END-TO-END CONGESTION CONTROL OVER WIRELESS NETWORKS: ANALYSIS AND ENHANCEMENTS

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Abstract

The success of the Internet can be attributed to the large number of useful applications running on it. Basically, these applications can be classified into two categories – bulk data transfer and streaming transfer – based on the different requirements on the data transmission. As we know, bulk data transfer applications, such as file transfers and emails, usually require high reliability of data transmissions while delays or delay variations caused are less concerned. On the other hand, streaming applications, including real-time audio and video, are very sensitive to delays and delay variations, while certain rate of packet losses is acceptable.

In order to provide the end-to-end congestion control for these two different types of applications, two end-to-end congestion control algorithms, namely TCP and TFRC, have been carefully designed over last years. To date, both algorithms work very well in wired networks. However, in wireless networks where random loss is rampant due to environmental noise, both TCP and TFRC suffer significant performance degradation unnecessarily, because they often misinterpret the random loss as the indication of network congestion.

This thesis studies such wireless issue for both TCP and TFRC. The analysis of TCP focuses on a novel wireless TCP variant, TCP Veno. Specifically, we develop a throughput model for TCP Veno, to provide theoretical analysis for TCP Veno evolution over wireless networks. The accordingly derived equation, called the “Veno equation”, characterizes TCP Veno congestion control algorithm in both wired and wireless situations. Our extensive simulations and the Internet experiments demonstrate the accuracy of such throughput prediction. At the same time, we also enhance TCP Veno in different ways, after observing that Veno behaves conservatively in certain wireless environments.

Very importantly, the analysis of TCP Veno provides us the clues to further improve TFRC performance over wireless networks. Based on the Veno equation obtained, we
innovatively propose an enhancement of TFRC called TFRC Veno. TFRC Veno intelligently makes use of the Veno equation to calculate the sending rate of the streaming data, instead of the original Reno equation. Simulations and the Internet experiments have proved that, TFRC Veno has up to 300% throughput improvement over the original TFRC in wireless environments. Meanwhile, such improvement does not cause any bias to the existing TFRC or other TCP-friendliness connections. The salient feature is that TFRC Veno only requires a simple modification at the sender side of TFRC and there is no change at the receiver protocol stack or intervention at any of the intermediate network nodes. Therefore, we can say such enhancement can be easily deployed in the current Internet.

At last, with the help of Veno state differentiator, the thesis also explores how to enhance TFRC performance over wireless networks in another way.
Chapter 1
Introduction

1.1 Research Background

The Internet is a collection of interconnected networks that offers a best-effort data transmission service to users using the Internet Protocol (IP) [1]. The IP protocol provides the means to connect diversity of networks such as Ethernet, TokenRing, FDDI, SONET, HDLC and ATM. The advantages of IP connectionless design, flexibility and robustness have led a great success of the Internet over the past 20 years. However, the lack means to limit the packet sending rate in the IP protocol can also result in severe service degradation or “Internet meltdown”, when the users of the network collectively demand more resources of link bandwidth and buffer space than the network has to offer. This phenomenon was first observed during the early growth phase of the Internet of the mid 1980s [2], and is technically called “congestion collapse”. Congestion collapse motivated the use of end-to-end congestion control in the Internet to avoid such performance degradation [3].

The original end-to-end congestion control algorithm was provided by Van Jacobson [4] in 1988, and later turned into the dominating transport protocol in the Internet, TCP [5]. Based on the principle of additive-increase/multiplicative-decrease (AIMD), TCP probes for extra bandwidth by increasing its congestion window linearly with time; once it detects the network congestion signalled by packet loss, it reduces its window by a factor of two. Here congestion window is used in TCP to determine the number of
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packets that can be outstanding at any time. Under certain assumptions of synchronized feedback, the AIMD control scheme has been shown to converge to a stable and fair operating point [6]. Because TCP only involves end hosts (sender and receiver) in its congestion control mechanisms, it is also called end-to-end congestion control.

TCP and its variants have enjoyed much success in the Internet as they can support all bulk data transfer applications such as file transfers, web services, emails, and interactive terminals. However, TCP is not well-suited for several emerging applications called streaming applications, such as real-time audio and video. This is because the reliability and ordering semantics in TCP increases end-to-end delays and delay variations. Furthermore, many of these streaming applications do not react well to the large and abrupt reductions in transmission rate on packet losses, which is caused by the TCP-style AIMD algorithms [7][8]. As a result, another kind of end-to-end congestion control mechanisms, which targets the low end-to-end delay, small delay variation, and smooth sending rate, was proposed to support these streaming communications. TCP-Friendly Rate Control (TFRC) [9][10] is one of the most popular protocols. It is equipped with an equation that can model the throughput of TCP [11] in the same network situations. During transmission, TFRC adjusts its sending rate according to the calculated result from the equation, making its instant throughput similar to that of TCP under the same circumstances. Thus TFRC does not bias any other existing TCP-like traffics during transmission.

In 2000, S. Floyd proposed the principles of designing end-to-end congestion control [12], and pointed out three goals of end-to-end congestion control: 1) Preventing congestion collapse. This is the initial reason leading to end-to-end congestion control, and has been elaborated in the previous part of this section. 2) Fairness. Flows belonging to the same end-to-end congestion control should share the network sources (i.e., bandwidth or buffer space) fairly. Furthermore, new end-to-end congestion control algorithm should
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not bias existing TCP-like traffics. The latter is also called TCP-friendliness. 3) Optimizing performance regarding throughput, delay, and loss. That is to say, end-to-end congestion control should not “overact” in the prevention of congestion collapse, which results in poor utilization on available network resources.

1.2 Problem Statement and Motivation

End-to-end congestion controls for both bulk data transfer and streaming are originally designed for wired networks. However, wireless communication technology has been making significant progress in the recent past and will be playing a more and more important role in access networks, as evidenced by the widespread adoption of wireless local area network, wireless home networks, and cellular networks. Wireless networks have introduced new issues into end-to-end congestion controls.

1.2.1 Wireless Issues in End-to-end Congestion Controls for Bulk Data Transfer

It is known that wireless networks may suffer high packet loss rate. As described in Section 1.1, TCP always treats the occurrence of packet loss as a manifestation of network congestion. This assumption does not apply to networks with wireless channels, in which packet loss is often induced by noise, link error, or reasons other than network congestion. As in [13], we refer to such packet loss as random packet loss. Misinterpretation of random loss as an indication of network congestion causes TCP to back down the sending data rate unnecessarily, resulting in significant performance degradation [13][14][15].

Over the past decade, there have been many efforts to enhance TCP throughput by eliminating negative impacts from random losses. These enhancements can be divided into two categories: 1) Network-support enhancements, which need the support from the intermediate nodes in the network, such as routers, proxies, access points, or any other
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kinds of devices. Snoop TCP [16], I-TCP [17], TCP with ECN [18], TCP with EBSN [19], and TCP with ELN [14] belong to this category. 2) End-to-end enhancements, which modified TCP mechanism only in the end nodes (sender or receiver or both). Typical examples are TCP Veno [20], TCP Westwood [21], TCP Jersey [22], TCP Santa Cruz [23], WTCP [24], TCP with FEC/ARQ [25], TCP-AP [26], C³TCP [27], and TCP-HO [28]. As compared to network-support enhancements, end-to-end enhancements do not need to add any new functions to the intermediate nodes in the network. This will avoid deployability problem in the real networks since it is costly and inconvenient to change the existing devices in the Internet.

As one of the end-to-end enhancements for TCP, TCP Veno provides efficient improvement in throughput by simply modifying the sender-side protocol stack. Many experiments have shown its superior performance, either in simulations or real networks [20][29]. TCP Veno is also incorporated by Linux Kernel since version 2.6.18 [30]. However, there is no theoretical analysis for TCP Veno to give an in-depth view on its behavior over wired and wireless networks. Moreover, whether its performance can be further improved is also an interesting issue.

1.2.2 Wireless Issues in End-to-end Congestion Controls for Streaming

Similar to TCP, TFRC also suffers severe performance degradation from wireless links. This is because TFRC is equipped with the equation modeling TCP throughput, and thus inherits the similar behaviors of TCP over wireless networks: TFRC interprets every packet loss as the indication of congestion and reduces its sending rate every time, resulting in unnecessary throughput degradation in wireless networks since a high percentage of losses are not caused by congestion.

Over recent years, some researchers have studied the throughput degradation problem of TFRC over wireless networks, and tried to eliminate such suffering from different ways.
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These efforts can also be divided into two categories: 1) Network-support enhancements, which need the support from the intermediate nodes in the network. Typical examples are ECN-based TFRC [31], Proxy-based TFRC [32], WM-TFRC [33], and AED-based TFRC [34]. 2) End-to-end enhancements, which modified TCP mechanism only on the end nodes (sender or receiver or both). MULTFRC [35] belongs to this category. Similarly, end-to-end enhancements do not need to add any new functions to the intermediate nodes in the network as compared to network-support enhancements, and thus avoid deployability problem in the real networks.

While MULTFRC achieves much higher throughput than TFRC does over wireless networks, it suffers poor fairness and poor TCP-friendliness during transmission. This will be elaborated later in Chapter 4. How to enhance TFRC in throughput while not sacrifice other advantages of TFRC such as good fairness and good TCP-friendliness is still an ongoing issue.

1.3 Summary of Contributions

This thesis considers the wireless issues in end-to-end congestion controls, and provides analysis and enhancements for both bulk data transfer and streaming. The series of analysis and enhancements is based on TCP Veno. Specifically, we first derive a throughput model for TCP Veno. After the in-depth analysis on TCP Veno, we propose two enhancements to further improve its performance over wireless networks. Finally, we introduce the core idea of TCP Veno into TFRC, the end-to-end congestion control for streaming, and propose two enhancements for TFRC over wireless networks. Based on the difference of target applications, the contributions can be divided into two areas: works for bulk data transfer and works for streaming.
1.3.1 Analysis and Enhancements of End-to-end Congestion Controls for Bulk Data Transfer over Wireless Networks

• Throughput Model for TCP Veno

This work is the footstone of the thesis. It develops an analytic approach in packet level to characterize TCP Veno behavior in both wired and wireless situations. Being different from the equation of TCP Reno, a more general closed form equation called “Veno equation” is derived, taking into account of the dynamic congestion window reductions in Veno over wireless networks [36]. Extensive experiments, including simulations and live Internet tests, demonstrate that the Veno equation is able to accurately predict TCP Veno throughput over different network scenarios, ranging from very low lossy links to very heavy lossy links.

The throughput model provides a theoretical way to analyze TCP Veno behaviors during transmission. Moreover, the Veno equation can be used as a more advanced equation in TFRC to enhance its performance over wireless networks (seeing the following section).

• TCP Veno Enhancement over Light-load Wireless Networks

After the comprehensive analysis, TCP Veno is found to still miss much available bandwidth over wireless transmissions, especially when the network load is light and random losses are pervasive. A new variable, named congestion loss rate, is introduced in this work [37]. It helps Veno act more efficiently in response to random losses and grab more available bandwidth on the link, without any fairness or friendliness sacrificed. Simulations and real network experiments have shown the better performance of this enhancement as compared to the original Veno.

• TCP Veno Enhancement over High Bandwidth-Delay Product Networks
Another enhancement of TCP Veno aims to alleviate random loss impacts on TCP Veno over high Bandwidth-Delay Product (BDP) networks. This work is based on the observation that higher BDP of the network makes random loss cause more negative impact on TCP throughput [13]. An approximated variable called $pBDP$ is introduced to inform Veno situations of the network and refine its congestion window evolution [38]. Both simulations and real network experiments are used to verify its performance.

1.3.2 Analysis and Enhancements of End-to-end Congestion Controls for Streaming over Wireless Networks

- **Veno Equation Based TFRC Enhancement**

TFRC employs a throughput equation to control its sending rate on streaming. However, this equation is derived from TCP Reno (hereinafter, called the “Reno equation”), and thus unavoidably makes TFRC suffer severe performance degradation from wireless links. In this work, we explore to enhance TFRC over wireless networks using a more advanced equation – Veno equation – to replace the Reno equation. The new protocol is also referred as TFRC Veno [39].

Our extensive experiments including simulations and live Internet measurements demonstrate that TFRC Veno can achieve significant throughput improvement over TFRC while keep the same nice fairness, TCP-friendliness and smoothness as TFRC. Furthermore, TFRC Veno only involves modification of TFRC on the sender side without requiring any changes of the receiver protocol stack or intervention of the intermediate network nodes. As compared to another sender-modified enhancement MULTFRC, our proposal is much more friendly to other network traffic when achieving the same throughput improvement.
• **Veno State Differentiator Based TFRC Enhancement**

We also propose another way to introduce Veno idea into TFRC mechanism [40]. Specifically, it utilizes Veno’s state differentiator to distinguish congestion losses and random losses during transmission. By discounting the impact of those non-congestion losses on the throughput calculation, it can effectively alleviate the throughput degradation caused by wireless links. The simulation results have shown that, our proposal can achieve better throughput as compared to the original TFRC. Meanwhile, it maintains other merits of the original TFRC, such as sending rate smoothness, fairness, and TCP-friendliness.

### 1.4 Outline of The Thesis

The rest of the thesis is organized as follows:

Chapter 2 reviews the two categories of end-to-end congestion controls: for bulk data transfer and for streaming. In the bulk data transfer part, we mainly introduce the primary version of TCP, TCP Reno, and one of its enhancements over wireless networks, TCP Veno. Other TCP variants for wireless networks are also brieﬁed. In the streaming part, we describe the mechanisms of TFRC and one of its wireless variants, MULTFRC, in detail. Other wireless enhancements of TFRC are also introduced.

Chapter 3 presents the detailed work on end-to-end congestion controls for bulk data transfer. We first describe the derivation of the throughput model for TCP Veno, and then provide two enhancements for TCP Veno to further enhance its performance over two different types of wireless networks. Simulations and real network experiments results are also given in this chapter.

Chapter 4 describes our work on end-to-end congestion controls for streaming. Two proposals are provided, both based on the core idea of TCP Veno: one is equipped with the Veno equation, which can model TCP Veno throughput; another uses Veno’s state
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differentiator. Simulations and real network experiments results for these two proposals are also shown in this chapter.

Chapter 5 concludes the thesis with a summary of the works. We also give suggestions for future work and possible directions.
Chapter 2

Literature Review

This chapter reviews the end-to-end congestion controls for bulk data transfer and for streaming respectively. In the bulk data transfer part, we mainly introduce the primary version of TCP, TCP Reno, and one of its enhancements over wireless networks, TCP Veno. Other TCP variants for wireless networks are also given. In the streaming part, we describe the mechanisms of TFRC and one of its wireless variants, MULTFRC, in detail. Other wireless enhancements of TFRC are introduced as well.

2.1 End-to-end Congestion Controls for Bulk Data Transfer

As the end-to-end congestion control for bulk data transfer, TCP and its variants have made great success in the Internet. This section gives an overview on TCP, including the primary version, TCP Reno, and some enhancements for wireless networks, e.g., TCP Veno.

2.1.1 TCP Reno

TCP Reno [41] is the primary version of TCP, which contains all necessary algorithms of TCP such as slow start (SS), additive-increase/multiplicative-decrease (AIMD), fast retransmit and fast recovery (FF), and retransmission timeout (TO). With reference
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to Figure 2.1, Reno starts with the slow start (SS) phase and its congestion window \((cwnd)\) is incremented exponentially until \(cwnd\) reaches a threshold called slow start threshold \((ssthresh)\). After that, the congestion window increases linearly till a packet loss is experienced, at which point the window is halved. This phase is called additive-increase/multiplicative-decrease (AIMD) phase. In order to retransmit the lost packet efficiently, Reno enters fast transmission and fast recovery (FF) phase until the packet is successfully received. However, sometime the sender is unable to know whether the packet is lost or not after a long time (e.g., network breakdown). In this case, Reno enters retransmission timeout (TO) phase to retransmit the timeout packets and restarts with the slow start phase. These four key algorithms will be described in detail in the following parts.

![Four key algorithms of TCP Reno.](image)

2.1.1.1 Slow Start (SS)

Slow start is used either at the start of a TCP connection or after a retransmission timeout (TO). The objective of slow start is to enable a TCP sender to discover the available bandwidth by increasing the amount of data injected into the network from initial window size of one segment, which prevents the TCP sender from congesting the network with a large burst of data.
Figure 2.2: Slow start.

Figure 2.2 shows the procedure of the slow start. In TCP, every time a data packet arrives at the receiver side, the receiver responds with an acknowledgement (ACK). The sender starts by transmitting one segment and waits for its ACK. When that ACK is received, the congestion window is incrementated from one to two, and two segments can be sent. When both of these two segments are acknowledged, the congestion window is increased to four. This provides an exponential growth of congestion window. In particular, the pseudo-code of the slow start is shown in Listing 2.1.

Listing 2.1: Slow Start algorithm

```c
// when an ACK arrives, cwnd = cwnd + 1;
```

By slow start Reno can quickly reach or approximate the equilibrium point of network system, where the capacity of the Internet is fully occupied. After reaching the initial ssthresh or getting an packet lost, Reno ends the slow start phase and comes into the AIMD phase.

The initial value of ssthresh is usually preset in Reno. However, many researchers have proposed to use estimated initial ssthresh to reduce or overcome the severe buffer overflow caused by the exponential growth of the congestion window [42][43][44][45]. M. Allman et al. also observed that the accuracy of sender-side estimated ssthresh is greatly limited, as compared to the estimation at receiver side, which results in a scalability problem while yielding better estimation than the sender-side estimation [46].
2.1.1.2 Additive-increase/multiplicative-decrease (AIMD)

Once slow start is completed, the network is expected having entered an equilibrium status, and then Reno uses the AIMD scheme to adjust its $cwnd$. The AIMD algorithm has two adjustment phases: 1) when the traffic load is steady or decreasing, the algorithm is in the probing phase. The internal traffic increases linearly to absorb released resources and probes the state of the bottleneck. 2) When a congestion signal (three duplicate ACKs, which will be discussed later) is received, the algorithm decreases the window size by half. This pattern of continually increasing and decreasing the congestion window continues throughout the lifetime of the connection.

As illustrated in Figure 2.1, the value of $cwnd$ as a function of time in AIMD follows a “sawtooth” pattern. The important thing to understand about AIMD is that the source is willing to reduce its congestion window at a much faster rate than it is willing to increase its congestion window. The details of the AIMD algorithm are described in Listing 2.2.

```c
// when an new ACK arrives,
cwnd = cwnd + 1/cwnd;
// otherwise a segment is lost,
ssthresh = cwnd/2;
cwnd = ssthresh;
```

Here the $ssthresh$ is used as an estimator of system equilibrium point. Its choice is key to the performance of AIMD algorithm. Note that AIMD is sometime also called congestion avoidance (CA) in some literature.

2.1.1.3 Fast Retransmission and Fast Recovery (FF)

Before describing the FF algorithm, we take a closer look at the operations of acknowledgement (ACK) in TCP receiver. It is known that the receiver responds with a new ACK when a packet arrives. However, if the packet arrives out of order, the receiver
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will re-send the same ACK it sent last time. This second transmission of the same ACK is called a duplicate ACK. TCP generates duplicate ACKs (DUPACK) for every out-of-order segment received.

![Diagram of TCP acknowledgement operations](image)

**Figure 2.3:** Operations of acknowledgement in TCP receiver.

As an example in Figure 2.3, the destination receives packets 1 and 2, but packet 3 is lost in the network, making packet 4, 5, and 6 out-of-order segments. Thus, the destination will send a duplicate ACK for packet 2 when packet 4 arrives, again when packet 5 arrives, and so on. When the sender sees the third duplicate ACK for packet 2, it notices the packet loss and retransmits packet 3. Note that when the retransmitted copy of packet 3 arrives at the destination, it then sends a cumulative ACK (ACK 6) for everything up to and including packet 6 back to the source.

The fast retransmission and fast recovery algorithms are usually implemented together. When three duplicate ACKs are received, the fast retransmission algorithm is triggered to retransmit the lost packet. After that, the fast recovery algorithm governs the transmission of new data until a non-duplicate ACK arrives. Specifically, 1) it sets
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$ssthresh$ to one half of the current $cwnd$, and set $cwnd$ to $ssthresh$ plus 3. 2) Each time another duplicate ACK arrives, increase $cwnd$ by one, and transmit a new segment if the $cwnd$ is larger than the number of unacknowledged segments. 3) When the next ACK arrives that acknowledges new data, set $cwnd$ to $ssthresh$ (the value set in step 1). Listing 2.3 is the pseudo-code of these operations.

Listing 2.3: Fast retransmission and fast recovery algorithm

```plaintext
1) Retransmit the missing packet, and
   $ssthresh \leftarrow cwnd/2$;
   $cwnd \leftarrow ssthresh + 3$;
2) Each time another dup ACK arrives,
   $cwnd \leftarrow cwnd + 1$;
3) When the next ACK acknowledging new data arrives,
   $cwnd \leftarrow ssthresh$;
```

In some extent, FF can be seen as an extension of MD (multiplicative-decrease) algorithm since it just follows MD phase. After the end of FF phase, the AI (additive-increase) algorithm takes over starting from latest setting $ssthresh$.

2.1.1.4 Retransmission Timeout (TO)

Reno holds a variable $RTO$ (Retransmit Time Out), which maintains the value of the time to wait for an ACK after sending a segment. If the waiting time exceeds $RTO$, Reno enters the retransmission timeout (TO) phase, which resets $cwnd$ to 1 and enters the slow start phase again.

The value of $RTO$ depends on another variable $RTT$, which is the round-trip time of the connection, i.e., the time it takes for the segment to travel from the sender to the receiver plus the time it takes for the ACK to travel from the receiver to the sender. Because real $RTT$ may vary every time, Reno estimates the mean (smoothed) $RTT$ via a low-pass filter. If we assume that $S.RTT$ is this smoothed $RTT$ estimate and $M.RTT$ is the most recently measured $RTT$, $RTO$ can be calculated as Listing 2.4.

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Listing 2.4: RTO estimation

$$RTO = S_{RTT} + 4 \times v;$$

where,

$$S_{RTT} = \alpha \times M_{RTT} + (1 - \alpha) \times S_{RTT};$$

$$v = \beta \times |M_{RTT} - S_{RTT}| + (1 - \beta) \times v;$$

Here $v$ is the smoothed mean deviation of $RTT$. $\alpha$ is usually set to $1/8$, and $\beta$ is set to $1/4$.

2.1.1.5 TCP Reno Variants

In Reno, if one packet is lost in a single window of data, after the lost packet is retransmitted to the receiver, the sender will receive a new ACK which acknowledges all of the packets transmitted before Fast Retransmit was entered, as shown in Figure 2.3. However, in the case of multiple packets dropped from a single window of data, this new ACK will only acknowledge the packets that are received before the second loss. The left packets transmitted before Fast Retransmit is not acknowledged by this ACK. We call this kind of ACK a partial ACK. Partial ACKs bring the sender out of fast recovery in Reno, resulting in a timeout. This will cause poor throughput in Reno. This throughput degradation has been observed in [47][48]. In order to quickly recover multiple packet losses, there are a number of enhanced loss recovery algorithms proposed such as NewReno [49], SACK [48][50], and FACK [51].

- TCP NewReno

In NewReno, when a sender receives a partial ACK, it does not come out of fast recovery. Instead, it assumes that the segment immediately after the most recently acknowledged segment has been lost, and hence the lost segment is retransmitted. Thus, in a multiple segment loss scenario, NewReno does not wait for a retransmission timeout and continues to retransmit lost segments every time it receives a partial ACK. This brings more efficiency in packet loss recovery.
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- **TCP SACK**

  Another way to deal with multiple segment losses is to tell the sender which segments have arrived at the receiver. Selective Acknowledgments (SACK) does exactly this. The receiver uses ACK with SACK option to indicate to the sender one contiguous block of data that has been received out of order at the receiver.

  When SACK blocks are received by the sender, they are used to maintain an image of the receiver queue, i.e., which segments are missing and which have made it to the receiver. Scoreboard is set up to track transmitted and received packets according to the previous information of the SACK option. For each transmitted segment, scoreboard records its sequence number and a flag bit that indicates whether the segment has been “SACKed”. A segment with the SACKed bit turned on does not require to retransmit, but segments with the SACKed bit off and sequence number less than the highest SACKed segment are available for retransmission. Whether a SACKed segment is on or off, it is removed from the retransmission buffer only when it has been cumulatively acknowledged.

- **TCP FACK**

  Forward Acknowledgments (FACK) is built on top of SACK. The name “Forward Acknowledgements” comes from the fact that the algorithm keeps track of the correctly received data with the highest sequence number. FACK provide a better way to estimate outstanding data more precisely, thus it is less bursty than SACK and can recover from episodes of heavy loss better than SACK. It also modifies the slow start phase of Reno, which decreases the congestion window more when a loss occurs during or near the slow-start period. Furthermore, it proposes a better way of halving the congestion window when congestion is detected: instead of immediately halving, the congestion window is decremented gradually.
2.1.1.6 Throughput Models for TCP Reno and its variants

Over last years, there have been many researchers working on providing analytical models for TCP Reno and its variants. Here we gives a brief review on these previous efforts.

In 1997, M. Mathis et al. [52] studied the long-term behavior of a TCP connection, and provided a simple throughput equation for TCP:

\[ B(p) = S \frac{k}{RTT \sqrt{p}} \]  \hspace{1cm} (2.1)

where \( B(p) \) is the throughput, \( S \) is the segment size, \( RTT \) is the round-trip time, \( K \) is a constant, and \( p \) is the packet loss rate. Note that this model only considers the AIMD algorithms during TCP communications.

In 1998, J. Padhye et al. derived a more complete model that includes the two types of loss indications used by TCP, namely triple duplicate ACK loss and timeout loss. Furthermore, it also considers the receiver window limitation during transmission. The derived equation is the Reno equation we mentioned in Chapter 1:

\[ B(p) \approx \min \left( \frac{W_{\text{max}}}{RTT} \sqrt{\frac{2p}{b}}, +T_0 \min \left( 1, \sqrt{\frac{3p}{2}} \right) p(1+32p^2) \right) \]  \hspace{1cm} (2.2)

where \( B(p) \) is the throughput; \( W_{\text{max}} \) is the receiver window limitation; \( RTT \) is the round-trip time; \( b \) is a constant; \( p \) is the packet loss; \( T_0 \) is the RTO value. This equation is also used by S. Floyd in TFRC as a rate control protocol for streaming applications.

This model is further extended in [53] and [54] by including startup effects such as the connection establishment and the slow start, which have a significant impact on the latencies of short TCP transfers. Another model for short live TCP flows is also studied by [55].

All the above models are based on the drop-tail packet loss model [56]. The drop-tail packet loss model means: 1) the first packet in a round is lost with probability \( p \); 2) if the previous packet was not lost, the packet is lost with probability \( p \); 3) if a packet is lost,
then all subsequent packets in the round are lost. There is also another packet loss mode called the correlated packet loss. It means: 1) the first packet in a round is lost with probability \( p \); 2) if the previous packet was not lost, the packet is lost with probability \( p \); 3) if the previous packet was lost, the packet is lost with probability \( q \). Papers [57] and [58] studied models based on this kind of packet loss.

There are also other efforts on modeling TCP performance. For example, paper [59] uses jump process driven Stochastic Differential Equations to model interactions of a set of TCP flows and Active Queue Management (AQM) routers such as Random Early Detection (RED) [60]; paper [61] provides a duality model for the interactions of TCP flows and network resources.

2.1.2 TCP Veno

Despite of the sophisticated recovery algorithms employed, TCP Reno and its variants do not modify the AIMD algorithms during transmission. That is to say, they always halve their window size whenever packet loss is detected by fast retransmit. This has caused negative effect of unnecessarily lowering the throughput when the loss is due to bit errors rather than network congestions. Particularly, in wireless networks with high bit error rates, this unnecessary throughput degradation is obvious.

TCP Veno deals with this wireless issue by modifying Reno’s AIMD phase based on its state differentiator. Such state differentiator distinguishes congestive state and non-congestive state, and employs refined AIMD algorithms for different states.

2.1.2.1 Veno State Differentiator

Veno’s state differentiator comes from the congestion detection method in TCP Vegas [62]. TCP Vegas is a quite different congestion control as compared to TCP Reno or its variants, because it can detect the early stage of congestion before packet loss occurs. However, Vegas has not been widely employed because it behaves conservatively in
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throughput when competing with the co-existing Reno-style connections [63][64]. Despite of its conservative behavior in congestion control, Vegas’ congestion detection method does provide a clue on how to differentiate between congestive and non-congestive state.

The congestion detection method of Vegas works as follows. The sender measures the so-called *Expected* and *Actual* rates:

\[
\begin{align*}
\text{Expected} &= \frac{cwnd}{\text{BaseRTT}} \quad (2.3) \\
\text{Actual} &= \frac{cwnd}{\text{RTT}} \quad (2.4)
\end{align*}
\]

where \(cwnd\) is the current congestion window size; \(\text{BaseRTT}\) is the minimum of the measured round-trip time collected so far, and is reset whenever packet loss is detected either due to time-out or duplicated ACKs; \(\text{RTT}\) is the smoothed round-trip time measured. The *Expected* rate can be seen as the best possible throughput the link can offer, and the *Actual* rate is the measured instant throughput during transmission. The difference of the rate is:

\[
\text{Diff} = \text{Expected} - \text{Actual} \quad (2.5)
\]

Vegas then uses the variable \(N = \text{Diff} \times \text{BaseRTT}\) as a congestion detector. Specifically, if \(N < \alpha\), Vegas considers the link is far away from congestion; if \(N > \beta\) (\(\beta > \alpha\)), the link is going to be congestive; if \(\alpha \leq N \leq \beta\), then the link is in an equilibrium point.

The implicit idea of Vegas is to maintain the queuing data in the bottleneck router buffer \(N\) to be between the lower threshold (\(\alpha\)) and upper threshold (\(\beta\)). This can be seen as follows. When the queue size is zero, the \(\text{RTT}\) is \(\text{BaseRTT}\). \(\text{BaseRTT}\) is basically the sum of the transmission delays and propagation delays throughout the forward and reverse paths (with queuing delay equal to zero). Then, when the packets of the connection accumulate at the queue, \(\text{RTT} > \text{BaseRTT}\), and we have:

\[
\text{RTT} = \text{BaseRTT} + \frac{N}{\text{Actual}} \quad (2.6)
\]
Here $N$ is the queuing data size. That is, the extra delay of $RTT$ is due to the queuing delay. Rearranging, we have:

$$N = Actual \times (RTT - \text{BaseRTT})$$

$$= Diff \times \text{BaseRTT}$$

which is just the variable used in Vegas.

Fundamentally different from TCP Vegas, which use $N$ to constraint window fluctuation around the system equilibrium, TCP Veno only uses it to estimate network state – congestive or non-congestive. Specifically, if $N < \beta$ we can declare that the connection is not experiencing congestion and the network is in non-congestive state. Packet losses incurred in this state are most likely due to non-congestion reasons, e.g., bit errors. If $N \geq \beta$, on the other hand, the network is in congestive state and packet losses are considered to be induced by congestion reason.

By using this implicit state indication, TCP Veno adopts a more flexible AIMD mechanism than Reno does, which can improve TCP throughput over wireless networks.

### 2.1.2.2 Modified Additive-increase (AI) Algorithm in TCP Veno

Veno refines the additive-increase algorithm of Reno by forcing the TCP connection to stay longer at the operating region in which the bandwidth is fully utilized. Specifically, if $N < \beta$ where the network is detected as in non-congestive state and the available bandwidth is considered under-utilized, Veno increases its congestion window by $1/cwnd$ for every new ACK received; if $N \geq \beta$ where the network is congestive and the available bandwidth is considered fully utilized, then Veno increase its congestion window by $1/cwnd$ for every other new ACK received. The pseudo-code is shown in Listing 2.5.

```
if $(N < \beta)$
    \text{cwnd} = \text{cwnd} + 1/cwnd \text{ for every new ACK received};
else if $(N \geq \beta)$
    \text{cwnd} = \text{cwnd} + 1/cwnd \text{ for every other new ACK received};
```

Listing 2.5: Modified AI phase in TCP Veno
2.1.2.3 Modified Multiplicative-decrease (MD) Algorithm in TCP Veno

The main improvement of Veno in throughput over wireless networks attributes to the modified multiplicative-decrease algorithm. Veno makes the congestion window cut down less (i.e., 1/5) when packet loss occurs in the non-congestive state, thus it improves Reno’s throughput during transmission. Specifically, if \( N < \beta \) where the network is in non-congestive state, packet losses incurred in this state are most likely due to non-congestion reasons, so Veno sets \( ssthresh \) to be \( 4/5 \) of \( cwnd \) and enters MD and FF phases; if \( N \geq \beta \) where the network is in congestive state, packet losses are considered to be induced by congestion reasons, so Veno sets \( ssthresh \) to be \( 1/2 \) of \( cwnd \) and enters MD and FF phases. The pseudo-code is shown in Listing 2.6. Note that here the code includes MD and FF phases.

Listing 2.6: Modified MD phase in TCP Veno

1) Retransmit the missing packet, and
   - if \( (N < \beta) \)
     \[ ssthresh = cwnd \times \frac{4}{5} \]
   - else if \( (N \geq \beta) \)
     \[ ssthresh = cwnd \times \frac{1}{2} \]
   \[ cwnd = ssthresh + 3 \]
2) Each time another dup ACK arrives,
   \[ cwnd = cwnd + 1 \]
3) When the next ACK acknowledging new data arrives,
   \[ cwnd = ssthresh \]

2.1.2.4 TCP Veno Variants

Since Veno only modifies the AIMD phase of Reno, it can also be integrated with different recovery schemes such as those in NewReno, SACK or FACK. These variants, namely NewVeno, SACK Veno, and FACK Veno, are all elaborated in [65]. Experiments have shown that these variants can further improve Veno’s throughput.

2.1.3 Other TCP Variants over Wireless Networks

- Snoop TCP
Snoop TCP requires to install a module called snoop agent at the base station (BS), which resides on the border between wired and wireless networks. The snoop agent monitors TCP traffic passing through it and caches the new packets. Once a packet is informed by the receiver to be lost (e.g., three duplicated ACKs) and the agent is aware that the packet is lost between itself and the wireless host, it will quickly retransmit the lost packet cached before and suppress the duplicated ACKs from the sender. This prevents unnecessary congestion control invocations at the sender and improves the whole throughput of TCP.

- **TCP with ECN**

Explicit Congestion Notification (ECN) [66] is an extension proposed to RED. RED is an active queue management mechanism in routers, which detects congestion before the queue overflows and provides an indication of this congestion to the end nodes. A RED router signals incipient congestion to TCP by dropping packets probabilistically before the queue runs out of buffer space. RED router operates by maintaining two levels of thresholds minimum ($\text{minth}$) and maximum ($\text{maxth}$). It drops packets probabilistically if and only if the average queue size lies between the $\text{minth}$ and $\text{maxth}$. ECN relies on extension to RED, which marks a packet instead of dropping it when the average queue size is between $\text{minth}$ and $\text{maxth}$. Upon receipt of congestion marked packet, the TCP receiver informs the sender (in the subsequent acknowledgement) about incipient congestion, which in turn will trigger the congestion avoidance algorithm at the sender. With ECN notification, TCP only concerns the congestion losses, eliminating impacts from random losses. Thus its throughput can be improved over wireless networks.

- **TCP Westwood**
TCP Westwood is a sender-side modification of the TCP congestion window algorithm that improves upon the performance of TCP Reno in wired as well as wireless networks. The general idea of Westwood is to use the bandwidth estimate $BWE$ to set the congestion window ($cwnd$) and slow start threshold ($ssthresh$) after a congestion episode. Specifically, $BWE$ is estimated by the sender along a TCP connection by measuring and averaging the rate of returning ACKs. Whenever the sender perceives a packet loss (i.e. a timeout occurs or three duplicate ACKs are received), the sender uses the $BWE$ to properly set the $cwnd$ and the $ssthresh$, rather than blindly halving the congestion window as Reno does. In this way TCP Westwood avoids overly conservative reduction of $cwnd$ and $ssthresh$ and thus improves its performance.

TCP Westwood has some variants such as TCP Westwood+ [67] and TCP Westwood-A [68]. TCP Westwood+ improves Westwood bandwidth estimation algorithm to avoid inaccurate estimation of $BWE$ in the presence of reverse traffic due to ACK compression. TCP Westwood-A introduces TCP Westwood with agile probing, which helps Westwood deal well with highly dynamic bandwidth, large propagation time/bandwidth, and random loss in the current and future heterogeneous Internet.

- **WTCP**

Wireless Transmission Control Protocol (WTCP) is quite different to standard TCP algorithms. It uses rate control instead of window-based congestion control. Specifically, the sender transmits data at an advertised rate that is checked and controlled by the receiver. The receiver checks the traffic rate by measuring the ratio of the inter-packet arrival time at the receiver to inter-packet sending time at the sender. When congestion becomes imminent, this period will become longer,
and the traffic rate can be altered accordingly through the receiver’s acknowledge-
ments. WTCP can also distinguish between congestion losses and random losses
by checking the inter-packet delay of packets arriving close to the lost ones. Thus
it can achieve better performance over wireless networks.

- **TCP-AP**

TCP-AP (TCP with Adaptive Pacing) was proposed to improve the throughput
of TCP over ad hoc networks. It paces the packets allowed to be sent out by the
congestion window adaptively. To configure the pacing, a metric of the 4-hop prop-
agation delay is evaluated. It is the time between the transmission of a packet by
the TCP source node and its reception by the node four hops downstream. The
4-hop propagation delay is chosen because a transmission currently in progress is
assumed to interfere within a range of four hops, given the theoretical and exper-
imental analysis. TCP-AP can be regarded as a hybrid between a window-based
and a rate-based approach, adding rate-based mechanisms to TCP in order to avoid
large bursts of packets. As a result, it reduces the packet losses due to the interfere
within adjacent nodes, and improve the throughput.

- **C³TCP**

C³TCP (Cross-layer congestion control) is also an enhanced congestion control for
ad hoc networks. In C³TCP, the information of bandwidth and delay within an
end-to-end link is obtained by cumulating the intermediate hops’ measurements.
So C³TCP needs a feedback field added to the link layer header. When an ACK
is generated at the destination node, the measurements are gathered hop by hop
from the intermediate nodes and stored in the feedback field. There is an additional
module located under TCP in the sender’s protocol stack, which can modify the
advertised window field in the ACK to dynamically limit the sender’s sending rate
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based on the measurements collected. By gathering sufficient information about the network, $C^3$TCP avoids sending burst traffic to the ad hoc network and thus improve its throughput.

- **TCP-HO**

TCP-HO (TCP HandOff) was proposed to deal with heterogenous mobile networks, where different kinds of handoffs may often occur. TCP-HO assumes that a mobile host is able to detect the completion of handoff immediately and has a coarse estimation of new wireless links bandwidth. When a mobile host detects handoff completion, it will immediately notify the server through two duplicate ACKs, whose TCP option also carries the bandwidth of new wireless link. After receiving this notification, the server begins to transmit immediately and keeps updating ssthresh according to the bandwidth from mobile host and its new RTT samples. This updating will be stopped after four RTT samples or after congestion is detected. TCP-HO avoids unnecessary timeout in TCP when handoff occurs and shifts TCP’s live connection to the new network quickly. Experimental results show that TCP-HO improves TCP performance without adversely affecting cross traffic in a heterogeneous mobile environment.

2.2 End-to-end Congestion Controls for Streaming

Streaming applications are more sensitive to delays than to packet loss. This makes TCP unsuitable for them, because the reliability and ordering semantics in TCP increase end-to-end delays and delay variations during transmission. Moreover, the AIMD algorithms in TCP cause large and abrupt reductions in transmission rate on packet losses, which also brings problems in playing real-time video/audio. Unlike TCP that has dominated end-to-end congestion control for bulk data transfer, congestion control for streaming is
still an emerging and ongoing research. There have been many studies on equation based and other congestion control mechanisms [69][70][71][72][73]. This thesis studies one of the most popular and recent protocols – TCP-Friendly Rate Control (TFRC), which is based on the Reno equation and has been standardize in RFC 3448. This section also reviews some enhancements of TFRC over wireless networks, e.g., MULTFRC.

### 2.2.1 TFRC

In general, TFRC is an equation based rate control protocol. Figure 2.4 gives an overview of TFRC mechanisms.

![TFRC Diagram](image)

Figure 2.4: Overview of TFRC mechanisms.

TFRC lets receiver estimate the packet loss rate $p$ on the link (the estimation method will be detailed latter). The estimated loss rate $p$ is returned to the sender through ACKs. After receiving the value of $p$, the sender adjusts its sending rate based on the Reno equation that can model TCP Reno throughput:

$$ T_{\text{calc}} = \frac{S}{RTT \sqrt{\frac{2p}{3}} + 3RTO \sqrt{\frac{3p}{8}} p (1 + 32p^2)} $$

where $T_{\text{calc}}$ is the expected sending rate in bytes/sec; $S$ is the packet size in bytes; $RTT$ is the round-trip time; $p$ is the steady-state rate of loss event; $RTO$ is the retransmit timeout value. Note that, TFRC sender does not need to retransmit any missing packets during the transmission.

Following parts describe some other important algorithms in TFRC.
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2.2.1.1 Slow Start

Slow start in TFRC occurs when TFRC is initialized or restarted after an idle time. In the slow start procedure, TFRC doubles its sending rate of each $RTT$, until packet losses are informed by the receiver through ACKs. Once a loss occurs, the slow start is terminated and TFRC uses the Reno equation (2.8) to calculate its sending rate.

2.2.1.2 Estimations of $RTT$, $RTO$, and $p$

In order to use the Reno equation, several variables in the equation are required, such as $RTT$, $RTO$, and $p$.

- **Estimations of $RTT$ and $RTO$**

  The measurement of $RTT$ in TFRC is similar as that in TCP, that is, TFRC uses a low-pass filter to estimate the smoothed round-trip time. However, the estimation of $RTO$ in TFRC is quite simple because $RTO$ only critically affects the sending rate when the packet loss rate is very high. TFRC calculates $RTO$ as $RTO = 4 \times RTT$.

- **Estimation of $p$**

  The packet loss rate $p$ is another critical variable in the Reno equation. It is estimated at the receiver side as follows.

  The receiver judges a packet is lost when three packets with higher sequence number than itself arrive. The term loss event refers to several packets lost within one round-trip time. Note that the subsequent losses following the first loss in the round-trip time are ignored, i.e., at most one loss event in one round-trip time. The term loss interval is defined as the number of packets between loss events. The value of a loss interval is obtained by subtracting the sequence number of the first
lost packet in a loss event from the sequence number of the first lost packet in the subsequent loss event. Figure 2.5 illustrates the relationship between loss events and loss intervals.

![Diagram of loss events and intervals](image)

**Figure 2.5: Estimation of loss events.**

Let $A_i$ be the $i$th recent loss interval, and let $w_i$ be the weight of $A_i$. The receiver calculates the average loss interval of the recent $n$ loss intervals as follows:

$$A(1, n) = \frac{\sum_{i=1}^{n} w_i A_i}{\sum_{i=1}^{n} w_i}$$

(2.9)

In TFRC, the default value of $n$ is 8, and $w_1 = w_2 = w_3 = w_4 = 1, w_5 = 0.8, w_6 = 0.6, w_7 = 0.4, w_8 = 0.2$. Let $w = 1/\sum_{i=1}^{n} w_i = \frac{1}{6}$, we can rewrite equation (2.9) as

$$A(1, n) = w \sum_{i=1}^{n} w_i A_i$$

(2.10)

In calculating $A(1, n)$, equation (2.9) or (2.10) ignores the most recent loss interval $A_0$, which is the number of received packets since the latest loss event. If there is no loss event in one round-trip time, $A_0$ will continue to increase. If $A_0$ is large enough, it means that there are no packet losses for a long time; therefore the average loss interval should be increased accordingly. Thus, TFRC also calculates $A(0, n - 1)$,
which covers $A_0$, according to

$$A(0, n - 1) = w \sum_{i=0}^{n-1} w_{i+1} A_i$$  \hspace{1cm} (2.11)

The refined value of average loss interval, $A$, is determined by

$$A = \max(A(1, n), A(0, n - 1))$$  \hspace{1cm} (2.12)

Finally, the loss event rate, $p$, is given by

$$p = \frac{1}{A}$$  \hspace{1cm} (2.13)

2.2.1.3 No Feedback Timeout

TFRC maintains a timer called the nofeedback timer. This timer is reset whenever the sender receives feedback (ACK) from the receiver. If there is no feedback arrival until it expires, the timer triggers no feedback timeout, and TFRC cuts the sending rate in half. However, if the sender has been idle for a long time and the sending rate is reduced to two packets per $RTT$, the halving operation is stopped. This ensures that the sending rate is never less than two packets per $RTT$ as a result of an idle period. Once a new feedback resumes reporting no losses, the sender enters slow start procedure and doubles its sending rate each $RTT$, as described in Section 2.2.1.1.

2.2.2 MULTFRC

TFRC inherits TCP throughput degradation over wireless networks due to its embedded Reno equation. MULTFRC is a kind of sender-side modifications of TFRC, which only changes the sender protocol stack of TFRC and provides significant improvement in throughput over wireless networks.

MULTFRC borrows the idea from MULTCP [74], which was originally used to improve TCP performance in high bandwidth-delay product networks. The authors of
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MULTFRC find that using one TFRC connection in streaming applications results in under-utilization of the wireless bandwidth. They then propose the use of multiple simultaneous TFRC connections for a given wireless streaming application. The basic idea behind MULTFRC is to measure the round-trip time and adjust the number of connections accordingly. Specifically, MULTFRC increases the number of connections $n$ by $\alpha/n$ or decrease it by $\beta$ depending on the $rtt$ measurements, that is,

$$n = \begin{cases} n - \beta, & \text{if } ave_{rtt} - rtt_{min} > \gamma rtt_{min} \\ n + \alpha/n, & \text{otherwise} \end{cases}$$

where $\alpha$, $\beta$ and $\gamma$ are preset constant parameters, and the default values are 1, 1, and 0.2 respectively; $ave_{rtt}$ is the average round trip time; for a given route, $rtt_{min}$ is the minimum round trip time for that route, i.e. with no queuing delay. As a result, under ideal conditions MULTFRC keeps increasing the number of connections to make $ave_{rtt}$ as close as possible to $(1 + \gamma)rtt_{min}$ without exceeding it. Ideally, $ave_{rtt}$ becomes larger than $rtt_{min}$ if and only if the link is fully utilized, and the queue on bottleneck link router is built up, introducing additional queuing delay. Thus by evaluating the difference between $ave_{rtt}$ and $rtt_{min}$, MULTFRC can detect whether link is fully utilized or not and adjust the number of connections accordingly. The authors also provide an enhancement of MULTFRC, called AIO-TFRC [75] to overcome the complexity issue of the MULTFRC implementation and undesirable results of quantization effect in MULTFRC. However, in spite of high throughput achieved, both MULTFRC and its variants suffer from poor fairness and poor TCP-friendliness, as what MULTTCP has encountered.

2.2.3 Other TFRC Variants over Wireless Networks

- ECN-based TFRC

ECN-based TFRC uses intermediate routers to detect incipient network congestion and inform the receiver using Explicit Congestion Notification (ECN) marking.
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The receiver then measures the marking event rate $p$ and feeds this information back to the sender. In this way, ECN-based TFRC ignores the losses occurring over the wireless hop or the effect of packet reordering encountered in the Internet, and only accounts congestion losses for sending rate adjustment.

- **Proxy-based TFRC**

Proxy-based TFRC employs a proxy to split the client-server connection, and uses TFRC only over the shared part of the client-server connection, i.e., the part between server and proxy. In the part between client and proxy, it tunnels RTP via a TCP which is TCP-friendly by definition. This method improves the average throughput and wireless link utilization.

- **WM-TFRC**

WM-TFRC uses the access point (AP) in wireless LAN to measure the rate of wireless loss events (i.e., loss events caused by bit error) $p_w$ and feeds back its value to the sender periodically. Meanwhile, the receiver also provides feedback about the rate of total loss events (including wireless loss and congestion loss) $p$ to the sender. Therefore, the sender can deduce the rate of congestion loss events (i.e., loss events caused by congestion) $p_{con}$ by subtracting $p_w$ from $p$. In this case, the sending rate of WM-TFRC, which is deduced from $p_{con}$ instead of $p$, can be improved.

- **AED-based TFRC**

AED-based TFRC uses a new differentiation scheme called Accurate and Explicit Differentiation (AED) to improve the performance of TFRC. AED assumes that agents are deployed before and after each wireless link. The agents snoop through each packet and detect a loss by finding a packet with an out-of-order sequence number. Agents at the end of wired networks treat a packet loss as congestion
loss and agents after a wireless link treat a packet loss as wireless loss. The agents then mark the packets that are not lost with information about the lost packets (i.e., whether those packets were lost due to congestion or bit error). Thus, when receiving these marked packets, the receiver can calculate the correct loss event rate and provide feedback to the sender.
Chapter 3
Analysis and Enhancements of End-to-end Congestion Controls for Bulk Data Transfer over Wireless Networks

This chapter presents our work on end-to-end congestion controls for bulk data transfer. We first derive a throughput model for TCP Veno, giving an in-depth analysis on TCP Veno behavior. Then we provide two enhancements for TCP Veno to further enhance its performance over two different types of wireless networks. Simulations and real network experiments results are also demonstrated.

3.1 Throughput Model for TCP Veno

In this work, we develop a throughput model for TCP Veno congestion control. The derived throughput equation, called Veno equation, can model TCP Veno throughput for bulk data transfer over different network scenarios, ranging from very low lossy links to very heavy lossy links. The simulation and live Internet results verify the effectiveness of such TCP Veno equation in the throughput modeling under different lossy situations.
3.1.1 Veno Equation

As described in Section 2.1.2, TCP Veno only refines the AIMD mechanism of Reno. The other parts of Reno, including the slow start, fast retransmit, fast recovery, and computation of the retransmission timeout, remain intact.

Here we simplify the AIMD mechanism of TCP Veno as follows:

in the AI phase,

\[
cwnd = \begin{cases} 
cwnd + \frac{1}{cwnd}, & \text{for every new ACK} \\
cwnd + \frac{1}{cwnd}, & \text{for every other new ACK} 
\end{cases} 
\]

and in the MD phase,

\[
cwnd = \begin{cases} 
cwnd \times \theta_1, & N < \beta \\
cwnd \times \theta_2, & N \geq \beta
\end{cases}
\]

Note that here we use \(\theta_1\) and \(\theta_2\) to replace the original fixed value in TCP Veno (i.e., 4/5 and 1/2), in order to derive a more general model which can characterize the effect of different congestion window drop factors.

![Figure 3.1: Congestion window evolutions of Veno and Reno.](image)

As shown in Figure 3.24, although the AI phase in Veno brings certain improvement in throughput of Reno, the most enhancement is attributed to Veno’s smart MD phase with dynamic window reductions. To simplify the problem, we ignore the slightly different AI phase between Veno and Reno in the following derivation. Thereafter, Veno acts quite
simply as Reno, and the only difference between Veno and Reno is how much the cwnd value should be cut down when losses occur: Reno always decreases its cwnd with factor $\frac{1}{2}$ while Veno sometimes decreases it with factor $\theta_1$ (random loss), and sometimes with factor $\theta_2$ (congestion loss). In this case, we introduce a random variable $\lambda$ to represent a dynamic window drop factor, rather than a fixed factor of $\frac{1}{2}$ seen in [11]. That is, $\lambda$ can be $\theta_1$ or $\theta_2$ with certain probability. In fact, $\lambda$ reflects how Veno takes window multiplicative reduction in response to packet loss.

With reference to [11], we also divide the derivation of the Veno equation into three steps: firstly we only consider the situation where packet losses are indicated by the triple duplicated ACKs. We call these losses triple-duplicated (TD) losses; then we include packet losses indicated by time outs, which are called time-out (TO) losses; finally we extend the previous model to include the impact of the advertisement window limitation.

3.1.1.1 Veno Equation for TD Losses Only

Following [11], we consider a TD period (TDP) in Veno, which is a period between two TD losses. As shown in Figure 3.2, the y-axis represents the number of packets sent during the current TDP, and the x-axis represents the number of rounds in the current TDP. Here a “round” starts with the back-to-back transmission of $W$ packets, where $W$ is the current size of the TCP congestion window. Once all packets falling within the congestion window have been sent in this back-to-back manner, no other packets are sent until the first ACK is received for one of these $W$ packets. This ACK reception marks the end of the current round and the beginning of the next round. In Figure 3.2 the number of rounds in a TDP is represented by $X_i$, $b$ is the number of packets that are acknowledged by a received ACK. Many TCP receiver implementations send one cumulative ACK for two consecutive packets received (i.e., delayed ACK), so $b$ is typically 2. If $W$ packets are sent in the current round and are all received and acknowledged correctly, then $W/b$
ACKs will be received. Since each ACK increases the congestion window size by $1/W$, in the current round the windows size is incremented by $W/b \times 1/W = 1/b$. $W_{i-1}$ and $W_i$ are the congestion window sizes at the ends of the previous TDP and the current TDP respectively. According to the Veno algorithm, the current TDP starts with the congestion window size of $\lambda_iW_{i-1}$, where $\lambda_i$ is either $\theta_1$ or $\theta_2$. After that, the window size is incremented by $1/b$ at each round until the end of the TDP.

*Figure 3.2: Packets sent during a TD period.*

If $Y_i$ is the number of packets sent in the period, and $A_i$ is the duration of the period, then the long-term steady-state Veno throughput can be shown to be:

$$B = \frac{E[Y]}{E[A]}$$  \hspace{1cm} (3.1)

It is observed in [11] that:

$$E[Y] = \frac{1-p}{p} + E[W]$$  \hspace{1cm} (3.2)

and,

$$E[A] = RTT(E[X] + 1)$$  \hspace{1cm} (3.3)

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where $p$ is the probability that a packet is lost, given that either it is the first packet in its round or the preceding packet in its round is not lost. Note that $p$ includes both congestion and random losses. $RTT$ is the average value of round trip time. Detailed derivation can be found in [11]. According to Figure 3.2 we have:

$$W_i = \lambda_i W_{i-1} + \frac{X_i}{b} \quad (3.4)$$

Then the packets transmitted between these two TD losses are:

$$Y_i = \sum_{k=0}^{X_i/b - 1} (\lambda_i W_{i-1} + k)b + \eta_i$$

$$= \lambda_i W_{i-1} X_i + \frac{X_i}{b}(X_i - 1) + \eta_i$$

$$= \frac{X_i}{2}(\lambda_i W_{i-1} + W_{i-1} - 1) + \eta_i \quad (3.5)$$

where $\eta_i$ is the number of packets sent in the last round, as seen in Figure 3.2. $\eta_i$ is considered as a uniformly distributed random variable between 1 and $W_i$ in [11], here we inherit this distribution in the Veno derivation.

From (3.4), (3.5) and (3.2), we have:

$$(1 - E[\lambda])E[W] = \frac{E[X]}{b} \quad (3.6)$$

and,

$$\frac{1 - p}{p} + E[W] = \frac{E[X]}{2}((1 + E[\lambda])E[W] - 1) + E[\eta] \quad (3.7)$$

where $E[\eta] = \frac{E[W]}{2}$.

Let $\gamma = E[\lambda]$, and from (3.6) and (3.7) we can get:

$$E[W] = \frac{b(1 - \gamma) + 1}{2b(1 - \gamma^2)} + \sqrt{\frac{2(1 - p)}{b(1 - \gamma^2)p} + \left(\frac{b(1-\gamma)+1}{2}\right)^2} \quad (3.8)$$

Then from (3.6) and (3.3), we have:

$$E[X] = \frac{b(1 - \gamma) + 1}{2(1 + \gamma)} + \sqrt{\frac{2b(1 - \gamma)(1 - p)}{(1 + \gamma)p} + \left(\frac{b(1-\gamma)+1}{2}\right)^2} \quad (3.9)$$
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and,

\[ E[A] = RTT \left( \frac{b(1-\gamma)+1}{2(1+\gamma)} + \sqrt{\frac{2b(1-\gamma)(1-p)}{(1+\gamma)p}} + \frac{(b(1-\gamma)+1)^2}{(1+\gamma)^2} + 1 \right) \]  

(3.10)

Finally, substituting \( E[A] \) and \( E[W] \) into (3.1) and (3.2) we have:

\[ B(p) = \frac{\frac{1-p}{p} + \frac{b(1-\gamma)+1}{2b(1-\gamma)^2} + \sqrt{\frac{2b(1-\gamma)(1-p)}{b(1-\gamma)^2p}} + \frac{(b(1-\gamma)+1)^2}{(1-\gamma)^2p} + 1}{RTT\left(\frac{b(1-\gamma)+1}{2(1+\gamma)} + \sqrt{\frac{2b(1-\gamma)(1-p)}{(1+\gamma)p}} + \frac{(b(1-\gamma)+1)^2}{(1+\gamma)^2} + 1\right)} \]  

(3.11)

For small values of \( p \), the above can be approximated by:

\[ B(p) \approx \frac{1}{RTT} \sqrt{\frac{1 + \gamma}{2b(1-\gamma)p}} \]  

(3.12)

3.1.1.2 Veno Equation for Both TD and TO Losses

So far equation (3.12) gives the Veno equation when only TD losses occur. Actually there are many TO losses over real communications, and they must be included in the model. According to [11], the TCP throughput with TO losses can be written as follows:

\[ B = \frac{E[Y] + Q \times E[R]}{E[A] + Q \times E[Z_{TO}]} \]  

(3.13)

where \( E[Y] \) and \( E[A] \) have been derived in the previous section. \( Q \) is the probability that a loss indication ending a TD period is a TO, \( E[R] \) is the mean number of packets sent during TO periods, and \( E[Z_{TO}] \) denotes the mean duration of TO periods.

As we have pointed out before, Veno does not change Reno’s algorithms on dealing with TO losses, such as the computation of the timeout, and the backoff algorithm. We can easily derive the equation following the same routine described in [11], because the expressions of \( Q, E[R], \) and \( E[Z_{TO}] \) are the same in both Veno and Reno. Therefore, we have:

\[ B(p) = \frac{\frac{1-p}{p} + E[W] + \hat{Q}(E[W])\frac{1}{1-p}}{RTT(E[X] + 1) + \hat{Q}(E[W])T_{V}f_{\frac{1}{1-p}}} \]  

(3.14)
where $E[W]$ and $E[X]$ have been given in the previous section. Note that, both $E[W]$ and $E[X]$ now include a new parameter $\gamma$ here. $T_0$ denotes the period the sender will wait for to retransmit non-acknowledged packet,

$$
\hat{Q}(w) = \min(1, \frac{(1 - (1 - p)^3)(1 + (1 - p)^3(1 - (1 - p)^w))}{1 - (1 - p)^w})
$$

and,

$$
f(p) = 1 + p + 2p^2 + 4p^3 + 8p^4 + 16p^5 + 32p^6
$$

For small values of $p$, (3.14) is reduced to:

$$
B(p) \approx \frac{1}{\text{RTT} \sqrt{\frac{2M(1-\gamma)p}{1+\gamma}} + T_0 \min \left(1, \sqrt{\frac{b(1-\gamma)p}{2}} \right) p(1+32p^2)}
$$

### 3.1.1.3 Veno Equation with Impact of Window Limitation

Finally, we consider the Veno equation with the impact of advertisement window limitation. In such scenario, the congestion window will start to increase after the previous loss, and will keep constant when it reaches the advertisement window limitation. Let $W_{max}$ denote the advertisement window limitation, $U_i$ denote the number of rounds where the congestion window increases, and $V_i$ denote the number of rounds where the congestion window keeps constant. Due to the new MD phase in Veno, $W_{max}$ is given by [11]:

$$
W_{max} = \lambda_i W_{max} + \frac{U_i}{6}
$$

where $\lambda_i$ is the proportion of $cwnd$ value after the previous TD loss ($\theta_1$ or $\theta_2$), as defined in Section 3.1.1.1. We assume $E[U] = \frac{bW_{max}}{2}$, same as that in [11].

Then considering the number of packets sent in the $i$th TD period, we have:

$$
Y_i = \frac{U_i}{2}(\lambda_i W_{max} + W_{max}) + V_i W_{max}
$$


and then,

\[
E[Y] = \left( \frac{1 + \gamma}{2} \right) W_{\text{max}} E[U] + W_{\text{max}} E[V]
\]

\[
= \frac{b(1 + \gamma)}{4} W_{\text{max}}^2 + W_{\text{max}} E[V]
\]

(3.18)

After getting \(E[Y]\), we can easily derive the final throughput equation when the advertisement window is limited, following the steps of [11]:

\[
B(p) = \frac{1-p}{p} + W_{\text{max}} + \bar{Q}(W_{\text{max}}) \frac{1-p}{1-p}
\]

\[
\text{RTT} \left( \frac{b(1+\gamma)}{4} W_{\text{max}} + \frac{1-p}{pW_{\text{max}} + 2} + \bar{Q}(W_{\text{max}})T_0 \frac{1-p}{1-p} \right)
\]

(3.19)

Combine equation (3.14) and (3.19) we can get the final approximated result for small values of \(p\):

\[
B(p) \approx \min \left( \frac{W_{\text{max}}}{\text{RTT}}, \frac{1}{\text{RTT}} \sqrt{\frac{2b(1-\gamma)p}{1+\gamma}} + T_0 \min \left( 1.3 \sqrt{\frac{b(1-\gamma)p}{2}} p(1+32p^2) \right) \right)
\]

(3.20)

3.1.1.4 Expression of \(\gamma\)

Now we need to determine the value of \(\gamma\). From our definition:

\[
\gamma = E[\lambda] = \theta_1 \times P(N < \beta) + \theta_2 \times P(N \geq \beta)
\]

\[
= \theta_1 \times P(N < \beta) + \theta_2 \times (1 - P(N < \beta))
\]

\[
= \theta_2 + (\theta_1 - \theta_2) \times P(N < \beta)
\]

(3.21)

According to [20]:

\[
N = \frac{W}{\text{RTT}} \times (\text{RTT} - \text{BaseRTT})
\]

(3.22)

where \(W\) is the current congestion window size, \(\text{BaseRTT}\) is the minimum of the measured round-trip time, and \(\text{RTT}\) is the smoothed measured round-trip time. For simplicity, we assume the value of congestion window at losses (including congestion and
random losses), \( W \), is uniformly distributed between 0 and \( W_{\text{max}} \), where \( W_{\text{max}} \) is the maximum congestion window size during the transmission. Then we have:

\[
P(N < \beta) = P(W < \beta \times \frac{\text{RTT}}{\text{RTT} - \text{BaseRTT}})
\]

\[
= \min \left( 1, \beta \times \frac{\text{RTT}}{(\text{RTT} - \text{BaseRTT}) \times W_{\text{max}}} \right)
\]

(3.23)

From (3.21) and (3.23) we have:

\[
\gamma = \min \left( \theta_1, \theta_2 + (\theta_1 - \theta_2) \times \frac{\beta \times \text{RTT}}{(\text{RTT} - \text{BaseRTT}) \times W_{\text{max}}} \right)
\]

(3.24)

Equation (3.20) with (3.24) is the derived Veno equation.

3.1.1.5 Discussion

The Veno equation has a variable \( \gamma \in [\theta_2, \theta_1] \) instead of the fixed factor \( \frac{1}{2} \) in Reno equation. Here we call \( \gamma \) the “Veno parameter”. To determine an accurate expression for \( \gamma \) is still an ongoing work. Here we assume \( W \) is uniformly distributed. The validation on NS-2 simulation [76][77] and the live Internet experiments show that our assumption can approximate the experimental results quite well. We also assume that the refined AI phase in Veno has negligible improvement in TCP throughput as compared to the refined MD phase, which has been explained at the beginning of Section 3.1.1.

Other assumptions, same as those in [11], have also been made during our derivation: 1) our model does not include the slow start and fast recovery algorithm of Veno. We believe the impacts of both phases are small, and can be ignored in the model; 2) we assume that packet losses within a round are correlated. This is caused by the drop-tail policy of intermediate routers: if one packet is dropped by the full buffer, it is likely the following packets in the round will be dropped too; 3) we assume losses in one round are independent of losses in other rounds, because packets in different rounds are separated by one RTT and more, and thus they are likely to encounter different buffer states; 4) we assume the round trip time is independent of the window size.
3.1.2 Validation of Veno Equation

The correctness of the Veno equation is verified under both NS-2 simulations and the live Internet experiments. These quantitative results show that our Veno equation can accurately model TCP Veno’s throughput.

3.1.2.1 Simulation Results

Figure 3.3 gives the topology of the NS-2 simulation experiment. The left side of the network has wired links with bandwidth of 10 Mbps, delay of 1 ms and buffer size of 50 packets. The right side of the network has wireless links with bandwidth of 10 Mbps, delay of 1 ms and buffer size of 50 packets. The bottleneck link between these two networks is a wired link and its bandwidth ($W_b$) and delay ($Delay_b$) will be varied in the experiments. Its buffer size is set to be 20 packets. TCP packets are transferred from the wired network to the wireless network. Packet size is 1000 bytes. All the simulation will run for 300 s and will be repeated 20 times. We plot the average of these results in the following figures and use the error bars to represent 95% confidence intervals.

![Figure 3.3: Topology of NS-2 simulation.](image)

In order to fully verify whether the Veno equation is able to accurately predict TCP Veno throughput or not, our simulation uses two different random loss models – exponential distribution and two-state distribution – in the wireless links. In the exponential
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Throughput (packet/s) vs. Packet Loss Rate for exponential distribution case:
- $\gamma = \frac{4}{5}$
- $\gamma = \frac{1}{2}$

Equations and Experiments for various delay times:
- Equation(40ms)
- Experiment(40ms)
- Equation(80ms)
- Experiment(80ms)
- Equation(160ms)
- Experiment(160ms)
- Equation(320ms)
- Experiment(320ms)

3.4.a: $W_b = 2Mbps$, $Delay_b = 80ms$.
3.4.b: $W_b = 2Mbps$, different delays.

Figure 3.4: Simulation results under exponential random loss: single flow.

distribution, packet loss rate ranges from 0.0001 to 0.1. In the two-state distribution, the good state has packet loss rate of 0.00001 and period of 8 s, and the bad state has packet loss rate ranging from 0.0001 to 0.1 and period of 4 s. Furthermore, the good state has 95% probability to transit to the bad state after every period, so does the bad state.

At first one TCP Veno connection is established. Figure 3.4.a shows the validation results under exponential random losses. Here the bandwidth of the bottleneck link is set to 2 Mbps and the delay is 80 ms. We also plot the curves when $\gamma = \theta_1 = \frac{4}{5}$ and $\gamma = \theta_2 = \frac{1}{2}$, which are the maximum and minimum boundaries of our equation respectively. Note that in this and the following experiments, we collect the dynamic values of $RTT$ and $(RTT - BaseRTT)$ at every TD or TO, to calculate Veno parameter $\gamma_i$ using equation (3.24). Finally we use the average of $\gamma_i$ to calculate the result. As shown in Figure 3.4.a, most experimental results fall between the maximum and minimum boundaries, and they can be accurately predicted by the Veno equation. Figure 3.4.b plots the results under different bottleneck link delays.

Figure 3.5 with the two-state random loss model demonstrate that the theoretical results are quite consistent with the simulation results.
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3.5.a: \( W_b = 2 \text{Mbps}, \text{Delay}_b = 80 \text{ms} \).

3.5.b: \( W_b = 2 \text{Mbps}, \) different delays.

Figure 3.5: Simulation results under two-state random loss: single flow.

3.6.a: UDP background traffic.

3.6.b: TCP Sack background traffic.

Figure 3.6: Simulation results under exponential random loss: background traffic.

Next, we introduce multiple UDP or TCP SACK connections as background traffic into our test. When UDP flows are added, there is only one TCP Veno connection. The bandwidth of the bottleneck link is set to 2 Mbps and the delay is 80 ms. UDP flows follow pareto distributions [78]. The packet size of UDP is 512 bytes. The burst time and idle time are both 100 ms. The rate of each connection varies from 250 Kbps to 1 Mbps. When TCP Sack flows are added, they range from 2 to 16, and there are equal numbers of TCP Veno flows running simultaneously. The bandwidth of the bottleneck
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3.7.a: UDP background traffic.

3.7.b: TCP Sack background traffic.

Figure 3.7: Simulation results under two-state random loss: background traffic.

link is set to 12 Mbps and the delay is 80 ms.

Carefully observing Figure 3.6 (with exponential random losses model introduced), and Figure 3.7 (with two-state random losses model introduced), we conclude that the theoretical analysis indeed match well with the simulation results, regardless of the types of losses.

Figure 3.8: More complex topology of NS-2 simulation.

To compare the theoretical analysis and simulation results in a more realistic scenario, a more complicated simulation topology is used here to validation the Veno equation.

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Figure 3.8 illustrates the topology, where eight TCP Veno flows (Flow 1 ~ 8) run from the left side network to the right side network through the bottleneck link. The left side network has wired links with bandwidth of 10 Mbps, delay of 1 ms. The right side network is a heterogeneous network with wired and wireless links, whose bandwidths are all 10 Mbps. The delays of the wired links are \((20 \times i - 19)\) ms, \(i = 1, 2, ..., 8\), and the delays of the wireless links are all 1 ms. Random losses in wireless links follow exponential distribution and the packet loss rate ranges from 0.0001 to 0.1. The bottleneck link between these two networks is a wired link with bandwidth of 12 Mbps and delay of 10 ms. Its packet buffer size is set to be 40. Packet sizes of TCP Veno flows are 1000 bytes. We also set up 8 UDP connections as background traffic in our test. These UDP flows follow pareto distributions. The burst time and idle time are both 100 ms. The rate of each connection is 300 Kbps and the packet size is 512 bytes.

Figure 3.9.a gives the validation result on one of the TCP Veno flows – Flow 5. The theoretical results are quite consistent with the experimental results. Figure 3.9.b plots the validation results of Flow 2, 4, 6, 8. Obviously, all these experimental results can be accurately predicted by the Veno equation.

3.9.a: A single flow. 3.9.b: Multiple flows.
3.1.2.2 The Live Internet Experiments

In the live Internet experiments, we set up a wireless TCP Veno client at Nanyang Technological University (NTU) in Singapore, and a TCP Veno server at The University of Hong Kong (HKU) in Hong Kong. Thus the client and the server is connected via cross-country WAN, as shown in Figure 3.10.

![Figure 3.10: Topology of the Internet experiments.](image)

One TCP Veno flow runs over the link from the server to the client, and the packet size is 512 bytes. In the wireless side, the client is a Compaq laptop with W200D wireless card. The wireless Access Point is Cisco Linksys WRT54G. They are placed in the different corners of a room and the distance between them is about 8 meters. The bandwidth of the wireless network is 11 Mbps and the frequency is 2.4 GHz. TCP Veno is implemented in FreeBSD 4.3. The measurement method in the Internet experiments is same as that described in [11]. The 1hr flow trace is divided into 36 consecutive 100-second intervals, and we count the number of packets transmitted, $RTT$, $BaseRTT$, and the packet loss rate $p$ in every interval. Note that $p$ is given as the total number of loss indications (TD and TO) divided by the number of packets transmitted. Then, we put the average values of $RTT$ and $BaseRTT$ into the Veno equation to calculate the theoretical result.
Due to the space constraints, we only give four samples collected recently, as shown in Figure 3.11. The results demonstrate that the Veno equation can model TCP Veno’s throughput fairly well.

### 3.1.3 Summary

In this work, we provide a model for TCP Veno algorithm, to characterize its dynamic reductions of congestion window over wireless networks. A simple throughput equation called Veno equation is derived. We introduce a Veno parameter $\gamma$ into previous work on Reno [11]. Such parameter accurately characterizes the refined MD phase of Veno. Our
extensive experiments on both simulation and the Internet have validated the correctness of the equation. In the future work, further efforts are needed to refine the simple assumption about the uniformly distributed $W$ in the deduction, and a more accurate expression of the Veno parameter $\gamma$ will be studied.

3.2 TCP Veno Enhancement over Light-load Wireless Networks

This work observes that TCP Veno behaves conservatively over light-load wireless networks. A new variable, congestion loss rate, is introduced into Veno’s algorithm. It helps Veno deal with random loss more intelligently, by keeping its congestion window increasing if the link load is in light state. The simulation and real network results demonstrate that, such enhancement can improve Veno’s throughput up to 60% without any fairness or friendliness sacrificed.

3.2.1 TCP Veno With Congestion Loss Rate

Many experiments have proved Veno’s better performance over Reno in wireless environments [20][29][79][80]. In this section, however, we present results showing that Veno still misses much available bandwidth, especially when the network load is light and random losses are pervasive. Further analysis points out that such conservation is due to Veno’s “blind” reduction of congestion window when random loss occurs. A new variable, named congestion loss rate, is introduced in our proposal. It helps Veno act more appropriately in response to random losses and grab more available bandwidth on the link, without any fairness or TCP-friendliness sacrificed.

3.2.1.1 Conservation of TCP Veno over Light-load Wireless Networks

Figure 3.12 gives a NS-2 simulation topology. The left side network has wired links with bandwidth of 10 Mbps, delay of 1 ms and buffer of 50 packets. The right side network
has wireless links with bandwidth of 10 Mbps, delay of 1 ms and buffer of 50 packets. Random loss in wireless links follows exponential distribution. The bottleneck link’s bandwidth and delay will change in our experiments. Its buffer is set to be 20 packets. TCP packets are transferred from the wired network to the wireless network. Packet size is 1000 bytes.

![Figure 3.12: Topology in NS-2 experiments.](image)

To simulate the light load of the bottleneck link, we let only one TCP Veno connection run over it. The delay of the bottleneck link is set to be 80 ms, and the bandwidth is 2 Mbps and 8 Mbps respectively.

Table 3.1 presents Veno’s throughputs over 2 Mbps and 8 Mbps bottleneck link respectively, under different packet loss rates (Pr). It shows that when the packet loss rate is larger than 0.005, Veno running over 8 Mbps bottleneck link has almost the same throughput as that over 2 Mbps one. In other words, additional 6 Mbps bottleneck link is not used by Veno even when there are no other competing connections.

<table>
<thead>
<tr>
<th>Pr=0.001</th>
<th>Pr=0.005</th>
<th>Pr=0.01</th>
<th>Pr=0.05</th>
<th>Pr=0.1</th>
</tr>
</thead>
<tbody>
<tr>
<td>Over 2Mbps (packets/s)</td>
<td>192.46</td>
<td>142.65</td>
<td>100.91</td>
<td>27.83</td>
</tr>
<tr>
<td>Over 8Mbps (packets/s)</td>
<td>320.12</td>
<td>153.19</td>
<td>104.95</td>
<td>28.45</td>
</tr>
</tbody>
</table>

The reason why Veno behaves so conservatively in bandwidth utilization can be easily found in its congestion window evolution, as shown in Figure 3.13 (the dotted line). In
light-load wireless networks, congestion losses are rare and random losses are dominating. These random losses make Veno frequently cut down its congestion window by $\frac{1}{5}$, and keep $cwnd$ value at a low level. In other words, such frequent random losses prevent Veno touching the equilibrium point during transmission. As a result, Veno wastes much available bandwidth during transmission.

![Figure 3.13: Window evolution of Veno-LL and Veno, packet loss rate = 0.01, bottleneck bandwidth = 8Mbps.](image)

According to the above analysis, a “blind” reduction of congestion window for random loss regardless of its congestion context is harmful to the improvement of Veno’s throughput. A more “intelligent” algorithm is needed to deal with these random losses, rather than always cutting down $\frac{1}{5}$ congestion window.

### 3.2.1.2 Congestion Loss Rate

We keep Veno’s method to estimate the backlog ($N$) at the link and deduce a packet loss is congestion loss ($N \geq \beta$) or random loss ($N < \beta$). However, a new variable, called
congestion loss rate, is introduced to help Veno adjust its congestion window in smarter way. Note that congestion loss rate is calculated as follows:

Considering a sequence of random losses during transmission \( \{T_i\} \), where \( T_i \) is the time at which random loss occurs, we count the number of congestion losses occurring between two consecutive random losses \( T_{i-1} \) and \( T_i \), called \( C_i \), as shown in Figure 3.14. Then congestion loss rate at this moment \( \text{con} \cdot r_i \) can be calculated as:

\[
\text{con} \cdot r_i = \frac{C_i}{T_i - T_{i-1}}
\] (3.25)

If \( \text{con} \cdot r_i > \text{con} \cdot r_{i-1} \), which means the rate of congestion loss occurrence increases since the last time, then we assume the network state is becoming congestive, and cut down \( \frac{1}{5} \) of congestion window at random loss \( T_i \). Otherwise if \( \text{con} \cdot r_i \leq \text{con} \cdot r_{i-1} \), which means the congestion of networks is not becoming worse, then we keep the value of congestion window unchanged at \( T_i \).

In summary, our enhancement of Veno for light-load networks (hereinafter, called “Veno-LL”) acts as follows:

\[
cwnd = \begin{cases} 
\text{cwnd}, & N < \beta \text{ and } \text{con} \cdot r_i \leq \text{con} \cdot r_{i-1} \\
\text{cwnd} \times \frac{4}{5}, & N < \beta \text{ and } \text{con} \cdot r_i > \text{con} \cdot r_{i-1} \\
\text{cwnd} \times \frac{1}{2}, & N \geq \beta 
\end{cases}
\]

Figure 3.14: Calculation of congestion loss rate.
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As shown in Figure 3.25 (the solid line), over a light-load network $con.r$ is always 0 during transmission, except of occasional pulses. Thus $cwnd$ value will keep increasing at most random losses and help Veno-LL absorb as much available bandwidth as possible.

It is worthy to emphasize that, as the network becomes more and more congestive, and congestion becomes the main reason of packet losses, our proposal turns to behave similarly as original Veno does. This can be observed in the performance evaluation part, where more detailed experiments are conducted.

3.2.2 Performance Evaluation

In this section, we evaluate Veno-LL’s performance in throughput, fairness, and TCP-friendliness, as compared to Veno. The evaluation is based on both NS-2 simulations and real network experiments.

3.2.2.1 Simulation Results

The topology and settings are same as those in Figure 3.12. The bandwidth of bottleneck link is set to be 12 Mbps and the delay is 80 ms. Random loss rate in wireless links changes from 0.0001 to 0.1 in packet unit.

- Throughput

In each experiment, we at first set up $N$ Veno connections running over the network for 300s, and calculate their average throughput. Here the throughput is counted as the number of packets sent divided by the transmission time. Then we set up the same number of Veno-LL connections running over the network for the same duration, and also calculate their average throughput. In order to better compare the difference of their throughput, we define the normalized throughput $TH_n$ as following:

$$TH_n = \frac{TH_{Enhanced}}{TH_{Veno}}$$

(3.26)
where, $TH_{Enhanced}$ and $TH_{Veno}$ are the average throughput of Veno-LL flows and Veno flows respectively. This metric illustrates the percentage of improvement Veno-LL can obtain as compared to Veno. The higher $TH_n$ is, the greater the improvement is.

The number of connections $N$ varies from 1 to 64 in the experiments, to make the bottleneck link more and more congestive.

![Normalized Throughput](image)

Figure 3.15: Throughput improvement of Veno-LL under different number of connections ($N$) and different random loss rates ($Pr$).

As shown in Figure 3.15, our proposal can effectively improve Veno’s throughput when connection number is small (the improvement can be up to 60% when random loss is around 0.01). However, as the connection number increases, which means the network load increases, our proposal performs more and more similarly as Veno ($TH_n$ is around 1) except under very heavy random losses. In other words, our proposal is a feasible enhancement of Veno over light-load wireless networks.

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• Fairness

Fairness means the same kind of flows should share the total bandwidth fairly. To reflect the fairness more clearly, we use the Jain’s Fairness Index $f$ which is define in [81]:

$$
f = \frac{\left( \sum_{i=1}^{n} x_i \right)^2}{n \sum_{i=1}^{n} x_i^2}$$  \hspace{1cm} (3.27)

where, $n$ is the number of connections, $x_i$ is the throughput of the $i$th connection. The closer $f$ is to 1, the more fairness that kink of flows has.

Figure 3.16: Fairness comparison between Veno-LL and Veno.

In each experiment, we at first set up $N$ Veno connections running over the network for 300s, calculate the throughput of each connection, and get their Fairness Index using equation 3.27. Then we set up the same number of Veno-LL connections running over the network for the same duration, and also calculate their Fairness Index. The number of connections $N$ varies from 1 to 64 in the experiments.
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Figure 3.16 plots the Fairness Index of Veno-LL and Veno respectively. The values of Veno-LL and Veno in the figure are almost same in each scenario, which proves that our enhancement does not harm original Veno’s fairness.

- **TCP-friendliness**

Veno has shown its compatibility to the dominating version of TCP today [20]. Here we study whether our proposal has the same TCP compatibility as Veno. We first set up certain numbers (ranging from 2 to 64) of TCP Sack flows over the link, and calculate their average throughput $T_1$. Then we replace half of them with the objective flows (Veno-LL or Veno) and recalculate the average throughput of the left TCP Sack flows $T_2$. We define the normalized throughput $F$ as follows:

$$F = \frac{T_2}{T_1} \quad (3.28)$$

![Figure 3.17: Friendliness comparison between Veno-LL and Veno.](image)
If $F$ is 1, it means Sack flows are not affected by the objective flows, and thus the objective flows are totally friendly. The closer $F$ is to 1, the more friendliness the objective flows have.

As illustrated in Figure 3.17, the value $F$ of Veno-LL is always around that of Veno in different scenarios, which means our proposal does not sacrifice Veno’s TCP-friendliness when improving its throughput.

### 3.2.2.2 Real Network Experiments

We also conduct the experiment in a wireless local area network (WLAN). The topology of the WLAN is shown in Figure 3.18. The TCP senders use FreeBSD 4.3, in which Veno and Veno-LL are implemented. The TCP receivers are normal linux machines with kernel 2.6.12. They are connected by a Gateway with FreeBSD 4.9. Dummynet [82] is used in this Gateway to shape the bandwidth and set the random loss rate.

![Topology of WLAN experiments.](image)

The bandwidth of the bottleneck link is set to be 5 Mbps, and the delay is 60 ms. Furthermore, the random loss rate is set to be 0.005. Two cases are studied here based on different traffic load: light load case where there is no other background traffic, and heavy load case where there is an co-exist 3 Mbps UDP flow during transmission. In each case, different types of flows are tested for 300 s respectively: 1) four Veno connections (two in each sender PC); 2) two Veno and two Veno-LL connections (one Veno and one Veno-LL in each sender PC); 3) four Veno-LL connections (two in each sender PC). Figures 3.19 and 3.20 plot their sequence numbers against the time in each case.
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Figure 3.19: Experiments over WLAN: light load case.

3.20.a: Four Veno connections. 3.20.b: Two Veno and two Veno-LL connections. 3.20.c: Four Veno-LL connections.

Figure 3.20: Experiments over WLAN: heavy load case.

From these figures, three important observations can be made. The first observation is about the fairness. Either in Figure 3.19.c or 3.20.c, the increasing rate in sequence number of each Enhance Veno flow is similar. This proves Veno-LL can share the network source fairly. The second observation is, Veno-LL has a similar increasing rate as Veno does when the network is in heavy-load state (Figure 3.20.b), but has a much higher increasing rate when the network is in light-load state (Figure 3.19.b). This proves Veno-LL has a better throughput than Veno over light-load wireless networks. The third observation is about the TCP-friendliness. Although Veno-LL has a higher increasing rate in the light-load networks, it does not affect the increasing rate of Veno running with them. This can be observed by comparing between Figure 3.19.a and 3.19.b. This
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proves Veno-LL does not sacrifice any TCP-friendliness.

3.2.3 Summary

In this work, we introduce a new variable called congestion loss rate into TCP Veno. Congestion loss rate can indicate the trend of congestion occurrence on the link and avoid “blind” reduction of congestion window at random losses, especially when the network load is light. Simulation and real network results demonstrate such enhanced TCP Veno can obtain significant throughput improvement over TCP Veno without any fairness or TCP-friendliness sacrificed.

3.3 TCP Veno Enhancement over High Bandwidth-Delay Product Networks

This work focuses on another situation TCP Veno may encounter during transmission – high Bandwidth-Delay Product (BDP) network. We find that TCP Veno suffers severe throughput degradation under random losses in a high BDP network. To alleviate such suffering, a new variable named $pBDP$ is proposed to indicate the BDP state of the network. It helps Veno deal with random losses more intelligently, by keeping its congestion window increasing if the link is in the high BDP state. The experiments based on the simulation and local area network demonstrate that, such enhancement can improve Veno’s throughput up to 80% without any fairness or friendliness sacrificed.

3.3.1 TCP Veno with $pBDP$

The BDP is loosely defined as the product of the round-trip delay for a data connection and the capacity of the bottleneck link in its path. Note that here we use the term “high BDP” as that in [13], i.e., the BDP is larger than the buffer size on the bottleneck link. This definition covers very large scale in different types of networks, besides of the high
speed networks. In a high BDP network, our study shows that TCP Veno suffers severe throughput degradation under random losses. Our further analysis points out that, such throughput degradation in Veno is due to its “blind” reduction of congestion window when random losses occur, without concerning the specific state of the link at that time. Based on the observation in [13], a new variable named $p_{BDP}$ is introduced into Veno. It indicates the BDP state of the link and helps Veno act more appropriately in response to random losses.

It is worthy to emphasize that, this work only focuses on alleviating the random loss impact on Veno over a high BDP network. Other applicable enhancements for high BDP networks, such as a faster additive-increase (AI) and a less multiplicative-decrease (MD) during the congestion window evolution [83][84], are out of the scope. However, our scheme can seamlessly cooperate with these enhancements.

### 3.3.1.1 Random Loss Impacts on TCP Veno over High BDP Networks

Figure 3.21 gives the topology of the NS-2 simulations used in this section. The left side network has wired links with bandwidth of 100 Mbps and delay of 1 ms. The right side network has wireless links with bandwidth of 100 Mbps and delay of 1 ms. Random loss in wireless links follows exponential distribution. The bottleneck link’s bandwidth and delay are varied in our experiments. Its buffer is set to be 50 packets. TCP packets are transferred from the wired network to the wireless network. Packet size is 1000 bytes. The duration time of each experiment is 300 s.

One TCP Veno connection is established in the experiment and its throughput is studied in three cases: at first we set the bottleneck bandwidth to be 2 Mbps, and its delay to be 50 ms; then we extend the delay to 300 ms; finally we set the bandwidth to be 12 Mbps and the delay to be 50 ms. The BDP of the bottleneck link is 25, 150, and 150 respectively. As compared to the buffer size (50), the latter two cases can be
regarded as the high BDP links. Note that, flow control is disabled in our experiment, thus the congestion window can increase as large as possible.

Figure 3.22 presents the results of Veno’s throughputs in these three cases under different random loss rates from 0.0001 to 0.1. As observed in the figure, when the random loss rate increases, Veno’s throughput decreases much faster in the two high BDP cases. For example, when the random loss rate increases from 0.0001 to 0.001, the decrease degree of Veno’s throughput is only 3% in the first case (BDP = 25), but is 52% and 47% in the latter two cases (BDP = 150) respectively.

The reason why Veno encounters such severe throughput degradation can be easily found in its congestion window evolution. Figure 3.23 plots Veno’s congestion window evolution over 2 Mbps, 50 ms link and 2 Mbps, 300 ms link respectively. According to [20], Veno cuts down $\frac{1}{5}$ of the window size in response to random losses. When the random loss rate is 0.0001 where the random losses are rare, the congestion window can always touch the equilibrium point and cause congestion losses. However, when the random loss rate is 0.001, many random losses prevent this happening and keep the congestion window size in a lower level than that under random loss rate of 0.0001. This makes Veno miss available bandwidth of the link. As shown in Figure 3.23.b, the gap between
the average window sizes under different loss rates becomes much larger when BDP is high. In other words, although Veno cuts down less window size ($\frac{1}{2}$) under random losses than Reno does, such action is still conservative in high BDP networks.

Figure 3.22: Random loss impacts on Veno in three BDP cases.

Figure 3.23: Congestion window evolutions of TCP Veno.

3.23.a: Over 2Mbps, 50ms bottleneck link. 3.23.b: Over 2Mbps, 300ms bottleneck link.
Therefore, we propose a new variable to indicate Veno about the BDP state of the network. It can help Veno deal with random losses more intelligently, rather than always cutting down $\frac{1}{5}$ of the congestion window.

### 3.3.1.2 \( pBDP \)

The random loss impact on TCP Reno over high BDP networks has been studied by [13] in both analytical and experimental way. It observes that, random losses lead to significant throughput deterioration when \( p(\mu T) \) is large. Here \( p \) is the random loss probability, \( \mu \) (packet/s) is the capacity of the bottleneck link, and \( T = (\tau + \frac{1}{\mu}) \) is the sum of propagation delay (\( \tau \)) and service time (\( \frac{1}{\mu} \)). Typically \( \tau \gg \frac{1}{\mu} \), so \( T \approx \tau \). Following this work, we can use the variable \( p(\mu T) \) to indicate the BDP state of the link. If it is larger than certain threshold, we consider the link a high BDP link and do not cut down the congestion window size when a random loss occurs. Otherwise, we cut down the congestion window by $\frac{1}{5}$ as the original Veno does. Note that here “BDP state” means the cooperative impacts from both BDP and random losses. For example, if the random losses is heavy, Veno’s throughput will go down severely even the BDP is low.

In a real network, however, the TCP sender is unable to exactly know the values of \( p, \mu, \) and \( \tau \). An alternative method is proposed here to approximate the value of \( p(\mu T) \).

The random loss rate \( p \) can be estimated by using a similar method as that in TFRC [9], as described in Section 2.2.1.2. Veno considers a packet loss is random loss when \( N < \beta \). Then we define the term random loss event as that in TFRC. With reference to Figure 2.5, let \( A_i \) be the \( i \)th recent random loss interval, and let \( w_i \) be the weight of \( A_i \), then the receiver calculates the average random loss interval of the recent \( n \) loss intervals as follows:

\[
A = \frac{\sum_{i=1}^{n} w_i A_i}{\sum_{i=1}^{n} w_i} \quad (3.29)
\]
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The default value of \( n \) is 8, and \( w_1 = w_2 = w_3 = w_4 = 1, w_5 = 0.8, w_6 = 0.6, w_7 = 0.4, w_8 = 0.2 \). Finally, the estimated random loss event rate, \( \bar{p} \), is given by

\[
\bar{p} = \frac{1}{A}
\]  

(3.30)

The value \( \mu \tau \) can be estimated as follows. Consider a network with a constant propagation delay \( \tau \) and a buffer size \( B \). If we ignore the processing time at the routers and the end hosts, the round-trip time (RTT) of a TCP flow is roughly the sum of the propagation delay and the queuing delay, i.e.,

\[
RTT \approx \tau + \frac{b}{\mu}
\]

(3.31)

where \( b \) is the number of packet cumulating in the bottleneck link buffer, and \( 0 \leq b \leq B \).

Thus we have

\[
RTT_{\text{min}} \approx \tau
\]

(3.32)

and,

\[
RTT_{\text{max}} \approx \tau + \frac{B}{\mu}
\]

(3.33)

On the other hand, according to [13], the maximal size of a TCP congestion window is

\[
W_{\text{max}} = \mu \tau + B + 1 \approx \mu \tau + B = \mu (\tau + \frac{B}{\mu}) = \mu RTT_{\text{max}}
\]

(3.34)

From (3.32) and (3.34), we have

\[
\mu \tau \approx \frac{W_{\text{max}} RTT_{\text{min}}}{RTT_{\text{max}}}
\]

(3.35)

where \( W_{\text{max}}, RTT_{\text{min}}, \) and \( RTT_{\text{max}} \) can be easily monitored in the TCP sender.

Figure 3.24 gives the validation of equation (3.35) under a simulation over 12 Mbps, 50 ms networks. Note that, the value of \( W_{\text{max}} \) will be reset to 0 after each time-out occurs. This makes our scheme adaptable to the dynamic of the network. As observed
in the figure, the large random loss rate (0.01) will cause $\mu\tau$ to be underestimated. This is because the frequent random losses prevent the congestion window touching its maximal size, thus make the monitored $W_{max}$ always lower than it should be. Although the underestimated value of $\mu\tau$ may make our scheme misinterpret some high BDP links to be normal links and obtain no improvement in these cases, it also avoids the trend of the scheme to be aggressive in most cases.

From (3.30) and (3.35), the variable indicating the BDP status of the network, named $pBDP$, is obtained by:

$$pBDP = \bar{p}\left(\frac{W_{max}RTT_{min}}{RTT_{max}}\right)^2$$  \hspace{1cm} (3.36)

In summary, our enhancement of Veno for high BDP networks (hereinafter, called “Veno-HB”) acts as follows:

$$cwnd = \begin{cases} 
  cwnd, & N < \beta \text{ and } pBDP \geq \alpha \\
  cwnd \times \frac{4}{5}, & N < \beta \text{ and } pBDP < \alpha \\
  cwnd \times \frac{1}{2}, & N \geq \beta 
\end{cases}$$

According to [20] and [13], the default values of $\beta$ and $\alpha$ are set to 3 and 10 respectively.

---

Figure 3.24: Estimation of $\mu\tau$ under different random loss rates.
3.3.2 Performance Evaluation

We compare our Veno-HB with the original Veno mechanism based on the NS-2 simulations and the real network experiments.

3.3.2.1 Simulation Results

The experiment environments are same as those in Figure 3.21. Throughput, fairness, and friendliness of Veno-HB are evaluated here, as compared to those of Veno.

- **Throughput**

  At first only a single connection is set up. Figure 3.25 presents the congestion window evolutions of Veno and Veno-HB over a bottleneck link of 2 Mbps, 300 ms. The random loss rate is 0.001. As shown in the figure, Veno-HB avoids many unnecessary window cuts under random losses, and keeps its congestion window in a high level near the BDP.

  More complete study is shown in Figure 3.26. It plots the throughput enhancements of our proposal under different BDPs and different random loss rates from 0.0001 to 0.1. The normalized throughput ($TH_n$) is defined as equation (3.26). As shown in the figure, Veno-HB has a good improvement (up to 80%) in throughput, especially over high BDP networks.

  Next, we increase the number of connections of Veno and Veno-HB from 1 to 64, and investigate the throughput enhancement under multiple connections scenarios. The bottleneck link here is set to be 12 Mbps and 50 ms. As shown in Figure 3.27, our proposal can effectively improve Veno’s throughput when connection number is small. As the connection number increases, network resources are shared by multiple flows and now the network is no longer a high BDP network. Thus Veno-HB performs more and more similarly as Veno does ($TH_n$ is around 1) except under very heavy random losses.
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Figure 3.25: Congestion window evolutions of Veno-HB and Veno over 2Mbps, 300ms bottleneck link.

Figure 3.26: Throughput improvement of Veno-HB under different BDP cases and different random loss rates (Pr).
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![Normalized Throughput](image)

Figure 3.27: Throughput improvement of Veno-HB under multiple connections (N) and different random loss rates (Pr).

- **Fairness**

  In the experiments, multiple Veno-HB and Veno connections (ranging from 1 to 64) are established respectively. The bottleneck link here is set to be 12 Mbps and 50 ms. To reflect the fairness of Veno-HB and Veno, we also use Jain’s Fairness Index $f$ which is define as equation (3.27). Figure 4.16 plots the Fairness Index of Veno-HB and Veno respectively. The values $f$ of Veno-HB and Veno in the figure are almost same in each scenarios, which proves that our enhancement does not harm original Veno’s fairness.

- **TCP-friendliness**

  We first set up certain numbers (ranging from 2 to 64) of TCP Sack flows over the link, and calculate their average throughput $T_1$. Then we replace half of them with the objective flows (Veno-HB or Veno) and recalculate the average throughput of
Figure 3.28: Fairness comparison between Veno-HB and Veno.

the remaining TCP Sack flows $T_2$. We also define the normalized throughput $F$ as equation (3.28), to reflect the TCP-friendliness of Veno-HB and Veno. Note that the bottleneck link here is set to be 12 Mbps and 50 ms.

As observed in Figure 4.17, the value $F$ of Veno-HB is always around that of Veno in different scenarios, which means our proposal does not sacrifice Veno’s friendliness when improving its throughput.

3.3.2.2 Real Network Experiments

We also implement this Enchanced Veno into FreeBSD 4.3, and run the experiment in a wireless LAN. The topology of the wireless LAN is same as that in Figure 3.18. The TCP senders use FreeBSD 4.3, in which Veno and Veno-HB are implemented. The TCP receivers are normal linux machines with kernel 2.6.12. They are connected by a Gateway with FreeBSD 4.9. Dummynet is used in this Gateway to shape the bandwidth and set the random loss rate.
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![Normalized Throughput](image)

Figure 3.29: Friendliness comparison between Veno-HB and Veno.

The random loss rate of the bottleneck is set to be 0.005 and the buffer size is set to be 50 packets. Three cases are studied here based on different pairs of bandwidth and delay: (2 Mbps, 40 ms), (2 Mbps, 240 ms), and (12 Mbps, 40 ms). In each case, different types of flows are tested for 300 s respectively: 1) four Veno connections (two in each sender PC); 2) two Veno and two Veno-HB connections (one Veno and one Veno-HB in each sender PC); 3) four Veno-HB connections (two in each sender PC).

From these figures, three important observations can be made. The first observation is about the fairness. Either in Figure 3.30.c, 3.31.c, or 3.32.c, the increasing rate in sequence number of each Enhance Veno flow is similar. This proves Veno-HB can share the network source fairly. The second observation is, Veno-HB has a similar increasing rate as Veno does when the network is not a high BDP network (Figure 3.30.b), but has a much higher increasing rate when the network is a high BDP network (Figure 3.31.b and 3.32.b). This proves Veno-HB has a better throughput than Veno over high BDP networks. The third observation is about the TCP-friendliness. Although Veno-HB has
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Figure 3.30: Sequence numbers against time over a 2Mbps, 40ms link.

3.31.a: Four Veno connections. 3.31.b: Two Veno and two Veno-HB connections. 3.31.c: Four Veno-HB connections.

Figure 3.31: Sequence numbers against time over a 2Mbps, 240ms link.

3.32.a: Four Veno connections. 3.32.b: Two Veno and two Veno-HB connections. 3.32.c: Four Veno-HB connections.

Figure 3.32: Sequence numbers against time over a 12Mbps, 40ms link.
a higher increasing rate in the high BDP networks, it does not affect the increasing rate of Veno running with them. This can be observed by comparing between Figure 3.31.a and 3.31.b, or between Figure 3.32.a and 3.32.b. This proves Veno-HB does not sacrifice any TCP-friendliness.

### 3.3.3 Summary

This work introduces a new variable called $pBDP$ into TCP Veno. It indicates the BDP state of the bottleneck link and avoids Veno’s “blind” reduction of congestion window at random losses. The simulations and local area network experiments demonstrate that, such enhanced TCP Veno can obtain significant throughput improvement over TCP Veno without any fairness or TCP-friendliness sacrificed. Furthermore, $pBDP$ can seamlessly cooperate with further enhancement of Veno over high BDP networks, such as a faster additive-increase (AI) and a less multiplicative-decrease (MD) during the congestion window evolution. A complete optimization of Veno over high BDP networks is under the future investigation.
Chapter 4

Analysis and Enhancements of End-to-end Congestion Controls for Streaming over Wireless Networks

In this chapter, we describe our work on the end-to-end congestion controls for streaming. We propose two enhancements of TFRC to improve its performance over wireless networks. One enhancement is equipped with the Veno equation rather than the Reno equation to adjust its sending rate, and another one uses Veno’s state differentiator to alleviate the impact of random losses on the sending rate. Simulations and real network experiments results for these two proposals are also shown in this chapter.

4.1 Veno Equation Based TFRC Enhancement

As described in Section 1.2.2, TFRC inherits TCP throughput degradation over wireless networks due to its embedded Reno equation. In this thesis, a more advanced equation, called Veno equation, has been derived in Section 3.1.1. Note that the Veno equation is derived from the wireless TCP (TCP Veno) rather than the wired TCP (TCP Reno), thus it can model TCP behavior more appropriately over wireless networks. In this work, we choose to make use of the Veno equation to enhance TFRC over wireless networks. The enhancement is referred as TFRC Veno. Our extensive experiments including simulations
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and the Internet measurements demonstrate that, TFRC Veno obtains two achievements: 1) it has a significant throughput improvement – in wireless networks with 10% loss rate, it can obtain 300% improvement over the original TFRC; 2) it inherits the desirable features of TFRC, namely, good fairness, nice TCP-friendliness and smoothness of sending rate. Moreover, TFRC Veno only needs to modify the sender-side protocol of TFRC while the receiver-side or intermediate node protocol stack remains intact.

4.1.1 TFRC Veno

Different from all the previous efforts that keep the Reno equation in TFRC intact, TFRC Veno tries to use the Veno equation to replace the Reno equation. The basic mechanism of TFRC has been described in Section 2.2.1. TFRC Veno modifies TFRC by simply replacing the Reno equation with the Veno equation, that is, the sender now adjusts its sending rate based on the following equation:

\[ T_{calc} = \frac{S}{RTT \sqrt{\frac{2b(1-\gamma)p}{1+\gamma}} + 3RTO \sqrt{\frac{b(1-\gamma^2)p^2}{2}p(1+32p^2)}} \]  (4.1)

where \( T_{calc} \) is the expected sending rate in bytes/sec; \( S \) is the packet size in bytes; \( RTT \) is the round-trip time; \( p \) is the steady-state rate of loss event; \( RTO \) is the retransmit timeout value.

Note that here we ignore the effect of the maximum congestion window size in the Veno equation (3.20). Furthermore, the original Veno equation calculates throughput in packets/sec, while here we multiply it by the packet size \( (S) \) to calculate throughput in bytes/sec. These two changes are the same as those of the Reno equation when used in TFRC. The other mechanisms in TFRC, such as slow start, packet loss estimation, and time-out mechanism, are not changed in TFRC Veno. Here \( \gamma \) is in the range \([\theta_1, \theta_2]\), with the settings of \( \theta_1 = 0.95 \) and \( \theta_2 = 0.5 \).
4.1.2 Performance Evaluation

We conduct various experiments including NS-2 simulation and live Internet experiments, in order to comprehensively evaluate the performance of TFRC Veno in terms of throughput, fairness, TCP-friendliness, and sending rate variation. We also compare our proposal with the original TFRC and another sender-modified enhancement MULTFRC [35]. The results in the experiments show that, TFRC Veno combines both advantages of TFRC and MULTFRC while avoiding their respective drawbacks, that is to say, TFRC Veno has much better throughput than TFRC while keeping the nice features of fairness, TCP-friendliness, and sending rate variation; TFRC Veno has much better fairness and TCP-friendliness than MULTFRC while having the same throughput as MULTFRC can reach.

4.1.2.1 Simulation Results

The topology of the simulation experiments is depicted in Figure 4.1. The left side of the network has wired links with bandwidth of 10 Mbps, delay of 1 ms and buffer size of 50 packets. The right side of the network has wireless links with bandwidth of 10 Mbps, delay of 1 ms and buffer size of 50 packets. The bottleneck link between these two networks is a wired link with bandwidth of 12 Mbps, delay of 80 ms. Its buffer size is set to be 20 packets. Three types of flows – TFRC, MULTFRC, and TFRC Veno – are transferred from the wired network to the wireless network. Packet sizes are all 1000 bytes. Furthermore, we set up 8 UDP connections with pareto distribution over the bottleneck link as the background traffic. The burst time and idle time are both 100 ms. The packet size of UDP is 512 bytes and the rate of each connection is 300 Kbps. Each experiment lasts 300 s, and is repeated by 20 times. We plot the average of these results in the following figures and use the error bars to represent 95% confidence intervals.

In order to fully evaluate the performance of TFRC Veno, our simulation uses two different random loss models – exponential distribution and two-state distribution – in
the wireless links. In the exponential distribution, packet loss rate ranges from 0.0001 to 0.1. In the two-state distribution, the good state has packet loss rate of 0.00001 and period of 8 s, and the bad state has packet loss rate ranging from 0.0001 to 0.1 and period of 4 s. Furthermore, the good state has 95% probability to transit to the bad state after every period, so does the bad state.

- **Throughput**

  Firstly we let 8 target (TFRC or MULTFRC or TFRC Veno) flows run in the experiment and calculate their average throughputs. Here the throughput is counted as the number of packets sent divided by the transmission time. Figure 4.2.a and 4.2.b plot the throughput comparisons of these three protocols – TFRC, MULTFRC, and TFRC Veno – under different types of random losses. Note that the packet loss rate in Figure 4.2.b corresponds to the bad state packet loss rate in the two-state distribution. As shown in these two figures, TFRC Veno has a similar throughput as MULTFRC: its throughput is a little lower than that of MULTFRC when random loss follows the exponential distribution while a little higher when random loss follows the two-state distribution. Furthermore, both TFRC Veno and MULTFRC has much better throughput than TFRC. In particular, the throughput improvement can be up to 300% over TFRC when the packet loss rate is 0.1.
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4.2.a: Exponential distribution case.

4.2.b: Two-state distribution case.

Figure 4.2: Throughput comparisons under different random loss rates.

Secondly, we vary the number of target flows from 4 to 32 and study their performance. The random packet loss rate is set to 0.01. Observing Figure 4.3, our proposal also has a similar performance as MULTFRC under different numbers of connections. As the number of flows increases, which means the congestion is dominating the network, both TFRC Veno and MULTFRC performs increasingly similar to TFRC.

4.3.a: Exponential distribution case.

4.3.b: Two-state distribution case.

Figure 4.3: Throughput comparisons under multiple connections (packet loss rate is 0.01).

• Fairness
Fairness means the same kind of flows should share the total bandwidth fairly. Figure 4.4.a and Figure 4.4.b plot the sequence numbers of 16 MULTFRC flows and 16 TFRC Veno flows respectively, when they run over a network where random losses follow the exponential distribution and the loss rate is 0.01. Comparing these two figures, TFRC Veno flows share the bandwidth equally while the evolution of the flows among MULTFRC connections may have significant difference.

Figure 4.4: Sequence number against time, 16 connections, random losses follow exponential distribution, packet loss rate is 0.01.

In order to have a better fairness comparison among TFRC, MULTFRC and TFRC Veno, Jain’s Fairness Index $f$ [81] is adopted here:

$$f = \left( \sum_{i=1}^{n} x_i \right)^2 / \left( n \sum_{i=1}^{n} x_i^2 \right)$$  \hspace{1cm} (4.2)

where $n$ is the number of connections, $x_i$ is the throughput of the $i$th connection. The closer $f$ is to 1, the more fairness the target flows enjoy. It should be pointed out that $f$ is not a sensitive function. That is to say, a quite different fairness situation can only result in a little variation of $f$. For example, $f$ of MULTFRC in Figure 4.4.a is only 0.03 less than that of TFRC Veno in Figure 4.4.b, but the gap
between the fastest flow and the slowest flow of MULTFRC is 6 times larger than that of TFRC Veno.

We set the number of target flows to 8, and study the fairness of the three protocols under different random packet losses. As seen in Figure 4.5, TFRC Veno has a good fairness as in TFRC (the fairness index is near 1). Both TFRC Veno and TFRC have better fairness than MULTFRC.

![Figure 4.5: Fairness comparisons under different random loss rates.](image)

4.5.a: Exponential distribution case. 4.5.b: Two-state distribution case.

Then we set the packet loss rate to 0.01, and vary the number of target flows from 4 to 32. As the number increases, TFRC Veno keeps a good fairness as in TFRC. However, the fairness of MULTFRC becomes increasingly poor. Figure 4.6 plots these results.

- **TCP-friendliness**

TCP-friendliness measures whether the target flows are aggressive or not when competing with TCP flows. It can be studied as follows: we first set up a certain number of TCP Sack flows over the link, and calculate their average throughput $T_1$. Then we replace half of them with the target flows and recalculate the average
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4.6.a: Exponential distribution case. 4.6.b: Two-state distribution case.

Figure 4.6: Fairness comparisons under multiple connections (packet loss rate is 0.01).

throughput of the remaining TCP Sack flows $T_2$. If $\frac{T_2}{T_1}$ equals to 1, it means Sack flows are not affected by the target flows, and thus the target flows are totally friendly. The closer $\frac{T_2}{T_1}$ is to 1, the more friendly the target flows are.

4.7.a: Exponential distribution case. 4.7.b: Two-state distribution case.

Figure 4.7: TCP-friendliness comparisons under different random loss rates.

Figure 4.7 shows the results when there are 8 connections, and random packet loss ranges from 0.0001 to 0.1. Figure 4.8 plots the results under different numbers (ranging from 4 to 32) of connections, where the random packet loss rate is 0.01. Note that in all experiments, we always replace half of the TCP Sack connections
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Figure 4.8: TCP-friendliness comparisons under multiple connections (packet loss rate is 0.01).

with target (TFRC or MULTFRC or TFRC Veno) connections. As shown in these figures, TFRC Veno flows are not harmful to the existing TCP Sack flows when random loss follows the exponential distribution ($T_2 / T_1$ is always around 1), and are acceptable when random loss follows the two-state distribution ($T_2 / T_1$ is close to 0.9). On the other hand, MULTFRC is always very aggressive in either cases (because $T_2 / T_1$ is less than 0.6).

• Sending Rate Variation

As a transport protocol for streaming multimedia transmission, the sending rate of the target flows should not vary excessively. Here, we use the coefficient of variation (CoV) to measure this variation. CoV is defined as the standard deviation divided by the mean:

$$CoV = \frac{\sigma}{\mu},$$  \hspace{1cm} (4.3)

$$\mu = \frac{1}{N} \sum_{i=1}^{N} x_i,$$  \hspace{1cm} (4.4)
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\[ \sigma = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (x_i - \mu)^2} \]  

(4.5)

where, \( x_i \) is the \( i \)th sample, and \( N \) is the number of samples.

In the experiment, a certain number of the target flows run with the same number of TCP Sack flows. For each target flow, we collect the number of packets it sends in every 0.1 second as the samples \( (x_i) \), and use these samples to calculate its CoV after transmission. The smaller CoV is, the smoother the sending rate it has.

Figure 4.9 shows the average results when 4 target flows run with 4 TCP Sack flows, and Figure 4.10 plots the average results when the total numbers of flows (half are the target flows) increases from 4 to 32 where the packet loss rate is 0.01. The results demonstrate that MULTFRC has a smoother sending rate in most cases. However, TFRC Veno can also keep CoV at small values, which are almost the same as those of TFRC.

In summary, from the comparisons of throughput, fairness, TCP-friendliness, and sending rate smoothness we can conclude that, TFRC Veno is an efficient enhancement.
4.10.a: Exponential distribution case.  

4.10.b: Two-state distribution case.

Figure 4.10: Sending rate variation comparisons under multiple connections (packet loss rate is 0.01).

of TFRC since the significant throughput improvement in TFRC Veno does not sacrifice the other desirable features of TFRC such as fairness, TCP-friendliness, and sending rate smoothness.

4.1.2.2 Internet Experiments

We use the same link between NTU and HKU as described in Section 3.1.2.2. As shown in Figure 4.11, the TFRC/TFRC Veno client is at NTU and the TFRC/TFRC Veno server is at HKU.

At first, we report the single connection results. We select six time slots (1 hour period) over a day. In each time slot, we first let a TFRC flow run from the server to the client for 5 minutes, and then let a TFRC Veno flow run in the same direction for 5 minutes. After a 2-minute break, we repeat such test. Thus totally there are five tests in each time slot, and we calculate the average results of TFRC and TFRC Veno, respectively. Three measurements are calculated here: the number of packets sent, the receiving ratio, and the CoV. Note that the packet size is 500 bytes during transmission. 

We performed five days of testing from Monday to Friday. Here we only present the
results of one day because the general trend of the results in each day is similar. As depicted in Table 4.1, TFRC Veno can obtain throughput improvements of up to 70% over TFRC, with the same receiving ratio.

Table 4.1: The Internet experiment results on single flow.

<table>
<thead>
<tr>
<th>Time Slots</th>
<th>TFRC/TFC Veno</th>
</tr>
</thead>
<tbody>
<tr>
<td>9:00−10:00</td>
<td>2916.8/4128.6</td>
</tr>
<tr>
<td>11:00−12:00</td>
<td>2458.3/3249.7</td>
</tr>
<tr>
<td>13:00−14:00</td>
<td>2037.3/2981.3</td>
</tr>
<tr>
<td>15:00−16:00</td>
<td>2165.3/2640.9</td>
</tr>
<tr>
<td>17:00−18:00</td>
<td>3471.2/3507.4</td>
</tr>
<tr>
<td>22:00−23:00</td>
<td>1058.3/1812.7</td>
</tr>
</tbody>
</table>

Table 4.2: The Internet experiment results on coexisting flows.

<table>
<thead>
<tr>
<th>Time Slots</th>
<th>TFRC Veno/TCP Sack</th>
<th>TFRC/TCP Sack</th>
<th>TCP Sack/TCP Sack</th>
</tr>
</thead>
<tbody>
<tr>
<td>10:30−11:30</td>
<td>1469.8/1132.3</td>
<td>1462.3/1340.0</td>
<td>1390.0/1021.8</td>
</tr>
<tr>
<td>13:30−14:30</td>
<td>1657.5/1253.3</td>
<td>1159.3/1179.8</td>
<td>1328.3/1347.5</td>
</tr>
<tr>
<td>16:30−17:30</td>
<td>2038.7/1202.5</td>
<td>1496.8/1343.3</td>
<td>1364.8/1404.8</td>
</tr>
<tr>
<td>19:30−20:30</td>
<td>2212.3/1357.8</td>
<td>1609.8/1525.3</td>
<td>1299.0/1312.8</td>
</tr>
<tr>
<td>21:30−22:30</td>
<td>1762.5/1224.5</td>
<td>1435.5/1349.0</td>
<td>1225.3/1458.8</td>
</tr>
</tbody>
</table>

We also test TFRC Veno’s performance when coexisting with one competing TCP Sack flow. We select five time slots (1 hour period) over a day. In each time slot, we
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sequentially let one TFRC Veno flow, one TFRC flow, and one TCP Sack flow run with an existing TCP Sack flow for 5 minutes. This test will be repeated four times in each time slot, and we calculate the average number of packets transmitted in each flow. Note that all flows have same packet sizes of 500 bytes and are transferred from the same server to the same client. Since the general trend of the results in each day of the entire five-day period is similar, we only present the results of one day. As shown in Table 4.2, the TFRC Veno flow does not cause any bias degradation to the competing TCP Sack flow, and obtains higher throughput while compared to the TFRC flow.

4.1.3 Summary

TFRC is designed to mainly provide TCP-friendly service in transport layer for streaming applications. In wireless environments TFRC suffers throughput degradation due to its embedded Reno equation. In this work, we replace the Reno equation by the Veno equation, which is derived from the throughput model of TCP Veno, to enhance TFRC performance. The comprehensive experiments show that our enhancement can improve TFRC performance significantly over wireless networks for multimedia transmission. As compared to another sender-modified enhancement MULTFRC, our proposal is more friendly to other network traffic when achieving the same high throughput.

4.2 Veno State Differentiator Based TFRC Enhancement

In this work, we study another possible way to improve TFRC performance – the Veno state differentiator based enhancement. Specifically, our enhancement applies the state differentiator of TCP Veno [20] to distinguish non-congestion losses and congestion losses. Non-congestion losses will make less contribution to the calculation of the sending rate than congestion losses do. Our simulation results show that, such Veno state differentiator
based modification can improve TFRC performance up to 70% over wireless networks. Furthermore, this proposal is an end-to-end enhancement, which does not require any supports from the intermediate nodes.

### 4.2.1 TFRC with Veno State Differentiator

As described in Section 2.1.2.1, the Veno state differentiator uses the measurement of $N = \frac{cwnd}{RTT} \times (RTT - BaseRTT)$ as an indication of whether the network is in congestive state or non-congestive state. Specifically, if $N \geq \beta$, Veno deduces the link is in congestive state and considers the packet loss is congestion loss. Otherwise if $N < \beta$, the link is in non-congestive state and the packet loss is non-congestion loss.

In our proposal, we let the TFRC receiver distinguish non-congestion losses from congestion losses based on the Veno state differentiator. So the receiver needs to know the measurement $N$, which can be obtained from the sender:

$$N = T_{current} \times (RTT - BaseRTT)$$

Here $T_{current} = \frac{cwnd}{RTT}$, which can be regarded as the current sending rate. The sender measures the $BaseRTT$ and $RTT$, and then calculates the value of $N$ using $T_{current}$. After that, it sends $N$ to the receiver.

When the receiver detected a new loss event $i$, it will calculate the loss interval $\overline{A}_i$ based on the current value of $N$. If $N \geq \beta$, it deduces such loss event is congestion loss event, and calculate the loss interval $\overline{A}_i$ as described in Section 2.2.1.2, i.e., $\overline{A}_i = A_i$. If $N < \beta$, however, it deduces the loss event is non-congestion loss event. To reduce the contribution of non-congestion loss event to the whole loss event rate $p$, we extend the current loss interval $A_i$ by certain times $\alpha$, i.e., $\overline{A}_i = \alpha A_i$. According to equation (2.10)–(2.13), the larger loss interval ($\alpha > 1$) results the smaller loss event rate $p$. 

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To determine the value of $\alpha$, consider the following extreme scenario: if all loss events are non-congestion loss events, then

$$\overline{A}_i = \alpha A_i \quad \forall i \quad (4.6)$$

According to equation (2.10)–(2.13) we have

$$\overline{A}(1, n) = w \sum_{i=1}^{n} w_i \overline{A}_i = \alpha w \sum_{i=1}^{n} w_i A_i = \alpha A(1, n) \quad (4.7)$$

$$\overline{A}(0, n-1) = w \sum_{i=0}^{n-1} w_{i+1} \overline{A}_i = \alpha w \sum_{i=0}^{n-1} w_{i+1} A_i = \alpha A(0, n-1) \quad (4.8)$$

$$\overline{A} = \max(\overline{A}(1, n), \overline{A}(0, n-1)) = \alpha \max(A(1, n), A(0, n-1)) = \alpha A \quad (4.9)$$

and,

$$\overline{p} = \frac{1}{\overline{A}} = \frac{1}{\alpha A} = \frac{1}{\alpha \overline{p}} \quad (4.10)$$

According to [36], the throughput of TCP Veno is $\sqrt{3}$ times of that of TCP Reno if all the losses are non-congestion losses and all the time-outs are ignored. Then we have:

$$\overline{T} = \frac{1}{RTT} \sqrt{\frac{3}{2 \overline{p}}} = \frac{1}{RTT} \sqrt{\frac{3\alpha}{2\overline{p}}} = \sqrt{3} \overline{T} = \frac{1}{RTT} \frac{3}{\sqrt{2\overline{p}}} \quad (4.11)$$

From equation (4.11) we get:

$$\alpha = 3 \quad (4.12)$$

In summary, the loss interval $\overline{A}_i$ is calculated as follows:

$$\overline{A}_i = \begin{cases} A_i, & N \geq \beta \\ 3A_i, & N < \beta \end{cases}$$

### 4.2.2 Performance Evaluation

The evaluation of our enhancement (hereinafter, called “Enhancement”) is based on NS-2 simulation experiments. Different aspects of performance such as throughput, receiving
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ratio, sending rate variation, fairness, and TCP-friendliness, have been extensively studied in the following section.

Note that in TCP Veno, $\beta$ is normally set to be 3. However, after a complete evaluation on different values of $\beta$ in the following simulation experiments, we find $\beta = 1$ is a better choice for our TFRC enhancement on achieving a good tradeoff between throughput and TCP-friendliness.

The topology of our experiment on NS-2 is shown in Figure 4.12.

The left side network has wired links with bandwidth of 10 Mbps, delay of 1 ms and packet buffer size of 50. The right side network has wireless links with bandwidth of 10 Mbps, delay of 1 ms and packet buffer size of 50. Random losses in wireless links follow exponential distribution and the packet loss rate ranges from 0.0001 to 0.1. The bottleneck link between these two networks is a wired link. The bandwidth is 5 Mbps and delay is 80 ms. Its packet buffer size is set to be 20. TFRC packets are transferred from the wired network to the wireless network. Packet size is 1000 bytes.

- **Throughput**

  At first we let a single TFRC or Enhancement flow run over the network. Meanwhile, we set up three UDP connections with pareto distribution over the bottleneck
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link as the background traffic. The packet size of UDP is 512 bytes. The burst time and idle time are both 100 ms. The rate of each connection is set to 500 Kbps. The experiment time is 300 s. Figure 4.13.a plots the throughput results of TFRC and Enhancements with different values of $\beta$ (1, 2, and 3). As shown in the figure, different values of $\beta$ bring a little difference in the throughput for each Enhancement. However, all of these three Enhancements improve much throughput (up to 70%) over the original TFRC.

![Figure 4.13.a: Single flow.](image1)

![Figure 4.13.b: Multiple flows, random loss rate is 0.01.](image2)

Figure 4.13: Throughput comparison.

Then we increase the number of flows from 4 to 64 and compare the throughputs of TFRC and Enhancement when random loss rate is 0.01. Figure 4.13.b observes that, as the number of connections increases, the network is dominated by congestion and Enhancement performs similarly as TFRC does finally.

- Receiving ratio

We also compare the receiving ratio between our enhancement and TFRC during the experiments in throughput before. Receiving ratio is defined as follows:

$$\text{Receiving Ratio} = \frac{\text{Number of packets received}}{\text{Number of packets sent}}$$
which indicates how many packets are lost during transmission.

4.14.a: Single flow. 4.14.b: Multiple flows, random loss rate is 0.01.

Figure 4.14: Receiving ratio comparison.

From Figure 4.14.a and 4.14.b we can see, Enhancement almost keeps the same receiving ratio as that of TFRC under single connection. However, as the number of connections increase, Enhancement becomes losing a few more packets than TFRC does. This is because of the accuracy of Veno’s state differentiator. As stated in [85], Veno’s state differentiator works well under light load networks, but will misinterpret some congestion losses as non-congestion losses when the network load is heavy. The larger the threshold $\beta$ is, the more misinterpretations Veno’s differentiator has. In this case, our proposal based on Veno’s state differentiator will overestimate the sending rate and thus causes slightly more packet losses during transmission. Furthermore, the larger the threshold $\beta$ is, the higher the overestimated sending rate is, and the more packet losses our enhancement has.

- **Sending rate variation**

We also use coefficient of variation (CoV) defined in 4.3 to measure the sending smoothness of our proposal.
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In the experiment, at first 4 TFRC flows run with 4 TCP SACK flows for 300s. Then we replace 4 TFRC flows with Enhancement flows and redo the experiment. We collect the number of packets the target flow sends in every 0.1s as samples, and calculate its CoV. Figure 4.15 plots the average results. As we see, our proposal does not sacrifice the smoothness of the sending rate.

- Fairness

We continue to use the Jain’s Fairness Index $f$ to reflect the fairness:

$$ f = \frac{\left(\sum_{i=1}^{n} x_i\right)^2}{n \sum_{i=1}^{n} x_i^2} $$

where, $n$ is the number of connections, $x_i$ is the throughput of the $i$th connection. The closer $f$ is to 1, the more fairness the flow has. Here we let 4 to 64 Enhancement connections run over the bottleneck link. As shown in Figure 4.16, Enhancement

Figure 4.15: CoV comparison.
can keep good fairness under different packet loss rates, no matter which threshold \( \beta \) is used.

![Graph showing fairness of Enhancement with different values of \( \beta \)]

4.16.a: \( \beta = 3 \).

4.16.b: \( \beta = 2 \).

4.16.c: \( \beta = 1 \).

Figure 4.16: Fairness of Enhancement with different values of \( \beta \).

- **TCP-friendliness**

  We first set up certain number (ranging from 4 to 64) of TCP SACK flows over the link, and calculate their average throughput \( T_1 \). Then we replace half of them with Enhancement flows and recalculate the average throughput of the left TCP SACK flows \( T_2 \). The closer \( \frac{T_2}{T_1} \) is to 1, the more friendliness Enhancement flows have. As illustrated in Figure 4.17, as the threshold value \( \beta \) increases, Enhancement flows becomes more and more aggressive, especially over heavy load networks. The reason is the same as that described in receiving ratio – the overestimated throughput makes Enhancement aggressive. However, the friendliness of Enhancement when \( \beta = 1 \) is acceptable since \( \frac{T_2}{T_1} \) is always around 85% even under heavy load.

### 4.2.3 Summary

This work modifies TFRC protocol based on the Veno state differentiator. Such differentiator is used here to distinguish congestion loss events and non-congestion loss events. Our proposal discounts the impact of non-congestion loss events by extending their loss.

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4.17.a: $\beta = 3$.  
4.17.b: $\beta = 2$.  
4.17.c: $\beta = 1$.

Figure 4.17: Friendliness of Enhancement with different values of $\beta$.

intervals by 3 times. Simulation results have shown the better performance our enhancement achieves. Meanwhile, we also study how the threshold $\beta$ used in the Veno state differentiator can influence the performance of our proposal. After a complete evaluation, we find $\beta = 1$ is a better empirical value on achieving a good tradeoff between throughput and TCP-friendliness.

4.3 TFRC Veno Implementation

As an important work for TFRC Veno, we implemented it into Linux Kernel 2.6.12. This work is based on the Linux Kernel implementations of TFRC and DCCP [86][87] maintained by Arnaldo C. Melo and Ian McDonald et al [88]. Here DCCP can be seen as a framework to provide unreliable data transmission, and TFRC is one optional choice of congestion control algorithms in DCCP [89]. In the following part, we will first review the DCCP implementation in Linux Kernel, and then show how to modify the TFRC module in DCCP to incorporate our TFRC Veno algorithm. Note that the implementations of TFRC and DCCP are only the experimental versions in Linux Kernel 2.6.12.

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4.3.1 DCCP in Linux Kernel

4.3.1.1 Structures

DCCP has its own protocol control block in the Kernel. Its structure in the protocol
switch table is shown in Listing 4.1.

Listing 4.1: DCCP protocol control block

```c
static struct inet_protosw dccp_v4_protosw = {
    .type = SOCK_DCCP ,
    .protocol = IPPROTO_DCCP ,
    .prot = & dccp_v4_prot ,
    .ops = & inet_dccp_ops ,
    .capability = -1 ,
    .no_check = 0 ,
    .flags = 0 ,
};
```

Here SOCK_DCCP is the type of the DCCP protocol, which is used when creating socket
for DCCP, e.g., `socket(AF_INET,SOCK_DCCP,0)`. The structure `dccp_v4_prot` defines all func-
tions the DCCP protocol supports, such as connecting, sending, and receiving, as shown
in Listing 4.2. When a system call from socket layer comes down, the Kernel can use this
structure to find the corresponding functions. For example, in response to the `connect`
system call, the Kernel will call `dccp_v4_connect` to do the remaining work.
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Listing 4.2: DCCP protocol functions

```c
struct proto dccp_v4_prot = {
    .name = "DCCP",
    .owner = THIS_MODULE,
    .close = dccp_close,
    .connect = dccp_v4_connect,
    .disconnect = dccp_disconnect,
    .ioctl = dccp_ioctl,
    .init = dccp_v4_init_sock,
    .setsockopt = dccp_setsockopt,
    .getsockopt = dccp_getsockopt,
    .sendmsg = dccp_sendmsg,
    .recvmsg = dccp_recvmsg,
    .backlog_rcv = dccp_v4_do_rcv,
    .hash = dccp_v4_hash,
    .unhash = dccp_v4_unhash,
    .accept = inet_csk_accept,
    .get_port = dccp_v4_get_port,
    .shutdown = dccp_shutdown,
    .destroy = dccp_v4_destroy_sock,
    .max_header = MAX_DCCP_HEADER,
    .obj_size = sizeof(struct dccp_sock),
    .sk_prot = &dccp_request_sock_ops,
    .orphan_count = &dccp_orphan_count,
};
```

4.3.1.2 Functions

Figure 4.18 gives the overview of the DCCP functions during data transmission. The sender sends data to the receiver and the receiver feeds back with acknowledgements (ACKs). In the sender side, when the DCCP socket has packets to send, it calls `dccp_sendmsg`. Then `dccp_sendmsg` calls `ccid_hc_tx_send_packet` to send out the packet. The function of `ccid_hc_tx_send_packet` is to choose one of the congestion control algorithms (CCID 1, CCID 2, ...) to determine the sending rate of the outgoing packets. As mentioned before, TFRC is one of these congestion control algorithms, and its congestion control identification is CCID 3. In this example, the TFRC module is selected. To determine the sending rate of the data, the TFRC module needs the information in the ACKs from the receiver. When the sender receives these ACKs, `dccp_v4_rcv` is called to process them. After that, `ccid_hc_tx_packet_rcv` is called to denote which congestion control module needs these ACKs. In this example, the TFRC module (CCID 3) receives...
them, and calculates the sending rate based on the Reno equation.

In the receiver side, `dccp_v4_rcv` is called when the data is received. Then `dccp_v4_rcv` calls `ccid_hc_rx_packet_recv` to select the congestion control module to process the data. In this example, the TFRC module is chosen, and it calculates the packet loss rate and the receiving rate according to the data received so far. Besides sending the data to the socket layer, the TFRC module also calls `dccp_send_ack` to feed back the information such as the packet loss rate and the receiving rate to the sender in ACKs.

### 4.3.1.3 CCID 3: TFRC Module

DCCP in the Linux Kernel holds a pointer array to store its congestion control modules. As shown in Figure 4.19, `ccid[255]` is an array storing the pointers which point to the congestion control modules in DCCP. Listing 4.3 shows the structure of the TFRC module (CCID 3).
Chapter 4. Analysis and Enhancements of End-to-end Congestion Controls for Streaming over Wireless Networks

Figure 4.19: CCID pointer array.

Listing 4.3: TFRC protocol functions

```c
static struct ccid ccid3 = {
    .ccid_id = 3,
    .ccid_name = "ccid3",
    .ccid_owner = THIS_MODULE,
    .ccid_init = ccid3_init,
    .ccid_exit = ccid3_exit,
    .ccid_hc_tx_init = ccid3_hc_tx_init,
    .ccid_hc_tx_exit = ccid3_hc_tx_exit,
    .ccid_hc_tx_send_packet = ccid3_hc_tx_send_packet,
    .ccid_hc_tx_packet_sent = ccid3_hc_tx_packet_sent,
    .ccid_hc_tx_packet_recv = ccid3_hc_tx_packet_recv,
    .ccid_hc_rx_init = ccid3_hc_rx_init,
    .ccid_hc_rx_exit = ccid3_hc_rx_exit,
    .ccid_hc_rx_parse_options = ccid3_hc_rx_parse_options,
    .ccid_hc_rx_packet_recv = ccid3_hc_rx_packet_recv,
};
```

When ccid_hc_* functions are called, a pointer named ccid is used to point to the congestion control module needed. For example, if ccid_hc_tx_packet_recv is called, and it passes ccid to point to the TFRC module ccid3, then the function ccid3_hc_tx_packet_recv in the TFRC module is called to process the remaining work. In the following part, we only focus on the functions in the TFRC module, which are shown in Figure 4.20.

In the receiver side, ccid3_hc_rx_packet_recv is called in the TFRC module to process the data. Beside passing the data to the upper layer, it calls ccid3_hc_rx_send_feedback to
calculate the packet loss rate and the receiving rate. After that, it calls dccp_send_ack to send these information to the sender in ACKs.

In the sender side, the function ccid3_hc_tx_packet_recv processes the received ACKs. It updates the round-trip time (RTT) based on these ACKs, and calculates the new sending rate for the data by calling ccid3_hc_tx_update_x. It is known that TFRC calculates its sending rate mainly based on the Reno equation. In ccid3_hc_tx_update_x, this work is done by the function ccid3_calc_x. Besides the result from the Reno equation, ccid3_hc_tx_update_x also has the responsibility to check the result satisfying other constraints, e.g., the result should not exceed two times of the receiving rate reported by the receiver. The detailed constraints of the sending rate calculation can be referred to [10]. After getting the final sending rate, ccid3_hc_tx_update_x schedules the sending timer based on it. After that, ccid_hc_tx_send_packet uses this sending timer to send out the data.
It is worthy to note that the floating point type is not allowed in the Linux Kernel, thus the calculation of the Reno equation, i.e., \texttt{ccid3\_calc\_x}, must be carefully designed. The TFRC module in the Linux Kernel assumes $b = 1$ in the Reno equation (2.8). Considering $RTO = 4 \times RTT$, the Reno equation can be arranged as:

$$T_{calc} = \frac{S}{RTT \times f(p)}$$  \hspace{1cm} (4.13)

where,

$$f(p) = \sqrt{\frac{2p}{3}} + 12 \sqrt{\frac{3p^8}{8}(1 + 32p^2)}$$  \hspace{1cm} (4.14)

The TFRC module keeps a table \texttt{calcx\_lookup[500][2]} to store the static values of $f(p)$, where \texttt{calcx\_lookup[*][1]} stores the values of $f(p)$ when $p \in [0.0001, 0.05]$, and \texttt{calcx\_lookup[*][0]} stores the values of $f(p)$ when $p \in [0.002, 1]$, as shown in Figure 4.21. Given a loss rate $p$, the TFRC module converts it to an index, and looks up the corresponding value of $f(p)$ in the table directly: if $p < 0.05$, then it looks up in \texttt{calcx\_lookup[*][1]}; otherwise, it looks up in \texttt{calcx\_lookup[*][0]}. To avoid the floating point type variables, the values in $f(p)$ are multiplied by 1,000,000 times in the Linux Kernel.

Note that, the calculation of $f(p)$ in \texttt{calcx\_lookup[500][2]} is based on the real values of $p$, but not the variable $p$ indicating the packet loss rate in the Linux Kernel, because this variable is also multiplied by 1,000,000 times. To convert the multiplied variable $p$ to the real index for the table \texttt{calcx\_lookup[500][2]}, the TFRC module perform following work shown in Listing 4.4.
Chapter 4. Analysis and Enhancements of End-to-end Congestion Controls for Streaming over Wireless Networks

![Table for the values of $f(p)$](image)

**Figure 4.21:** Table for the values of $f(p)$.

**Listing 4.4:** Conversion from $p$ to index

```c
if (p < 50000)
    index = (p / 100) - 1;
else
    index = (p / 2000) - 1;
...

if (p >= 50000)
    f = calcx_lookup[index][0];
else
    f = calcx_lookup[index][1];
```

After getting the value of $f(p)$, the TFRC module can calculate the result of the Reno equation based on (4.13). Note that $RTT$ is in millisecond in the Kernel, thus the value of $S$ must be multiplied by $1,000,000,000$ times to make the final result correct.

### 4.3.2 TFRC Veno Module

As stated in Section 4.1, an attractive feature of TFRC Veno is that it needs only a minor modification in the TFRC sender. As for the TFRC module in the Linux Kernel, we only need to add a function called `tfrc_veno_calc_x` in `ccid3_hc_tx_update_x`, as shown in Figure 4.20. `tfrc_veno_calc_x` takes the responsibility to calculate the sending rate based on the
Chapter 4. Analysis and Enhancements of End-to-end Congestion Controls for Streaming over Wireless Networks

Veno equation instead of the Reno equation.

The calculation of the Veno equation is similar to the method described in Section 4.3.1.3. However, the situation is more complicated in the Veno equation because the Veno parameter $\gamma$ is added.

We also arrange the Veno equation (4.1) by considering $b = 1$ and $RTO = 4 \times RTT$:

$$T_{calc} = \frac{S}{RTT \times f(p)}$$  \hspace{1cm} (4.15)

where,

$$f(p) = a(p)c1(\gamma) + b(p)c2(\gamma)$$  \hspace{1cm} (4.16)

$$a(p) = \sqrt{2p}, \quad b(p) = 12 \sqrt{\frac{p}{2}} p(1 + 32p^2)$$  \hspace{1cm} (4.17)

$$c1(\gamma) = \sqrt{\frac{1 - \gamma}{1 + \gamma}}, \quad c2(\gamma) = \sqrt{1 - \gamma^2}$$  \hspace{1cm} (4.18)

That is to say, to obtain the result from $f(p)$, we must calculate $a(p)$, $b(p)$, $c1(\gamma)$, and $c2(\gamma)$ respectively. Thus we keep three static tables in the Kernel: `calcx_A[500][2]`, `calcx_B[500][2]`, and `calcx_C[451][2]`. Here `calcx_A[500][2]` and `calcx_B[500][2]` have the same structure as `calcx_lookup[500][2]` described in Section 4.3.1.3, which are used to map $p$ to the corresponding $a(p)$ and $b(p)$ respectively. `calcx_C[451][2]` is a new structure for the Veno parameter $\gamma \in [0.5, 0.95]$, where `calcx_C[*][0]` stores the values of $c1(\gamma)$, and `calcx_C[*][1]` stores the values of $c2(\gamma)$. Note that all these values are multiplied by 1,000,000 times. Figure 4.22 gives the detailed information.

To avoid the floating point type value of $\gamma$, its value is multiplied by 1,000,000 times. As similar to the TFRC module, we also need to convert the variable $p$ or $\gamma$ to the index for the table `calcx_A[500][2]` or `calcx_B[500][2]` or `calcx_C[451][2]`. The conversion and calculation of $f(p)$ is shown in Listing 4.5.
Chapter 4. Analysis and Enhancements of End-to-end Congestion Controls for Streaming over Wireless Networks

![Figure 4.22: Table for the values of \( c_1(\gamma) \) and \( c_2(\gamma) \).](image)

<table>
<thead>
<tr>
<th>index (c1)</th>
<th>index (c2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>2</td>
<td>0</td>
</tr>
<tr>
<td>449</td>
<td>0</td>
</tr>
<tr>
<td>450</td>
<td>0</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>index (c1)</th>
<th>index (c2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>2</td>
<td>0</td>
</tr>
<tr>
<td>449</td>
<td>0</td>
</tr>
<tr>
<td>450</td>
<td>0</td>
</tr>
</tbody>
</table>

Listing 4.5: Conversion from \( p \) and \( \gamma \) to index

```c
index = (\gamma / 1000) - 500;
...

c1 = calcx_C[index][0]; // look up \( c_1(\gamma) \)
c2 = calcx_C[index][1]; // look up \( c_2(\gamma) \)

if (p < CALCX_SPLIT)
   index = (p / 100) - 1;
else
   index = (p / 2000) - 1;
...

/* look up \( a(p) \) and \( b(p) \) */
if (p >= 500) {
   a = calcx_A[index][0];
   b = calcx_B[index][0];
}
else {
   a = calcx_A[index][1];
   b = calcx_B[index][1];
}

f = (u64)a * (u64)c1 + (u64)b * (u64)c2; // calculate \( f(p) \)
do_div(f, 1000000); // make \( f(p) \) smaller for the following work
```

Finally, the TFRC Veno module can calculate the result of the Veno equation based
4.3.3 Summary

The DCCP implementation in the Linux Kernel provides a framework for unreliable data transmission. The TFRC module is one of the congestion control algorithms (CCID 3) used in DCCP. Our TFRC Veno module only add a new calculation function in the TFRC module, to calculate the sending rate based on the Veno equation instead of the Reno equation. However, because the Linux Kernel does not allow the floating point type, these calculation functions must be carefully designed. Static tables are used in both TFRC and TFRC Veno modules to provide a method to calculate floating point values by integers.
Chapter 5

Conclusion and Future Work

Wireless networks bring performance degradation issue to end-to-end congestion control algorithms, either in bulk data transfer applications or in streaming applications. This thesis studies the wireless issue in these two different categories of end-to-end congestion control algorithms respectively. In the bulk data transfer category, the main object to study is TCP Veno, a novel TCP variant for wireless communications. In the streaming category, the main object to study is TFRC, a popular equation based rate control for multimedia applications. In the following part, we will summarize our research work on these two different protocols, and then give the possible directions for future work.

5.1 Summary

The research work on TCP Veno starts with the analytical model. We develop a throughput model for TCP Veno, to provide theoretical analysis on TCP Veno behavior over wireless networks. The derivation characterizes TCP Veno congestion control algorithm in both wire and wireless situations, considering the dynamic reduction on congestion window in TCP Veno for different types of packet losses. The derived equation, called “Veno equation”, is proved to be able to predict TCP Veno throughput accurately by extensive simulations and the Internet experiments. Here we choose PFTK method [11] to
Chapter 5. Conclusion and Future Work

derive the Veno equation, because the equation obtained by PFTK method can be seam-
lessly imported into TFRC protocol. In this way, the Veno equation not only gives an
in-depth view on TCP Veno about its behavior during network communications, but also
provides a simple but efficient way to enhance TFRC protocol for streaming applications,
i.e., TFRC Veno in this thesis.

The second issue of TCP Veno this thesis tries to explore is whether Veno has reached
its optimal performance over wireless networks. Through the analysis on Veno behav-
ior over different types of wireless networks, we find that the fixed reductions (1/5) of
congestion window on random losses without concerning the network circumstance will
constraint the possible improvement of Veno in throughput. This thesis studies two typi-
cal wireless networks that TCP Veno may experience – light load networks and high BDP
networks. Light load networks mean that the network resources are not fully occupied
and congestion losses are not pervasive during TCP communications. Wireless local area
networks (wireless LAN) usually belong to this category. A new variable, named conges-
tion loss rate, is introduced in this case. It helps Veno act more appropriately in response
to random losses and grab more available bandwidth on the link, while there is no fair-
ness or friendliness sacrificed. High BDP networks mean the BDP of the connection is
lager than the buffer size on the bottleneck link. This definition covers very large scale in
different types of networks, including the high speed networks and long delay networks.
An approximated variable called $pBDP$ is introduced to inform Veno situations of the
network and refine its congestion window evolution. This alleviates the negative impact
of random losses over the high BDP networks and improves Veno throughput.

TCP Veno provides an efficient solution for the end-to-end congestion control over
wireless networks. Although it is designed for bulk data transfer, its core ideas, such
as the Veno equation and the state differentiator, can also be integrated into congestion
controls for streaming. This thesis proposes two enhancements for TFRC based on the
Veno equation and the state differentiator respectively.
CHAPTER 5. CONCLUSION AND FUTURE WORK

The Veno equation based enhancement, also known as TFRC Veno, replaces the Reno equation with the Veno equation. Since the Veno equation models the behavior of TCP Veno, a wireless TCP variant, TFRC Veno can perform much better than the original TFRC does in wireless environments. The significant improvement obtained by TFRC Veno does not bias the fairness or TCP-friendliness of TFRC. This makes TFRC Veno outperforms MULTFRC, another TFRC enhancement which may cause intra- and inter-protocol fairness problems during transmission. Furthermore, TFRC Veno only involves a simple modification of TFRC on the sender side – replacement of the equation, and requires no changes on the receiver protocol stack or interventions of the intermediate network nodes. Thus it can be easily deployed in the current Internet.

The following table (Table 5.1) gives comparisons among TFRC, MULTFRC, and TFRC Veno over wireless networks based on our analysis and experiments.

<table>
<thead>
<tr>
<th></th>
<th>TFRC</th>
<th>MULTFRC</th>
<th>TFRC Veno</th>
</tr>
</thead>
<tbody>
<tr>
<td>Throughput</td>
<td>Bad</td>
<td>Good</td>
<td>Good</td>
</tr>
<tr>
<td>Fairness and TCP-friendliness</td>
<td>Good</td>
<td>Bad</td>
<td>Good</td>
</tr>
<tr>
<td>Sending rate smoothness</td>
<td>Good</td>
<td>Good</td>
<td>Good</td>
</tr>
<tr>
<td>Deployability</td>
<td>Good</td>
<td>Good</td>
<td>Good</td>
</tr>
</tbody>
</table>

Another way to enhance TFRC performance is to use the Veno state differentiator. This proposal keeps the Reno equation in TFRC intact, but uses the Veno state differentiator to distinguish congestion losses and random losses during transmission. By discounting the impact of those non-congestion losses on the throughput calculation, it can effectively alleviate the throughput degradation caused by wireless links, although
such improvement cannot reach as high as TFRC Veno does. It also maintains other merits of the original TFRC, such as sending rate smoothness, fairness, and TCP-friendliness. This enhancement is an end-to-end solution, which involves modifications of both sender and receiver sides.

5.2 Future Work

There are several interesting directions for the future work, based on the work we have done.

- **A More Accurate Veno Equation**

  The derivation of the Veno equation in this thesis makes two simplified assumptions in the Veno algorithm: 1) we assume that the refined AI phase in Veno has negligible improvement in TCP throughput as compared to the refined MD phase; 2) we assume the congestion window values under random losses are uniformly distributed. While extensive experiments have proved that our Veno equation has a fairly good fit to the throughput of Veno in different environments, it is worthy to obtain a more complete equation for TCP Veno, which includes the refined AI algorithm of Veno and the accurate distribution of congestion window under random losses. We believe this future work will help to explore the behaviors of Veno in a more detailed way, and improve the performance of our TFRC Veno further.

- **A More Flexible Veno Algorithm for Diverse Wireless Networks**

  This thesis has proposed two enhancements for TCP Veno to avoid its “blind” reduction of congestion window under random losses. Each of them uses an additional variable (\(\text{congestionlossrate}\) or \(\text{pBDP}\)) to indicate the network circumstance. This provides more detailed information about the network situations to Veno, and thus
make it act more appropriately on dealing with random losses. If we regard the original Veno using the state differentiator as "one-factor" algorithm for wireless networks, then these two enhanced Veno algorithms that use not only the state differentiator but also additional network situation indication can be seen as "two-factor" algorithms. However, these proposals are only the preliminary results of this research direction, since either enhancement can only handle its own special network situation (light load or high BDP). In the future work, two issues should be considered: 1) how to indicate Veno with network circumstance more accurately; 2) how to make Veno enhancement suitable for not only one special type of networks, but also complicated heterogenous networks. To realize these two objects, more useful variables should be studied to indicate the network situations. These variables may build up a "n-factor" Veno algorithm. Furthermore, these factors may contribute to the reduction of congestion window flexibly using a control method, e.g., fuzzy control [90][91][92][93].

- Congestion Controls in Multicast of Streaming over Wireless Networks

TFRC and TFRC Veno are both end-to-end congestion controls for streaming, which support single flow transmitted from one node to the other node in the network. Many multimedia applications such as IPTV or video conference, however, require flow delivery among several nodes at the same time in networks. This can be realized by multicast. As the reference to [94], multicast has two types of architectures – IP multicast [95] and overlay multicast [96][97]. IP multicast requires support from intermediate routers. Data originates from a sender, is replicated at intermediate routers if needed, and is forwarded until it reaches a destination. On the other hand, overlay multicast let end hosts perform all multicast functions, i.e., routers only perform unicast forwarding, and data is replicated at end hosts and
delivered to other hosts using end-to-end transmission. No matter which architecture is employed, congestion controls are needed in multicast to avoid network collapse during transmission. Similar to the end-to-end congestion controls, these multicast congestion controls also encounter performance degradation issue over wireless networks.

Congestion control over IP multicast has been studied for many years and some solutions have been proposed, such as Representative [98], RLA [99], Golestani [100], MTCP [101], TFMCC [102], and SMCC [103]. Among these proposals, TFMCC and SMCC are two mechanisms based on TFRC rate control protocol. TFMCC is a single rate multicast congestion control, which means all the nodes in the multicast receive data with a same rate. On the other hand, SMCC is a multiple rate multicast congestion control, that is, different nodes may receive data with different rates. Since we have obtained a wireless enhancement of TFRC – TFRC Veno, it is possible to import TFRC Veno into TFMCC and SMCC to replace TFRC protocol. In fact, some preliminary results have been obtained in our experiments. They have shown the significant throughput improvement of both TFMCC and SMCC over wireless networks. In the future work, more experiments and more results are needed to completely analyze the performance of TFRC Veno in multicast congestion control functions.

Overlay multicast is an emerging technology in the Internet. Since it remains the end-to-end semantic during multicast transmission, traditional congestion controls such as TFRC should work well in this architecture. However, it is an interesting issue on how random losses impact the performance of the whole system, when wireless networks are introduced. It can be predicted that TFRC Veno should perform better than TFRC does in these wireless environments. More study and work are needed to prove that in the future work.
**Chapter 5. Conclusion and Future Work**

- **Veno II: A Uniform Platform of End-to-end Congestion Controls for Heterogenous Networks**

Bulk data transfer and streaming compose the daily applications of the Internet. As wireless communication technology has been making significant progress in the recent Internet, congestion control algorithms meet new issues over wireless networks. This thesis covers the study of wireless issue in end-to-end congestion controls for both bulk data transfer and streaming. The wireless enhancements proposed in this thesis, either for bulk data transfer or for streaming, originate from the idea of TCP Veno. In the future work, we will explore the possibility to integrate these two categories of wireless enhancements into a uniform platform, named “Veno II”, which is being developed by our whole team.
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REFERENCES


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REFERENCES


REFERENCES


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