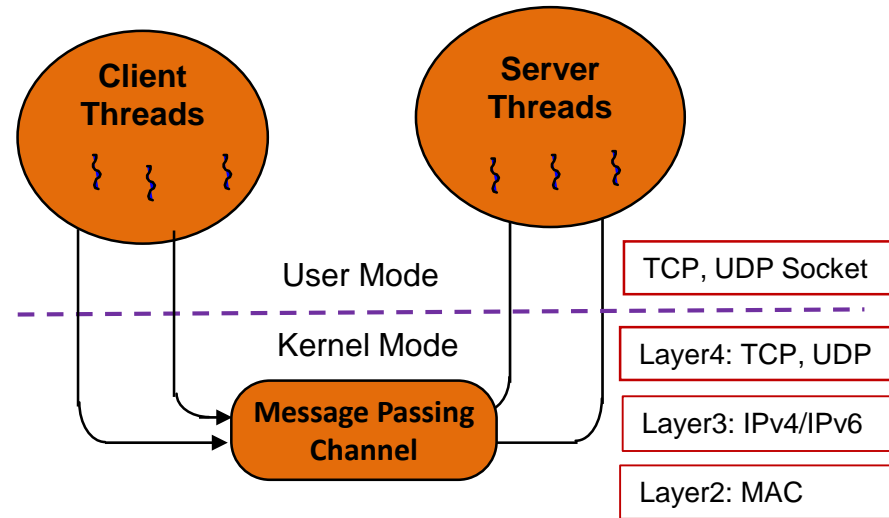


Fig.2. Multi-threaded Client-Server Communication





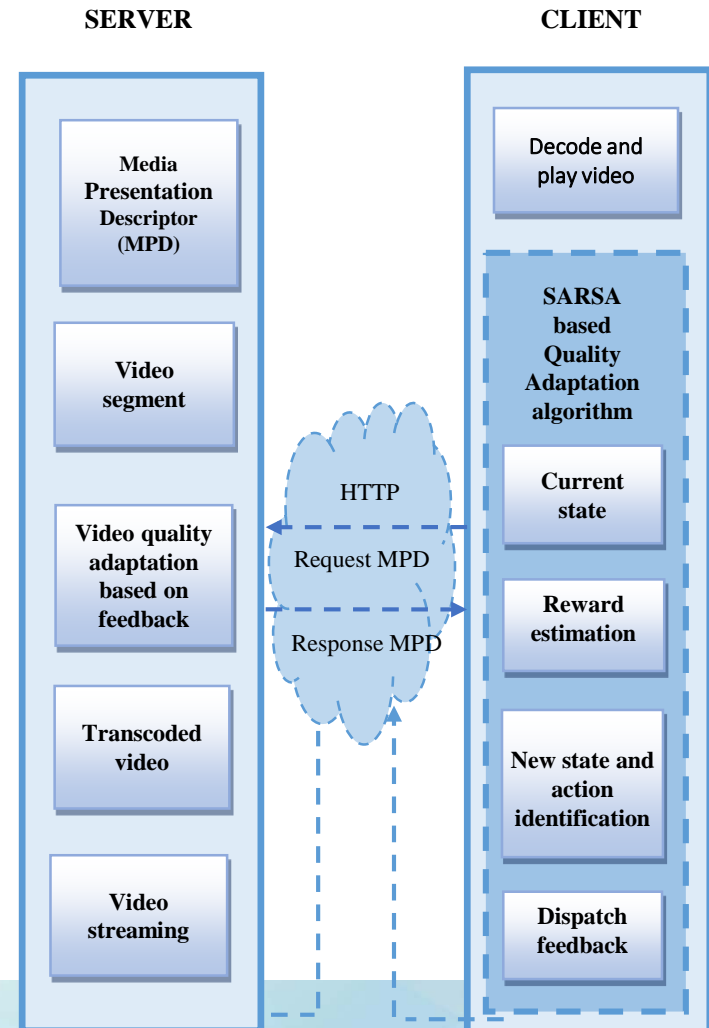
System Development

Adaptive Streaming System

- Media Capture
- Adaptation Procedure
- Content Encoding / Transcoding
- Stream Processing

Content Transcoding at Server

- ❖ Quality estimation through standard procedures (**ITU-T G.1070**) at client
- ❖ Trigger action based on machine learning (**State Action Reward State Action (SARSA)**) at client





State Action Reward State Action (SARSA) Approach

The Q-matrix is updated after each quality decision as follows.

$$Q[s_{cur}, a_{cur}] \leftarrow Q[s_{cur}, a_{cur}] + \alpha [V_q + \gamma Q(s_{new}, a_{new}) - Q(s_{cur}, a_{cur})] \quad (1)$$

where α denotes *learning rate*, γ *discount factor*, and v_q associated immediate *reward*

SARSA Base Quality Adaptation (SBQA):

(1) **Softmax Policy**: The action is chosen according to the resulting distribution, which is the Boltzmann distribution given by

$$P(a_j) = \frac{e^{-\frac{Q(s,a_j)}{r}}}{\sum_{i=1}^{|a_s|} e^{-\frac{Q(s,a_i)}{r}}} \quad (2)$$

(2) **ϵ -greedy Policy**: the agent selects at each time step a random action with a fixed probability, $0 < \epsilon < 1$

$$Action = \begin{cases} \text{random_action_from_}A(s), & \text{if } r < \epsilon \\ \text{argmax}_{a \in A(s)} Q(s, a), & \text{otherwise} \end{cases} \quad (3)$$

ITU-T G.1070 Based Video Quality Evaluation

The video quality (V_q) is represented as

$$V_q = 1 + I_c I_t \quad (4)$$

Where I_c represents the basic video quality resulting from the encoding distortion, I_t is the factor governed by degree of robustness due to packet loss.

I_c is expressed in terms of bit rate (b) and frame rate (f) according to equations (5 - 8) as follows.

$$I_c = I_0 e^{-\frac{(\ln(f) - \ln(f_0))^2}{2D_{Fr}^2}} \quad (5)$$

$$f_0 = v_1 + v_2 b \quad (6)$$

$$D_{Fr} = v_6 + v_7 b \quad (7)$$

$$I_0 = v_3 \left(1 - \frac{1}{1 + \left(\frac{b}{v_4}\right)^{v_5}} \right) \quad (8)$$

I_t depends on packet loss robustness factor ($D_{P_{plv}}$) and rate of packet loss (p) given by

$$I_t = e^{-\frac{p}{D_{P_{plv}}}} \quad (9)$$

$$D_{P_{plv}} = v_{10} + v_{11} e^{-\frac{f}{v_8}} + v_{12} e^{-\frac{b}{v_9}} \quad (10)$$



SBQA Using Softmax Policy (SBQA-SP) Algorithm

1. Initialize with number of packets N , learning rate α and discount factor γ , last state s_{last} and Q-matrix Q .
2. Compute throughput (Th) resulting from the capture of N packets.
3. Identify current state s_{cur} based on Th value.
4. Read the resolution res , and frame per second fps from the header in streamed video.
5. Determine current action a_{cur} based on the current quality segment.
6. While $s_{cur} < s_{last}$ // till last state reached
7. Read the encoded bit rate b , and compute frame loss percentage p .
8. Calculate the reward (video quality) V_q using (4)
9. Estimate the current throughput Th_{cur} and based on Th_{cur} identify new state s_{new} .
10. Compute new action $a_{new} \leftarrow \text{SoftMax}(Q,s)$.
// Exploration policy function to get best possible future action
11. Update the $Q[s_{cur}, a_{cur}]$ based on (1)
12. $s_{cur} \leftarrow s_{new}$ // Update new state to the current state
13. $a_{cur} \leftarrow a_{new}$ // Update new action to the current action
14. Assign action a_{cur} as feedback to the server
15. End
16. Go to Step 2 and continue till streaming occurs
17. End



SBQA Using ϵ -Greedy Policy (SBQA-GP) Algorithm

1. Initialize fixed probability ϵ and max // Store maximum value (max) in s^{th} row of Q -matrix.
2. Generate a random value ran in the range 0 to 1.
3. If $ran < \epsilon$
4. $selectedAction = -1$
5. Else
6. For $i = 1$ to Q_{length} // get Q_{length} from Q -matrix
7. If $Q[s,i] \geq max$
8. $selectedAction = i$ // action with max reward
9. $max = Q[s,i]$
10. End
11. If $selectedAction = -1$
12. Generate a random number r , in range of action.
13. $selectedAction = r$
14. Return $selectedAction$



Q-Learning Based Quality Adaptation (QBQA) Algorithm

1. Initialize the learning rate α , discount factor γ , Q-matrix, and optimal bandwidth value B_{opt} .
2. Read the current buffer occupancy level buf_k for k^{th} segment and quality level q_{k-1} for segment $k-1$
3. For $i = 1$ to t // *Training Phase*
4. Estimate the bandwidth bw_k
5. Assign $s_k = \{ bw_k, buf_k, q_{k-1} \}$ // *Current State*
6. $a_k = \text{Softmax}(Q, s_k)$ // *Exploration policy function to get best possible action.*
7. Calculate the quality factor related to bandwidth and buffer occupancy level using the equation

$$R_{quality} = -1.5 \cdot \left| bw_k \cdot \frac{1+(buf_k/B_{opt})}{3-(bw_1/a_1)} - a_k \right|$$
8. Calculate the quality factor related to switch in quality using the equation

$$R_{switches} = -|q_{k-1} - a_k|$$
9. Read the duration of video freeze t_{stall} , time elapsed from the last freeze t_{play} and number of freezes n

10. Calculate the quality factor related to video freezing using the equation

$$R_{freezes} = \begin{cases} -100 \cdot \left| \frac{a_k}{bw_k} \cdot \frac{e^{\frac{t_{stall}}{10}}}{\ln(t_{play}+1)} \right| a_k = a_1 \\ -100 \cdot \left| \frac{a_k}{bw_k} \cdot \frac{e^{n+\frac{t_{stall}}{10}}}{\ln(t_{play}+1)} \right| a_k \neq a_1 \end{cases}$$

11. Calculate

$$R_{total} = R_{quality} + R_{switches} + R_{freezes}$$

12. End

13. Determine the resultant state, s_{k+1} using $\{ bw_{k+1}, buf_{k+1}, q_k \}$

14. Update the Q-matrix using

$$Q[s_k, a_k] \leftarrow (1 - \alpha) Q[s_k, a_k] + \alpha [R_{total} + \gamma \max_b Q(s_{k+1}, b)]$$

15. Estimate the bandwidth bw_k // *Testing Phase begins*

16. Assign $s_k = \{ bw_k, buf_k, q_{k-1} \}$

17. $a_k = \max_a (Q(s_k, a))$

18. Send a_k as feedback to the server

19. Repeat from Step 15 until streaming occurs

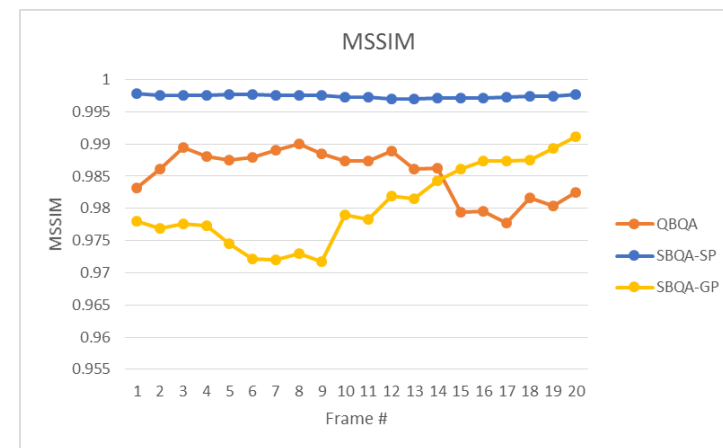
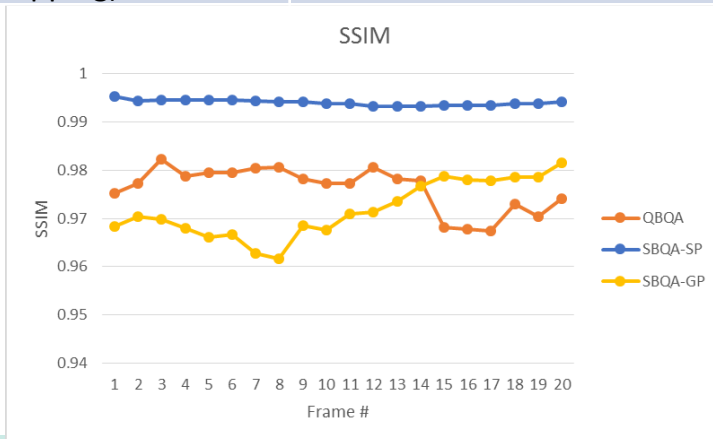
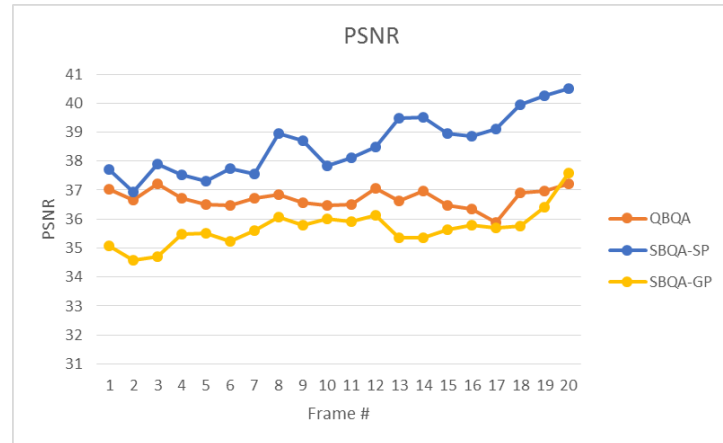




Experimental Results

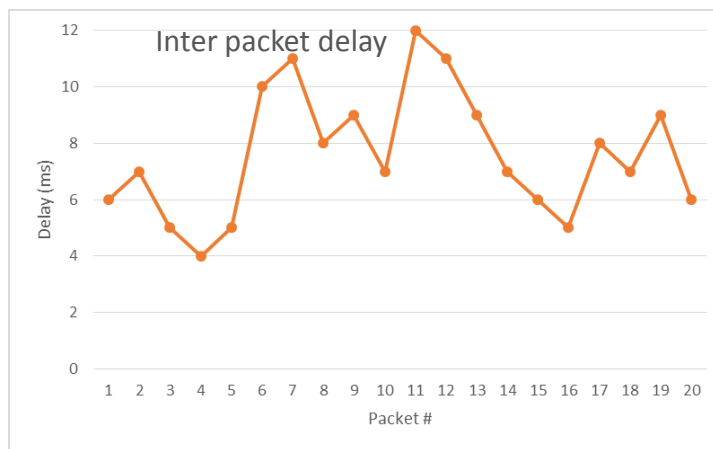
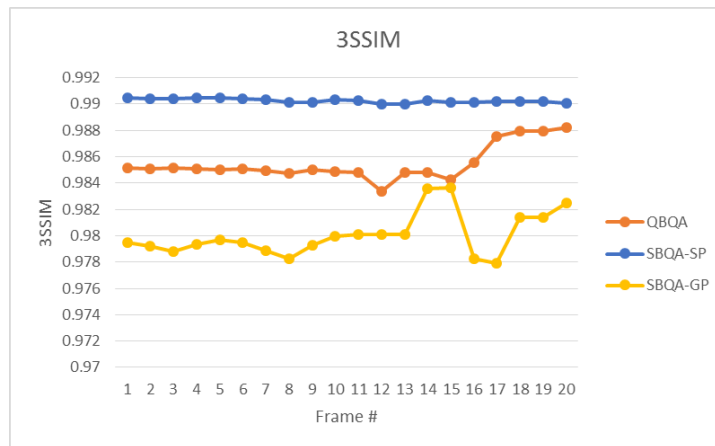
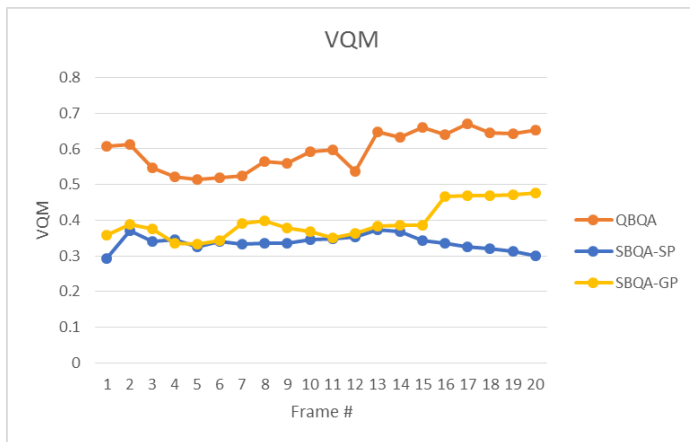
Test Parameters as per ITU-T J.247

S.no	Parameters	Values
1.	Transmission	Errors with packet loss
2.	Frame Rate	5 fps to 30 fps
3.	Video Codec	H.264/AVC (MPEG-4 part10), VC-1, Windows Media9, Real Video (RV 10), MPEG-4 Part 2
4.	Video Resolutions and bit rates	QCIF: 16 - 320 kbps CIF: 64 - 2000 kbps VGA: 128 - 4000 kbps
5.	Temporal errors (pausing with skipping)	Maximum of 2 seconds





Experimental Results cont.





Experimental Results cont.



Fig.(a) Original Live Video



Fig.(b) Decoded at Receiver



Conclusions

- ❖ On-the-top (OTT) implementation with the last mile connectivity through 4G mobile
- ❖ ITU-T G.1070 model embedded in the algorithm
- ❖ The State Action Reward State Action (SARSA) approach achieves the system goal
- ❖ Experiments conducted on live / stored video stream and results validated through FR metrics
 - ❖ PSNR, SSIM, MS-SSIM, VQM, and 3-SSIM
- ❖ Implementation of multi-casting service will require additional mechanism



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