DP-FEC: Dynamic Probing FEC for High-Performance Real-Time Interactive Video Streaming

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Abstract: High-quality and high-performance real-time interactive video streaming requires both keeping the highest data transmission rate and minimizing data packet loss to achieve the best possible streaming quality. TCP-friendly rate control (TFRC) is the most widely recognized mechanism for achieving relatively smooth data transmission while competing fairly with TCP flows. However, because its data transmission rate depends largely on packet loss conditions, high-quality real-time streaming suffers from a significant degradation of streaming quality due to both a reduction in the data transmission rate and data packet losses. This paper proposes the dynamic probing forward error correction (DP-FEC) mechanism that is effective for high-quality real-time streaming to maximize the streaming quality in a situation in which competing TCP flows pose packet losses to the streaming flow. DP-FEC estimates the network condition by dynamically adjusting the degree of FEC redundancy while trying to recover lost data packets. It effectively utilizes network resources and adjusts the degree of FEC redundancy to improve the playback quality at the user side while minimizing the performance impact of competing TCP flows. We describe the DP-FEC algorithm and evaluate its effectiveness using an NS-2 simulator. The results show that by effectively utilizing network resources, DP-FEC enables to retain higher streaming quality while minimizing the adverse condition on TCP performance, thus achieving TCP friendliness.

Keywords: real-time streaming, forward error correction, dynamic FEC, TCP-friendly rate control

1. Introduction

The evolution of the Internet has seen a rapid growth in both users and applications. Owing to the dissemination of high-speed broadband networks, users are expecting to fulfill an application demand more comfortably, and applications promote users to increase their motivation. High-quality and high-performance real-time applications are the primary ones. Engineers are developing high-performance streaming applications [5],[23] that can be used easily with consumer electronic devices and ordinary PCs for the purposes of e-learning, international symposiums, and telemedicine [2]. It is easily expected that as the Internet evolves to higher-speed networks, these video applications will become increasingly popular.

Typically, real-time interactive video streaming requires the minimization of packet losses while maintaining the highest data transmission rate under acceptable transmission delays, to produce the best possible streaming quality. However, because high-quality and high-performance streaming flows consume a large amount of network bandwidth for data transmission, they may compete for bandwidth with other data flows (e.g., transmission control protocol (TCP) and/or user datagram protocol (UDP) flows) in a shared best-effort network like the Internet, and cause packet losses that can severely impact playback quality. The quality degradation caused by packet loss is a critical problem, especially for mission-critical applications such as telemedicine. The type of packet loss caused by congested routers, which is characterized by near-random and unpredictable behavior, is inevitable even with the future Internet, and therefore presents a formidable challenge to high-performance streaming applications seeking to optimize streaming quality.

TCP-friendly rate control (TFRC) has become the de facto standard to control network congestion [10],[12]. In accordance with the definition of TCP friendliness, TFRC tries to maintain fairness with competing TCP flows in the same network condition while providing a promising mechanism for a smooth data transmission rate [17],[27]. The throughput of non-TCP flows is strongly affected by the average throughput of a conformant TCP connection under the same conditions. Since this constraint can force a high-performance streaming flow to reduce the data transmission rate at the expense of video quality, TFRC is not suitable for the flow to maximize the streaming quality. In addition, when the feedback delay (i.e., the round trip time) is large, a timely and correct estimation of network conditions becomes an impossible task. As a result, unacceptable conditions for video streaming (i.e., packet loss or data rate oscillations) continuously occur [24] and they can have a negative impact on the communication quality of other competing flows (e.g., throughput). In this context, existing approaches do not properly control congestion to fulfill the demands of end-users wanting the highest streaming quality.

To address the problem that TFRC fails to maintain higher streaming quality, we propose the dynamic probing forward er-
ror correction (DP-FEC) mechanism that provides a promising method for achieving higher streaming quality while seeking the behaviors of competing TCP flows. DP-FEC estimates the network conditions by dynamically adjusting the degree of FEC redundancy while trying to recover lost data packets. By successively observing variation in the intervals between packet loss events, DP-FEC effectively utilizes network resources. It optimally adjusts the degree of FEC redundancy to recover as many lost data packets as possible while minimizing the performance impact of competing TCP flows. We evaluated the DP-FEC algorithm using an NS-2 simulator, and showed that it allows streaming flows to retain higher streaming quality and minimizes the impact of FEC on TCP performance to achieve TCP friendliness.

The remainder of this paper is organized as follows: Section 2 examines the requirements for high-performance real-time video streaming. Section 3 gives an overview of the DP-FEC algorithm. Section 4 evaluates the effectiveness of DP-FEC using an NS-2 simulation. Section 5 describes related work, and in Section 6, we present our conclusions.

2. High-Quality Real-Time Video Streaming

2.1 Background and Problem Statements

Most of the proposed congestion control mechanisms for real-time video streaming applications try to achieve TCP-friendliness according to its definition [1], [10], [11], [17], [27]; TFRC thus enables a streaming flow to maintain a desired smoothness of data transmission rate and fairness with coexisting TCP flows. However, because high-performance video streaming requires a large amount of bandwidth that cannot be utilized by TCP, throughput regulation based on TCP-friendliness reduces the data transmission rate at the expense of a reduction in video quality (i.e., the transmission rate is adjusted by changing the parameters for encoding the video). Video quality is increasingly reduced as the round trip time (RTT) increases, because TCP is inefficient under network conditions in which the RTT is large. In addition, because TFRC examines packet loss conditions as an indicator of congestion, it often delays the action for rate control and causes the degradation of playback quality. Estimated RTT variations are also utilized as a congestion indicator to avoid packet losses caused by traffic congestion. Yet, delay-based congestion control mechanisms [13], [14] (e.g., TCP Vegas) react to RTT variations incorrectly and cause a severe quality degradation, especially in high bandwidth paths [15], [16].

To clarify the specific problem of competition between a high-performance real-time streaming flow and TCP flows, we used an NS-2 simulator to simulate a single bottleneck with a 100 Mbps link capacity. The round-trip propagation delay varied from 1 to 200 ms. DropTail queue is used and the queue size is set to max\{100, Bandwidth Delay Product (BDP)\}. To create network congestion, we generate short-lived TCP flows using a Poisson process with an average rate of 30 flows per second, and the size of a TCP flow follows Pareto distribution with an average of 180 packets and a shape parameter of 1.5. As a high-performance real-time streaming, we used 1) a TFRC flow, 2) a TCP-Vegas [13] flow, 3) a High-speed TCP (HSTCP) [19] flow, 4) a Scalable TCP [20] flow, or 5) a normal UDP flow. We assume that the maximum data rate of a streaming flow is 30 Mbps.

Figure 1 (a) shows the average throughput of a single streaming which adopts TFRC/TCP-Vegas/HSTCP/Scalable-TCP. The maximum average throughput of a TCP-Vegas flow under the RTTs becomes about 25 Mbps. The maximum average throughputs of both a HSTCP and Scalable TCP flow become around 30 Mbps. Because of losses caused by short-lived TCP flows, each of these TCP flows causes the regulations and fluctuations of the data transmission rate. Since these behaviors lead to a severe degradation of video quality, these TCP protocols are not suitable for a high-performance real-time streaming application. A TFRC flow under the RTTs achieves the average throughputs between 5 and 10 Mbps. Despite reducing the transmission rate, as shown in Fig. 1 (b), the average data loss rates of a TFRC flow under the RTTs becomes more than 0.1%, which is not much different from that of normal UDP flow. In addition, a normal UDP flow does not have a negative impact on the time necessary for TCP flows to complete data transfer. In this context, a TFRC flow with relatively high available bandwidth tends to suffer from a degradation of streaming quality caused by packet losses that TCP flows induce.

2.2 Requirement

As described and shown in Section 2.1, high-performance real-time streaming using TFRC suffers from an unnecessary reduction in the data transmission rate (i.e., video quality) due to improper congestion control. Instead of using TFRC, an alternative protocol is indispensable for high-performance real-time stream-
ing to preserve the highest data transmission rate and avoid a
degradation of playback quality caused by packet losses in con-
gested networks, especially when the physical network band-
width is much larger than the total consumption bandwidth of
high-performance streaming flows\textsuperscript{2}. From this point of view, it
is important to protect playback quality from packet loss while
executing effective congestion control to achieve TCP friendli-
ness.

An application level forward error correction (AL-FEC) ap-
proach prevents retransmission delays by preventively adding re-
dundancy packets to the streaming data flow. Thanks to this re-
dundancy, a certain number of missing data packets can be re-
covered. Even if a certain degree of FEC redundancy is applied
to a high-performance streaming flow with TFRC, the flow can-
not improve the video quality in congested networks as described
before. However, ascertaining and controlling optimal FEC re-
dundancy at the highest data transmission rate is a real challenge,
because 1) it is difficult for a sender to determine the packet loss
pattern at each moment (as there is a feedback delay) and to pre-
dict the future packet loss pattern, and 2) increasing FEC re-
dundancy may disturb both streaming and competing flows such as
TCP when the conditions of competing flows are sensitively os-
cillated in the network. The following factors are important when
enhancing FEC redundancy.

- Because failure to appropriately adjust FEC redundancy
  (such that FEC cannot recover lost data packets) will lead
to non-recoverable data loss, FEC redundancy must be ad-
justed to optimize the recovery of lost data packets.
- Controlling FEC redundancy is not inconsequential to the
  behavior of other flows. It is important, therefore, to esti-
mate the network conditions and the impact of FEC on the
communication quality of other flows while controlling FEC
redundancy.

3. Design of Dynamic Probing FEC Mechanism

In this section, we propose the dynamic probing forward er-
ror correction (DP-FEC) mechanism to satisfy the requirements
aforementioned. According to the network condition, the mecha-
nism changes “FEC window size” (packets) to make a packet
loss tolerance, while considering the impact of FEC on the per-
formances of competing TCP flows.

3.1 Overview

Figure 2 illustrates the DP-FEC overview. DP-FEC operates
mainly at the sender. A sender transmits data and FEC repair
packets over a real-time transport protocol (RTP)\textsuperscript{3} carried on
top of the Internet Protocol (IP) and UDP. The DP-FEC collects
the feedback information transmitted over the real-time trans-
port control protocol (RTCP) delivered by a receiver, then adjusts
the degree of FEC redundancy using feedback information about
packet losses. The data transmission rate, which depends on the
video format preliminarily assigned, is maintained during trans-
mission.

DP-FEC leverages an application level forward error correc-
tion (AL-FEC). In AL-FEC, $n-k$ repair packets are added to a block of $k$
data packets. We consider maximum distance separa-
able (MDS) codes such as Reed-Solomon codes\textsuperscript{4} which can re-
cover all of the missing packets from any set of exactly $k$ packets.
Here, we define the number of repair packets as the “FEC window
diameter,” indicated by “$n-k$”. If the code parameters, the FEC win-
dow size and $n$, are appropriately set in the event of packet losses,
a receiver may recover all of the missing data packets within a
block. DP-FEC changes the FEC window size in a fixed FEC
encoding block length $n$ during transmission. Furthermore, all of
the generated data packets are stored once in the FEC encoding
block buffer. Depending on the FEC window size, the number of
data packets stored in the FEC encoding block buffer will vary.

DP-FEC is designed as an end-to-end model in consideration of
the following criteria:

- (1) Utilizing network resources effectively, the FEC window
  size during data transmission is increased to make a packet
  loss tolerance as high as possible. At the same time, network
  estimation for congestion control can be conducted.
- (2) In accordance with the network estimation, DP-FEC adjusts
  the FEC window size to minimize the impact of FEC on the
  performances of competing TCP flows.

3.2 DP-FEC Algorithm

The DP-FEC algorithm consists of three components: 1) the
congestion estimation function, 2) repeating congestion estima-
tion, and 3) the FEC window size adjustment function. Next, we
describe the algorithm in detail and the parameters used.

3.2.1 Congestion Estimation Function

The decision to adjust the FEC window size can be summa-
rized as follows:

- If no congestion is sensed, the FEC window size is increased
  except when the FEC window size is already at its maxi-

- If congestion is sensed, the FEC window size is decreased
  except when the FEC window size is at its minimum.

The DP-FEC recognizes above a certain degree of FEC impact
on competing TCP performances as an indicator of network con-
gestion. This means that the DP-FEC does not just use packet
loss as a congestion indicator, but allows a certain level of packet
loss during transmission. Here, we assume that a DP-FEC flow
competes with TCP flows in the same bottleneck link, and de-
fine FEC impact as a ratio of $FEC_{bw}$ to $TCP_{bw}$; $FEC_{bw}$ is the
consumption bandwidth of FEC repair packets that a DP-FEC
flow adds per unit time, and $TCP_{bw}$ is the available bandwidth
that existing TCP flows can utilize per unit time. Thus, in compe-
tition with $N$ TCP flows, the change in the degree of FEC impact
($\Delta FEC_{impact}$) caused by added $\Delta FEC_{bw}$ is given by:

$$\Delta FEC_{impact} = (\Delta FEC_{bw}/TCP_{bw})$$

$$= (\Delta FEC_{bw}/N)/(TCP_{bw}/N)$$

$$= (\Delta FEC_{bw}/N)/(TCP_{bw})$$ (1)

\textsuperscript{2} When the physical bandwidth is notably less than the total consumption
bandwidth of high-performance streaming flows, end-users should re-
duce the sending data rate at the cost of video quality. In this study, we
do not assume such a situation.
where $TCP_{bw}$ is the data bandwidth that one TCP flow consumes. According to $TCP_{bw}$ and $N$, the FEC impact on the TCP performances varies. $TCP_{bw}$ can be estimated as follows[10]:

$$TCP_{bw} = \frac{\text{PacketSize}}{\text{RTT} \cdot \frac{\text{RTT}}{2} + \text{RTO}} \left(3 \cdot \frac{\text{RTT}}{4} p(1 + 32p^2)\right)$$

where $\text{PacketSize}$ is the size of sending packet (bytes); $p$ is the loss event rate; RTT is the Round Trip Time (sec); RTO is the retransmission timeout value (sec). If we can know $N$, $\Delta FEC_{impact}$ can be estimated. However, $N$ changes from moment to moment, and it is naturally hard for an end system on an end-to-end model to precisely know $N$ at the right time.

DP-FEC approximates $\Delta FEC_{bw}/N$ by successively changing $FEC_{bw}$ and observing the loss interval ($LI$) defined as the interval of time between packet loss events. An occurrence of more than one packet loss within a specific time is recognized as a single packet loss event. In order to approximate $\Delta FEC_{bw}/N$, we now focus on the steady-state behavior of one TCP flow adopting the well-deployed AIMD algorithm[30] known as a mechanism for the remarkable stability of the Internet[11]. We assume that when a TCP flow observes more than one packet loss event during an RTT that results in halving the congestion window size, a DP-FEC flow also observes the same packet loss event. Since the TCP flow adopts the AIMD algorithm, $LI$ can be approximated as follows:

$$LI \approx \frac{(TCP_{bw}/N) \cdot \text{RTT}^2}{2 \cdot \text{MSS}}$$

where MSS is the TCP maximum segment size (bytes). When DP-FEC adds $\Delta FEC_{bw}$, $\Delta LI$ can be approximated using Eq. (3), as follows:

$$\Delta LI \approx \frac{(TCP_{bw}/N) \cdot \text{RTT}^2}{2 \cdot \text{MSS}} - \frac{[\text{TCP}_{bw} - \Delta FEC_{bw}/N] \cdot \text{RTT}^2}{2 \cdot \text{MSS}}$$

$$\Delta LI = \frac{\text{TCP}_{bw}/N \cdot \text{RTT}^2}{2 \cdot \text{MSS}} - \frac{\text{TCP}_{bw}/N \cdot \text{RTT}^2}{2 \cdot \text{MSS}}$$

Using Eqs. (1), (2) and (4), we can approximate $\Delta FEC_{impact}$ as follows:

$$\Delta FEC_{impact} = \frac{\Delta LI}{TCP_{bw} \cdot \frac{2 \cdot \text{MSS}}{\text{RTT}^2}}$$

DP-FEC estimates $\Delta FEC_{impact}$ using an observed $\Delta LI$. An observation of a larger value of $\Delta LI$ implies that the increasing FEC window size has a more negative impact on TCP performances by reducing $TCP_{bw}$. DP-FEC recognizes network congestion when an estimated $\Delta FEC_{impact}$ exceeds Threshold FEC Impact. The value of Threshold FEC Impact in DP-FEC is set to 0.1. Since an observed $\Delta LI$ is influenced by packet loss patterns that vary according to ever-changing network conditions (e.g., the number of competing TCP flows and the available bandwidth), precisely estimating $\Delta FEC_{impact}$ is naturally difficult. Therefore, by successively observing $\Delta LI$ while changing the FEC window size, DP-FEC evaluates whether $\Delta FEC_{impact}$ exceeds Threshold FEC Impact or not. Although this strategy leads to low responsiveness to a change in network conditions, DP-FEC can effectively utilize network resources by adding FEC repair packets while considering the impact of FEC on TCP flows. The decision frequency is described in next Section 3.2.2.

As Eq. (5) shows, the estimation of $\Delta FEC_{impact}$ depends largely on RTT that competing TCP flows have, since TCP performance generally relies on RTT. If the RTT value is large, DP-FEC tends to conservatively behave and keep a small FEC window size even in slightly congested networks. This situation causes a number of non-recovered data packets without effectively utilizing network resources. To avoid such a behavior, DP-FEC estimates $\Delta FEC_{impact}$ assuming that all competing TCP flows have $RTT = 0.01$ (sec). Thus, in competition with TCP flows with $RTT > 0.01$, the DP-FEC tends to keep a larger FEC window size due to the underestimation of $\Delta FEC_{impact}$.

3.2.2 Repeating Congestion Estimation

A DP-FEC receiver sends feedback information at constant time interval (SYN time), which denotes whether packet loss happened or not during the SYN time. Multiple packet losses in the same SYN time are considered as a single loss event. As described before, DP-FEC assumes that 1) both a DP-FEC flow and TCP flow observe the same packet loss event during an RTT and 2) all competing TCP flows have $RTT = 0.01$ (sec). Thus, the value of SYN time in DP-FEC is set to 0.01 (sec). A sender receives feedback information at the SYN time interval, and calculates the sample loss interval ($SampleLI$), the minimum value of which is equivalent to 0.01 (sec). Using the Exponential Weighted Moving Average (EWMA), current $LI$ is calculated as follows: $LI = \alpha \cdot LI + (1 - \alpha) \cdot SampleLI$. DP-FEC uses the
smoothing factor $\alpha = 0.9$, and estimates $\Delta FEC_{\text{impact}}$ with the observed variation of $LI$ ($\Delta LI$) at the constant interval ($ResTime$). The $ResTime$ relates to the tradeoff between the aggressiveness (i.e., increasing rate of FEC window size) and estimation accuracy of $FEC_{\text{impact}}$ (i.e., TCP-friendliness). In an ever-changing network condition, the appropriate setting of $ResTime$ value is quite difficult. In this paper, we set $ResTime$ to 0.16 (sec) based on our comprehensive simulation results as of now.

Figure 3 shows the relation between the current $LI$ and maximum acceptable $\Delta LI$, where MSS=1,460 (bytes), RTT=0.01 (sec). If an observed $\Delta LI$ exceeds the maximum acceptable $\Delta LI$, a sender recognizes an occurrence of network congestion (i.e., $\Delta FEC_{\text{impact}} > 0.1$). Note that if a network condition drastically becomes worse (e.g., due to an occurrence of bursty traffic), $LI$ severely decreases. If this situation continues, DP-FEC needs to decrease the FEC window size in order to avoid congestion collapse. Therefore, DP-FEC also recognizes a network congestion when $LI$ becomes less than the minimum value ($MinLI$). Every time a sender receives feedback information, the current $LI$ is calculated and judged to be below $MinLI$. When DP-FEC uses an extremely-low value of $MinLI$, the attainable FEC window size becomes large even in highly congested networks due to the slow response to the network congestion. Such a situation causes a significant degradation of the performances of competing TCP flows. To avoid it, we set $MinLI$ to 0.035 (sec) based on our simulation experiments as of now.

3.2.3 FEC Window Size Adjustment Function

DP-FEC recognizes network congestion when estimated $\Delta FEC_{\text{impact}}$ exceeds threshold $FEC_{\text{impact}}$ or current $LI$ falls below $MinLI$. In DP-FEC, AIMD algorithm is used to adjust the FEC window size. FEC window size ($Fund(t)$) in packets is given by:

$$Fund(t) = \begin{cases} 
Fund(t - 1) + 1 & \text{per } ResTime \\
\beta Fund(t - 1) & \text{per congestion}
\end{cases}$$

\[(6)\]

During no congestion, $Fund$ is linearly increased. During congestion, DP-FEC immediately decreases $Fund$ using multiplicative factor $\beta$. DP-FEC with a low value of $\beta$ tends to fail to recover lost data packets in continuously congested networks, because the attainable FEC window size becomes small. Since 1) DP-FEC should try to recover data loss as much as possible, and 2) by estimating and controlling $\Delta FEC_{\text{impact}}$, DP-FEC can avoid a negative impact on TCP performances in network conditions where added FEC redundancy significantly disturbs TCP performances, DP-FEC sets $\beta$ to a relatively high value, 0.8 (i.e., more than 0.5).

4. Evaluation

4.1 Simulation Setup and Performance Metrics

Using an NS-2 simulator extended with DP-FEC module, we performed experiments using the dumbbell topology shown in Fig. 4. The bandwidth and propagation delay of the bottleneck link between $B1$ and $B2$ were set to 1 Gbps and 10 ms, respectively. A drop-tail queuing was used at the router, and the queue length size in packets was set to 128. Each sender and receiver were connected to the bottleneck link through the 1 Gbps access link with a propagation delay of 1 ms. The packet size was set to 1,500 bytes for all connections. Each simulation ran for about 120 seconds.

In all simulation experiments, a DP-FEC sender takes the same algorithm for sending data packets. It transmits smoothed data packets at a rate of 30 Mbps. As a practical matter, both the FEC encoding block length and the maximum FEC window size should be determined by the acceptable FEC encoding/decoding time for the streaming application. Here, we set the FEC encoding block length to 127 and the maximum FEC window size to 63, modeling a real environment.

As the DP-FEC performance metrics, we observed 1) the average residual data loss rate of DP-FEC flows (i.e., the rate of non-recovered data packet) and 2) the average throughput of competing TCP flows. The metric of the average residual data loss rate of DP-FEC was compared with those observed when we used a normal UDP flow at a rate of 30 Mbps or a TFRC flow [10] under the same network condition. As TCP friendliness index ($TFindex$), we use the result of the average throughput of TCP flows observed when they compete with TFRC flows, and define $TFindex$ as follows:

$$TFindex = \frac{TCP\text{ throughput}_{\text{Target}}}{TCP\text{ throughput}_{\text{TFRC}}}$$

\[(7)\]

where $TCP\text{ throughput}_{\text{Target}}$ is the result of the average throughput of TCP flows competing with normal UDP flows or DP-FEC flows; $TCP\text{ throughput}_{\text{TFRC}}$ is the result of the average throughput of TCP flows competing with TFRC flows under the same network condition.

We used two types of TCP flows using a NewReno, 1) Short-lived TCP flows and 2) Long-lived TCP flows. Short-lived TCP flows observed when they compete with TFRC flows, and define $TFindex$ as follows:
flows arrive at the bottleneck link, as a Poisson process with an average rate of \( r_{tcp} \) flows per second. The size of each TCP flow follows a Pareto distribution with an average of \( s_{tcp} \) packets and shape parameter 1.5. Here, we define the load of short-lived TCP flows as \( \rho_{tcp} = r_{tcp} \times s_{tcp} \). Long-lived TCP flows are persistent in the network.

4.2 Competition with Only Short-Lived TCP Flows

We evaluated the DP-FEC performance in competition with only short-lived TCP flows. The load of short-lived TCP flows (\( \rho_{tcp} \)) was set to 25%, 50% and 75% of the bottleneck link capacity. In addition, the number of UDP streaming flows (i.e., TFRC, normal UDP or DP-FEC flows) was changed from 1 to 10.

As shown in Fig. 5 (a), the average data loss rates of DP-FEC flows under 25% load of TCP flows are considerably improved by added FEC redundancy, compared to those of TFRC or normal UDP flows. In addition, Fig. 5 (b) shows that the TCP friendliness indexes of both normal UDP and DP-FEC flows are maintained at more than approximately 0.9, which means that DP-FEC flows add a well-coordinated degree of FEC redundancy by effectively utilizing network resources that competing TCP flows cannot consume. In Fig. 5 (c), we can see that the FEC window sizes are almost the same (between 17 and 20) regardless of the number of competing DP-FEC flows, because FEC impacts that each DP-FEC flow estimates during transmission are not largely different (i.e., there is little difference in the available bandwidth for TCP flows in each test). Because of an occurrence of packet losses caused by a behavior in slow start phase of generated short-lived TCP flow, the average FEC window size is suppressed to below 20. TFRC flows in each test also observe packet losses caused by a slow start phase of generated short-lived TCP flow, and reduce the data transmission rates to about 6 Mbps. As opposed to a TFRC flow, a DP-FEC flow suppresses the average data loss rate and preserves the highest data transmission rate while achieving high TCP friendliness index.

When \( \rho_{tcp} = 50\% \) of the bottleneck link capacity, the average data loss rates of TFRC and normal UDP flows becomes more than 1%, as shown in Fig. 6 (a). On the other hand, owing to added FEC redundancy, the average data loss rates of DP-FEC flows are suppressed to below 1%. In addition, Fig. 6 (b) shows that although the TCP friendliness indexes of both normal UDP and DP-FEC flows decrease with an increase in the number of streaming flows, DP-FEC flows in each test retains more than approximately 0.85% of the TCP friendliness index. In Fig. 6 (c), we can see that as the number of DP-FEC flows increases, the average FEC window size gradually decreases to about 9. This is because, under 50% load of TCP flows, DP-FEC flows tend to experience continuous packet loss and estimate higher FEC impact with an increase in the number of DP-FEC flows. The continuous packet loss leads to an increase in average data packet loss rate, as shown in Fig. 6 (a).

When \( \rho_{tcp} = 75\% \) of the bottleneck link capacity, the average loss rates of DP-FEC flows are slightly improved except when the number of DP-FEC flows is more than 8, as shown in Fig. 7 (a). In Fig. 7 (b), we can see that as the number of streaming flows increases, the TCP friendliness indexes of both flows of Normal UDP and DP-FEC severely decrease because of a reduction in the available bandwidth for heavy short-lived TCP traffic. Due to the continuous packet loss caused by heavy short-lived TCP traffic, the average FEC window sizes are suppressed to between 5 and 10, as shown in Fig. 7 (c). Since DP-FEC is designed to minimize FEC impact defined as a ratio of the consumption bandwidth of FEC repair packets to the available bandwidth for TCP flows, the TCP friendliness indexes of DP-FEC flows become approximately the same as those of normal UDP. Therefore, in a situation
in which a normal UDP flow severely decreases the TCP friendliness index, a DP-FEC cannot retain a higher TCP friendliness index.

4.3 Competition with Both Short-Lived TCP Flows and Long-Lived TCP Flows

We evaluated the DP-FEC performance in competition with both short-lived TCP flows and long-lived TCP flows. The load of short-lived TCP flows ($\rho_{tcp}$) was set to 25% of the bottleneck link capacity, and the number of long-lived TCP flows was changed from 1 to 101. The number of UDP streaming flows was set to 3.

Figure 8 (a) shows that whereas the average data loss rates of TFRC and normal UDP flows become more than 1% in competition with more than 20 of long-lived TCP flows, those of DP-FEC flows are suppressed to below 1% in competition with less than 60 of long-lived TCP flows. As the number of long-lived TCP flows increases, the average data loss rates of DP-FEC flows also increase and approach those of TFRC and normal UDP flows. This is because many long-lived TCP flows probe available bandwidth during the congestion avoidance phase and incur more continuous packet loss. Due to the continuous packet loss, DP-FEC flows recognize higher FEC impact, resulting in the reduction of average FEC window sizes, as shown in Fig. 8 (c). However, the aver-
Fig. 8 DP-FEC Performances and average FEC window size in competition with long-lived TCP flows, under 25% load of TCP flows.

age FEC window sizes are kept between 15 and 20 when the number of long-lived TCP flows is less than or equal to 60. This condition is almost the same as in the case of competition with only short-lived TCP flows causing 25% load, because a DP-FEC flow recognizes that there is a certain level of available bandwidth by observing the interval between packet loss events while adjusting the FEC window size (i.e., the available bandwidth for TCP flows is not largely different despite increasing in FEC window size). Figure 8(b) shows the TCP friendliness indexes of normal UDP flows and DP-FEC flows. The TCP friendliness index is distinguished by the type of TCP flow. Although DP-FEC flows retain the average FEC window size between 15 and 20 in competition with less than 60 of long-lived TCP flows, the TCP friendliness indexes for long-lived TCP flows always become more than 0.9. Whereas a TFRC flow reduces the data transmission rate to below about 6 Mbps under the influence of packet loss conditions, a DP-FEC flow utilizes network resources effectively by adding FEC repair packets which contribute to the recovery of data packet losses. On the other hand, the TCP friendliness indexes of both normal UDP and DP-FEC flows for short-lived TCP flows become more than 0.9 when the number of long-lived TCP flows is below 25. When the number of long-lived TCP flows becomes more than 25, the TCP friendliness indexes for short-lived TCP flows becomes between 0.7 and 0.9. This is because the total consumption bandwidth of long-lived TCP flows increases, which deteriorates the performances of short-lived TCP flows. However, as described in Section 4.2, the TCP friendliness indexes of DP-FEC flows for short-lived TCP flows become approximately the same as those of normal UDP, since DP-FEC flows try to suppress the impact of FEC on the TCP performances.

### 4.4 Effect of the Value of Threshold FEC Impact

The value of Threshold FEC Impact (the default value is 0.1) is considerably important for DP-FEC to recognize network congestion and adjust the FEC window size. Thus, we evaluated the performances of DP-FEC with Threshold FEC Impact of 0.3 under TCP load of 25%/50%/75%, and compared the results with those of DP-FEC with Threshold FEC Impact of the default value.

Figure 9(a) shows the average loss recovery rates of DP-FEC flows (the cases of Threshold FEC Impact of 0.1/0.3) under 25%/50%/75% load of TCP flows. The loss recovery rate is defined as the ratio of the number of the recovered data packets to that of the lost data packets in the network. We can see that under 25% load of TCP flows, DP-FEC flows with Threshold FEC Impact of 0.1 maintain the average loss recovery rates of more than 0.98. However, under both 50% and 75% load of TCP flows, the average loss recovery rates decrease because the attainable FEC window size of each DP-FEC flow tends to become small as described in Section 4.2. On the other hand, the average loss recovery rates of DP-FEC flows with Threshold FEC Impact of 0.3 almost become 100% except when the number of DP-FEC flows under 75% load of TCP flows is more than 6. This is because a DP-FEC flow with Threshold FEC Impact of 0.3 keeps larger FEC window size than that of a DP-FEC flow with Threshold FEC Impact of 0.1.

Figure 9(b) shows the corresponding TCP friendliness indexes of DP-FEC flows. When the number of competing DP-FEC flows is the same, the TCP friendliness indexes of DP-FEC flows with Threshold FEC Impact of 0.3 becomes lower than those of DP-FEC flows with Threshold FEC Impact of 0.1. Under 25% load of TCP flows, the difference of TCP friendliness indexes is not large regardless of the number of DP-FEC flows. However, as the number of DP-FEC flows increases under 50% or 75% load of TCP flows, the difference becomes large (i.e., the TCP friendliness indexes of DP-FEC flows with Threshold FEC Impact of
0.3 largely decrease). This is because, since DP-FEC flows with Threshold FEC Impact of 0.3 tends to keep a larger FEC window size, the total bandwidth of FEC repair packets increases considerably. Since the value of Threshold FEC Impact relates to the tradeoff between efficiency (i.e., loss recovery rate) and TCP friendliness, it is of greater importance for DP-FEC to appropriately set the value.

5. Related Work

Many congestion control mechanisms for real-time streaming have been proposed and analyzed. Rejaie et al. [1] developed a rate-based congestion control mechanism that employs an additive-increase, multiplicative-decrease (AIMD) algorithm to achieve TCP-like behavior. The datagram congestion control protocol (DCCP) [26] was proposed by Kohler et al. to provide TCP-friendly rate control [10], [12]. The mechanism function is designed to behave fairly with coexisting TCP flows in terms of the consumption bandwidth. Papadimitriou et al. have proposed a rate control mechanism from the perspective of inter-protocol fairness [11]. Using a packet loss as a congestion indicator, this mechanism decreases the data transmission rate by a previously determined degree. If no congestion is sensed, it periodically increases the data transmission rate in incremental steps. Feng et al. [27] defined the new TCP friendliness necessary to achieve a desired constant data rate while maintaining fairness with coexisting TCP flows. However, because these mechanisms utilize packet loss as a congestion indicator and severely reduce the data transmission rate in response, they are not suitable for high-quality real-time video streaming that needs to maintain a higher data rate and avoid a degradation of playback quality, as described in Section 2.

Various numerical studies, using actual experimental measurement or analytical models, have evaluated the effectiveness of FEC. Bolot et al. [21] measured audio packets transmitted via the Internet, and analyzed the packet loss patterns. The results suggested that FEC is effective for low bandwidth streaming, such as an audio application, even when the network conditions are unstable because of network congestion. However, Altman et al. [25] proposed a detailed queuing analysis based on a ballot theorem, and concluded that FEC caused only a slight improvement in the audio streaming quality under any amount of redundant FEC data. In Ref. [6], an analysis of FEC effectiveness in ATM networks indicated that performance gains quickly diminish when all traffic sources employ FEC and the number of sources increases. Shacham et al. [7] used both analytic and simulation models to evaluate a situation using redundant parity packets, residual packet-loss rates after FEC decoding were reduced by up to three orders of magnitude. However, Xunqi et al. [8] indicated that the simplified analysis in Ref. [7] overestimated the performance of FEC because it assumed that the packet loss process is independent. As an alternative approach, Xunqi et al. [8] proposed a recursive algorithm based on the algorithm proposed by Eidon et al. [9] to compute block error density, and analyzed FEC performance in various FEC parameters (i.e., the encoding length, code rate, and interleaving depth). They concluded that an interleaving method increases FEC effectiveness for real-time streaming over congested networks. Although these numerical results vary depending on the loss model that is used or the testbed network, they provide insights into FEC recovery capability that can be applied to audio and video streaming.

A number of mechanisms for adjusting FEC redundancy have been proposed. Bolot et al. [31] mentioned that the congestion losses were modeled with a two-state Gilbert model, and the proposed scheme determined the FEC redundancy together with a TCP-friendly rate control scheme. Wu et al. [22] described that bandwidth constraints derived by the TFRC equation were used to optimally adjust FEC redundancy and the data transmission rate to maximize the video and playback quality. Seferoglu et al. [17] proposed TFRC with FEC to deal with packet losses induced by competing TCP flows. The mechanism utilizes the correlation between packet losses and the estimated RTT fluctuation as a network indicator. Moreover, they provided a rate-distortion optimized way to decide the best allocation of the available TFRC rate between source and FEC packets, and performed a significantly extended performance evaluation [18]. However, such a delay-based control which susceptibility reacts to RTT variations leads to a severe quality degradation, especially in high bandwidth paths [15], [16]. In addition, because these proposed mechanisms depend largely on the TFRC throughput equation, they do not effectively utilize network resources, which results in a failure to maintain higher video quality. In contrast, the approach of our proposed DP-FEC does not rely on TFRC rate to avoid a degradation of video quality caused by the regulation of the sending rate of source packets. We previously proposed an adaptive rate control mechanism that dynamically adjusts both the FEC redundancy and the data transmission rate [29]. It is an effective
technique for a single stream. However, it does not cooperate with other competing flows. This type of self-organized method adversely increases traffic congestion and disturbs other communication qualities. However, our proposed DP-FEC attempts to utilize the network resource effectively to optimize the recovery of lost data while minimizing the impact of FEC on other competing flows.

6. Conclusion and Future Work

In this paper, we proposed the DP-FEC that effectively achieves streaming quality for a high-performance real-time streaming. This mechanism tolerates packet loss in the network through the adjustments of the FEC window size while examining network congestion. By successively observing variation in the intervals between packet loss events, DP-FEC effectively utilizes network resources and adjusts the degree of FEC redundancy while minimizing the impact of FEC on the performance of competing TCP flows. We verified the efficiency of DP-FEC using an NS-2 simulator, and recognized that DP-FEC retains high streaming quality.

In our future work, we will make more investigation to optimally set a parameter such as Threshold FEC impact according to network conditions. Then, we will evaluate the DP-FEC mechanism competing with more high-performance streaming and TCP flows in complex and higher bandwidth networks. This evaluation will improve the DP-FEC algorithm, and then accelerate deployment of high-quality streaming applications over future heterogeneous networks.

Reference


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