A Receiver-Centric Rate Control Scheme for Layered Video Streams in the Internet

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Abstract

We present a new end-to-end protocol, namely *Dynamic Video Rate Control* (DVRC), which operates on top of UDP and enables the adaptive delivery of layered video streams over the Internet. The protocol optimizes the performance on video delivery with concern to friendliness with interfering traffic. DVRC enables a closed-loop control between server and client, where the receiver detects the state of congestion, determines the proper transmission rate, and eventually opts for the optimal number of layers that should be delivered according to this rate. The protocol relies on a hybrid *Additive Increase Additive Decrease* (AIAD)/*Additive Increase Multiplicative Decrease* (AIMD) algorithm (namely AIAMD) that manages to differentiate congestive and non-congestive loss by utilizing history in its control rules. AIAMD combines the most desirable features of AIAD and AIMD, reacting gently to random loss and more aggressively to congestion, adapting effectively to the dynamics of the network. Therefore, DVRC enables the desired smoothness for video streaming applications and at the same time avoids significant damage during congestion. Exploring DVRC's potential through extensive simulations, we identify notable gains in terms of bandwidth utilization and smooth video delivery. Furthermore, our results indicate that the protocol allocates a well-balanced amount of network resources maintaining friendliness with corporate TCP connections.

Keywords: *end-to-end protocols, rate control, Quality of Service, video streaming*

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1. Introduction

Multimedia applications gain popularity and streaming media data is expected to compose a considerable portion of the overall data traffic traversing the Internet. These applications generally prefer timeliness to reliability. Video streaming, in particular, calls for strict requirements on end-to-end latency and delay variation. Long end-to-end delays commonly affect the timely delivery of video-data causing data unavailability and unintelligible real-time interaction with frustrating consequences to the end-user. Furthermore, reliability parameters, such as packet loss and bit errors, usually compose an impairment factor, since they cause a perceptible degradation on video quality. Unlike bulk data transfers, video streaming seeks to achieve smooth playback quality rather than simply transmit at the highest attainable bandwidth.

Multimedia applications commonly *rely* on the unreliable transport services provided by *User Datagram Protocol* (UDP). UDP is a fast, lightweight protocol without any transmission or retransmission control. The protocol does not have functionality to override application characteristics, such as its transmission rate. It simply transmits at application rate and pattern. However, the lack of congestion control poses a threat to the network: had such applications dominated the Internet, it would have faced risk of congestion collapse. In this context, Internetworking functionality evolves towards punishing free-transmitting protocols.

On the other hand, *Transmission Control Protocol* (TCP), based on the principles of congestion management [14], *Slow-Start* [32], and *Additive Increase Multiplicative Decrease* (AIMD) [6], provides a reliable data delivery service to Internet applications and is in large part responsible for the remarkable stability of the Internet. However, the protocol introduces arbitrary delays, since it enforces reliability and in-order delivery. Furthermore, the process of probing for bandwidth and reacting to the observed congestion induces oscillations in the achievable transmission rate. The variations in sending rate can be theoretically smoothed out with application-level buffering, but this could result in huge client buffers and unacceptable end-to-end delays depending on the extent of fluctuations.

Several TCP protocol extensions [1, 15, 21, 27, 34] have emerged to overcome the standard TCP limitations providing more efficient bandwidth utilization and sophisticated mechanisms for congestion control, which preserve the fundamental *Quality of Service* (QoS) guarantees for multimedia traffic. *TCP-friendly* protocols, proposed in [10, 36, 37], achieve smooth window adjustments while they manage to compete fairly with TCP flows. In order to achieve smoothness, they use gentle backward adjustments

upon congestion. However, this modification has a negative impact on protocol responsiveness. In [35, 23] we showed that TCP-friendly protocols are unable to effectively recover from excessive congestion incidents resulting in increased packet drops, which eventually degrade application performance. Equation-based rate control [10], which can be construed as the opposite end of the spectrum of linear congestion control algorithms (e.g. AIMD), has been promoted as an attractive option for multimedia Internet transmission. It has the desirable feature of delivering maximally smooth, TCP-fair transmission rate. However, it is characterized by very slow responsiveness to network dynamics, as well.

Considering TCP's limitations and the impending threat of unresponsive UDP traffic, we need sophisticated congestion control that interacts efficiently with other flows on the Internet. An overview of Internet's current congestion control paradigm reveals that routers play a relatively passive role: they merely indicate congestion through packet drops or *Explicit Congestion Notification* (ECN). It is the end-systems that perform the crucial role of responding appropriately to these congestion signals. Furthermore, since video encoding involves inter-frame dependencies, the random dropping of packets by routers can seriously degrade video quality. In MPEG, for example, dropping packets from an independently encoded I (*intra* picture) frame causes the following dependent P (*predictive*), and B (*bidirectional*) frames not to be fully decodable. In practice, inter-frame dependencies may render a 3% packet loss rate up to a 30% frame loss rate [2].

Without explicit feedback, end-to-end congestion measure can only be loss probability, as used in TCP Reno and *TCP-friendly Rate Control* (TFRC) [10], or queuing delay, as used in TCP Vegas [3] and FAST TCP [15]. However, recent studies [20] uncovered that delay and packet loss can have a weak correlation, especially when packet losses occur due to other reasons than buffer overflow (i.e. wireless errors). As a result, using delay as a measure of congestion may cause undesirable effects in terms of bandwidth utilization, as well as network stability, especially if it is not augmented with loss information.

At the same time, numerous video-streaming applications have implemented their own congestion control mechanisms, usually on a case-by-case basis on top of UDP. Most applications exploit *Real-Time Control Protocol* (RTCP) [31] and *Real-Time Streaming Protocol* (RTSP) [30] to enable controlled delivery of video. RTCP, in particular, allows the video application to exploit feedback of reception statistics (i.e. received and lost packets, jitter, round-trip delay) and adjust the sending rate accordingly. In the presence of such an end-to-end protocol, a video streaming server can adapt the quality of its transmission. The manipulation of a compressed video stream can be attained by techniques, such as *temporal* and *quality* scalability [19]. According to temporal scaling, the streaming server selectively discards frames prior to transmission. *Simulcast* and *layered adaptation* are the most remarkable quality

scalability techniques, which directly adjust the bitrate of the transmitted video stream. Despite the presence of such mechanisms, implementing application-level congestion control is still difficult and eventually is not part of most applications needs. We believe that a new transport protocol is needed, which would combine unreliable datagram delivery with built-in congestion control. This protocol would act as an enabling technology: new and existing applications could use it to timely transmit data without destabilizing the Internet.

In this context, we have been working on a rate control scheme for adapting outgoing video streams to the characteristics of the end-to-end network path. Rate adaptive video streams offer the clients the benefit of being resilient to changing network conditions and allow a large number of streams to concurrently share network resources. Multimedia streams can be adaptive, since user-perceived QoS is often satisfactory over a range of stream compression levels. Although this adaptivity is limited (i.e. multimedia streams have minimum subscription levels, below which service quality is unacceptable), they have the capability of adjusting their subscription levels in response to congestion, much as elastic flows do. Along these lines, we designed a new end-to-end protocol, namely *Dynamic Video Rate Control* (DVRC), which operates on top of UDP and desirably provides out-of-order delivery. DVRC is intended to interact with a plethora of multirate streaming applications that rely on cumulative layered transmission [19]. This approach is based on information decomposition. The video stream is encoded at a *base layer* and one or more *enhancement layers*, which can be combined to render the stream at high quality. Layered adaptation is performed by adding or dropping enhancement layers depending on the prevailing network conditions.

DVRC enables a closed-loop control between server and client, monitoring the transport's progress separately for transient and persistent rate changes, and in response actuating the video stream quality-rate trade-off. The protocol employs a receiver-centric congestion control mechanism and does not rely on QoS functionality in routers, such as *Random Early Drop* (RED), ECN or other *Active Queue Management* mechanisms. Inline with [13, 27, 34], certain protocol functionalities can be moved from the sender to the receiver. The sender merely acts based on the requests from the receiver. Generally, delegating the protocol's control functions to the receivers composes an elegant and functional approach in several occasions:

- Receiver-oriented error control incarnates the property of the receiver to determine with better accuracy the data delivery rate and the potential level of data loss. This abrogates the impact of false assessments at the sender due to lost or delayed acknowledgements.

- In wired/wireless environments, the receiver is adjacent to the wireless last-hop and has first knowledge about the characteristics of the wireless links.

- Receiver-centric transport protocols can significantly reduce the complexity of the server implementation, since they distribute the state management across a large number of clients.

DVRC relies on a hybrid *Additive Increase Additive Decrease* (AIAD)/AIMD scheme, namely *AIAMD*, in order to adapt its sending rate. In contrast to the *memory-less* AIAD and AIMD algorithms, AIAMD utilizes *history* information in its control rules. A very limited number of proposals relying on linear congestion control schemes exploit history information (e.g. [16]). In summary, AIAMD has the following salient attributes: (i) utilizes history of receiving rates in order to distinguish between congestion-induced and random loss, (ii) reacts gracefully to random loss in order to keep the sending rate variation to minimum, but reacts quickly to the onset of congestion, and (iii) is both efficient and fair to the target environment. As a result, DVRC enables the desired smoothness for video streaming applications and at the same time avoids significant damage during congestion. Our simulations demonstrate that DVRC provides low variation in the transmission rate in steady state, and at the same time is reactive and TCP-friendly. We also show that satisfactory performance can be achieved with a small number of layers (4-5 layers).

The remainder of this paper is organized as follows. Section 2 provides an overview of related work, while in Section 3 we discuss the design and implementation details of the proposed rate control scheme. In Section 4, we present the parameters of our evaluation methodology, followed by Section 5, where we demonstrate conclusive performance studies based on extensive simulations. Finally, in Section 6 we highlight our conclusions.

2. Related Work

The literature includes numerous studies and proposals towards efficient rate/congestion control for multimedia applications in the Internet. Generally, we can distinguish two approaches for rate control of multimedia traffic in the Internet:

- TCP-like protocols (e.g. [26, 27]): Rate control is performed in TCP fashion, where packet loss infers congestion, and subsequently the transmission rate is reduced multiplicatively.

- Equation-based protocols (e.g. [10]): The available bandwidth is estimated based on statistics of the *Round Trip Time* (RTT) and packet loss probability. In response to the bandwidth estimates obtained, the source adjusts the transmission rate accordingly.

Rate Adaptation Protocol (RAP) [26] is a rate-based protocol that employs an AIMD algorithm for the transmission of real-time streams. The sending rate is continuously adjusted by RAP in a TCP-friendly fashion by using feedback from the receiver. However, since RAP employs TCP's congestion control parameters (i.e. 1, 0.5), it causes short-term rate oscillations, primarily due to the multiplicative decrease. Furthermore, RAP occasionally does not result in inter-protocol fairness. *TCP Emulation at the Receivers* (TEAR) [27] enables the receiver to emulate the congestion window modifications of a TCP sender, based on the congestion signals observed at the receiving host. The receiver maintains an exponentially weighted moving average of the congestion window, and divides this amount by the estimated RTT to obtain the TCP-friendly sending rate.

TFRC [10] is a representative equation-based protocol, which adjusts its transmission rate in response to the level of congestion, as estimated based on the calculated loss rate. Multiple packet drops in the same RTT are considered as a single loss event by TFRC and hence, the protocol follows a more gentle congestion control strategy. More precisely, the TFRC sender uses the following TCP response function:

$$T(p, RTT, RTO) = \frac{1}{RTT\sqrt{\frac{2p}{3}} + RTO\ (3\sqrt{\frac{3p}{8}})\ p\ (1+32p^2)} \quad (1)$$

where p is the steady-state loss event rate and *RTO* is the retransmission timeout value. Equation (1) enforces an upper bound on the sending rate T. However, the throughput model is quite sensitive to parameters (e.g. p, RTT), which are often difficult to measure efficiently and to predict accurately. Also, the long-term TCP throughput equation does not capture the transit and short-lived TCP behaviors, and it is less responsive to short-term network and session dynamics. According to [10], TFRC's increase rate never exceeds 0.14 packets per RTT (or 0.28 packets per RTT when history discounting has been invoked). In addition, the protocol requires 5 RTTs in order to halve its sending rate. Consequently, the instantaneous throughput of TFRC has a much lower variation over time. TFRC eventually achieves the smoothing of the transmission gaps and therefore, is suitable for applications requiring a smooth sending rate. However, this smoothness has a negative impact, as the protocol becomes less responsive to bandwidth availability [35].

Datagram Congestion Control Protocol (DCCP) [17] is a new transport protocol that provides a congestion-controlled flow of unreliable datagrams. DCCP is intended for delay-sensitive applications which have relaxed packet loss requirements. The protocol aims to add to a UDP-like foundation the minimum mechanisms necessary to support congestion control. DCCP provides the application with a choice of congestion control mechanisms via *Congestion Control IDs* (CCIDs), which explicitly name standardized congestion control mechanisms. CCIDs are negotiated at connection startup. Currently, two CCIDs have been developed supporting TCP-like and TFRC congestion control, respectively.

Since TCP is rarely chosen to transport delay-sensitive traffic over the Internet, numerous TCP-friendly protocols [36, 37] constitute an elegant framework for multimedia applications. We consider as TCP-friendly any protocol whose long-term arrival rate does not exceed the one of any conformant TCP in the same circumstances [9]. *GAIMD* [37] is a TCP-friendly protocol that generalizes AIMD congestion control by parameterizing the additive increase rate α and multiplicative decrease ratio β. For the family of AIMD protocols, authors in [37] derive a simple relationship between α and β in order to be friendly to standard TCP:

$$\alpha = \frac{4(1-\beta^2)}{3}$$

Based on experiments, they propose an adjustment of $\beta = 0.875$ as an appropriate smooth decrease ratio, and a moderated increase value $\alpha = 0.31$ to achieve TCP friendliness.

TCP Westwood [21] is a TCP-friendly protocol that emerged as a sender-side-only modification of TCP Reno congestion control. TCP Westwood exploits end-to-end bandwidth estimation in order to adjust the values of slow-start threshold and congestion window after a congestion episode. The protocol incorporates a recovery mechanism which avoids the blind halving of the sending rate of TCP Reno after packet losses and enables TCP Westwood to achieve high link utilization in the presence of wireless errors. The specific mechanism considers the sequence of bandwidth samples *sample_BWE[n]* obtained using the *ACKs* arrivals and evaluates a smoothed value, *BWE[n]*, by low-pass filtering the sequence of samples, as described by the following pseudocode:

Algorithm 1. TCP-Westwood

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if an ACK is received

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sample_BWE[n] = (acked * pkt_size * 8) / (now - last_ACK_time)
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BWE[n] = (1 - beta) * (sample_BWE[n] + sample_BWE[n - 1]) / 2 + beta * BWE[n - 1]

end if

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where *acked* is the number of segments acknowledged by the last *ACK*; *pkt_size* is the segment size in bytes; *now* is the current time; *last_ACK_time* is the time the previous *ACK* was received; *beta* is the pole used for the filtering (a value of 19/21 is suggested). TCP Westwood+ is a recent extension of TCP Westwood that computes one sample of available bandwidth every RTT, using all data acknowledged in the specific RTT [11].

TCP-Real [34] is a high-throughput transport protocol that incorporates congestion avoidance mechanism in order to minimize transmission-rate gaps. The protocol approximates a receiver-oriented approach beyond the balancing trade of the parameters of additive increase and multiplicative decrease. TCP-Real introduces another parameter, namely γ, which determines the window adjustments during congestion avoidance. This parameter can be adaptive to the detected conditions. Generally, TCP-Real can be viewed as a TCP (α, β, γ) protocol, where γ captures the protocol's behavior prior to congestion when congestion boosts up.

Numerous studies for adaptive video delivery appear in [7, 8, 19, 25, 29]. An overview of existing solutions for video adaptation is presented in [19]. Authors in [8] analyze the impact of selected congestion control algorithms on the performance of streaming video delivery. They concentrate on binomial congestion control [1] and especially on SQRT, which responds to packet drops by reducing the congestion window size proportional to the square root of its value instead of halving it. However, binomial schemes are not able to achieve TCP-friendliness independent of link capacity [4]. In [7] a *Real-Time Control* (RTP) [31] compatible protocol (i.e. SR-RTP) is proposed, which adaptively delivers high quality video in the face of packet loss. SR-RTP effectively enables selective reliability, retransmitting only the important data. Finally, authors in [25] propose a layered mechanism to adapt the quality of congestion-controlled video. The mechanism is able to control the level of smoothing in order to improve the quality of the delivered video stream.

3. Design and Implementation

The design principles of the proposed rate control scheme mainly rest on the assumption that user's perception is sensitive to smooth and timely playback of the received video. Despite the degradation in

visual quality, smooth video of lower bitrate is considered more preferable than inconsistent and jerky video of higher rate. In this context, the primary goal of DVRC is to deliver the optimal number of layers that the client can manage according to the prevailing network conditions. Layered video adaptation is performed in terms of user-perceived quality, as well as from the perspective of inter-protocol friendliness.

3.1 Sender and Receiver Functionality

DVRC, in a complementary role, operates on top of UDP and supports rate control relying on sender and receiver interaction. DVRC acknowledges each datagram received by transmitting a control packet. The protocol does not integrate reliability to UDP datagrams, so control packets do not trigger retransmissions. Although DVRC uses the control packets to update the sending rate, data transmissions themselves are not directly triggered by control packets, but instead are sent out based on the determined rate at the receiving end. Therefore, DVRC does not employ a self-clocking mechanism, inline with most rate-based protocols (e.g. TFRC, RAP, TEAR). The presence of self-clocking would enforce DVRC to reduce its transmission rate drastically when the available bandwidth has decreased. Such behavior is more suited to TCP which commonly transports bulk-data traffic without timing considerations. In the case of DVRC, the magnitude of rate decrease on the occurrence of packet loss is directly controlled by AIAMD, as we discuss in the following subsection.

We have encapsulated additional header information to UDP datagrams (Fig. 1), including *packet type*, *length* and *sequence number*, *frame type*, *timestamp* and *video layer*. The DVRC header includes all necessary fields to add RTP functionality, and to enable rate control. Therefore, we do not use RTP avoiding its overhead. *Packet type* field denotes whether a segment with video-data or a control packet is transmitted. *Video layer* field includes the cumulative number of layers that should be delivered based on current conditions. *Timestamp* field is used to handle RTT computation. More precisely, when the sender transmits a video-packet, it updates the specific field with current time Ts_n. As soon as the receiver acquires the packet, it generates a control packet attaching Ts_n to the *timestamp* field. Upon the receipt of the corresponding feedback, the sender subtracts the included *timestamp* from current time in order to estimate the RTT sample. If Ts_n' denotes the time the control packet has been received, the sender gets the observed value of each RTT, as follows:

$$\text{SampleRTT} = Ts_n' - Ts_n$$

Therefore, DVRC obtains an accurate approximation of RTT, which does not require synchronization between sender's and receiver's clocks.

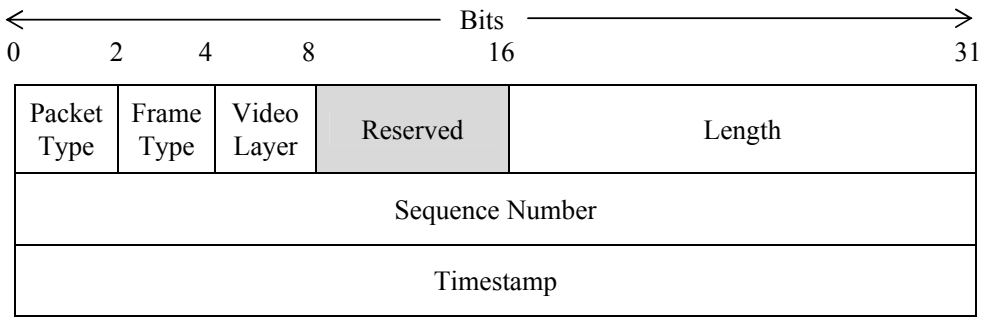


Figure 1. DVRC Header

Since UDP is a protocol without reliability, some datagrams may be lost due to congestion or inability of the receiving host from reading the packets rapidly enough. The receiver uses packet drops or re-ordering as congestion indicator. Consequently, congestion control is triggered, when:

- a packet is received carrying a sequence number greater than the expected sequence number
- the receiver does not acquire any packets within a timeout interval

Along these lines, the proper adjustment of the timeout interval is critical. A timeout interval that is set too short will claim false packet losses resulting in a wasteful reduction of the transmission rate. On the other hand, a long and consequently conservative timeout interval will inevitably impact the protocol responsiveness. In order to properly adjust the timeout, we exploit RTT measurements (*SampleRTT*) and based on this quantity we further compute the weighted average of RTT:

$$\text{EstimatedRTT} = \gamma \times \text{EstimatedRTT} + (1 - \gamma) \times \text{SampleRTT} \quad (2)$$

setting the smoothing factor γ to 0.9. After RTT estimation, the timeout interval for DVRC (DTO) can be calculated by:

$$\text{DTO} = \text{EstimatedRTT} + \delta \times \text{Deviation} \quad (3)$$

where δ is set to 4 and *Deviation* is the smoothed estimation of the variation of RTT. Deviation is expressed as:

$$\text{Deviation}_n = \varepsilon \times \text{Deviation}_{n-1} + (1 - \varepsilon) \times | \text{EstimatedRTT} - \text{EstimatedRTT}' | \quad (4)$$

where $Deviation_{n-1}$ and $EstimatedRTT'$ are the variation of RTT and the estimated RTT in the last round respectively, while ε is set to 0.25.

3.2 Rate Adjustment

The rate adjustment of the video stream is performed on a receiver-oriented fashion. The sender merely acts based on the requests from the receiver. The selection of the rate control parameters is based on the following directions: (i) to enable the desired smoothness for video streaming applications, (ii) to avoid significant damage during congestion, and (iii) to maintain friendliness with coexisting TCP flows. Although some damage is inevitable at periods of congestion, a smooth backward adjustment has the potential to enhance application performance. On the contrary, multiplicative decrease causes transmission gaps that hurt the performance of multimedia applications, which subsequently experience jitter and degraded throughput.

According to the model in [6], we consider n users sharing a single bottleneck link. In order to simplify our analysis and focus on the behavior of AIAMD control, our model inherits the assumptions of *Chiu-Jain* model: (i) all users have the same RTT and (ii) adjust their loads simultaneously. In [35] we extended the *Chiu-Jain* model by taking into consideration the role of the bottleneck queue. Furthermore, authors in [12] provide an extension of this model for flows with different RTTs. Incorporating such extensions into the AIAMD control is not in the scope of this paper and will compose future work.

Following the *Chiu-Jain* model, we consider a discrete timescale where every instant t corresponds to the moment when each user adjusts its load. If during time slot t, the i^{th} user's load is $x_i(t)$, then the total load at the bottleneck resource would be:

$$x(t) = \sum_{i=1}^{n} x_i(t) \quad (5)$$

The network provides the users with a binary feedback y(t), which indicates whether the total load x(t – 1) after the previous adjustment exceeds an optimal value X_{goal}:

$$y(t) = \begin{cases} 1 & \text{if } x(t-1) > X_{goal} \\ 0 & \text{if } x(t-1) \le X_{goal} \end{cases} \quad (6)$$

Linear congestion control algorithms are governed by the following update function:

$$x_i(t) = \begin{cases} \alpha_I + \beta_I\, x_i(t-1), & \text{if } y(t) = 0 \\ \alpha_D + \beta_D\, x_i(t-1), & \text{if } y(t) = 1 \end{cases} \tag{7}$$

In the case of AIAD ($\beta_I = \beta_D = 1$), user *i* responds to binary feedback y(t), as follows:

$$x_i(t) = \begin{cases} \alpha_I + x_i(t-1), & \text{if } y(t) = 0 \\ \alpha_D + x_i(t-1), & \text{if } y(t) = 1 \end{cases} \tag{8}$$

where $\alpha_I > 0$ and $\alpha_D < 0$. Following AIMD ($\beta_I = 1$, $\alpha_D = 0$), the corresponding response for user *i* is:

$$x_i(t) = \begin{cases} \alpha_I + x_i(t-1), & \text{if } y(t) = 0 \\ \beta_D\, x_i(t-1), & \text{if } y(t) = 1 \end{cases} \tag{9}$$

where $\alpha_I > 0$ and $0 < \beta_D < 1$. We note that neither of these controls utilizes history information; the increase and decrease rules depend solely on current load and parameters α_I, α_D, β_I and β_D.

Before analyzing the behavior of AIAMD, we investigate the efficiency of hybrid congestion controls algorithms in terms of fairness. Consider a system in a steady state with synchronous congestion signals and static bandwidth. Following a linear congestion control algorithm, each flow would experience a periodic cycle of increases and decreases in its rate *R*. We assume that such a sequence consists of S_I increases followed by S_D decreases. Based on the generalized model of linear congestion control, as expressed in (7), we formulate an expression for the rate variation in steady state. After a sequence of S_I increases, the rate evolves at $\alpha_I \sum_{h=1}^{S_I} \beta_I^{\,h} + \beta_I^{\,S_I} R$, while after S_D decreases, it becomes $\alpha_D \sum_{h=1}^{S_D} \beta_D^{\,h} + \beta_D^{\,S_D} R$. Since the transmission rates before and after the sequence of S_I increases and S_D decreases are equal, rate *R* in steady state can be expressed by the following equation:

$$\begin{aligned} R &= \alpha_D \sum_{h=1}^{S_D} \beta_D^{\,h} + \beta_D^{\,S_D} \left(\alpha_I \sum_{h=1}^{S_I} \beta_I^{\,h} + \beta_I^{\,S_I} R \right) = \\ &= \alpha_D \sum_{h=1}^{S_D} \beta_D^{\,h} + \alpha_I \beta_D^{\,S_D} \sum_{h=1}^{S_I} \beta_I^{\,h} + \beta_D^{\,S_D} \beta_I^{\,S_I} R = \\ &= A + B\,R \end{aligned} \tag{10}$$

where $A = \alpha_D \sum_{h=1}^{S_D} \beta_D^{\,h} + \alpha_I \beta_D^{\,S_D} \sum_{h=1}^{S_I} \beta_I^{\,h}$ and $B = \beta_D^{\,S_D} \beta_I^{\,S_I}$. We differentiate three cases:

(i) A = 0 which gives $\alpha_I = \alpha_D = 0$ (Multiplicative Increase/Multiplicative Decrease). Therefore equation (10) is written as R = B R. Since R > 0, B = 1 allowing any value of R.

(ii) B = 1, thus $\beta_I = \beta_D = 1$ (AIAD) or a specific case of $\beta_I \beta_D = 1$. Equation (10) now gives R = A + R which enforces A = 0. Likewise, there is no restriction to the value of R.

(iii) $A \neq 0$ and $B \neq 1$. In this case, R = A/(1 - B), and subsequently there is a single value of R in steady state where all flows can converge.

Following these observations, hybrid (linear) controls achieve fairness and can be safely used without implications on network stability.

DVRC employs a hybrid AIAD/AIMD (AIAMD) scheme beyond the conventional approach of purely additive increase and multiplicative decrease. According to AIAD, in the absence of packet loss, the rate is gracefully increased in order to probe for additional bandwidth; otherwise, the transmission rate is gently decreased in order to alleviate congestion. The graceful rate adjustments of AIAD overcome a number of problems associated with AIMD rate control. More precisely, AIAD is less susceptible to random loss[1], results in higher link utilization and does not induce significant fluctuations in the transmission rate. However, AIAD's responsiveness is poor upon sudden congestion, due to its additive decrease policy. More precisely, invoking an additive decrease in response to severe congestion, the sender will not throttle aggressively enough and the loss will persist. In such case, AIMD responds more aggressively in order to confine packet loss. The integrated rate control algorithm combines the strengths of AIAD and AIMD reacting gently to random loss and more aggressively to congestive loss, adapting effectively to the dynamics of the network.

AIAMD is able to differentiate congestive and non-congestive loss by maintaining a history of the receiving rates throughout the connection. Observations of the network dynamics and event losses are frequently assumed within a time period of an *epoch*. We consider an epoch as the time period between two observed loss events (i.e. during an epoch the transmission rate evolves uninterrupted). The receiving rates at the end of each epoch are useful, since they compose a good predictor for the congestion state for the following epochs. We define the state variable $\Re$, which is the receiving rate updated at the end of each epoch. DVRC monitors the progress of the receiving rate, and subsequently keeps track of the profile of $\Re$ by maintaining two variables: the moving average of the receiving rate $\overline{\Re}$ and the average deviation $\overline{D}$ of the receiving rate. We also use a constant *n* which gives a pre-determined weight to $\overline{D}$.

[1] Random loss is not uncommon in wireless/mobile systems, due to temporary loss of lower-level connectivity (e.g. fading, handoffs).

Based on the progress of $\overline{\Re}$ and $\overline{D}$, $\overline{\Re} - n\overline{D}$ can be used as a threshold between congestion and random loss. On the occurrence of packet loss, the instant value of $\overline{\Re} - n\overline{D}$ is compared to the currently measured rate, determining the appropriate recovery strategy. A receiving rate within the $\overline{\Re} - n\overline{D}$ bound indicates a small deviation from the average rate $\overline{\Re}$, allowing the interpretation of a temporary loss. In this case, AIAMD invokes an additive decrease in the transmission rate. On the other hand, a measured rate below $\overline{\Re} - n\overline{D}$ indicates a considerable rate decrease due to a congestion incident. Subsequently, the protocol infers congestion and triggers a multiplicative decrease. Fig. 2 illustrates the loss differentiation applied by AIAMD, where $\overline{\Re}_e$ and $\overline{D}_e$ denote the average receiving rate and the average deviation of that rate at the end of e^{th} epoch, respectively.

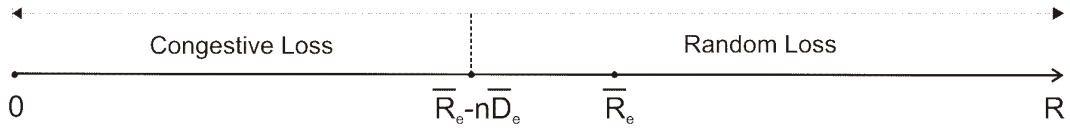


Figure 2. AIAMD Loss Differentiation

With respect to equations (5)-(9), the transmission rate for DVRC is determined based on the following algorithm:

$$R(t) = \begin{cases} \alpha_I + R(t-1), & \text{if } y(t) = 0 \\ \alpha_D + R(t-1), & \text{if } y(t) = 1 \text{ and } R(t-1) \geq \overline{\Re} - n\overline{D} \\ \beta_D R(t-1), & \text{if } y(t) = 1 \text{ and } R(t-1) < \overline{\Re} - n\overline{D} \end{cases} \quad (11)$$

We adopt the conventional AIAD/AIMD parameters, $\alpha_I = 1$, $a_D = -1$, and $\beta_D = 0.5$. We have also set n experimentally to 1.5; however, it can be adjusted differently in order to modify the transient behavior of the control. AIAMD tries to preserve AIAD's property of gentle variations in the transmission rate for random loss, enabling the desired smoothness for video streaming applications. At the same time, AIAMD reacts more aggressively in response to the reduction of network resources or the advent of new connections. Therefore, the protocol prevents multimedia applications from significant damage during congestion.

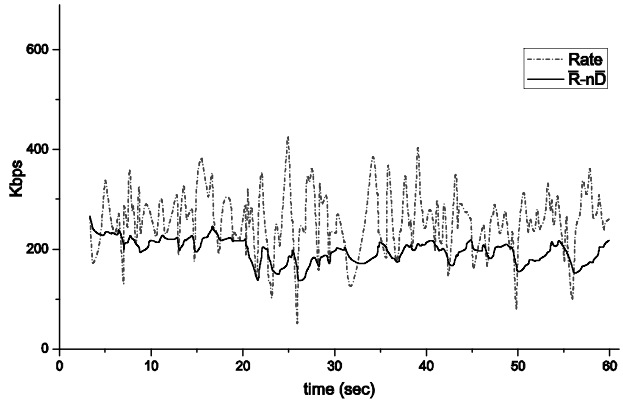


Figure 3. Receiving Rate and $\overline{\Re} - n\overline{D}$ Variation

In order to provide more insight to the behavior of AIAMD control, Fig. 3 shows the variation in the receiving rate and in the $\overline{\Re} - n\overline{D}$ bound. In this simulation, an AIAMD flow shares an 1 Mbps bottleneck link with one TCP flow. The AIAMD flow experiences infrequent congestion events where the measured rate falls below $\overline{\Re} - n\overline{D}$. At the same time, simulated random loss with 3% packet error rate enforces several loss events which are accurately interpreted as error-induced, since the measured rate exceeds the current value of $\overline{\Re} - n\overline{D}$.

3.3 Delivery of Video Layers

DVRC is optimized for layered video streaming. The server encodes raw video into k cumulative layers using a layered coder: layer 1 is the base layer and layer k is the least important enhancement layer. The layer rates are given by l_i, i = 1, 2, …, k. Let c_j denote the cumulative layer rate up to layer j, i.e. $c_j = \sum_{i=1}^{j} l_i$, j = 1, 2, …, k, while $r_k = (c_1, c_2, ..., c_k)$ denotes the rate vector of the cumulative layer rates. With the cumulative subscription policy, this discrete set offers all possible video rates that a client could receive, and the maximum number of layers that can be delivered to a receiver with an expected bandwidth B is $K = \max\{j : c_j \le B, c_j \in r_k\}$. DVRC can effectively interact with static layer-rate allocation schemes, as well as with dynamic source allocation, i.e. dynamically allocating the layer rates.

DVRC enables a closed-loop control between server and client, where the receiver detects the state of congestion, determines the proper transmission rate and eventually opts for the optimal number of layers

that should be delivered according to this rate. More precisely, the protocol performs the following control loop in order to actuate the video stream adaptation:

1) the receiver acquires the rate vector r_k [2] and monitors the progress of reception statistics $\overline{\Re}$ and $\overline{D}$
2) it determines the subsequent transmission rate R, based on equation (11)
3) it calculates K using $K = \max\{j : c_j \leq R(t), c_j \in r_k\}$
4) it generates a control packet attaching K to the *video layer* field
5) upon receiving the control packet, the sender joins or leaves layers until the subscription level is *K*

In order to explore the available bandwidth, transmission initiates with the lowest rate that can accommodate the base layer (i.e. $R(t_0) = c_1$). Although DVRC does not specify any particular coding algorithm in the application layer, the protocol can interact efficiently with coders of wide dynamic range and fine granularity in terms of rate control, such as *Fine Granularity Scalability* (FGS) [18].

In addition, exploiting the field *frame type* in the DVRC header, the protocol can optionally deliver layered video using a prioritized transmission. According to this scheme, the sender assigns different priorities to the layers according to their levels of importance, and during congestion the routers discard low priority packets first, securing the successful delivery of the most important video frames. In this paper, DVRC does not utilize any type of prioritization.

4. Evaluation Methodology

4.1 Experimental Settings

The evaluation plan was implemented on the *NS-2* network simulator [22]. In order to assess the performance of DVRC, we implemented an experimental MPEG-4 video streaming server which is capable of distributing layered video. The server transmits video data by dividing each frame into fixed size packets. The traffic generated closely matches the statistical characteristics of an original MPEG-4 video trace. MPEG-4 coding standard is based on I, P and B frames. The compression initiates by encoding a single I frame, followed by a group of P and B frames. P frames carry the signal difference between the previous frame and motion vectors, while B frames are interpolated; the encoding is based on

[2] It is not necessary for the receiver to periodically acquire the rate vector, when layer-rates have been allocated statically. In this case, the server is only required to advertise the rate vector to the client at the beginning of the connection.

the previous and the next frame. The model developed is based on *Transform Expand Sample* (TES). We used three separate *TES* models for modeling I, P and B frames, respectively. The resulting MPEG-4 stream is generated by interleaving data obtained by the three models. In the case of DVRC, the protocol header (Fig. 1) is encapsulated in each segment during the packetization process. Since the number of layers is quite limited in a practical layered coder, we maintain a base plus 4 enhancement layers (i.e. k = 5) that further refine video quality. The experimental layered adaptation scheme is illustrated in Fig. 4.

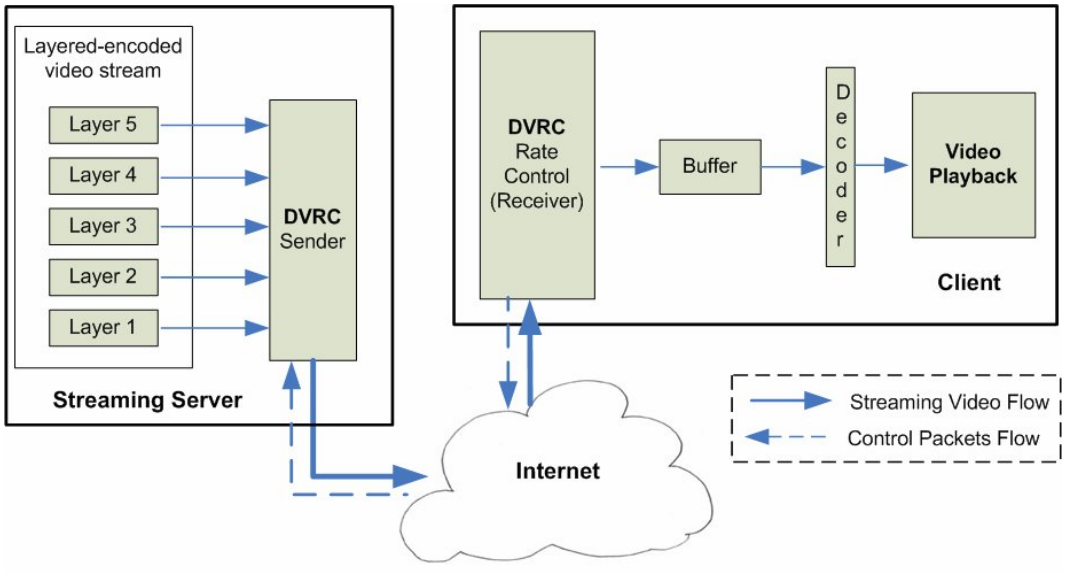


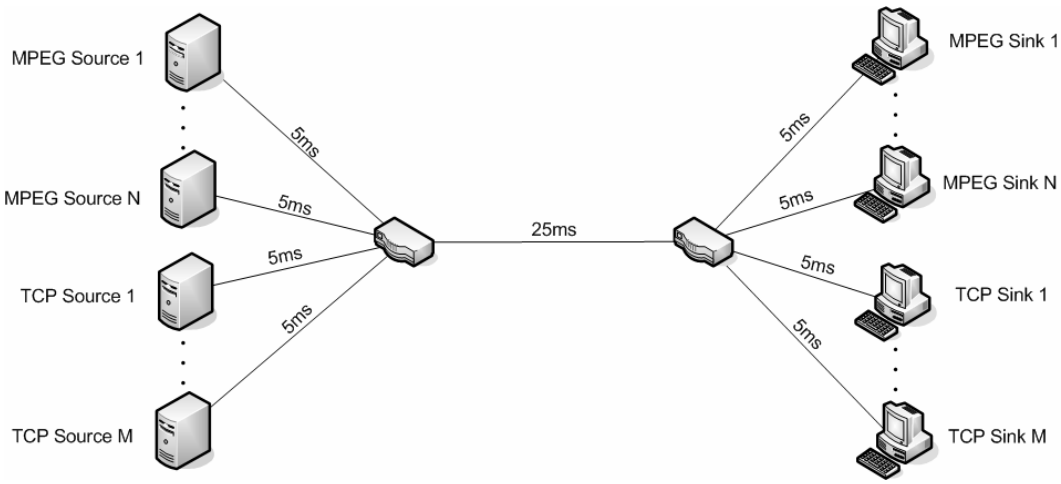
Figure 4. Experimental layered adaptation scheme under DVRC

We performed extensive evaluations of DVRC along with TFRC and the measurement-based TCP Westwood+ (TCPW+) [33]. TFRC, in particular, is designed for efficiency on media delivery over a wide range of network and session dynamics. We also included in our experiments the combined approach of RTCP over UDP. In this case, the receiver utilizes *RTCP Receiver Reports* in order to provide feedback to the sender at periodic intervals and whenever congestion is detected.

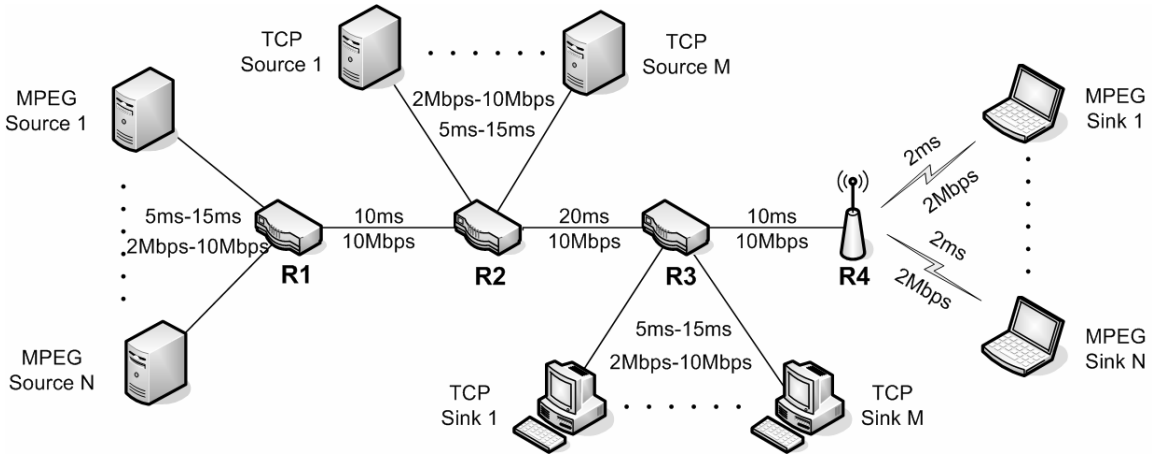
When the MPEG-4 server transmits over TFRC or TCPW+, the video is streamed at the optimal quality (i.e. adaptation is disabled). As a result, we evaluate the transmission rate control performed by each protocol at the transport layer. Since DVRC enables rate control based on layer distribution, the

streaming server performs layered adaptation based on DVRC's feedback loop. Likewise, RTCP/UDP is tested with the specific layered adaptation scheme.

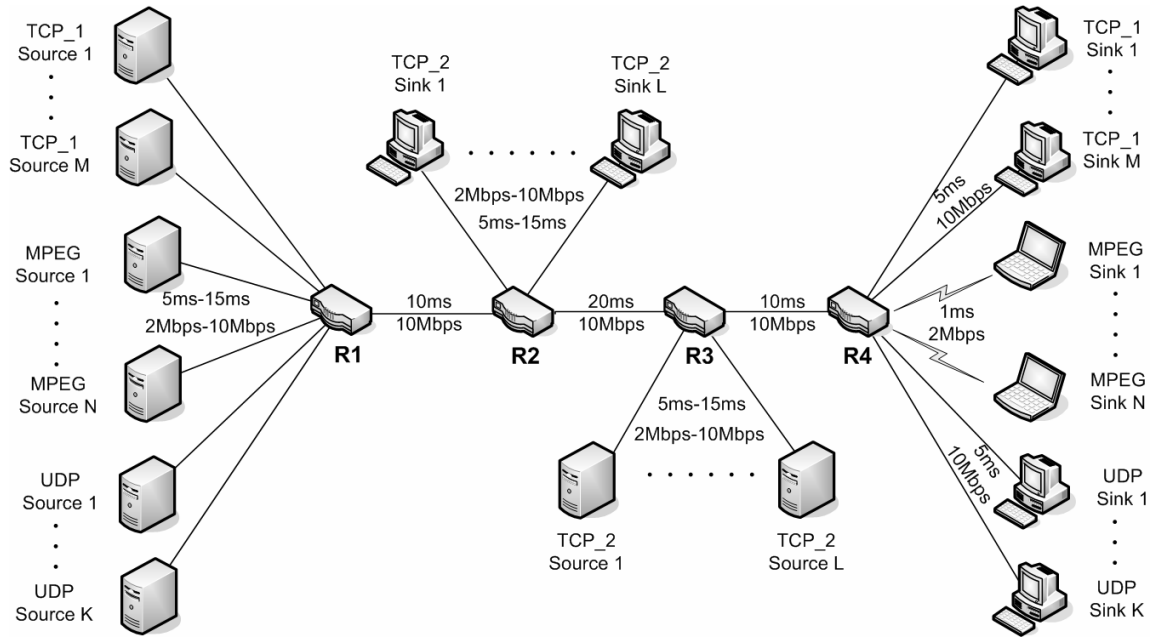
The experiments were conducted based on realistic scenarios which address the heterogeneity of the Internet. Initially, we enabled simulations on a single-bottleneck *dumbbell* topology with a round-trip link delay of 70 ms in order to evaluate single DVRC flow performance, as well as inter-protocol friendliness (Fig. 5a). The bottleneck link is shared by competing MPEG and TCP connections. The link capacities are configured depending on the experiment performed. Furthermore, we used a network topology (Fig. 5b) which includes multiple bottlenecks, cross traffic, wireless links and varying RTTs. The router *R1* is the bottleneck for MPEG traffic, while the router *R2* is another bottleneck for competing MPEG and cross traffic. Cross traffic includes a number of FTP connections over TCP Reno. The propagation delays of the access links from all the source nodes, as well as the links to the TCP sink nodes range from 5 ms to 15 ms, while the corresponding bandwidth capacities range from 2 Mbps to 10 Mbps. The capacity of all access links to the MPEG sink nodes is set to 2 Mbps. By randomizing RTTs, we avoid synchronization effects. Finally, we used the topology in Fig. 5c which employs the characteristics of the previous topology along with FTP/Reno flows both in the forward and backward direction. Random burst UDP traffic with the Pareto distribution has been also introduced to the forward direction.



(a) Dumbbell Topology



(b) Cross-Traffic Topology



(c) Reverse-Traffic Topology

Figure 5. Simulation Topologies

In cross- and reverse-traffic topologies, we used a link error model in the access links to the MPEG sink nodes. Error models were configured on both directions of the link traffic. [24] provides evidence for

correlated packet loss. In order to model temporally correlated loss observed in a fading wireless channel, we used the correlated *Bernoulli* model which characterizes the loss pattern as a Bernoulli distribution of loss rounds. Each round consists of a group of consecutive packets, the length of which is approximated by a geometric distribution. The first packet in the round is lost with probability p. Every other packet is lost with probability p, if the previous packet has not been lost; otherwise, the loss probability is q. In our simulations we adjusted $p = 0.01$ and $q = 0.15$.

In all topologies, the routers are drop-tail with buffer size adjusted in accordance with the *bandwidth-delay* product. We set the packet size to 1000 bytes for all system flows (with the exception of UDP flows in the reverse-traffic topology which have 250 bytes packet size) and the maximum congestion window to 64 KB for all TCP connections. Each simulation lasts for 60 sec, and diverse randomization seeds were used in order to reduce simulation dynamics. All the results are collected after 2 sec in order to avoid the skew introduced by the startup effect.

4.2 Measuring Performance

We hereby refer to the performance metrics supported by our simulation model. Since both topologies include competing MPEG and FTP connections, our performance metrics are applied separately to the MPEG and FTP traffic. Goodput is used to measure the overall system efficiency in bandwidth utilization and is defined as:

$$\text{Goodput} = \frac{\text{Original Data}}{\text{Connection Time}}$$

where *Original Data* is the number of bytes delivered to the high-level protocol at the receiver (i.e. excluding retransmitted packets and overhead) and *Connection Time* is the amount of time required for data delivery. Following the metric in [35], we use *Coefficient of Variation* (CoV) in order to gauge the throughput smoothness experienced by flow i:

$$\text{CoV}_i = \frac{\sqrt{E_t\{\text{throughput}_i^{\,2}(t)\} - E_t\{\text{throughput}_i(t)\}^2}}{E_t\{\text{throughput}_i(t)\}}$$

where $E_t\{\}$ denotes the computation of the mean along time. For the i^{th} flow, $throughput_i(t)$ is sampled at a time scale of a few RTTs throughout the entire connection. In our simulations, the sampling period is set to 150 ms. The frequency of throughput samples is dependant on RTT, as well as the type of network

traffic. Generally, a time period of 150 ms composes a plausible candidate for a minimum interval over which throughput variations would begin to be noticeable to multimedia users. Lower sampling periods can be considered for network paths with very low RTT, while higher intervals may conceal an amount of throughput variation. For a system with multiple flows, the system CoV is the average of CoVs of all flows.

Long-term fairness is measured by the *Fairness Index*, derived from the formula given in [6], and defined as:

$$\text{Fairness Index} = \frac{(\sum_{i=1}^{n} \text{Throughput}_i)^2}{n\sum_{i=1}^{n} \text{Throughput}_i^2}$$

where $Throughput_i$ is the throughput of the i^{th} flow and n is the total number of flows. As a supplementary fairness metric, we use *Worst-Case Fairness*:

$$\text{Worst Case Fairness} = \frac{\min_{1 \le i \le n} (\text{throughput}_i)}{\max_{1 \le i \le n} (\text{throughput}_i)}$$

in order to conduct a *worst-case* analysis and provide a tight bound on fairness. Similarly to *Fairness Index*, the range of *Worst-Case Fairness* is in [0, 1] with 1 representing the absolute fairness. Inter-protocol friendliness measurements were conducted based on *Normalized Throughput*, which is the ratio of the average throughput received by each flow over the bandwidth fair-share on each case.

The task of specifying the effects of network QoS parameters on video quality is challenging. Transmission rate fluctuations, increased delays, jitter and packet loss commonly deteriorate the perceived quality or fidelity of the received video content. We note that these network QoS parameters do not affect quality in an independent manner; they rather act in combination or cumulatively, and ultimately, only this joint effect is detected by the end-user. Delay jitter composes a critical factor in the performance of video delivery. According to [31] delay jitter is the value of packet spacing at the receiver compared with packet spacing at the sender for a pair of packets. Therefore, the value of jitter reflects packet-by-packet delay. Let S_i and R_i denote the sending and receiving time for packet i respectively; then for two packets i and j, packet jitter can be expressed as:

$$J(i, j) = (R_j - R_i) - (S_j - S_i) = (R_j - S_j) - (R_i - S_i) \quad (12)$$

According to equation (12), there is no delay jitter between packets *i* and *j*, if the value of J(i, j) is equal to zero.

Based on [23], we use a metric for the performance evaluation on video delivery, called *Video Delivery Index*, which captures the joint effect of jitter and packet loss on perceptual quality. The metric monitors packet inter-arrival times and distinguishes the packets that can be effectively used by the client application (i.e. without causing interruptions) from delayed packets according to a configurable packet inter-arrival threshold. The proportion of the number of delayed packets is denoted as *Delayed Packets Rate*. Video Delivery Index is defined as the ratio of the number of *jitter_free* packets over the total number of packets sent by the application:

$$\text{Video Delivery Index} = \frac{\#\,\text{jitter\_free packets}}{\#\,\text{sent packets}} \leq 1$$

According to streaming video guidelines, playback quality is notably degraded when delay variation exceeds 75 ms. Buffering can eliminate the effects of delay variation by smoothing out jitter; however, additional delays are incurred to the video playback. Furthermore, buffering exhibits certain limitations, such as application delay tolerance and buffer memory constraints. Along these lines, we adjusted the packet inter-arrival threshold at 75 ms, which specifies the point where delay variation becomes perceptible and possibly disturbing. Since MPEG traffic is sensitive to packet drops, we additionally define *Packet Drop Rate*, as the ratio of the number of lost packets over the number of packets sent by the application. For a system with multiple flows, we present the average of the Video Performance Index of each MPEG flow.

5. Performance Evaluation

In the sequel, we demonstrate conclusive performance studies based on extensive simulation results. More precisely, we evaluate the efficiency of DVRC in terms of video delivery and fairness, and we further examine DVRC's friendliness with interfering traffic.

5.1 Single DVRC Flow

Initially, we evaluate DVRC's efficiency in terms of smooth video delivery. We simulated a single MPEG flow over DVRC competing with two FTP flows of TCP Reno. We enabled the simulations on the dumbbell topology, where all link capacities are set to 1 Mbps. Fig. 6 illustrates the variation in the

sending rates of the 3 flows throughout the entire experiment. The transmission rate is averaged over 150 ms intervals, each one including approximately 2 RTTs for the specific topology. In the case of DVRC, the fluctuations in the sending rate are of low magnitude in comparison with the highly variable rates of TCP congestion control. Apart from the oscillations, Fig. 6 depicts occasional abrupt rate reductions induced by the TCP sources in response to coarse timeouts, which inevitably cause interruptions on data delivery. Generally, TCP's strategy of rapid backward and graduated upward adjustments hurts the performance of TCP-based applications, and especially damages delay-sensitive traffic. On the contrary, DVRC's gentle rate adjustments result in a smoothed flow that optimizes video delivery and playback on the receiver. Only in a few situations, such as the occasional bursts of a corporate TCP flow, AIAMD invokes a multiplicative decrease in order to confine packet loss for the MPEG flow. Generally, DVRC remains relatively immunized from the disturbances caused by the interfering TCP connections.

In order to quantify the smoothness observed by the end-user, we measured CoV for the single DVRC connection, which is: $CoV_{DVRC} = 0.1854$. In addition, we carried out the same experiment simulating the MPEG transfer with TFRC. The corresponding CoV measurement is: $CoV_{TFRC} = 0.2229$. A lower CoV expresses a lower variation in throughput rates, and consequently higher smoothness; hence, DVRC appears to be smoother than TFRC. In the remainder of the section, we provide more extensive performance studies, where we compare DVRC with TRFC, TCP Westwood+, and RTCP/UDP in dynamic networks with high link-multiplexing.

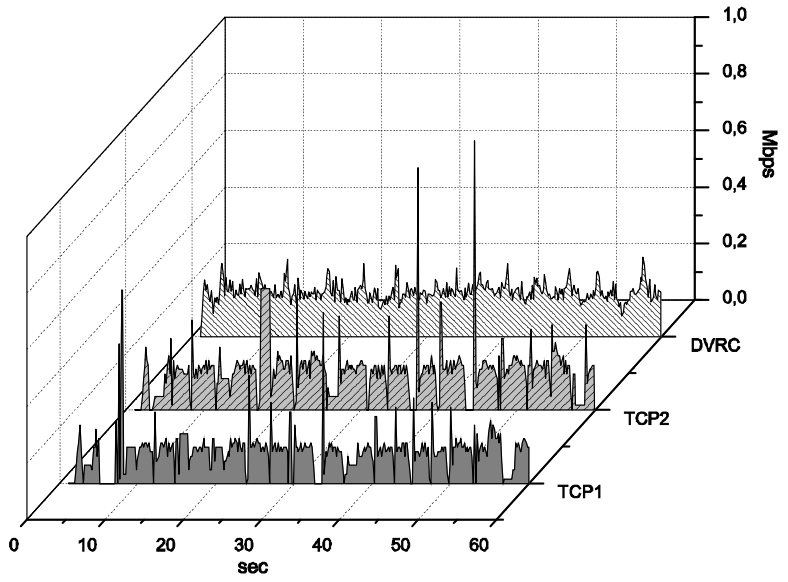


Figure 6. Sending Rates of Competing DVRC and TCP flows

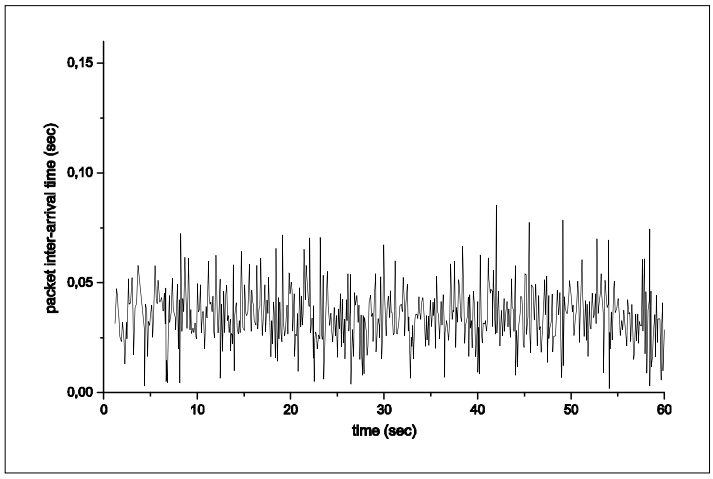


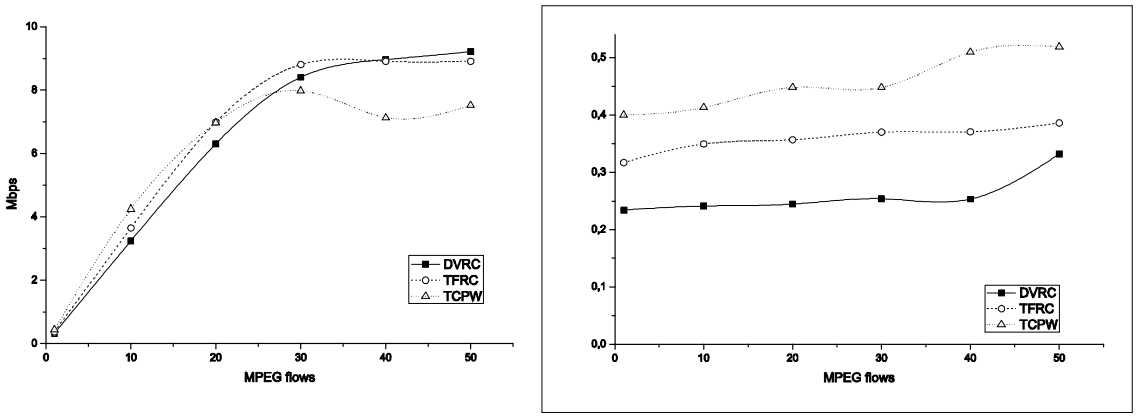
Figure 7. DRVC Delay Jitter

The performance on video delivery for the DVRC connection is depicted in Fig. 7. The DVRC flow shares the bottleneck link with two FTP connections, which are expected to cause notable disturbances on perceptual video quality. Nevertheless, delay jitter does not exceed the frustrating limit of 75 ms, while a high proportion of inter-arrival times are monitored with values below 50 ms, where jitter effects are not noticeable by the majority of multimedia users. Such a performance does not necessitate the use of deep

playback buffers in order to ameliorate the effect of jitter. Only in conditions of scarce bandwidth, a playback buffer may notably improve the delivered video quality and prevent potential interruptions on video playback.

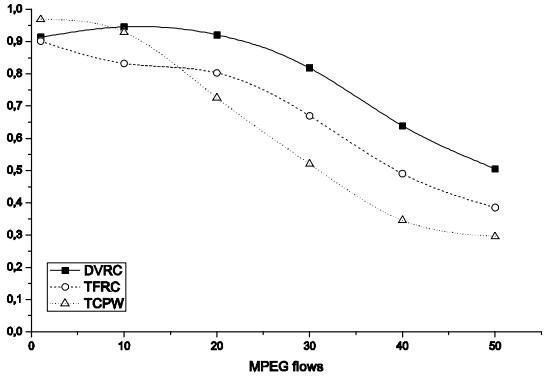
5.2 DVRC Performance with Multiple Flows

Departing from the performance study of a single DVRC flow, we carried out a series of experiments in order to assess the performance of our approach versus TCP-friendly traffic. We simulated a wide range of MPEG flows (1-50) of (i) DVRC, (ii) TFRC, and (iii) TCP Westwood+ competing with 10 FTP connections of TCP Reno, successively. The corresponding experiments were conducted on the cross-traffic topology. We measured *Goodput*, *Video Delivery Index*, *Fairness Index* and *Worst-Case Fairness,* and we additionally demonstrate statistics from delayed and lost packets, since both compose influencing factors for perceived video quality (Figs. 8-10). Furthermore, we present traces of the queue-length of router *R2* (Fig. 11) in the presence of 40 MPEG flows.

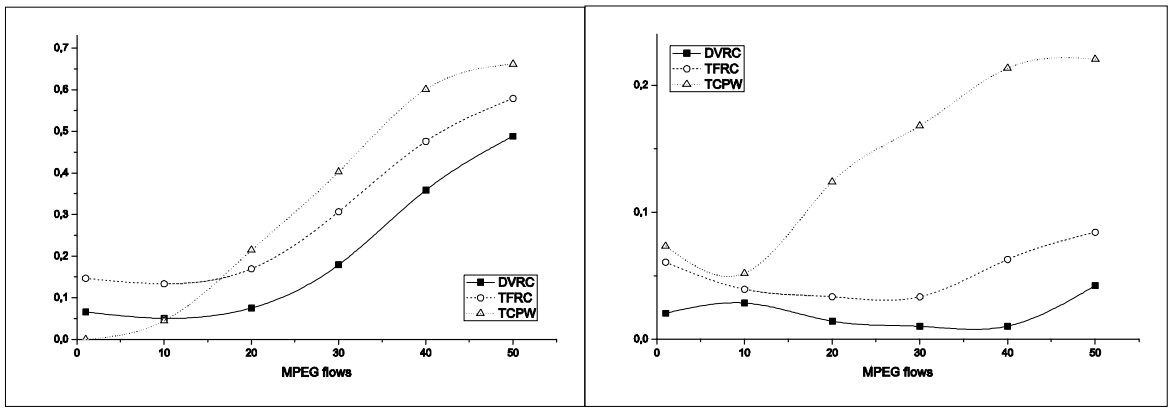


(a) Goodput of MPEG flows **(b)** CoV of MPEG flows

Figure 8. Protocol Performance

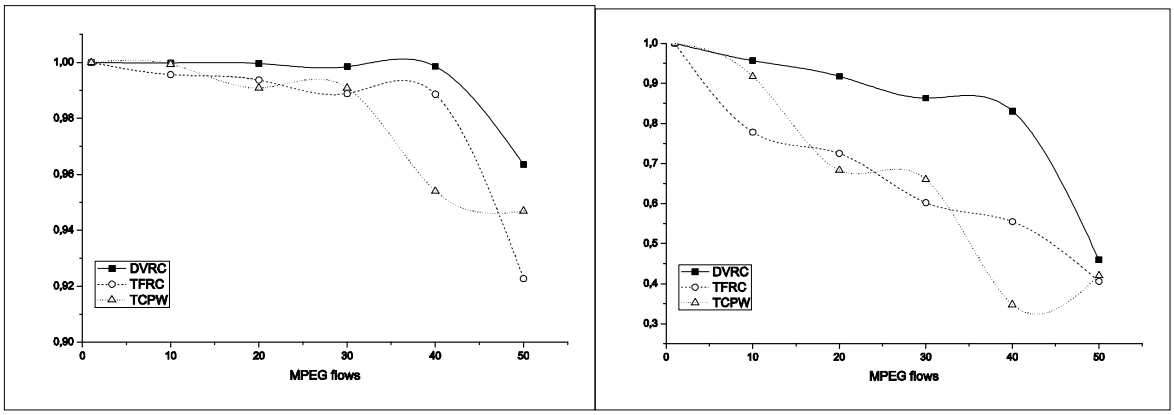


(a) Average Video Delivery Index



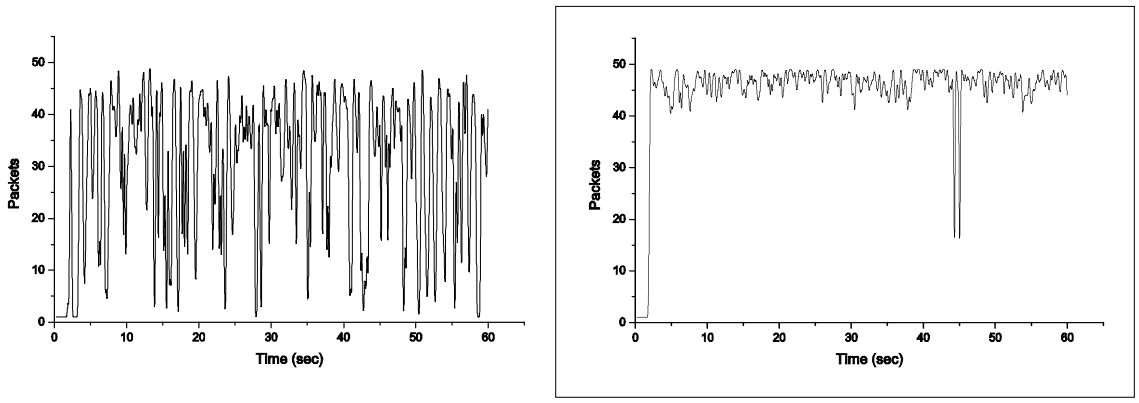
(b) Packet Drop Rate **(c)** Delayed Packets Rate

Figure 9. Performance on Video Delivery



(a) Fairness Index **(b)** Worst Case Fairness

Figure 10. Fairness



(a) DVRC **(b)** UDP

Figure 11. Queue Length of Router R2 (40 flows)

According to Fig. 8a, DVRC yields effective bandwidth utilization, especially for high link-multiplexing. The protocol exploits its rate control algorithm to react gently to the link errors across the last-hop wireless channel. Only congestion-induced loss enforces DVRC to significantly reduce its transmission rate in order to adapt to current network dynamics. Inline with DVRC, TFRC manages to utilize a high fraction of the available bandwidth. TCPW+ achieves remarkable goodput performance for relatively low link-multiplexing. However, at high contention (40-50 flows) the protocol does not achieve full utilization of the available bandwidth. Despite the improvements over the initial version of Westwood, TCP Westwood+'s algorithm occasionally underestimates the available resources, since the estimation filter is slow, needing time to converge to the available bandwidth. Apparently, TCPW+ would perform more efficiently in the presence of higher bandwidth, unfolding its potential.

Fig. 8b reveals that DVRC's transmission rates exhibit small fluctuations in average, achieving the desired smoothness required by most video streaming applications. The protocol adjusts the sending rate in response to the prevailing conditions, as well as congestion history, trying to preserve AIAD's property of graceful variation in the sending rate.

From the perspective of video delivery, DVRC exhibits remarkable efficiency, delivering smooth video which is slightly affected by contention (Fig. 9a). The protocol achieves the timely delivery of most packets, enabling stable playback quality without interruptions (Fig. 9c). In addition, the low packet drop rate (Fig. 9b) in conjunction with the relaxed packet loss requirements of streaming video justifies our choice not to integrate reliability to DVRC. AIAMD is therefore less susceptible to random packet loss and loss synchronization. Our simulations (including a more extensive set of experiments with a diverse

number of layers, not presented here) demonstrate that DVRC can achieve satisfactory performance with a small number of layers (4-5 layers).

According to Fig. 9c, TFRC delivers a smoothed flow and eventually confines the short-term oscillations in the sending rate, inline with DVRC. However, Fig. 9b illustrates relatively increased packet losses for TFRC, which deteriorate the perceptual video quality (Fig. 9a). In dynamic environments with wireless errors, TFRC occasionally fails to obtain accurate estimates of the loss event rate, invoking an inappropriate equation-based recovery, since TFRC's throughput model, as expressed in equation (1), is sensitive to the packet loss rate. TCPW+ exhibits increased packet loss (Fig. 9b), while a considerable proportion of the packets that are not dropped, reach the recipient later than required (Fig. 9c). The combined effect degrades the playback quality of the video stream (Fig. 9a). However, we note that TCPW+, as a modification of TCP Reno, is not optimized for multimedia applications, justifying its inferior performance on media delivery.

As shown in Fig. 10a, DVRC excels in bandwidths sharing, regardless of link-multiplexing. The AIMD-based responses during congestion enforce competing DVRC flows to converge to the fairness point. Similarly, TFRC and TCPW+ achieve relatively high levels of fairness, if assessed by the traditional *Fairness Index*. However, according to Fig. 10b the worst-case fairness for TFRC and TCPW+ fluctuates and gradually degrades, representing a notable difference between the minimum and maximum throughput rates achieved by both protocols. Therefore, in the case of TFRC and TCPW+, the system can be significantly unfair to a small fraction of flows. Generally, competing TFRC flows may have different throughput rates, since their observed loss event rates can differ. This observation is profound in the case of increased contention (Fig. 10b).

Fig. 11a illustrates that DVRC results in variable buffer conditions (due to the increased contention which inevitably leads to congestion), while UDP alone (i.e. without any supporting end-to-end protocol such as DVRC or RTCP) results in rapidly growing queues and eventually in buffer overflows (Fig. 11b). In the case of DVRC, the buffer is utilized efficiently and the oscillations are apparently of smaller magnitude than in conventional AIMD protocols. Furthermore, Fig. 11a reflects the overall behavior of DVRC: there are gentle reductions in the queue length (additive decrease), while DVRC occasionally enforces a persistent buffer draining phase by promptly responding to congestion (multiplicative decrease).

5.3 DVRC Performance with Heterogeneous Networks and Reverse Traffic

We enabled additional simulations on the reverse-traffic topology, which allowed us to draw further conclusions on DVRC efficiency at conditions of increased contention with various types of traffic, including TCP reverse flows and UDP burst traffic. We specifically simulated 1-50 MPEG flows competing with 15 TCP (5 in the forward and 10 in the reverse direction) and 10 UDP flows. The peak sending rates for UDP flows range from 64 to 256 Kbps, each one occupying up to 2.5% of the bottleneck bandwidth. The MPEG flows were simulated with DVRC, TFRC, TCPW+, as well as RTCP/UDP. Recall that DVRC and RTCP/UDP utilize layered adaptation, while TFRC and TCPW+ transfer MPEG streams at optimal quality.

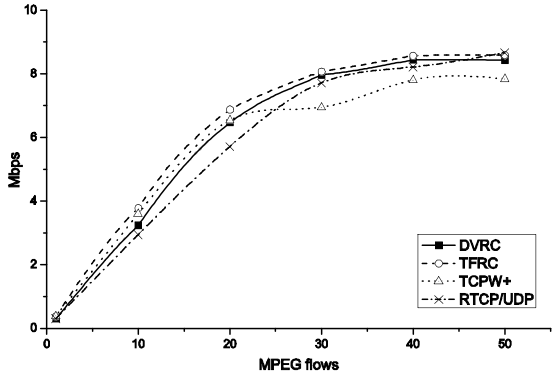
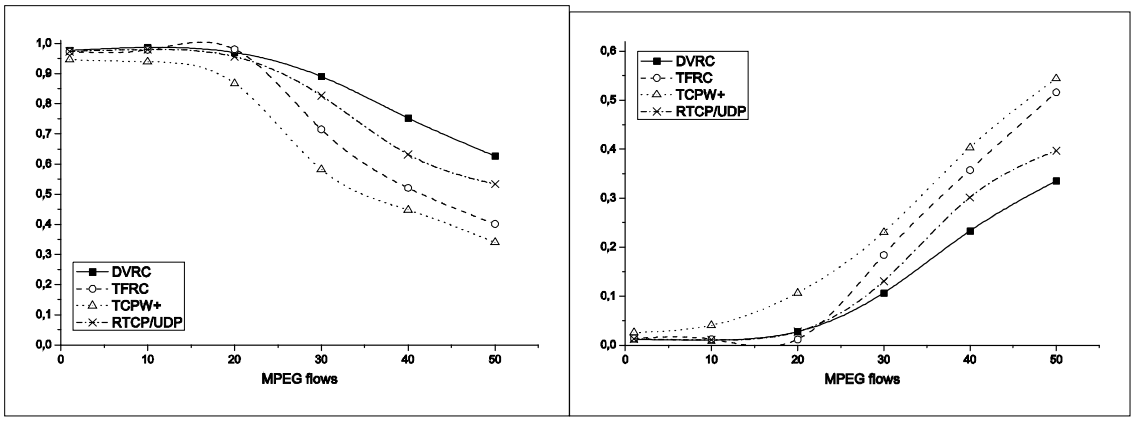
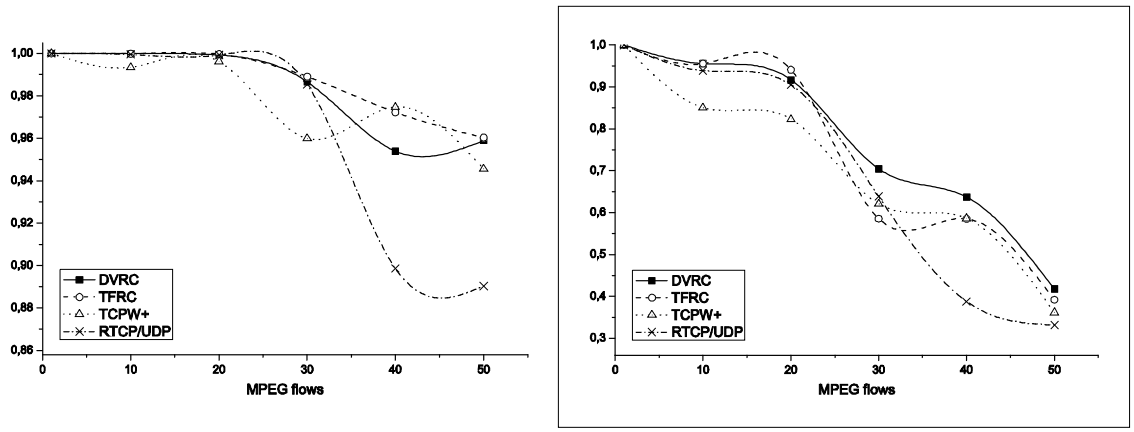


Figure 12. Goodput of MPEG flows



(a) Average Video Delivery Index **(b)** Packet Drop Rate

Figure 13. Performance on Video Delivery



(a) Fairness Index **(b)** Worst Case Fairness

Figure 14. Fairness

Fig. 12 shows that DVRC and TFRC achieve satisfactory bandwidth utilization, although the available resources are relatively limited due to the presence of interfering TCP and UDP flows. Furthermore, DVRC is not affected by reverse traffic, achieving high goodput rates even for high link-multiplexing. TCPW+ exhibits a perceptible deficiency for increased contention, where reverse traffic can slow down the rate of incoming acknowledgments (ACK) and subsequently the growth of the congestion window. An infrequent ACK rate may also result in timeout events for TCP protocols, since *Fast Retransmit* and *Fast Recovery* [32] are not triggered (i.e. the TCP sender does not receive 3 duplicate ACK) when the reverse path becomes congested.

According to Fig. 12, RTCP/UDP yields inferior performance in terms of goodput compared with DVRC and TFRC. Even if RTCP/UDP could compare more favorably with these protocols, its role is still limited: it simply provides information to the application allowing the source to reduce its transmission rate on the occurrence of congestion. It does not implement congestion control on top of UDP, as DVRC does. Hence, the task of rate adaptation is delegated to the application itself. The efficiency of RTCP/UDP is therefore highly dependant on the application-level rate control.

Fig. 13a illustrates the remarkable efficiency of DVRC from the perspective of video delivery. Inline with the corresponding results of previous section (Fig. 9), the protocol delivers video of acceptable quality regardless of link-multiplexing. DVRC's interaction with the layered scheme sustains relatively low packet loss, delivering a large amount of the video-data sent by the application without any retransmission effort (Fig. 13b). Apart from packet loss statistics, Video Delivery Index, as depicted in Fig. 13, reflects DVRC's smooth video delivery for a wide range of network dynamics. A combined

overview of Figs. 12, 13 reveals that competing flows in the forward and backward direction do not cause notable implications on DVRC efficiency and the perceptual video quality attained by the protocol. The rest of the protocols exhibit a noticeable degree of performance degradation (Fig. 13), which becomes more evident as contention increases. Apparently, packet loss composes a limiting factor both for TFRC and TCPW+. TFRC, in particular, yields inferior performance since it is not able to respond rapidly to the onset of congestion, decreasing its sending rate gently and allowing for a considerable amount of packet loss.

In terms of intra-protocol fairness, DVRC, TFRC and TCPW+ achieve a fair behavior among their flows (Fig. 14). Fig. 14b shows relatively small deviations between the minimum and maximum throughput rate of the DVRC flows. On the contrary, RTCP does not enable UDP to overcome its well-known fairness issues, as the unresponsive protocol does not employ any mechanism to converge to fairness. Furthermore, the presence of application-level rate control in collaboration with RTCP can hardly provide guarantees for fairness.

5.4 DVRC Friendliness

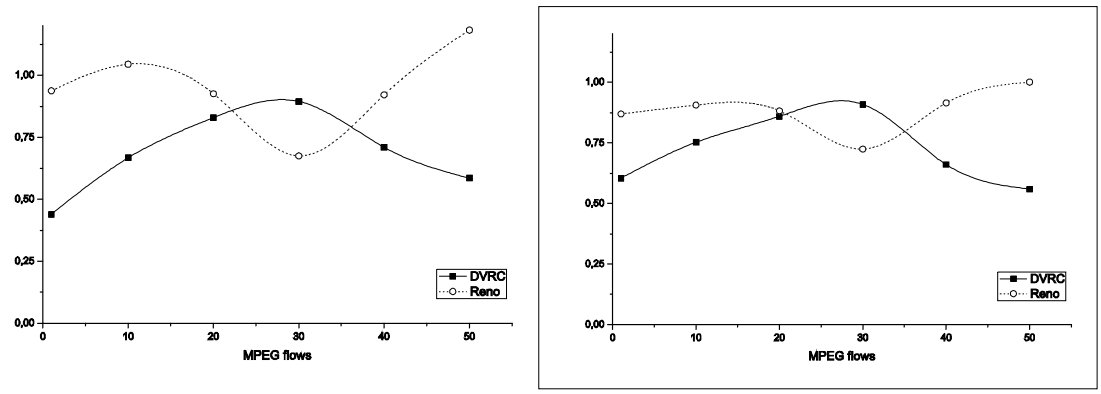
We conclude our performance studies investigating the interactions of a diverse range of MPEG flows (1-50) with interfering TCP traffic. We demonstrate results from DVRC, TFRC and TCP Westwood+ competing with FTP traffic over TCP Reno. We performed the simulations on the dumbbell topology configuring all link capacities at 10 Mbps. We demonstrate *Normalized Throughput* results from experiments with 20 and 40 FTP connections (Figs. 15-17).

Overcoming the *greedy* nature of UDP, DVRC monitors the prevailing network conditions and adjusts the video transmission rate maintaining friendliness with corporate flows. At periods of congestion, the integrated rate control algorithm behaves as AIMD (which is admittedly TCP-friendly), enforcing DVRC to relinquish a respectable amount of its resources. As depicted in Fig. 15, DVRC allows interfering TCP flows to obtain network resources close to the fair share of the link (i.e. *Normalized Throughput* of 1), which is critical in heterogeneous systems with multiple flows and different protocols. This observation is more profound in the situation of high link-multiplexing, where DVRC coexists fairly with TCP. In lightly loaded environments, the protocol allocates only the resources required to transmit all the video layers that can be accommodated to the receiver's bandwidth. In congested networks, DVRC simply relies on delivering the base layer, which corresponds to the minimum subscription level that service

quality is acceptable. DVRC performs its task without implications on background traffic, preventing the potential of a congestive collapse.

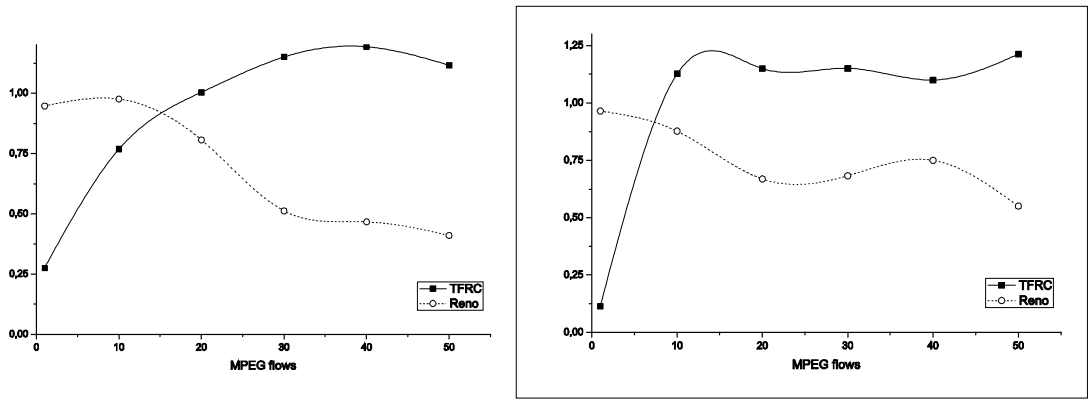
On the other hand, Figs. 16, 17 reveal a significant discrepancy between the throughput rates achieved by competing MPEG and TCP connections. Fig. 16 illustrates that TFRC flows behave more aggressively. TFRC sources achieve higher throughput rates than the link fair share with a consequent starvation of the TCP Reno sources. Competing TFRC and TCP flows may observe loss event rates that can be significantly different, especially if they have different sending rates. [28] studies analytically the long-term throughput imbalance between competing TFRC and TCP connections, reporting that the throughput difference can be further amplified, as long as coexisting TCP and TFRC flows experience different loss event rates. One fundamental factor for this throughput imbalance is the different RTO estimation schemes employed by TCP and TFRC. According to [28], competing TFRC and TCP connections may end up having different RTO values depending on network delays.

Similarly, TCPW+ utilizes increased network resources exceeding the link fair share. According to Fig. 17, TCPW+ connections behave more aggressively *stealing* bandwidth from the Reno flows, which are forced to invoke sharp backward window adjustments diminishing their throughput rates. Authors in [5] further explore the bandwidth estimation mechanism of TCPW+, as well as the protocol's behavior when TCPW+ coexists with TCP Reno.



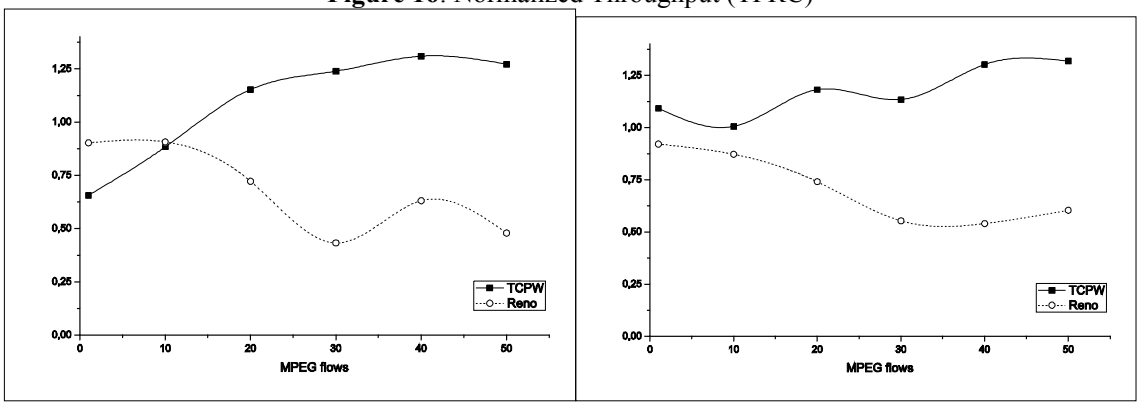
(a) 1-50 DVRC flows vs. 20 Reno flows **(b)** 1-50 DVRC flows vs. 40 Reno flows

Figure 15. Normalized Throughput (DVRC)



(a) 1-50 TFRC flows vs. 20 Reno flows **(b)** 1-50 TFRC flows vs. 40 Reno flows

Figure 16. Normalized Throughput (TFRC)



(a) 1-50 TCPW+ flows vs. 20 Reno flows **(b)** 1-50 TCPW+ flows vs. 40 Reno flows

Figure 17. Normalized Throughput (TCPW+)

6. Conclusions

We have presented a rate control scheme for the adaptive delivery of layered video streams over the Internet. DVRC enables a more sophisticated loss-recovery strategy adjusted dynamically to the particular characteristics of the underlying network. The integrated rate control algorithm (AIAMD) combines the most desirable features of AIAD and AIMD, a graceful variation in the transmission rate and sensitivity to the onset of sudden congestion. Through extensive simulations, we identified significant gains for DVRC in highly-multiplexed dynamic networks. The protocol is less susceptible to random packet loss, effectively adapts to the vagaries of the network and eventually delivers smooth video. The corresponding performance studies reveal that the proposed rate control scheme compares very favorably with

congestion control mechanisms that explicitly address delay-sensitive traffic, such as TFRC. DVRC has also demonstrated remarkable intra-protocol fairness, as well as friendliness with corporate TCP flows. DVRC is therefore a viable alternative to existing rate/congestion control schemes. In addition, DVRC can be integrated into existing protocols (e.g. DCCP), as a rate control option.

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