Dynamic-CBT – Router Queue Management for Improved Multimedia Performance on the Internet

by

Jae Won Chung

A Thesis

Submitted to the Faculty

of the

WORCESTER POLYTECHNIC INSTITUTE

In partial fulfillment of the requirements for the

Degree of Master of Science

in

Computer Science

by

May 2000

APPROVED:

Professor Mark Claypool, Major Advisor

Professor Robert Kinicki, Thesis Reader

Professor Micha Hofri, Head of Department

Abstract

The explosive increase in Internet traffic has placed a growing emphasis on congestion control and fairness in Internet routers. Approaches to the problem of congestion, such as active queue management schemes like Random Early Detection (RED) that are successful with TCP flows, use congestion avoidance techniques. Approaches to the problem of fairness, such as Fair Random Early Drop (FRED), punish misbehaved, non-TCP flows. Unfortunately, these punishment mechanisms result in a significant performance drop for multimedia flows that are well behaved. We propose a new active queue management mechanism as an extension to RED called Dynamic Class-Based Threshold (D-CBT) to improve multimedia performance on the Internet. Also, as an effort to further improve multimedia performance especially on jitter, we propose a lightweight packet scheduling called Cut-In Packet Scheduling (ChIPS) as an alternative to FIFO packet scheduling. The performance of our proposed mechanisms is measured, analyzed and compared with other mechanisms (RED and CBT) in terms of throughput, fairness and multimedia jitter through simulation using NS. The study concludes that D-CBT improves fairness among different classes of flows and ChIPS improves multimedia jitter without degrading fairness. The contributions we make are the design, implementation and evaluation of D-CBT and ChIPS, the first study of multimedia jitter due to network queue management policy, and a simulator implementation of CBT, D-CBT and ChIPS.

Table of Contents

Acknowledgements

I would like to express my gratitude to my academic and thesis advisor Professor Mark Claypool for his advise, support and help not only on this thesis but also on all academic and personal matters. His kindness, trust and support gave me another chance to further expose myself to the excitement of computer science and multimedia networking.

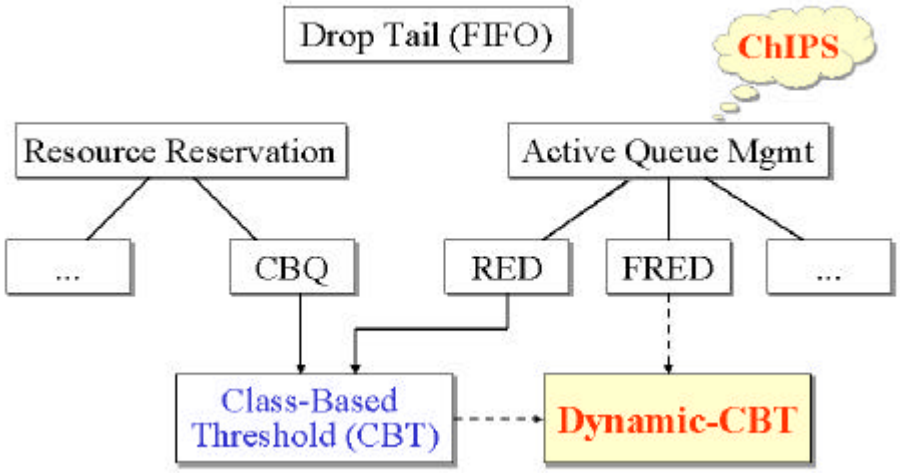
Also, I would like to specially thank to my reader Professor Robert Kinicki, who was previously my Major Qualifying Project (MQP) advisor. He kindly taught me how to research in my undergraduate years, introduced me to the joy of computer networks.

I would like to specially thank to my parents who love me and encourage me all the time, and supported me to pursue my graduate study. Especially, my father's love for knowledge and study, and my mother's pray was the driving force behind my current achievement. Also, I would like to thank with love to Youjin Ji who loves and cares about me so much.

This thesis is dedicated to my dear God who gave me insights, the power and ability to think, and always lead and guide me with his own arms.

1. Introduction

The Internet has moved from a data communication network for a few privileged professions to an essential part of public life similar to the public telephone networks, while assuming the role of the underlying communication network for multimedia applications such as Internet phone, video conferencing and video on demand (VOD). As a consequence, the volume of traffic and the number of simultaneous active flows that an Internet router handles has increased dramatically, placing new emphasis on congestion control and traffic fairness. Complicating traditional congestion control is the presence of multimedia traffic that has strict timing constraints, specifically *delay* constraints and variance in delay, or *jitter* constraints [lc95, GP97, AP97]. This paper presents a router queue management mechanism that addresses the problem of congestion and fairness, and improves multimedia performance on the Internet. Figure 1.1 shows some of the current and the proposed router queue mechanisms.



<Figure 1.1> Router Queue Mechanisms (shaded are proposed)

There have been two major approaches suggested to handle congestion by means other than traditional drop-tail FIFO queuing. The first approach uses packet or link scheduling on multiple logical or physical queues to explicitly reserve and allocate output bandwidth to each class of traffic, where a class can be a single flow or a group of similar flows. This is the basic idea of various Fair Queuing (FQ) disciplines [DH90] and the Class-Based Queuing (CBQ) algorithm [FJ95, Fl]. The main advantage of this approach is in its ability to assign network resources to classes of flows in a predefined manner. Moreover, when coupled with admission control, the mechanism not only suggests a solution to the problem of congestion but also offers potential performance guarantees for the multimedia traffic class. However, the explicit resource reservation approach would change the "best effort" nature of the current Internet, and the fairness definition of the traditional Internet may no longer be preserved. Adopting this mechanism would require a change in the network management and billing practices. Also, the algorithmic complexity and state requirements of scheduling make its deployment difficult [PJS99].

The second approach, called Active Queue Management, uses advanced packet queuing disciplines other than traditional FIFO drop-tail queuing on an outbound queue of a router to actively handle (or avoid) congestion with the help of cooperative traffic sources. This approach is based on the observation that packets from a connection usually take the same path, and depend on the ability of the traffic source to respond to the congestion in the network. In the Internet, TCP recognizes packet loss as an indicator of network congestion, and its back-off algorithm reduces transmission load when

network congestion is detected [Fl94]. Recently, active queue management has become the subject of interest within the Internet research committee [PJS99].

The most well-known and basic active queue management mechanism is Random Early Detection (RED), which prevents congestion through monitoring outbound buffers to detect impending congestion, and randomly chooses and notifies senders of network congestion so that they can reduce their transmission rate [FJ93, FF97]. More precisely, RED uses a weighted-average queue size and minimum-maximum thresholds to detect impending congestion, and randomly drops incoming packets according to the calculated rate based on the average queue size during impending congestion. While fairly handling congestion for only TCP flows, RED reveals the critical problem that non-TCP flows that are unresponsive or have greedier flow-control mechanisms than TCP take more share of the output bandwidth than TCP flows [LM97, PJS99]. In the worst case, it is possible for non-TCP flows, especially for unresponsive ones, to monopolize the output bandwidth while TCP connections are forced to transmit at their minimum rates. This unfairness occurs because non-TCP flows reduce transmission load relatively less than TCP flows or do not reduce at all, and the same drop rate is applied to every flow. This problem may become critical as the number of multimedia flows increases. Delay sensitive Internet multimedia applications use UDP rather than TCP because they require in-time packet delivery and can tolerate loss, rather than the guaranteed packet delivery with potentially unbounded end-to-end delay that TCP produces. Also, they prefer periodic packet transmission characteristics of UDP rather than bursty packet transmission characteristics of TCP that can introduce higher receiver side jitter. The multimedia UDP applications

either do not use any flow-control mechanism or use their own application-level flow control mechanisms that are rate-based rather than window based and tend to be greedier than that of TCP taking the multimedia Quality of Service (QoS) requirements into account.

In addressing the problem of fairness, there have been a strong argument that unresponsive or misbehaving flows should be penalized to protect well-behaved TCP flows[1]. As a result, various extensions to RED, such as Fair Random Early Drop (FRED), have been suggested [LM97]. FRED adds per-active-flow accounting to RED, isolating each flow from the effect of others. It enforces fairness in terms of output buffer space by strictly penalizing unresponsive or misbehaving flows to have an equal fair share while assuring packets from flows that do not consume their fair share are transmitted without loss. FRED serves its purpose not only in protecting TCP flows from unresponsive and misbehaving flows but also in protecting fragile TCP connections from robust TCP connections. However, the per-active-flow accounting is an expensive process that might slow down the performance of routers as the number of flow increase. FRED also has a potential problem that its TCP favored per-flow punishment could unnecessarily discourages flow-controlled interactive multimedia flows. Under FRED, incoming packets for a well-behaved TCP flow consuming more than their fair share are randomly dropped applying RED's drop rate. However, once a flow, although flow controlled, is marked as a non-TCP friendly flow, it is regarded as an unresponsive flow

[1] A *well-behaved flow* is defined as a flow that behaves like a TCP flow with a correct congestion avoidance implementation. A flow-controlled flow that acts different (or greedier) than well-behaved flow is a *misbehaving flow.*

and all incoming packets of the flow are dropped when it is using more than its fair share. As a result, a flow-controlled multimedia UDP flow, which may have a higher chance to be marked, will experience more packet loss than a TCP flow and be forced to have less than its fair share of bandwidth.

Recently, Jeffay *et al.,* [PJS99] proposes a new active queue management scheme called Class-Based Threshold (CBT), which releases UDP flows from strict per-flow punishment while protecting TCP flows by adding a simple class-based static bandwidth reservation mechanism to RED. In fact, CBT implements an explicit resource reservation feature of CBQ on a single queue that is fully or partially managed by RED without using packet scheduling. Instead, it uses class thresholds that determine ratios between the number of queue elements that each class may use during congestion. CBT defines three classes: tagged (multimedia) UDP[2], untagged (other) UDP and TCP. For each of the two UDP classes, CBT assigns a pre-determined static threshold and maintains a weighted-average number of enqueued packets that belong to the class. When a UDP packet arrives, the weighted-average for the appropriate class is updated and compared against the threshold for the class to decide whether to drop the packet before optionally applying the RED test. For the TCP class, CBT does not apply a threshold test but directly passes incoming packets to the RED test unit. Thus, by applying a threshold test to each UDP class, CBT protects TCP flows from unresponsive or misbehaving UDP flows, and also protects multimedia UDP flows from the effect of other UDP flows. CBT avoids congestion as well as RED, has less overhead and improves multimedia throughput and

packet drop rates compared to FRED. However, as in the case of CBQ, the static resource reservation mechanism of CBT could result in poor performance for rapidly changing traffic mixes and is arguably unfair since it changes the best effort nature of the Internet.

To eliminate the limitations due to the explicit resource reservation of CBT while preserving its good features from class-based isolation, we propose *Dynamic-CBT (D-CBT)*. D-CBT fairly allocates the bandwidth of a congested link to the traffic classes by dynamically assigning the UDP thresholds such that the sum of the fair share of flows in each class is assigned to the class at any given time. As in CBT, D-CBT categorizes flows into three classes. However, unlike the class categorization of CBT in which flow-controlled multimedia flows are not distinguished from unresponsive multimedia flows, we propose that only flow-controlled multimedia UDP flows are tagged leaving the unresponsive UDP flows untagged. We categorize UDP traffic this way, since multimedia applications are the primary sources of UDP traffic and generate high bandwidth flows. By differentiating flow-controlled multimedia flows from unresponsive flows, D-CBT not only protects flow-controlled multimedia flows from unresponsive multimedia flows, but also encourages multimedia applications to use congestion avoidance mechanisms.

In addition, as a means to improve multimedia jitter, we propose a lightweight multimedia-favored packet scheduling mechanism, *Cut-In Packet Scheduling* (ChIPS), as

[2] Tagged (multimedia) UDP flows can be distinguished from other (untagged) UDP flows by setting an

an alternative to FIFO packet scheduling under D-CBT and possibly under other RED like active queue management mechanisms. ChIPS monitors average enqueue rates of tagged and the other flows, and is invoked when the tagged flows are using a relatively smaller fraction of bandwidth than the TCP flows. On transient congestion in which the queue length is greater than the average queue length, ChIPS awards well-behaved (flow-controlled) multimedia flows by allowing their packets to "cut" in the line of queue where the average queue length points. We believe that the overhead of ChIPS will be small while significantly improving multimedia jitter when a relatively small number of multimedia flows are competing for the bandwidth with a large number of TCP flows.

To evaluate the proposed mechanisms, we use an event driven network simulator called NS (version 2) that simulates variety of IP networks [ns2]. NS implements most of common IP network components including RED. However, it does not have a traffic generator that simulates a flow-controlled multimedia application, which is essential to evaluate the proposed mechanisms. Therefore, we design and implement a general-purpose multimedia traffic generator and a trace-driven MPEG-1 video traffic generator that utilizes a rate-based flow control mechanism called media scaling, in which a transmission rate is determined by a media encoding and transmission policy pair that the source uses [DHH+93]. The multimedia applications, which show greedier bandwidth utilization behavior and better jitter performance than TCP under RED queue management, are used for the evaluation of D-CBT and ChIPS.

---

unused bit of the Type of Service field in the IP header (Version 4).

We also implemented CBT in NS, and ran a validation test that simulates an experiment on Jeffay's paper [PJS99]. By comparing our simulated result with Jeffay's experimental results, we not only validate that our CBT implementation is correct, but also confirm that the simulator implementations of RED and TCP behave similar to the ones used for the experiment. Then, we built D-CBT by extending the CBT implementation, add ChIPS into D-CBT, and compare the performance of D-CBT and D-CBT with ChIPS with that of RED and CBT. In the evaluation, our primary focus is on the effect of heterogeneously flow-controlled traffic on the behavior of the queue management mechanisms especially on fairness, and the effect of queue management on the performance of well-behaved (flow-controlled) multimedia flows.

The next chapter presents related work, and Chapter 3 presents our approaches in detail including the design and performance of the flow-controlled multimedia traffic generators, the validation of our CBT implementation on NS, and the details of D-CBT and ChIPS. Chapter 4 evaluates and analyzes D-CBT and ChIPS, Chapter 5 discusses future work, and Chapter 6 concludes the thesis.

2. Related Work

This chapter is intended to serve as a roadmap to work that is related to this thesis: flow control and perceptual quality of multimedia. We omitted router queue management mechanisms from this chapter, since they are introduced in Chapter 1 and also are discussed in detail throughout this thesis.

2.1 Flow Control

Van Jacobson and Michael J. Karels discussed the fundamentals of the TCP (Tahoe) flow control mechanism [JK88]. This paper presented the slow-start and congestion avoidance mechanisms of TCP, designed after the linear system theory that says that an unstable system can be stabilized by adding some exponential damping to its primary excitation. This paper also discussed the TCP's Round Trip Time (RTT) estimation method. For information on additional TCP (Reno) congestion control mechanisms, refer to RFC2001 [rfc2001].

Sally Floyd and Kevin Fall discussed potential unfairness and congestion collapse due to non-TCP friendly flows, defined as any flow with a long-term arrival rate that exceeds that of any conformant TCP in same circumstance [FF98], and insisted that these flows should be regulated. This paper also introduced methods to detect non-TCP friendly and unresponsive flows.

Hari Balakrishnan, Hariharan S. Rahul and Srinivsan Seshan introduced an end-system architecture, in which a Congestion Manager (CM) ensures proper congestion behavior and allows applications to easily adapt to network congestion [BRS99]. The main objective behind this work is to prevent users from misusing the network without proper congestion control mechanism by letting CM take care of flow control, while giving flexible options for the streams with different reliability requirements by separating congestion control from the function of loss recovery. Internally, CM used a window-based flow control algorithm, a scheduler to regulate transmissions and a lightweight protocol to elicit feedback from receivers. Their result showed that TCP connections using CM could effectively share bandwidth and obtain consistent performance, without adversely affecting other network flows. It also showed that CM enables audio applications to adapt to congestion conditions without having to perform congestion control or bandwidth probing on their own.

Luca Delgrossi *et al.*, presented a *media scaling* mechanism that is proposed to make the Heidelberg Transport System (HeiTS), a multimedia communication system for real-time delivery of digital audio and video, work with networks with no reservation mechanisms such as Ethernet [DHH+93]. Media scaling, which refers to a mechanism in which media encoding is modified according to the bandwidth available in the underlying networks, is essentially a rate-based flow control mechanism. However, in their study, media scaling was not used as a congestion control mechanism. Their paper discussed transparent and non-transparent scaling methods, and continuous and discrete scaling method that were used in the HeiTS system.

We used the above references to study the behavior of TCP and also to design a rate-based multimedia flow control mechanism, which uses media scaling as a mean to avoid congestion in response to network congestion. The simulator that we used to evaluate our proposed queue mechanisms does not support any flow-controlled multimedia traffic generators. Therefore, we built a general purpose and a trace-driven MPEG-1 traffic generators, of which the flow-control is modeled after fast recovery and congestion avoidance mechanisms of TCP (Reno), although it is rate-based rather than window-based. The design and implementation of our multimedia flow-control mechanism is presented in detail in Chapter 3.1.

2.2 Multimedia Performance

Jonathan Walpole *et al.,* introduced a media scaling mechanism that is based on an MPEG-1 video stream, which dynamically adapts to changes in available bandwidth [WKC+97]. The paper described in detail an MPEG-1 frame selection method, which carefully considered the frame dependencies, and discussed a resolution-changing method. However, their media scaling approach was more concerned about the end-system performance rather than its effect on the network. The purpose of their media scaling was described as to avoid waste of resources and improve perceptual quality by monitoring the end-to-end performance and selectively dropping packets at the sender, taking frame dependencies into account. It is reported that the method was effective in running streams over Internet connections of up to 28 hops and across networks with several orders of magnitude variation in available bandwidth. We used the MPEG-1

frame section policies described in their paper as the transmission policies for our MPEG-1 traffic generator, as described in Chapter 3.1.

Mark Handley measured Mbone multicast performance in 1997 [Ha97]. He examined routing tables to monitor route stability, and observed traffic as it arrived at sites (to which they had access) to look at individual packet losses. The research showed that 50% of receivers had a mean loss rate of about 10% or lower, while 80% reported a loss rate less than 20%. Around 80% of receivers have some interval during the day when no loss was observed. On the other hand, 80 % of sites reported some interval during the way when the loss rate was greater than 20%. About 30% of sites reported at least one interval where the loss rate was above 95% at some time during the day. The research also shows that although not as dominant as single loss, packet losses tend to occur in a bursty manner.

Vicky Hardman, Martina Angela Sasse and Isidor Kouvelas presented an audio piggybacking method, in which a low quality version of an audio frame is piggybacked onto the next frame so that it can be played when its high quality version of frame is lost [HSK98]. This study (a user study) shows that, using this method, a single copy of redundancy is good enough to provide packet loss protection for loss rates of up to 20-30%, assuming that packet loss follows a random pattern. Similarly, Yanlin Liu and Mark Claypool showed that video (MPEG-1) redundancy, which gives about 10% overhead, is a reasonable repair method when the loss rate is under 20% [LC00].

In this thesis, Handley's research result is used as a reference to real-world multimedia packet drop rates, and the result of the audio and video piggybacking redundancy studies is used as a rough guide line of what is the maximum packet drop rate that does not affect multimedia perceptual quality. We used these results to see whether the multimedia packet drop rate that D-CBT gives is reasonable, although the drop rate is also largely determined by multimedia traffic source behaviors and the amount of network traffic load.

Mark Claypool and Jonathan Tanner compared the effect of jitter and packet loss on perceptual quality [CT99]. In this study, packet loss (8%) and residual jitter (from the same trace) were induced into video clips based on Internet traces that sent simulated video from locations across the United States and New Zealand to Worcester, Massachusetts [GBC98], and perceptual quality was measured through a user study. The study showed that the effect of jitter on perceptual quality could be nearly as important as that of packet loss. Our approach to reduce multimedia jitter at network routers using ChIPS originates from and is supported by this user study.

3. Approach

Throughout the thesis, an event driven network simulator called NS (version 2) is used [ns2] to evaluate and compare the performance of Dynamic-CBT and ChIPS with that of RED and CBT. NS has most of the common IP network components implemented including TCP (Tahoe, Reno and Vegas) and UDP transport agents, and RED router queues. We implemented CBT, and our proposed Dynamic-CBT and ChIPS into NS. We also designed and implemented two traffic generators that simulate multimedia applications with an application-level flow control mechanism, and modified the IP packet header to represent tagged (multimedia) flows. This distinguishes well-behaved UDP flows (flow-controlled multimedia traffic) from unresponsive UDP flows at the IP level.

The first section of this chapter describes the design of the multimedia traffic generators, and shows their behavior when put together with TCP flows under RED queue management through simulation. This section also presents a simulation result showing that TCP is not the transport agent of choice for multimedia applications because it results in poor performance. The next section is devoted to describing and validating our CBT implementation. The last two sections describe in detail the design and implementation of Dynamic-CBT and ChIPS.

3.1 Flow-Controlled Multimedia

As mentioned briefly in Chapter 1, we are interested in the effect of queue management on performance of well-behaved (or flow-controlled) multimedia flows, as well as heterogeneously flow controlled traffics' effect on the behavior of the queue management mechanisms, especially on fairness. However, NS does not have a traffic generator that simulates a well-behaved multimedia application, which is essential to evaluate D-CBT. Also, the fact that evaluating the queue management mechanisms with well-behaved multimedia traffic generators would result in far more accurate than with unresponsive general-purpose CBR or VBR traffic generators already present in NS, pretending to be well-behaved, encourages us to build them in NS.

3.1.1 Design and Implementation

We designed and implemented in NS two slightly different multimedia traffic generators (or multimedia applications) that respond to network congestion using media scaling, based on the work in [DHH+93]. The two traffic generators have the same congestion control and avoidance (or flow control) mechanism, while the traffic they generate in response to congestion notification from the network is different. The first one, called MM-APP, reduces or increases transmission rate by decreasing or increasing the transmission interval with a fixed frame size. The second one, called trace-driven MPEG-APP (or MPEG-APP for short), changes the transmission rate by selecting frames to transmit from an input MPEG trace, where frame sizes vary while the transmission interval of frames in the file are fixed in terms of frames per second. The next couple of

paragraphs explain the congestion control and avoidance mechanism of the multimedia applications.

Before transmitting actual data, it is assume that the multimedia sender and the multimedia receiver agree on five scale values (0 to 4), each of which is assigned to a different media encoding method and transmission policy (i.e. which frame to transmit) pair. The scale value 0 is assigned to a set from which a predetermined minimum sustainable media quality can be achieved, the next value is assigned to sets from which a better media quality can be achieved, and so on. It is assumed that the media encoding and transmission policy sets are carefully chosen so that the transmission rates resulting from the sets increase linearly as the scale value increases. Table 3.1.1 shows an example assignment.

Scale Value	Media Encoding and Transmission Policy Set	Estimated Average Transmission Rate (Kbps)
4	A	1100
3	B	900
2	C	700
2	D	500
1	E	300

<Table 3.1.1> Example Media Scale Assignment

Thus, having five discrete and linearly increasing transmission rates assigned to the scale values, the sender starts from scale 0 transmitting at the lowest rate. In designing the multimedia applications, most of the congestion control and avoidance mechanisms are

placed as receiver side functions. More precisely, the receiver detects congestion, determines the next transmission rate of the sender in terms of scale value, and notifies the sender of this scale value. The sender, being notified of the scale value, simply changes the transmission rate by using media encoding and transmission policy assigned to the scale value.

In detecting congestion, the receiver uses frame loss as the network congestion indicator. There are two circumstances where the receiver claims frame loss. The first is when the receiver gets a frame whose sequence number is greater than the expected sequence number. The second is when the receiver times out by not receiving any frames within a timeout interval. Determining this timeout interval is often difficult in a dynamic network and also can greatly affect the behavior of the congestion control mechanism. A timeout interval that is set too short will claim false frame loss, which will make the sender to reduce the transmission rate needlessly. On the other hand, a timeout interval that is set too long will fail to detect multiple sequential frame loss effectively such that the sender reduces transmission rate later than other competing connections, which could result in an unfair allocation of bandwidth. In our implementation, when the Round Trip Time (RTT) of the connection is greater than the longest possible frame transmission interval (Max_Interval), the timeout interval (TOI) is set to RTT. Otherwise, TOI is set to Max_Interval + α, where α is a value between 0 and RTT.

```
RTT > Max_Interval: TOI = RTT
RTT ≤ Max_Interval: TOI = Max_Interval + α (where 0 < α < RTT)
```

The receiver, when detecting congestion, reduces its scale value to half (integer division) and notifies the sender of this value by sending a small packet. The receiver, when no network congestion is detected within a RTT from the last checkpoint, increases the scale value by one and notifies the sender of this value. This design of drop scale to half at congestion, and increase one scale up at a RTT is motivated by the fast recovery algorithm that is found in TCP Reno implementations [FF96] and the TCP-friendly definition: "TCP-friendly assumes TCP can be characterized by a congestion response of reducing its congestion window at least by half upon indication of congestion, and of increasing its congestion window by a constant rate of at most one packet per round trip time" [FF98].

In fact, the congestion control mechanism of our multimedia applications, although its exponential back off mechanism is conforming to linear system theory that says "if a system is stable, the stability is exponential" [JK88], could be greedier than that of TCP because of the following reasons. First, reducing the scale value to one-half at congestion is not necessarily dropping the transmission rate in half, and increasing one scale value at a RTT can increase more than one packet within a RTT. Second, the multimedia applications always transmit frames at least at the lowest rate and never time out. When the available network bandwidth is decreased, TCP reduces transmission rate by decreasing its congestion window size, which could make the connection fragile. Multimedia applications, on the other hand, as the available network bandwidth gets lower and lower, have a congestion control mechanism operating at low scales that does not effectively respond to network congestion, since it cannot decrease transmission rate

exponentially or at all. Because of the above difference, the multimedia streams can unfairly use bandwidth over the competing TCP connections.

The last reason is that the multimedia flow-control mechanism is based on the transmission rate while TCP is based on the congestion window. The rate-based multimedia applications could get greedier than TCP as round trip time (or end-to-end delay) is increased. Consider a situation that the two differently flow-controlled streams competing in an IP network with high end-to-end delay. When congestion occurs, the network notifies the senders of this event by dropping packets or marking the ECN bit [Fl94]. However, it will take a while (at least half the RTT) for the senders to be notified this event. Meanwhile, the multimedia applications with the rate-based mechanism keep sending packet at the last rate. However, the congestion window based TCP agents after sending packets up to the size of the congestion window wait for an ACK packet to arrive. Assuming that the end-to-end delay is large enough to make flows significantly bursty (less responsive) and to make the sender's waiting time significantly large, the flows with the rate-based mechanism transmit considerably more packets during each congestion notification delay than the TCP flows. Moreover, the additional multimedia packets injected in the network can cause greater congestion, in which bursty TCP flows could be more severely punished than rather continuous multimedia flows by sequential packet drops. This effect would be enlarged as the available bandwidth share is decreased due to the increased number of active flows, and as the ratio between the number of multimedia flows and the number of TCP flows is increased.

Now that the congestion control part of the multimedia traffic generators has been examined, the specifics of how MM_APP and MPEG_APP generate traffic associated with scales are presented below. As briefly mentioned earlier, MM_APP directly associates transmission rates to scale values without particular media encoding and transmission policy pairs in mind. It assumes that every media encoding and transmission policy pair associated to the scale values generates traffic with a fixed frame size. In other words, it assumes that the transmission intervals are the only factor that causes the rate changes. Therefore, the resulting traffic can be characterized by CBR traffic of fixed frame size and various transmission intervals associated with scale values. Although not targeted at a specific multimedia application, MM_APP is useful in that it is easy to change rate associated to scale values and to test the media scaling scheme eliminating the effect of a specific traffic characteristic of a particular application.

MPEG_APP, on the other hand, simulates a very specific video application that is based on MPEG-1 encoding scheme [MPFG97]. It implements five sets of MPEG-1 encoding and transmission policies and associates them with scale values – in fact, it only changes the frame transmission policy leaving the encoding scheme unchanged. MPEG-1 encodes motion pictures at a given interval and quality. It generates a stream of inter-dependent frames of types, I, P and B, in a specific pattern such as IBBPBBPBB. Among the three frame types, only I-frames can be decoded on their own. The decoding of a B-frame relies on a pair of I-frame and/or P-frame that comes before and after the B-frame and the decoding of a P-frame relies on an I-frame or P-frame that comes before the P-frame. MPEG_APP supports MPEG-1 streams using the common pattern of

IBBPBBPBB and IBBPBBPBBPBB. Table 3.1.2 shows the transmission policies on the two stream patterns, which are carefully selected keeping the frame dependencies in mind [WCK+97].

Scale	Transmission Policy (Pattern 1)	Transmission Policy (Pattern 2)
4	I B B P B B P B B I	I B B P B B P B B P B B I
3	I B P B P B I	I B P B P B P B I
2	I P P I	I P P P P I
1	I P I	I P I
0	I I	I I

<Table 3.1.2> MPEG Transmission Policy Associated with Scale Values

The MPEG_APP gets a MPEG-1 trace file that contains frame information for a stream with the maximum frame rate (scale 4) as input along with the maximum frame rate and the longest possible frame transmission interval that is used for congestion detection at the receiver side. At every scale 4 transmission interval, MPEG_APP reads the frame information from the input file, and determines whether or not to transmit this frame using the current scale value and the transmission policy associated with the scale value. Figure 3.1.1 (a) shows an example input file that contains frame information for 30 frames per second IBBPBBPBB pattern stream in which the sizes of I-frames P-frames and B-frames are 11 KB, 8 KB and 2 KB. The frame sizes used in this example are the mean frame size of each type obtained while playing a short, high-quality MPEG-1 news clip. Figure 3.1.1 (b) shows each transmission policy assigned to scale values with the estimated average transmission rate for the input trace file.

```
I     11000
B     2000
B     2000
P     8000
B     2000
B     2000
P     8000
B     2000
B     2000
I     11000
.     .
.     .
```

(a) Input Trace File (bytes)

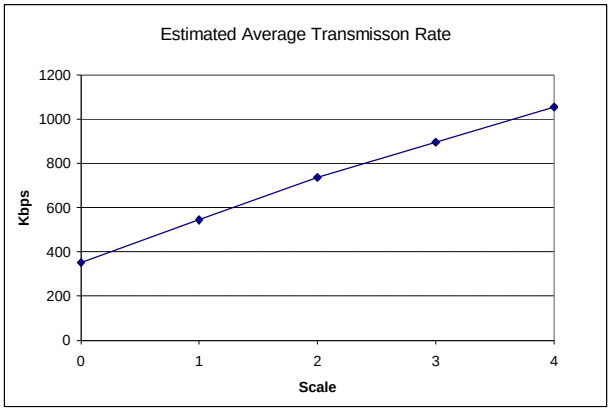
Maximum Frame Rate (Scale 4) = 30 frame/sec

Scale	Transmission Policy (Pattern 1)	Estimated Average Transmission Rate (Kbps)
4	I B B P B B P B B I	1056
3	I B P B P B I	896
2	I P P I	736
1	I P I	544
0	I I	352

(b) Transmission Policies and Associated Rates

<Figure 3.1.1> Example Input Trace File and Average Transmission Rates

Figure 3.1.2 visualizes the estimated transmission rate in Figure 3.1.1 (b). The almost linearly growing estimated average transmission rates shows that the assignment of the transmission policies to the scale values works well with the given example MPEG-1 stream. This is because the linear increment of scale results in a linear increment of transmission rate and the exponential decrement of scale results in exponential decrement of transmission rate.

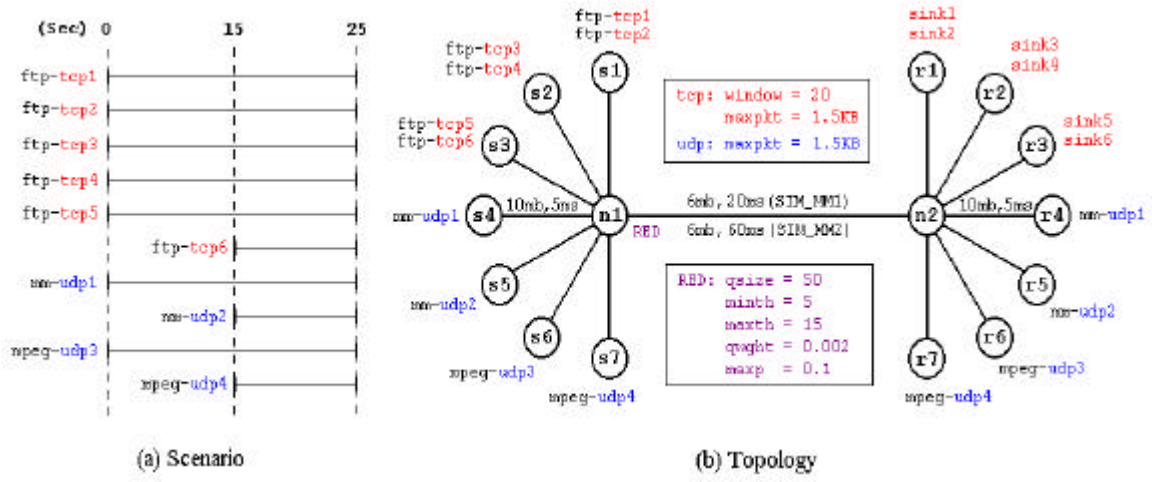


<Figure 3.1.2> Estimated Average Transmission Rate of the Example in Figure 3.1.1

The transmission rate resulting from the transmission policies assigned for the scale value for the given MPEG-1 encoded news clip stream grows linearly as scale value increases.

3.1.2 Simulation Test and Analysis

We ran a series of simulations to validate the NS implementation of MM_APP and MPEG_APP, and to measure their fairness when competing for bandwidth with TCP flows under RED queue management. Here, we present the results of two simulations that exhibit the behavior of MM_APP and MPEG_APP. The simulations, referred to as SIM_MM1 and SIM_MM2, are designed to show the effect of available bandwidth and end-to-end delay on fairness. Figure 3.1.3 shows the network topology and the application scenario used for the simulations. Note that the only difference between SIM_MM1 and SIM_MM2 is the delay for the link that connects the two network nodes (*n1-n2*).



<Figure 3.1.3> SIM_MM1 and SIM_MM2 Scenario and Topology

Each link that connects a source or destination node and a network node is set to have 10 Mbps link capacity and 5ms of delay. The link that connects the two network nodes is set to have 6 Mbps of link capacity and 20ms of delay for SIM_MM1 and 60ms of delay for

SIM_MM2. The network node *n1* uses RED queue management, for which the parameter set used, shown in Figure 3.1.3 (b), is chosen from one of the sets that are recommended by Floyd and Jacobson [RK99].

For traffic sources, 6 FTP, 2 MM_APP and 2 MPEG_APP traffic generators are used, where FTP uses TCP Reno and the others use UDP as the underlying transport agent. All the TCP agents are set to have a maximum congestion window size of 20 packets and a maximum packet size of 1500 bytes. The UDP agents are also set to have a maximum packet size of 1500 bytes. The MM_APP traffic generators use the transmission rates shown in Table 3.1.1, that is 300, 500, 700, 900, 1100 Kbps, for scale 0 to 4 transmission rates. The MPEG_APP traffic generators use the transmission policies and the trace file shown in Figure 3.1.1, which generates traffic rates of 352 Kbps to 1056 Kbps.

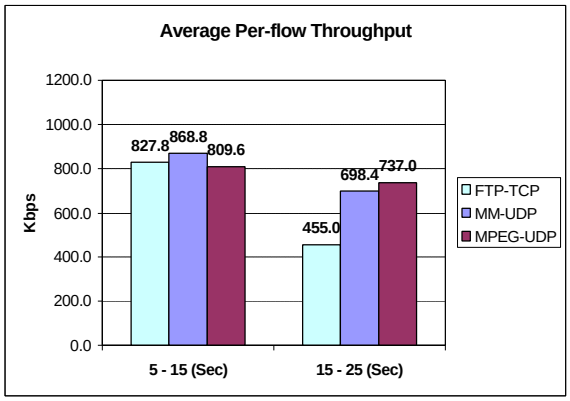
Both simulations start with five FTP, and 1 MM_APP and 1 MPEG_APP in action, and at 15 seconds the remaining traffic sources join. For the first 15 seconds, the available bandwidth share for each connection is about 857 Kbps, and for the next 10 seconds, the share goes down to 600 Kbps. Figure 3.1.4 and Figure 3.1.5 in the next two pages show the results of SIM_MM1 and SIM_MM2. For the throughput measurements, we omitted the first 5 seconds to eliminate the effect of the initially unstable TCP and RED behaviors on the fairness.

<Figure 3.1.4> SIM_MM1 – Simulation Test of MM_APP and MPEG_APP #1

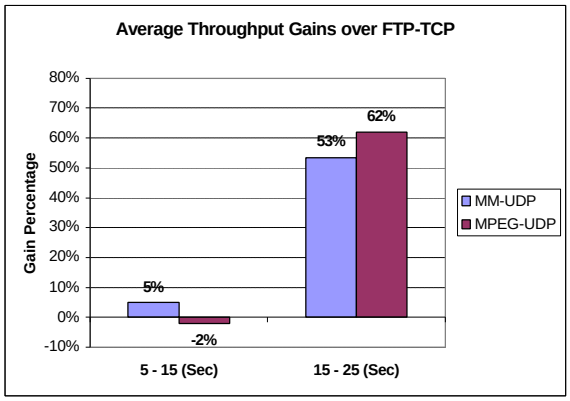
This simulation is set to have the FTP applications over TCP and the flow-controlled multimedia applications to fairly share the output bandwidth for the first 10 seconds (5 to 15 seconds). For the next 10 seconds when three more sources join decreasing the available bandwidth share, the multimedia applications get more bandwidth than the TCP flows.

	5 – 15 (Sec)	15 – 25 (Sec)
FTP-TCP1	733.2	505.2
FTP-TCP2	888.0	464.4
FTP-TCP3	717.6	552.0
FTP-TCP4	958.8	412.8
FTP-TCP5	841.2	428.4
FTP-TCP6		367.2
MM-UDP1	868.8	669.6
MM-UDP2		727.8
MPEG-UDP1	809.6	826.8
MPEG-UDP2		647.2
Fair Share	857.1	600.0

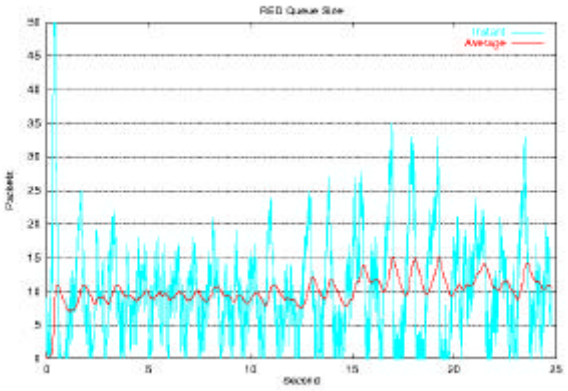
(a) Per-flow Throughput Data (Kbps)



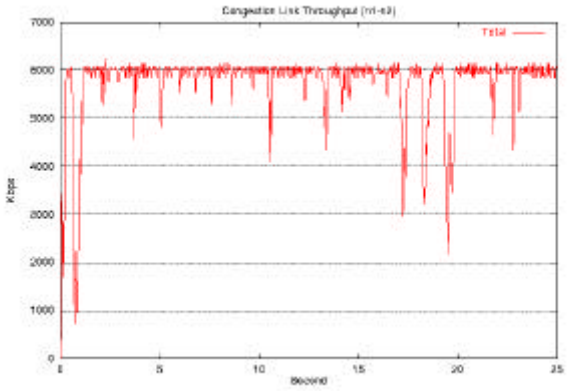
(b) Average Per-flow Throughput



(c) Throughput Gains over FTP-TCP



(d) RED Queue Size



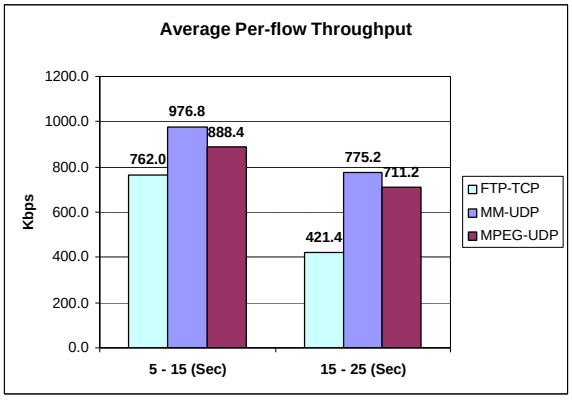
(e) Congested Link Throughput

<Figure 3.1.5> SIM_MM2 – Simulation Test of MM_APP and MPEG_APP #2

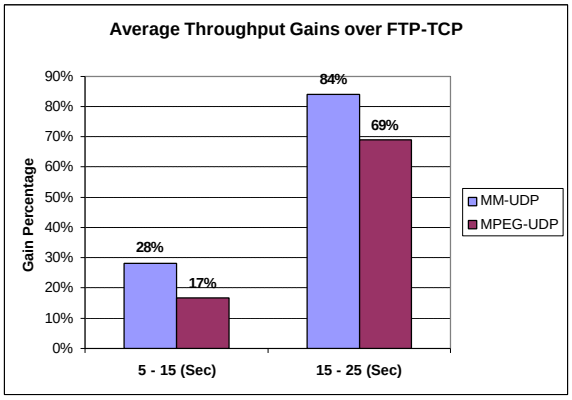
SIM_MM2 shows the effect of a longer end-to-end delay on fairness. The only difference from SIM_MM1 is that the network link is set to have longer delay. Comparing the throughput gain over FTP-TCP graph with that of Figure 3.1.4, one can easily see that the multimedia applications become greedier as the end-to-end delay is increased.

	5 – 15 (Sec)	15 – 25 (Sec)
FTP-TCP1	910.8	464.4
FTP-TCP2	734.4	453.6
FTP-TCP3	952.8	324.0
FTP-TCP4	493.2	465.6
FTP-TCP5	718.8	268.8
FTP-TCP6		552.0
MM-UDP1	976.8	804.0
MM-UDP2		746.4
MPEG-UDP1	888.4	699.2
MPEG-UDP2		723.2
Fair Share	857.1	600.0

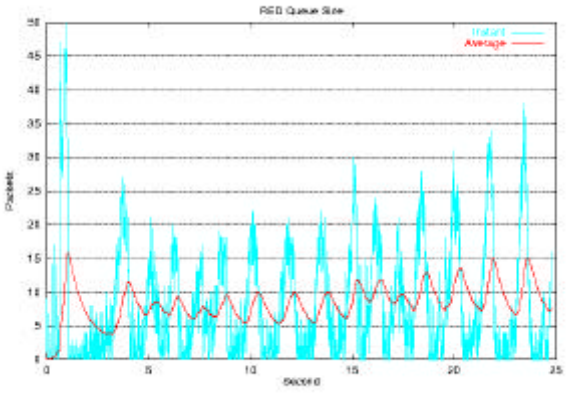
(a) Per-flow Throughput Data (Kbps)



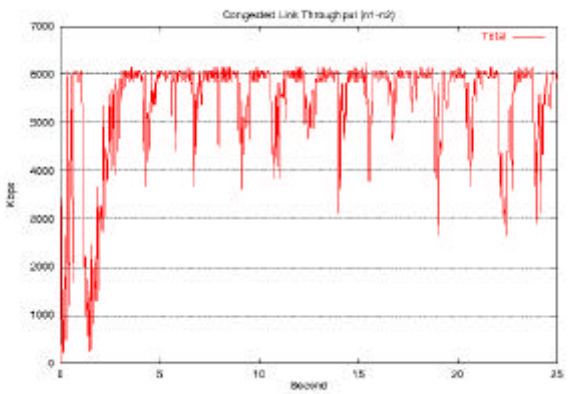
(b) Average Per-flow Throughput



(c) Throughput Gains over FTP-TCP



(d) RED Queue Size



(e) Congested Link Throughput

In SIM_MM1, as shown in Figure 3.1.4 (c), the multimedia streams that consume about the same amount of bandwidth as the TCP streams during the first 10 seconds become greedier as the available fair bandwidth share for each flow goes down to 600 Kbps for the next 10 seconds. Although difficult to pinpoint, the unfairness is mainly caused by the average reduction of TCP's congestion window size, and the widely fluctuating and higher RED's average queue length that is shown in Figure 3.1.4 (d). As discussed in the earlier section that describes the flow control mechanism of MM_APP and MPEG_APP, reducing the congestion window size makes the TCP connection more fragile. This could result in TCP (Reno) performing more slow starts than fast recovery, and even timeouts, especially when RED applies higher early-drop rates as the average queue length gets high.

SIM_MM2 exhibits the "end-to-end delay effect on fairness" discussed in the section that describes the multimedia flow control mechanism. As shown in the queue size graphs and congested link throughput graphs of SIM_MM1 and SIM_MM2 (graphs (c) and (d)), as the end-to-end delay is increased the router experiences more bursty packet reception, because the congestion notification delay is also increased. Comparing the throughput gain graphs in Figure 3.1.4 and Figure 3.1.5, one can easily see that the longer end-to-end delay significantly increases the unfairness. The following calculates and compares the number of transmitted packet during a congestion notification delay for a TCP and a MM_APP in SIM_MM2.

Network Packet Size (NPS) = 12 Kbits (1500 Bytes)
Network Link Capacity (NLC) = 6 Mbps
Average Queue Size (AQS) = 7 and 10 Pkts (for the first and second period)

Queuing Delay (QD) = AQS * NPS / NLC = 0.014 or 0.020 Sec
Total Link Delay (TLD) = Link Delay (SRC $\rightarrow$ DST) = 0.070 Sec
Feedback Link Delay (FLD) = Link Delay (Router $\rightarrow$ DST $\rightarrow$ SRC) = 0.135 Sec

End-to-End Delay (EED) = QD + TLD = 0.084 or 0.090 Sec
Congestion Notification Delay (CND) = QD + FLD = 0.149 or 0.155 Sec

TCP Congestion Window Size for *X* Kbps of transmission rate = CWS(*X*) = *X* * EED / NPS
MM Transmission Interval for *Y* Kbps of transmission rate = MTI(*Y*) = NPS / *Y*

Assuming that congestion occurred and notified to both the TCP and MM, the following is the number of packets each source injects into the network during the congestion notification delay, when they are transmitting at 857 Kbps and 600 Kbps.

# of TCP packets injected during CND (TR = 857Kbps) = CWS(857) = 6.0 Pkts
# of MM packets injected during CND (TR = 857Kbps) = CND / MTI(857) = 10.6 Pkts

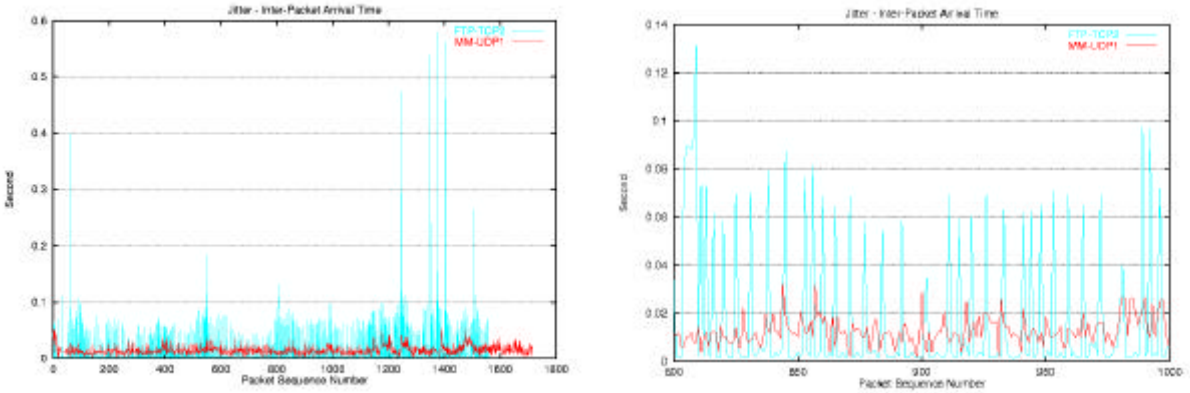
# of TCP packets injected during CND (TR = 600Kbps) = CWS(600) = 4.5 Pkts
# of MM packets injected during CND (TR = 600Kbps) = CND / MTI(600) = 7.8 Pkts

The following are the assumptions made in the calculation: First, the TCP and the MM_APP are currently transmitting at the same transmission rate, did not change the current transmission load in the near past, and also do not change the load before they are notified of congestion. Second, they reduce the same amount of transmission load in responding to the congestion, recover to the previous load after the same amount of delay, and never get additional congestion notification before fully recovering to the

previous state. Third, the TCP and the MM_APP are able to adjust the window size or the transmission interval exactly to the given transmission rate, although this is not possible since they discretely adjust window sizes or transmission intervals. Last, to make the calculation easy, it is assumed that ECN marking is used for congestion notification (although packet drop is used in the simulation). The calculation is made for the transmission rate of 857 Kbps that is the fair share during the first period and for the rate of 600 Kbps that is the fair share during the second period. Under SIM_MM2, whenever both the TCP and MM_APP packets are marked for the network congestion, MM_APP injected 4.6 and 3.3 more packets to the network during the notification delay before reducing the transmission load for the first and the second period. This gives an instant bandwidth gain of 370 Kbps and 255 Kbps for the multimedia applications during the notification delay of 0.149 and 0.155 second. Assume that the TCP and the MM_APP received the same number of congestion notifications at the same time, and the notification delay total takes half the total simulation time. The MM_APP will get 185 Kbps of bandwidth more than the TCP in the first period, and will get 146 Kbps more in the second period. Although this observation is based on the assumptions that would never happen in a real life, it gives an intuition of how end-to-end delay could contribute the unfairness.

SIM_MM1 and SIM_MM2 showed that the multimedia applications that use the rate-based flow control mechanism could be greedier than TCP agents that use window-based flow control mechanism. It is likely that this is not only the problem with the flow control mechanism we build, but with most of rate-based mechanisms. In general,

interactive multimedia applications today would not want to use TCP as the underlying transport agent, but use UDP with no or their own flow control mechanism that is possibly rate-based or window-based with transmission scheduling [BRS99]. Figure 3.1.6 that shows the TCP and MM_APP jitter in terms of inter-packet arrival time gives an explanation why.



<Figure 3.1.6> FTP-TCP and MM-UDP Jitter under RED (Inter-Packet Arrival Time)

TCP's bursty transmission policy that transmits data packets up to the minimum of congestion and receiver side window at a time without using any particular transmission scheduling introduces high jitter compared to the continuous transmission policy. Furthermore, from time to time, TCP's timeouts add huge peaks. Delay sensitive interactive multimedia applications degrade in quality under high jitter, since they need to play out the multimedia data at a specific time interval. They could use relatively large buffers to normalize the jitter effect, however this gives extra delays that might not satisfy the users of the application. In addition, the fact that TCP does not separate flow control from loss recovery discourages multimedia applications from using it [BRS99],

since this gives possibly unbounded transmission delays for a multimedia packet that is useless after a specific period due to the retransmission attempts for the previous packets.

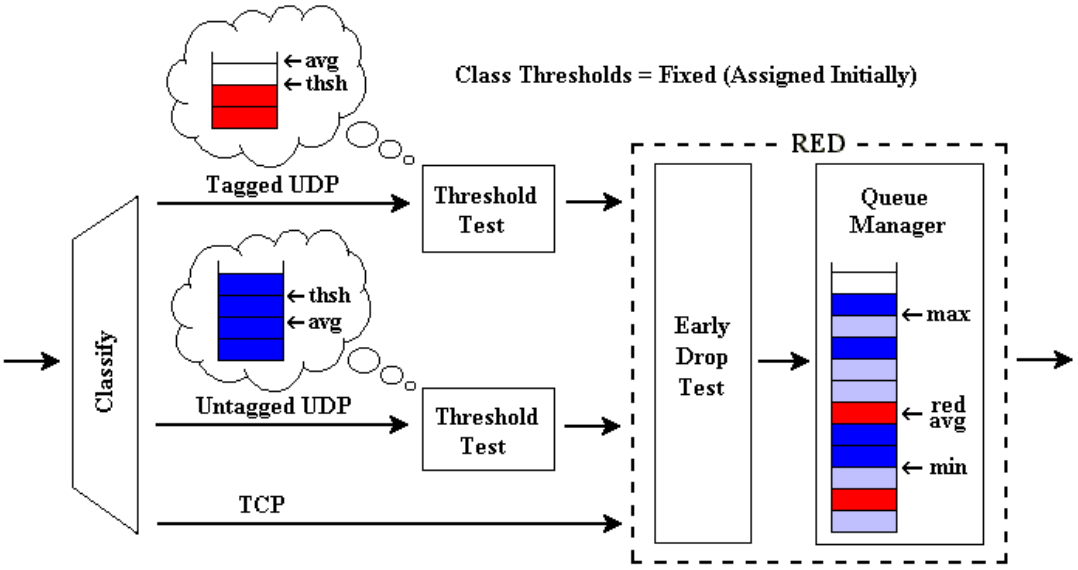
In this section, we described the design and showed the behavior of the multimedia applications (or traffic generators) that use a flow-control mechanism based on media scaling, which we built and used to evaluate the proposed router queue management and scheduling mechanisms. As shown in the simulation results, the multimedia applications become greedier than TCP when the round trip time is relatively high and the available bandwidth share is relatively low. This is mainly due to the transmission characteristics of the rate-based mechanism and the minimum bandwidth requirements. We also demonstrated that TCP is not the transport agent of choice for multimedia applications by showing its poor performance on jitter and delay that is due to its transmission policy. The next section presents the implementation and validation of CBT on NS.

3.2 CBT Implementation and Validation

As briefly mentioned in Chapter 1, the main idea behind the design of CBT [PJS99] is applying class-based isolation on a single queue that is fully or partially managed by RED without using packet scheduling. Instead of using packet scheduling on multiple logical queues, CBT regulates congestion-time output bandwidth for *n* classes of flows using a RED queue management mechanism and a threshold for each of the *n-1* classes of flows, which is average number of queue units that a class may use. CBT categorizes flows into three classes, which are TCP, tagged (multimedia) UDP and untagged (other) UDP, and assigns a pre-determined static threshold for each of the two UDP classes, assuming that UDP flows are mostly unresponsive or misbehaving and need to be regulated. When a UDP packet arrives, the weighted-average for the appropriate class is updated and compared against the threshold for the class to decide whether to drop the packet before passing it to the RED algorithm. For the TCP class, CBT does not apply a threshold test but directly passes incoming packets to the RED test unit. This is Jeffay's first design of CBT, called "CBT with RED for all". The conceptual view of the first CBT design is shown in Figure 3.2.1 in the next page.

In Jeffay's second design, called "CBT with RED for TCP", only TCP packets are subjected to RED's early drop test, and UDP packets that survive a threshold test are directly enqueued to the outbound queue that is managed by RED. Another difference from the first design is that the RED's average queue size is calculated only using the number of enqueued TCP packets. CBT with RED for TCP is based on the assumption that tagged (multimedia) UDP flows as well as untagged (other) UDP flows are mostly

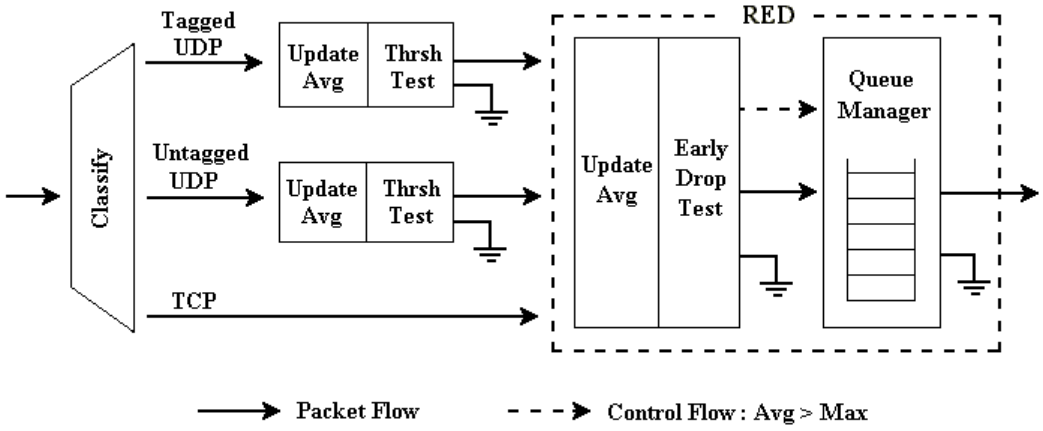
unresponsive, and it is of no use to notify these traffic sources of congestion earlier. CBT with RED for TCP improves the UDP packet drop rate by not applying the early drop test, and thus improves multimedia performance compared to the first design [PJS99]. We implemented CBT with RED for all into NS, from which our proposed mechanism D-CBT is extended, and validated the implementation by simulating an experiment from Jeffay's paper of which title is *"Lightweight Active Router – Queue Management for Multimedia Networking"* [PJS99]. This section presents the implementation and validation. In the rest of this thesis, CBT refers to CBT with RED for all.



<Figure 3.2.1> CBT (with RED for all) Conceptual View

CBT recognizes the three different classes of flows at the IP level. In IP, this can be done by using a protocol field that tells it what the transport protocol is and an unused bit of type-of-service field in IPv4 header [PJS99, NJZ97]. Likewise, we used the protocol field variable in the header structure to distinguish UDP packets from TCP packets.

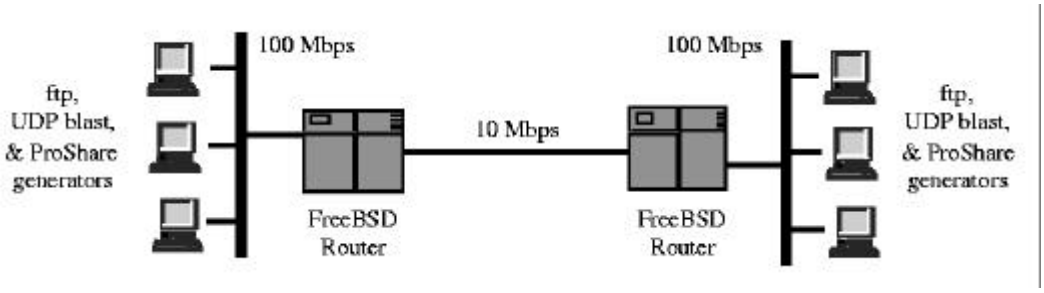
However, to distinguish tagged UDP from untagged UDP, we decided to use a priority variable, and modified the UDP agent to set the priority variable of the outgoing packets to 15 when requested by the multimedia application (or traffic generator). This was because the type-of-service field parameter was not defined in the original NS IP header definition, and we did not want to modify the header structure. Indeed, the NS IP header has the IPv4 parameters (but not all) and also has IPv6 parameters such as a priority field at the same time. Therefore, it is up to users which version of the header to assume.



<Figure 3.2.2> CBT (with RED for all) Implementation Diagram

We took the RED implementation in NS and extended it to CBT (with RED for all) as described in Jeffay's paper [PJS99]. The CBT implementation is shown in Figure 3.2.2. When a packet arrives, it is classified into one of the three classes. For an incoming UDP packet, an appropriate class average is updated and the class' threshold test is applied. If the packet survives the threshold test, it is given to the RED unit. An incoming TCP packet is directly given to the RED unit. As a packet, either a UDP packet that survived a

threshold test or a TCP packet, arrives the RED unit updates the average queue length and applies its early drop test on it, in which one of three possible decisions is made. First, when the average is less than the minimum threshold, the packet is given to the queue manager. Second, when the average is greater than the minimum and less than the maximum threshold and the packet is selected for an early drop, the early drop test unit drops the packet. Third, when the average is greater than the maximum threshold, the unit passes the packet to the queue management unit and signals congestion, where the queue management unit enqueues this packet at the end, randomly selects a packet in the queue and drops it [3]. When the actual queue overflows, the queue manager performs this random drop without a congestion signal from the test unit.

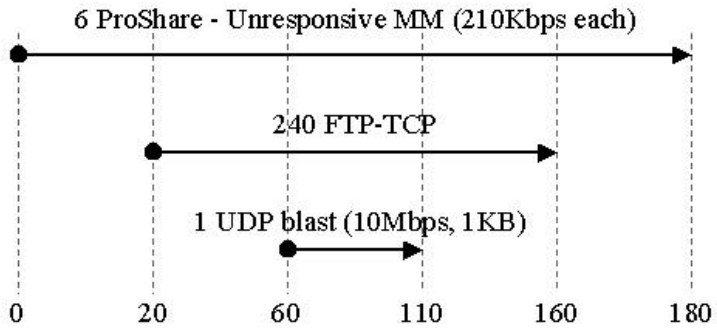


<Figure 3.2.3> Jeffay's Experimental Network

After the implementation, we validated the RED and the CBT implementations by simulating Jeffay's experiment and comparing his RED and CBT results with the simulated ones. Figure 3.2.3 shows his experimental network consisted of two switched 100 Mbps Ethernet LANs (source and destination LAN) that were interconnected by a

[3] This is not the behavior of standard RED. Standard RED drops all incoming packets when the average is greater than the maximum threshold.

full-duplex 10 Mbps Ethernet. For traffic sources, he used 6 ProShare multimedia applications that generate approximately 210 Kbps of traffic, 240 FTP-TCP bulk transfers and 1 UDP non-multimedia application that sent 1 Kbyte packets at 10 Mbps (the maximum capacity of the network), of which the scheduling is shown in Figure 3.2.4. To simulate a large RTT for the TCP connections, the kernel on each machine was modified to introduce a delay in transmitting packets from each connection. Refer to [PJS99] for the further experiment setup details.



<Figure 3.2.4> Traffic Source Schedules

Similarly, our simulation used the network topology of 100 Mbps source and destination Ethernet LAN, each of which consisted of 248 nodes (147 hosts and a router), connected by a 10 Mbps link. However, instead of introducing a large delay in the nodes, we give the 10Mbps link the delay of 20ms. For the traffic source, we assumed many of the parameter settings, since no detailed source behavior or parameter settings are described the paper. To simulate the ProShare multimedia application, we used a CBT traffic generator that transmitted 6 Kbytes frames at the rate of 210 Kbps, each of which was broken into six 1 Kbyte tagged packets at the underlying UDP layer. For the TCP (Reno)

connection, we set the maximum packet size to 300 Bytes and the maximum congestion window size to 10 packets.

- RED Settings:

  qsize = 60 pkts
  max-th = 30 pkts
  min-th = 15 pkts
  qweight = 0.002 (not specified in the paper)
  max-pro = 0.1 (not specified in the paper)

- CBT Settings: the above settings plus

  mm-th = 10 pkts
  udp-th = 2 pkts

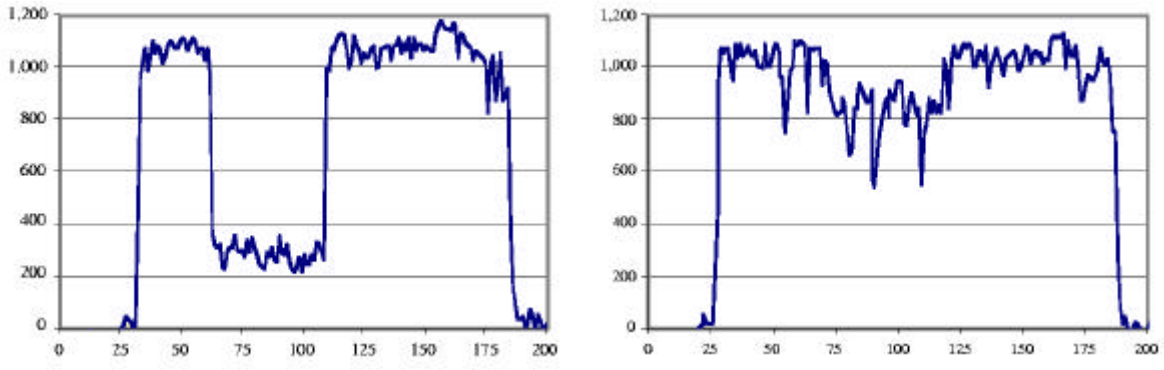
<**Table 3.2.1**> RED and CBT Settings

Table 3.2.1 shows the RED and CBT parameter settings used for the experiment and the simulation. Each router is set to have a 60 packet long output queue. For RED, the maximum and minimum thresholds were 30 packets and 15 packets, the weight was 0.002 for the weighted queue average calculation, and the maximum early drop probability was 0.1. For CBT, in addition to the RED settings, the tagged UDP class threshold (mm-th) was set to 10 packets, and the untagged UDP class threshold (udp-th) was set to 2 packets. These class thresholds were determined based on an assumption that the average of the average queue length would be about 26 packets throughout the experiments (or the simulation), and all packets are of the same size. By assigning the maximum average of a 10-packet space of the queue to the tagged flow class, it approximately gets the maximum bandwidth of 3.8 Mbps, 10/26 of the network link bandwidth, at congestion. This was enough for the 6 ProShare flows whose aggregated

average bandwidth requirement was approximately 1.3 Mbps. Similarly, for the untagged UDP flows, by assigning the average of a 2-packet space the class for congestion, approximately 0.8 Mbps of output was reserved for the class. From the above calculation, the aggregated throughput of TCP, which should have been about 8.7 Mbps (or 1088 Kbytes/Sec) when TCP flows are sharing the bandwidth with the ProShare flows, should have dropped down to approximately 7.9 Mbps (or 988 Kbytes/Sec) when the UDP blast joins. However, TCP flows got a little bit less bandwidth than the calculation due to their smaller packet size. By using smaller packets, TCP agents would operate in a larger congestion window size and transmit more packets into the network resulting in more TCP packet drops at the router. However, since the TCP bandwidth loss due to a packet drop was reduced, the TCP bandwidth loss was not that much.

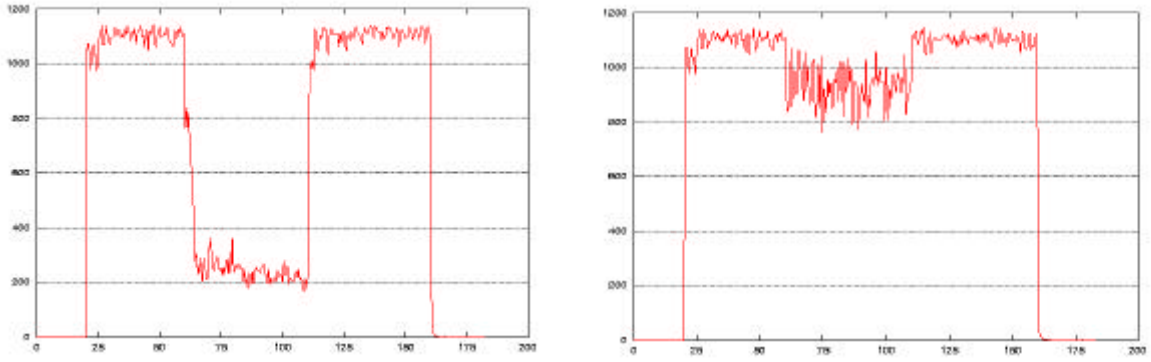
Figure 3.2.4 shows the aggregated TCP throughput (in Kbytes/Sec) under RED and CBT for the experiment and the simulation – the throughput graph for the experiments is copied from Jeffay's paper. By comparing the experimental result and the simulated result under RED, one can see that the traffic source parameter settings we guessed for the unspecified ones are similar to the settings used in the experiments. Also, the results imply that the TCP and RED implementation on NS are comparable to the ones used in the experiment. Next, the aggregated TCP throughputs under CBT for the experiment and the simulation show that our NS implementation is correct, although the experimental result shows several sharply dropping aggregated TCP throughputs during the period

when the UDP blast joins. We believe that this phenomenon is caused by experimental "noises" from the real world.



RED CBT with RED for all

(a) Jeffay's Experimental Results



RED CBT with RED for all

(b) NS Simulated Results

<Figure 3.2.4> Aggregate TCP Throughput (X-axis = Seconds, Y-axis = Kbytes/Sec)

Comparing the experimental and the simulated results carefully, one would notice the delayed throughput of the experiment at the beginning and at the end. This could be

simply from the fact that Jeffay used a transmission source scheduling that is slightly different than shown in Figure 3.2.3. Assuming that the same schedule is used, it is possible that this difference comes from the mechanism in the host kernels that introduces a delay for TCP packet transmissions could be set to have a higher delay than that of the network link in the simulation. Another possible factor is the routing delays of the routers in the experiment. In fact, the router implementation in NS does not have a routing delay and thus does not have an inbound queue. It appears that each router in the experiment is set to have a large inbound queue. In this situation, when a severe overload occurs in the inbound link that is considerably off the router's switching capacity (or processing delay), the incoming packets wait a significant time in the inbound queue to be processed or even dropped. At the period when TCPs start transmission the overload occurs due to the slow start algorithm with the relatively large congestion window sizes. Likewise, it seems that when the UDP blast starts, a very significant overload occurred at the inbound link. These effects, we believe, contributed to the delayed transmission at the router, and possibly the sharply dropping aggregated TCP throughput during the period of 60 to 110 seconds.

This section described our CBT implementation on NS, and the validation of the TCP, RED and the newly added CBT implementations. By simulating the Jeffay experiment and comparing the results with the experimental results, we showed that the TCP, RED, and the CBT implementations are comparable to the ones used in his experiments. The next section presents the design and implementation of our proposed mechanism, D-CBT and ChIPS.

3.3 Dynamic-CBT and Cut-In Packet Scheduling

Dynamic-CBT (D-CBT) is an active queue management mechanism that is derived from CBT and based on RED, and *Cut-In Packet Scheduling (ChIPS)* is a lightweight multimedia favored packet scheduling mechanism that can be used under RED and its variants. While the two mechanisms can be used separately, we implemented ChIPS into D-CBT and measured the performance of D-CBT w/o ChIPS and D-CBT w/ ChIPS. This section presents the design and implementation of D-CBT and ChIPS.

3.3.1 Dynamic-CBT (D-CBT)

D-CBT is an extension of CBT (with RED for all) that enforces fairness among classes of flows, and gives UDP classes better queuing resource utilization. Figure 3.3.1 in the next page shows the design of D-CBT. The key difference from CBT is (1) the dynamically moving fair thresholds and (2) the UDP class threshold test that actively monitors and responds to RED indicated congestion. To be more specific, by dynamically assigning the UDP thresholds such that the sum of the fair average queue resource share of flows in each class is assigned to the class at any given time, D-CBT fairly allocates the bandwidth of a congested link to the traffic classes. Also, the threshold test units, which are activated when RED declares impending congestion (i.e. *red_avg* > *red_min*), coupled with the fair class thresholds, allow the UDP classes to use the available queue resources more effectively than in CBT, in which each UDP class uses the queue elements an average of no more than its fixed threshold at any time. Looking at it from a

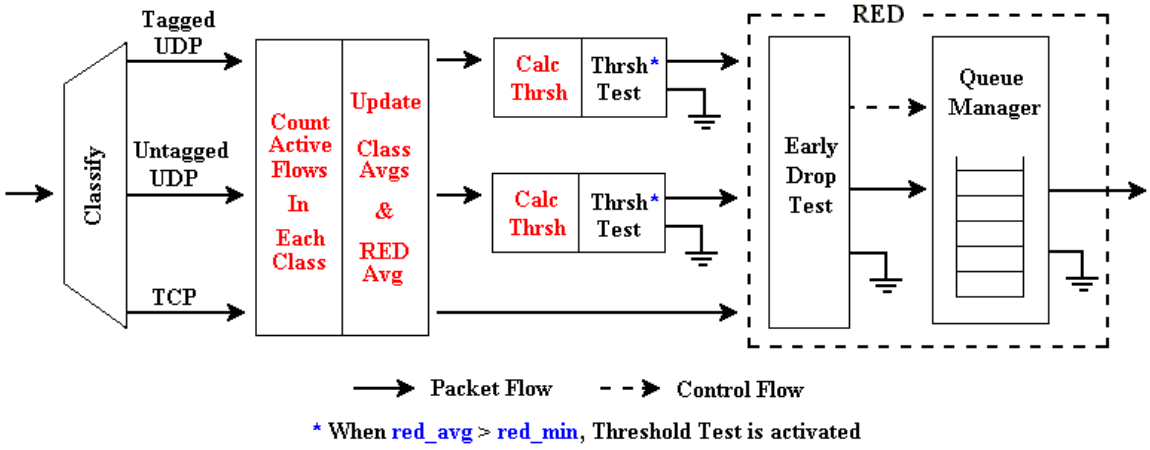
different view as shown in Figure 3.1.1, D-CBT can be thought of a Class-Based FRED-like mechanism that does *per-class-accounting* on the three classes of flows.

As in CBT, D-CBT categorizes flows into TCP, tagged UDP and untagged UDP classes. However, different from the class categorization of CBT in which flow-controlled multimedia flows are not distinguished from unresponsive multimedia flows (all tagged), D-CBT classifies UDP flows into flow-controlled multimedia (tagged) UDP and other (untagged) UDP. The objective behind this classification is to protect flow-controlled multimedia flows from unresponsive multimedia flows, and encourage multimedia applications to use congestion avoidance mechanisms, which may be different than those of TCP. We believe that there are advantages in categorizing UDP traffic in this way for the following reasons: First, multimedia applications are the primary flows that use UDP and generate high bandwidth. Second, by categorizing flows by their congestion responsiveness characteristic (i.e. TCP friendly, flow-controlled but misbehaving multimedia and unresponsive flows), different management can be applied to the classes of differently flow-controlled flows.

In fact, in determining the fair[4] UDP thresholds, D-CBT calculates the fair average output buffer share of the tagged UDP class from the average queue length that is maintained by RED, and that of untagged UDP class from the RED's minimum threshold (plus a small allowance). This is based on the assumption that tagged flows (or flow-controlled multimedia) can respond to network congestion and will actively try to lower the average

[4] Fair share of a class is the number of active flows in a class divided by total number of active flows.

length of a congested queue on notification of congestion. Therefore, they are allowed to use the impending congestion state queue buffers (i.e. *red_avg – red_min* when *red_avg* > *red_min*) up to their fair share of the average. However, unresponsive (untagged) flows, which have no ability to respond to network congestion, are not allowed to use the impending state queue buffers at impending congestion. Actually, we allow the unresponsive UDP class to use a small fraction of the impending state queue buffers, which is 10% of (*red_max – red_min*) * *untagged_UDP_share* when the maximum early drop rate is 0.1, to compensate the effect of needless additional early drops for the class.



<Figure 3.3.1> Design and Implementation of D-CBT

Figure 3.3.1 shows the design and also the implementation of D-CBT, in which the existence of the active flow counting unit is a big structural difference from CBT. In order to calculate a fair threshold (or average queue resource share) for each class, D-CBT needs class state information, and therefore keeps track of the number of active

flows in each class. Generally, as in FRED, active flows are defined as ones whose packet is in the outbound queue [LM97]. However, we took slightly different approach in detecting active flows, in that an active flow is one whose packet has entered the outbound queue unit during a certain predefined interval since the last time checked. In D-CBT, an active flow counting unit that comes right after the classifier maintains a sorted linked list, which contains a flow descriptor and its last packet reception time, and a flow counter for each class. Currently, the flow descriptor consists of a destination IP address and the flow ID (IPv6). However, assuming IPv4, this could be replaced by source and destination address, although this would redefine a flow as per source-destination pair.

For an incoming packet after the classification, the counting unit updates an appropriate data structure by inserting or updating the flow information and the current local time. When inserting new flow information, the flow counter of the class is also increased by one. The counting unit, at a given interval (set to 300ms in our implementation), traverses each class' linked list, deletes the old flow information and decreases the flow counter. The objective behind this probabilistic active flow counting approach is twofold: First, D-CBT does not necessarily require an exact count of active flows as do other queue mechanisms that are based on flow-based-accounting, although a more exact count is better for exercising fairness among flow classes. Second, it might be possible to improve the mechanism's packet processing delay by localizing the counting unit with the help of router's operating system and/or device. For example, the traversing delete is a garbage collection-like operation that could be performed at the router's idle time or

possibly processed by a dedicated processor in a multiprocessor environment. In our implementation, we used a sorted linked list data structure that has the inserting and updating complexity of *O(n)*, and the traversing complexity of *O(n)*, where *n* is the number of flows of a class. Assuming that a simple hash table is used instead, the complexity of inserting and updating operation drops to *O(1)*, while the complexity of the traverse delete will remain *O(n)*.

For a incoming packet that is updated or inserted for its flow identification to its class data structure at the counting unit, D-CBT updates the RED queue average, the tagged UDP average and the untagged UDP average, and passes the packet to an appropriate test unit as shown in Figure 3.3.1. Note that for every incoming packet all of the averages are updated using the same weight. This is to apply the same updating ratio to the weighted-averages, so that a snapshot in time at any state gives the correct average usage ratio among the classes. Using the three averages and the active flow count for each class, the UDP threshold test units calculate the fair thresholds for the tagged and untagged UDP classes as described earlier, and apply the threshold test to incoming packets of the class when the RED queue indicates impending congestion. The rest D-CBT work the same in CBT, which are explained in the previous section.

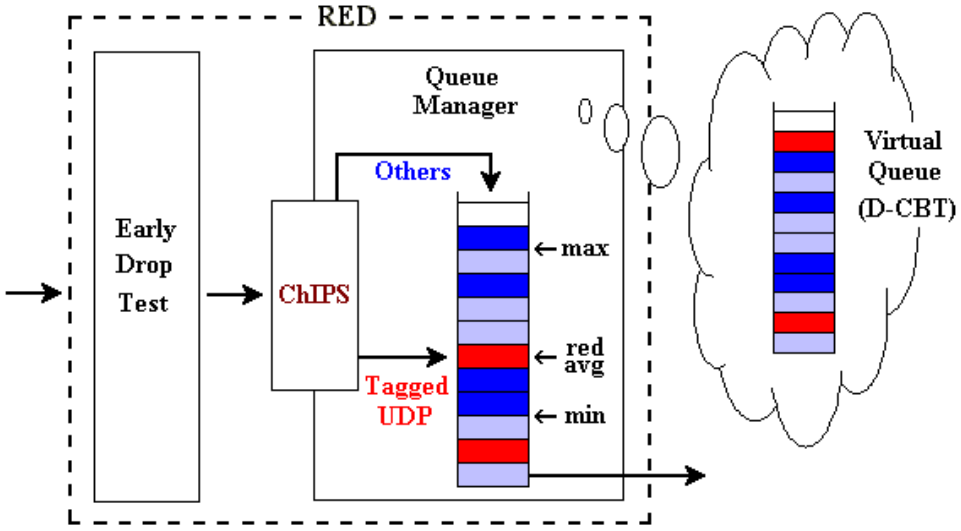
Thus, D-CBT is designed to exercise the traditional fairness between flows of different characteristics by classifying and applying different enqueue policies to them, and restrict each UDP class to use the queue buffer space up to their share in average. We hypothesize that the advantages of D-CBT are the following: First, D-CBT avoids

congestion as RED with the help of responsive traffic sources. Second, assuming that the flows in a class (especially the tagged UDP) use flow control mechanisms of which the congestion responsiveness characteristics are almost the same, D-CBT will fairly assign bandwidth to each flow with much less overhead than FRED, which requires per-flow state information. Even if the tagged flows do not use their fair share, D-CBT will still successfully assign bandwidth fairly to each class of flows, protecting TCP from the effect of misbehaving and unresponsive flows and also protecting the misbehaving (flow-controlled multimedia) flows from the effect of unresponsive flows. Lastly, D-CBT gives tagged (flow-controlled multimedia) flows better chance to fairly consume the output bandwidth than under FRED by performing per-class punishments instead of the strict per-flow punishment. Also, D-CBT may more flexibly assign the bandwidth share within the class.

3.3.2 Cut-In Packet Scheduling (ChIPS)

ChIPS is a light-weight multimedia favored packet scheduling mechanism that can replace the FCFS enqueue style packet scheduling of a RED-managed queue for CBT, D-CBT and possibly other RED-like mechanism, which is specifically targeted to improve multimedia jitter. ChIPS monitors average enqueue rates of tagged and the other flows, and is activated when the tagged flows are using a relatively smaller fraction of bandwidth than TCP flows. On transient congestion in which the queue length is greater than the average queue length, ChIPS awards tagged (flow-controlled multimedia) flows by allowing their packets to “cut” in the line of queue where the average queue length

points. Figure 3.3.2 in the next page shows the design of ChIPS (the virtual queue is described later).



<Figure 3.3.2> ChIPS (Tagged Packet Insertion on Transition Congestion)

By inserting tagged UDP packets at the average queue length on transition congestion, ChIPS improves flow-controlled multimedia jitter. However, this could harm TCP flows and even make them time out by introducing a large extra delay when the multimedia traffic is taking a considerable portion of the output bandwidth. Under the normal RED queue mechanism that has no means to regulate the queue buffer usage among the classes of flows, it is essential for ChIPS to monitor the average enqueue ratio between the tagged and other flows and turn on its function only when the ratio is small. However, under CBT, in which the tagged threshold can be explicitly set to use a small fraction of the available queue buffer, this automatic turn on/off function is not really necessary. When used with D-CBT, the ratio that turns off the ChIPS could be set relatively large

(tested for up to 50% in our simulations with the RED minimum threshold of 5 and the maximum of 15) without degrading the fairness because of the "self-adjusting" ability of D-CBT. When a relatively large number of tagged flows compete for the bandwidth with TCP flows, ChIPS could instantly lower the TCP throughputs. However, this will also lower the average queue length of the queue, and therefore the fair threshold for the tagged class will be reduced and the tagged class throughput will be reduced as well. Thus, ChIPS may cause the average queue length to fluctuate a bit more but does not reduce fairness significantly. Chapter 4 has more detailed results.

Another issue in implementing ChIPS is that the increment of the tagged packet dequeue rate caused by the insertion could degrade the fairness when the packet enqueue decision makes use of each class' buffer usage as in CBT and D-CBT. This faster tagged packet drain rate is not an issue for RED since it enqueue decision has nothing to do with the drain rate. However, in CBT and D-CBT, the faster drain rate lowers the average number of enqueued packets for the tagged class, which could result in the tagged class getting more bandwidth than its fair share. To prevent this effect, we used a virtual queue for counting the number of enqueued packets for the UDP classes, in which the class information of an enqueuing packet is always enqueued at the end, even though ChIPS cuts a tagged packet in the line of the real queue. In this way, a virtual queue can deceive the class average counting units by telling that the tagged packets that have been transmitted already are still in the queue. Thus, the actual tagged packet drain rate does not affect the calculation of the average number of enqueued packets for the tagged class.

Looking at the complexity of the design, ChIPS has *O(1)* behavior, since the insertion complexity is *O(1)* and the virtual queue maintenance complexity is also *O(1)*. We believe that ChIPS, which noticeably improves tagged flow (flow-controlled multimedia) jitter, along with D-CBT would further encourage multimedia applications to use a flow control mechanism. An important issue that is not addressed in this paper is who is going to monitor and tag the flow-controlled multimedia flows. However, we believe that this job has to be done in the Internet Service Provider (ISP) level or the local network management level at the gateways to the public networks, and leave the routers of the public networks free from this issue. The next chapter presents the simulated performance and evaluation of the proposed mechanisms.

4. Results and Analysis

To evaluate our research approach, the performance of RED, CBT, D-CBT, and D-CBT with ChIPS are measured and compared in terms of congested link throughput, fairness among the different classes of flows and multimedia jitter through a series of simulations, in which TCP, flow-controlled multimedia and unresponsive flows coexist. For the flow-controlled multimedia traffic sources we used the multimedia traffic generators described in Chapter 3.1. The first section of this chapter discusses the performance measurement metrics used for the evaluation, and the second section presents the simulation setup. The third and the last sections present the simulation results and analysis.

4.1 Performance Measurement Metrics

The objective behind the design of D-CBT and ChIPS is to improve fairness among different classes of flows and to reduce multimedia jitter. This section presents the fairness and jitter measurement metrics used to evaluate our proposed mechanism. To measure the fairness among the three different classes of flows and also to visualize the fairness among individual flows, we use the following two metrics. The first one is the *direct comparison of the average per-flow throughput in each class*, which is an average aggregated class throughput divided by the number of flows in the class. We believe that this is a good indicator of how fairly the output bandwidth is assigned to each class considering the number of flows in the class. However, it does not tell how fairly each individual flow is assigned of the output bandwidth. Therefore, we use a second fairness

measurement metric, called *Jain's fairness index*, to visualize the fairness among individual flows [Ja91]. Figure 4.1.1 shows the formula that calculates Jain's fairness.

$$f(x_0, x_1, x_2, \ldots, x_n) = \frac{(\sum_{i=0}^{n} x_i)^2}{n\sum_{i=0}^{n} x_i^2} \qquad \text{(Original)}$$

$$= f(x_0+d_0, x_0+d_1, x_0+d_2, \ldots, x_0+d_n) \qquad \text{(Offset Form)}$$

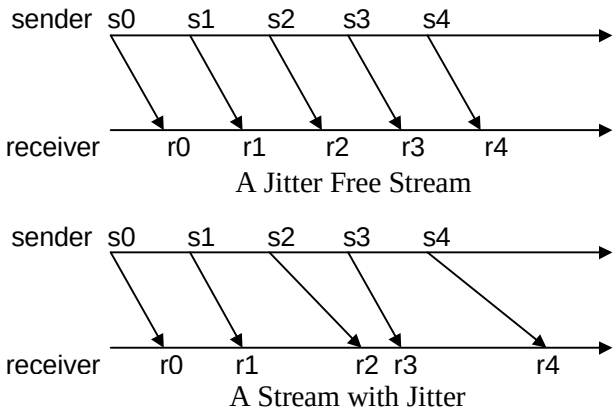
$$= \frac{(\sum_{i=0}^{n} x_i)^2}{(\sum_{i=0}^{n} x_i)^2 + \sum_{i=0}^{n-1}\sum_{j=i+1}^{n}(d_i - d_j)^2} \qquad \text{(Derived)}$$

$$0 \le f \le 1 \text{ (Greatest Fairness)}$$

<Figure 4.1.1> Jain's Fairness (f): x_i is the average throughput of i^{th} flow

Jain's fairness formula gets the average throughputs of the flows of which the fairness is measured as an input, and produces a normalized number between 0 and 1, where 0 indicates the total unfairness and 1 indicates the greatest fairness. However, looking at the original Jain's formula, which is used for our fairness calculations, it is hard to see what it really measures. Therefore, to better understand the formula, we rewrote the formula into more meaningful form by converting the throughputs of flows into an offset presentation, in which the throughput of the i^{th} flow is written as throughput of the 0^{th} flow (x_0) plus the displacement from x_0 that is denoted as d_i. The derived formula indicates that the variation on the throughput displacements decreases the index value by

increasing the denominator. Indeed, Jain's fairness index is a good indicator of how fairly each flow shares the output bandwidth, since it is the scaled displacement of each flow's bandwidth that determines the Jain's value. For example, when a couple of flows are severely punished and have far less throughput than the others, the Jain's fairness value will decrease dramatically indicating that the system is unfair, although the class average measurement would show that the system is fair. We decided to use Jain's fairness index along with the direct comparison of the per-flow average throughput in each class to visualize the system's fairness on each flow as well as on the classes.



<Figure 4.1.2> Multimedia Jitter – An Example: This figure models packet video between sender and receiver. Each s*i* is the time at which the sender transmits video frame *i*. Each r*i* is the time at which the receiver receives frame *i*. [CT99]

Another network system performance factor we want to measure especially for multimedia streams is the variance in packet delivery or *jitter*. In a wide area multimedia networking, jitter, which directly affects multimedia perceptual quality and also can indirectly increase end-to-end delay, is an important performance factor as well as end-to-end delay and packet loss [CT99]. Figure 4.1.2 depicts an example of jitter.

> "In the absence of jitter and packet loss, video frames can be played as they are received, resulting in a smooth playout, as depicted in Figure 4.1.2-top. However, in the presence of jitter, inter-arrival times will vary, as depicted in Figure 4.1.2-bottom. In Figure 4.1.2-bottom, the third frame arrives late at r*2*. In this scenario, the user would see the frozen image of the most recently delivered frame (frame two) until the tardy frame (frame three) arrived. The tardy frame (frame three) would then be played only briefly in order to preserve the timing for the subsequent frame (frame four)." [CT99]

Jitter introduced by the Internet could degrade a multimedia perceptual quality significantly especially on video. In order to ameliorate the jitter effect, a receiver could add a delay in the playout, which is called *delay buffering*. However, delay buffering results in increasing of end-to-end delay, which can degrade quality for interactive multimedia applications.

Now that the impact of jitter on multimedia applications has been discussed, the next issue is determining its measurement method. First, jitter can be measured in terms of *variance in inter-frame arrival time* at the receiver, which is a receiver-oriented observation on the variance. This is a useful and also an easy measurement method, since this is what a receiver application actually experiences and what can be easily measured at the end systems in a real network environment. However, it has a limitation that when a source application changes it transmission rate or the frame rate often as in the case of a flow-controlled multimedia application, what it measures may no properly visualize the effect of the underlying system on jitter. Actually, we used this method in comparing the performance of TCP and MM_APP on multimedia streams in Chapter 3.1, in which we showed that TCP is not the choice for the transport agent for the multimedia applications because of its bursty transmission characteristics. In this case, the method

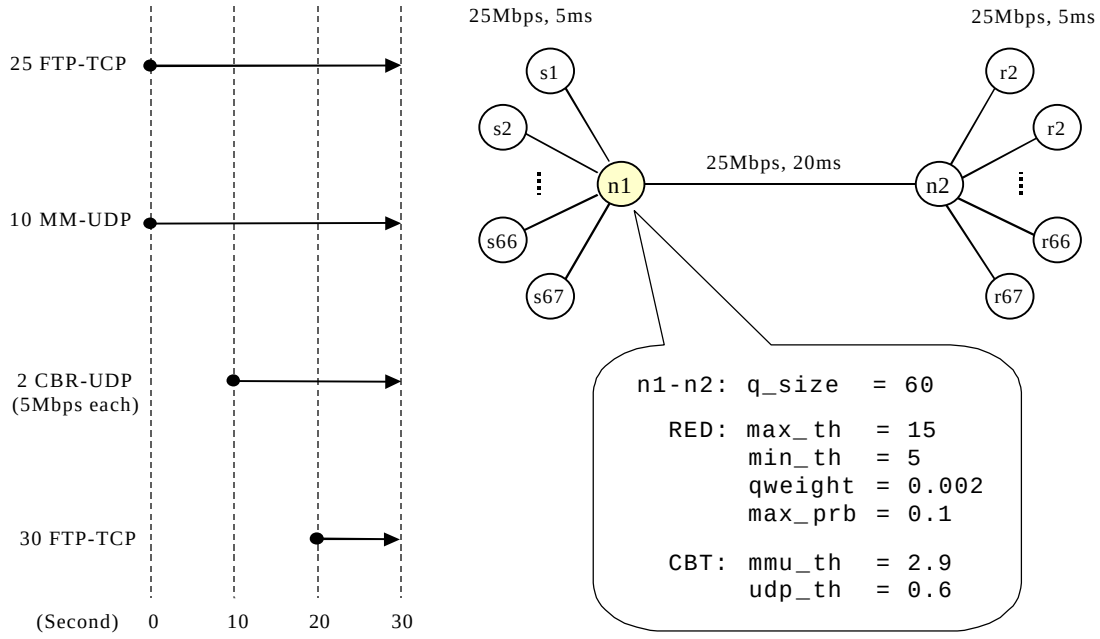
visualized the effect of TCP's transmission policy well, because the effect was tremendous. However, more often than not the underlying network system's effect on jitter would be relatively smaller than the one contributed by the flow-controlled application's transmission policy.

The second method of measuring jitter is in terms of *variance in end-to-end delay*. This is a more network-oriented observation of the variance and a direct indicator of a network, an operating system, or their subsystem's performance on multimedia streams compared to the first method, since it eliminates the inter-packet transmission periods of the source. However, in real environments, it is hard to measure jitter in this way because of asynchronized clocks of the source and the destination. However, in our simulation environment where only one logical clock is used for the whole system, it is very easy to measure actual variance end-to-end delay. Moreover, this method can even visualize the effect of queuing delays of a single router on jitter well. Therefore, we decided to use this second method in evaluating the effect of ChIPS on multimedia jitter.

Using the fairness metrics and jitter measurement metric discussed in this section, we evaluated RED, CBT, D-CBT and D-CBT with ChIPS in terms of fairness in output bandwidth allocation and multimedia jitter. We also measured the congested link throughput to visualize the effect of the mechanisms on the link utilization. However, since this is very primitive and well-known metric, it is not discussed in this section. The next section presents the simulation setup in detail.

4.2 Simulation Setup

The purpose of this simulation study is twofold. The first purpose is to measure the effect of RED, CBT and D-CBT on congested link utilization and fairness when different classes of flows coexist. The second purpose is, by comparing the performance of D-CBT and D-CBT with ChIPS, to see if ChIPS affects fairness and/or improves jitter. We ran a simulation for each of RED, CBT, D-CBT and D-CBT with ChIPS. Every simulation had the exactly same settings except for the network routers, each of which is set to use one of the above four outbound queue management mechanisms. The network topology and the traffic source schedules are shown in Figure 4.2.1.



<Figure 4.2.1> Simulation Scenario and Network Setup

In each simulation, we had 67 source nodes connected to one router and 67 destination nodes connected to the other router, which are interconnected by a link with 25Mbps bandwidth and 20ms of delay. Each link that connects a source (or destination) node and a router was set to have 25Mbps of bandwidth and 5ms of delay. For traffic sources, 55 FTP, 10 flow-controlled MM_APP (tagged) and 2 CBR (untagged) traffic generators were used, where FTP used TCP Reno and the others used UDP as the underlying transport agent. All the TCP agents were set to have a maximum congestion window size of 20 packets and maximum packet size of 1Kbyte, with which a TCP agent could transmit at the average rate of up to 4.6Mbps (the calculation is based on the average queue size of 15 packets). The UDP agents are also set to have maximum packet size of 1Kbyte, so that all the packets in the network are the same size. The MM_APP traffic generators (flow-controlled multimedia applications) used the transmission rates shown in Table 3.1.1 (Chapter 3.1), that is 300, 500, 700, 900, 1100 Kbps, for scale 0 to 4 transmission rates, with the fixed packet size of 1Kbyte. The CBR sources were set to generate 1Kbyte packets at 5Mbps.

We scheduled the traffic sources such that 25 TCP flows and 10 MM_APP flows are competing for the bandwidth during 0 to 10 seconds. At this period the fair bandwidth share for each connection is about 714Kbps (25Mbps / 35 flows). In the next period (10 to 20 seconds), the two high bandwidth CBR blasts join trying to aggressively use the output bandwidth of which the average fair share was about 675Kbps (25Mbps / 37 flows). Later at 20 seconds, 30 more TCP flows came into the network lowering the average fair share during the last 10 seconds to about 373Kbps (25Mbps / 67 flows).

Network routers were assigned of a 60-packet long physical outbound queue. For RED queue management, the maximum and minimum thresholds were set to 15 and 5 packets, the weight for the weighted queue average was set to 0.002, and the maximum early drop probability was set to 0.1. These RED parameters were chosen from one of the sets that are recommended by Floyd and Jacobson [RK99]. For CBT, beside the RED parameters, the tagged threshold (denoted as *mmu_th* in the figure) was set to 2.9 packets and the untagged threshold (denoted as *udp_th*) was set to 0.6 packets. These thresholds were selected to force the UDP classes to get about their fair bandwidth shares during 0 to 20 seconds. Assuming the average queue size is 10 packets, by reserving an average of 2.9-packet space, the tagged class could get an average bandwidth of 7250Kbps (25Mbps * 2.9 / 10) at congestion, which is about 10 times (10 tagged flows) the fair share during 0 to 10 seconds. Likewise, by reserving 0.6-packet space in the queue, the untagged class could get an average of 1500Kbps, that is little bit more than 2 times (2 untagged flows) the fair share during 10 to 20 seconds.

D-CBT also shares the RED settings, however, since each threshold is assigned dynamically to the fair share of each class, no threshold setup was necessary. Finally, ChIPS was set to turn off its cut-in scheduling feature when the ratio between the number of tagged flows and the other flows are greater than 50%. However, under our simulation, ChIPS was always on since the ratio was always under 50 %.

Thus, the simulations were designed to give an environment under which every queue management mechanism manages output bandwidth fairness during the first 10 seconds,

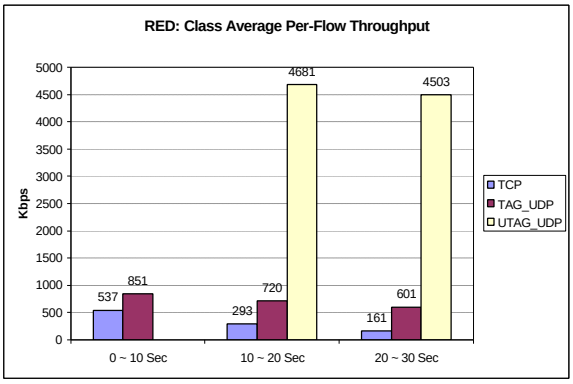
RED fails during the second 10 seconds, and CBT fails during the last 10 period. Then, we wanted to see if D-CBT could dynamically offer fair bandwidth allocation in every situation. Also, by comparing the results (fairness and jitter) of D-CBT with ChIPS with the basic D-CBT, we could measure the effect of ChIPS on fairness and multimedia jitter. The next two sections present the simulation results and analysis.

4.3 Throughput and Fairness

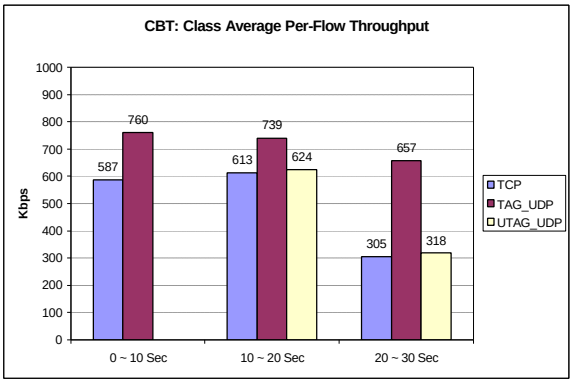
This section presents our simulation results and analysis on RED, CBT, D-CBT and D-CBT with ChIPS. In the first part of the section, the fairness of the mechanisms in managing bandwidth is shown using the fairness measurement metrics discussed in Section 4.1. Then, we showed the efficiency of the mechanisms in utilizing the output bandwidth by visualizing the congested link throughput.

4.3.1 Class' Average Per-Flow Throughput

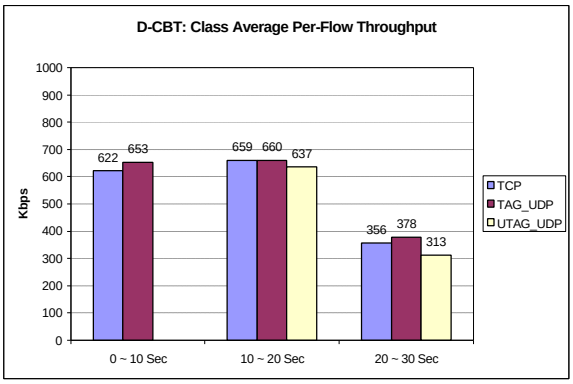
Figure 4.3.1 (a) through (d) in the next page compares the periodic (i.e., 0-10, 10-20 and 20-30 seconds) average per-flow throughput for each class under the four queue mechanisms.



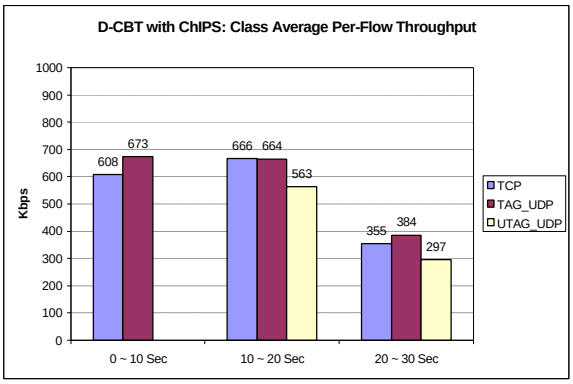
(a) RED



(b) CBT



(c) D-CBT



(d) D-CBT with ChIPS

<Figure 4.3.1> Average Per-Flow Throughput for TCP, Tagged UDP and Untagged UDP Classes under RED, CBT, D-CBT and D-CBT with ChIPS

As shown in Figure 4.3.1 (a), RED absolutely failed to assign bandwidth fairly to each class of flows from 10 seconds when the two high bandwidth untagged UDP flows (unresponsive CBR) join transmitting at the total of 10Mbps, about 40% of the link bandwidth. During 0-10 seconds, when 25 TCP and 10 tagged (flow-controlled MM_APP) flows are competing for the bandwidth, it was somewhat unfair as a tagged flow gets an average of 37% more bandwidth than a TCP flow, but RED was able to

manage the bandwidth. However, when the untagged UDP blast came into the system, RED was totally unable to manage bandwidth. The 2 untagged UDP flows got most of the bandwidth they needed (average of 4.68Mbps out of 5Mbps), and the remaining flows used the leftover bandwidth. Especially, the 25 TCP flows got severely punished and transmitted at an average of 293Kbps per flow. An interesting observation here is that the tagged (MM_APP) flows were not punished much but each flow in average was transmitting (720Kbps) over the fair share of the second period, that is 675Kbps. This was mainly because TCP (Reno) often failed to perform *fast recovery* and went back to *slow start* mode [rfc2001], and/or even timed out due to a high packet drop rate at the router. Fairness got worse as 30 more TCP flows joined at 20 seconds, and experienced starvation.

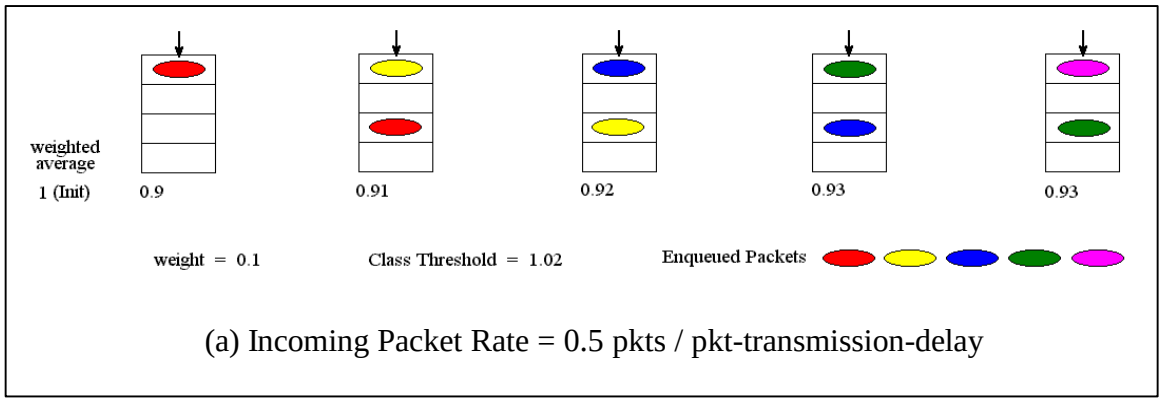
Figure 4.3.1 (b) shows that CBT can avoid the great unfairness of RED using fixed thresholds for the UDP classes. However, CBT was not assigning the output bandwidth to each class as expected. When designing the simulation, we set the UDP thresholds such that during 0-10 seconds each tagged UDP flow should get 725Kbps (7250Kbps / 10) in average. During 10-20 seconds, we expected that each tagged flow's average bandwidth would remain the same and each untagged UDP flow would get an average of 750Kbps (1500Kbps / 2). Also, we expected that during the last period (20-30 seconds), the tagged and untagged flows would get a large portion of the bandwidth that is close to the figure during 10-20 seconds and the TCP flows would get much less than the fair share during this period. However, the simulation result shows that the tagged UDP class

got more bandwidth than the expected values especially during the last period, while the untagged UDP class got much less bandwidth than expected.

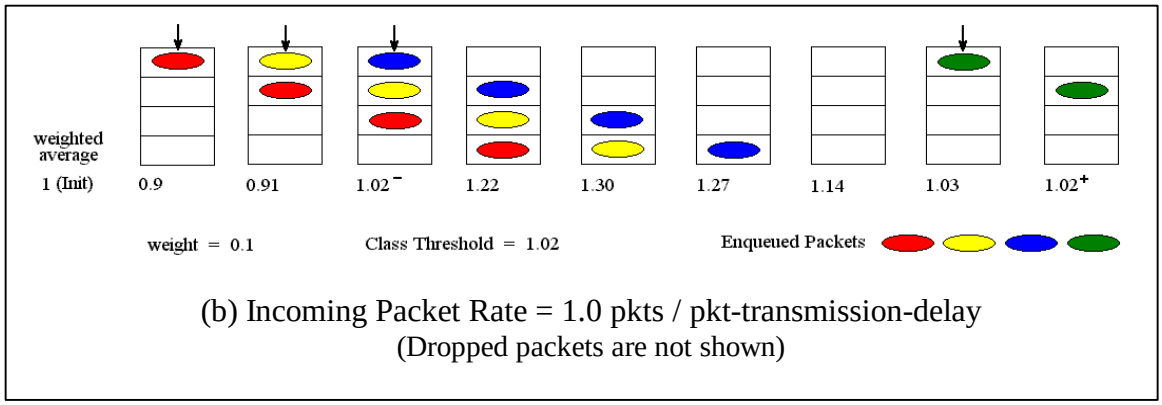
We found out that this is mainly due to how and when CBT updates each UDP class threshold and RED queue average. As shown in Figure 3.2.2 (Chapter 3.2), CBT updates each UDP class average only for incoming packets that belong to the class, and the RED unit updates its queue average for all incoming TCP packets and for UDP packets that passed an appropriate threshold test. Therefore, the class averages and the RED queue average are almost independently updated at different speeds that are closely related the number of incoming packets that belong to the class. In addition, the RED average has a higher chance of being updated faster than the UDP class averages. In this situation, which we call *unsynchronized weighted-average updates*, whoever (i.e. a class) updates its weighted-average more often will get less bandwidth by having a larger weighted-average than the average of others for the same amount of class output bandwidth, and the output bandwidth is controlled using the averages at the UDP threshold test units.

Figure 4.3.2 shows this effect by comparing two situations where a UDP class that has an initial class weighted-average of 1, a weight of 0.1 and a class threshold of 1.02, is experiencing two different incoming packet rates. Figure 4.3.2 (a) is the case when the incoming packet rate is 0.5 packets per packet transmission delay, and Figure 4.3.2 (b) is the case when the class is receiving packets at the rate of 1.0 packets per packet transmission delay. In this example, it is assumed that the traffic sources are unresponsive CBR applications. One thing to note in the figure is that the class average

shown at the left bottom of each queue in each state is its value before making the enqueue admission decision for an incoming packet at that state. As you can see in the figure, as the number of incoming packets for a class increases, packets are enqueued in a bursty manner, and more importantly, its class average gets larger. As the average is updated more frequently, not only is a newly enqueued packet added to the average (with the weight of 0.1), but also the existence of the other already enqueued packet are added to the average. For example, the existence of the first (red colored) packet is added to the average (with the weight of 0.1) 2 times more for the second situation than for the first situation.



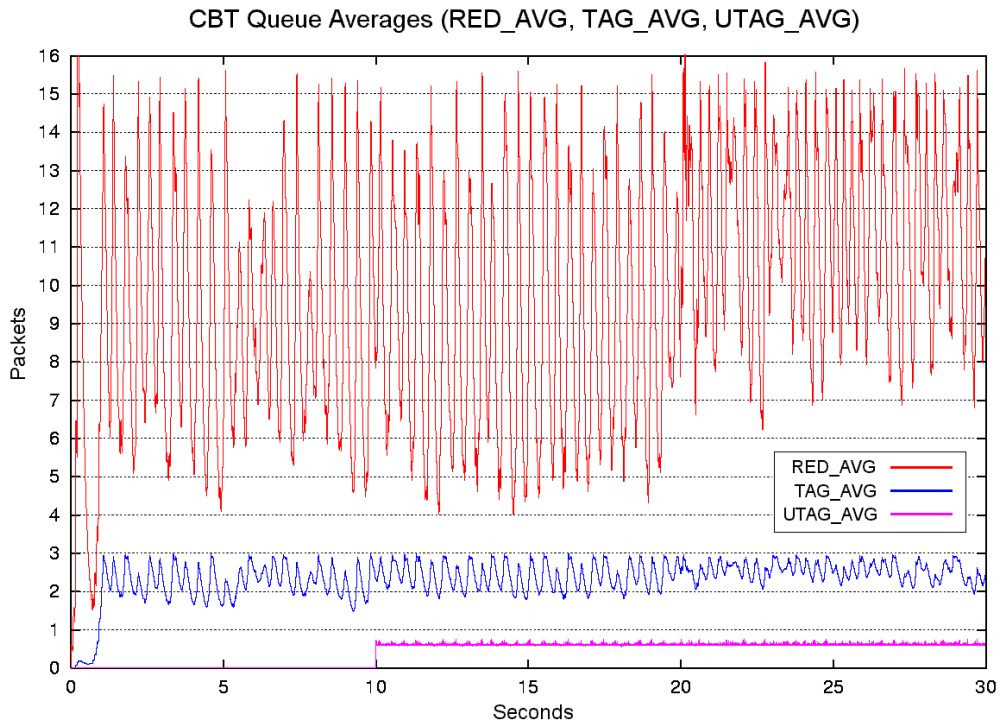
(a) Incoming Packet Rate = 0.5 pkts / pkt-transmission-delay



(b) Incoming Packet Rate = 1.0 pkts / pkt-transmission-delay
(Dropped packets are not shown)

<Figure 4.3.2> A CBT Class's ***Weighted Average*** under Two Different Incoming Packet Rates

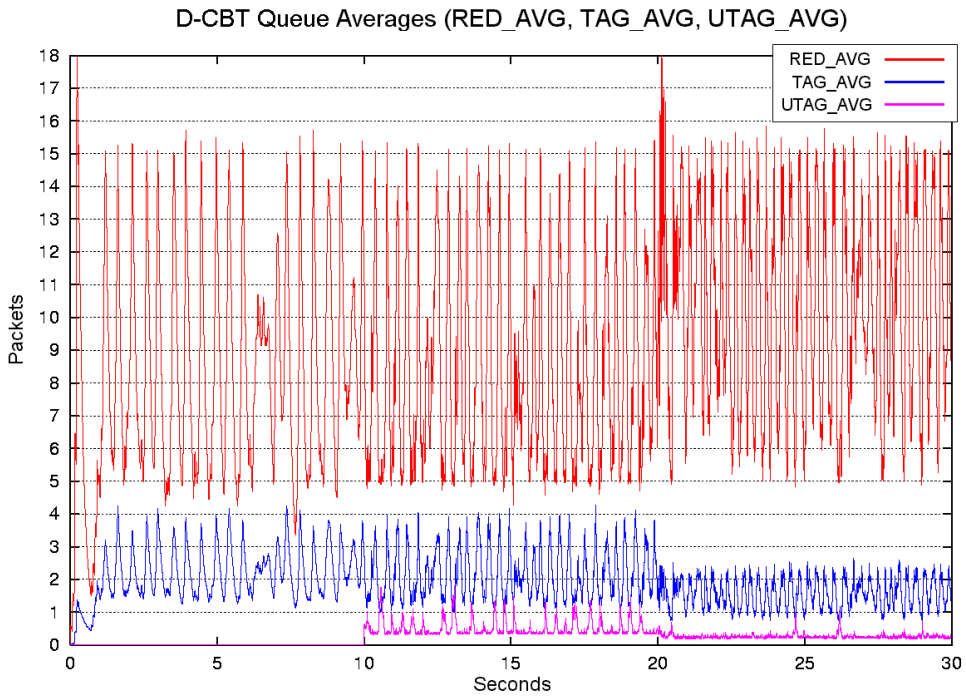
The weighted-average calculation method works fine when the purpose of measuring an average queue size is to detect impending congestion as in RED. However, when the method is used to assign bandwidth to different classes of flows by comparing each class' weighted-average number of enqueued packets, special care is needed in deciding how and when to measure the averages. We have determined that all the weighted-averages should be updated at the same time and at an equal frequency to give a correct output bandwidth utilization ratio among the classes. In the case of CBT, by measuring each UDP class average and the RED average independently, the classes' bandwidth utilization could not measured correctly by comparing the class averages. By comparing the fairness measurement in Figure 4.3.1 (b) and the CBT's outbound queue averages in Figure 4.3.3, especially for 20-30 seconds, one can easily see that CBT's attempt of using unsynchronously updated weighted-averages to regulate class bandwidth was misleading. Figure 4.3.3 indicates that during 20-30 seconds, 10 tagged flows used an average of about 2.5 packet-spaces in the queue that is 0.25 packet-spaces per each flow, and the 2 untagged flow used an average of about 0.6 packet-spaces that is 0.3 packet-spaces per each flow. However, as shown in Figure 4.3.1 (b), each tagged flow used about 657 Kbps of bandwidth and each untagged flow used about 318 Kbps, about one half of the per-flow bandwidth of a tagged flow.



<Figure 4.3.3> CBT Queue Averages: The ratio between the averages does not correctly indicate the ratio between each class' average bandwidth utilization because of the effect of unsynchronized weighted-average updates.

From the above observation, we conclude that the current CBT design can only prevent a great unfairness caused by unresponsive or misbehaving flows, and it needs some adjustment on weighted-average calculation. Indeed, we tried the average calculation method that is used in D-CBT in CBT and got a much better result, that is the ratio between the three averages indicates the ratio between the actual classes' bandwidth utilization. However, we did not include the result in this paper, since the method is used in only D-CBT and we are presenting D-CBT in the next paragraph.

Figure 4.3.1 (c) shows the D-CBT result, which indicates that D-CBT fairly manages bandwidth during all periods by dynamically allocating the right amount of output queue space to each flow class. It also shows that by updating each class and RED average at the same time in a synchronized manner, the ratio between the averages is a good indicator of the ratios between each class' bandwidth utilization. One thing to note in the figure is that although we strictly regulate the untagged class by assigning a fair threshold calculated from RED's minimum threshold, the untagged class did get most of its share. This is because the high bandwidth untagged (unresponsive) packets were allowed to enter the queue without its threshold test, when RED indicated no congestion. As a reference, we show the queue averages of D-CBT in Figure 4.3.4, in which each class average is actively moving proportional to RED's average.

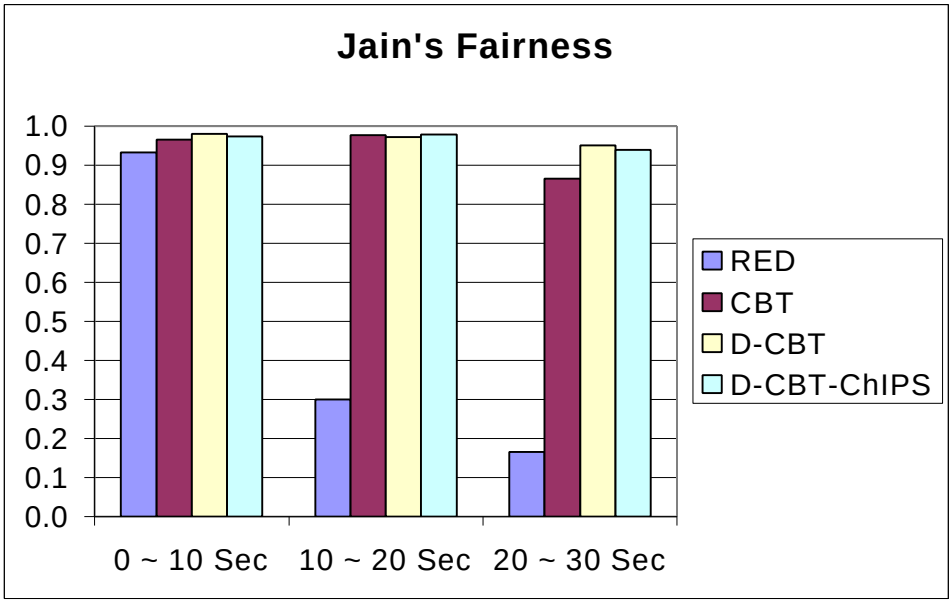


<Figure 4.3.4> D-CBT Queue Averages

Figure 4.3.1 (d) shows the result of D-CBT with ChIPS. The result confirms that ChIPS, when used with D-CBT, does not affect fairness between each class of flows, due to the virtual queue and D-CBT's self adjusting capability described in Section 3.3.3. In the simulation, the ratio between tagged flows and all flows was about 28% during 0-20 seconds, and was about 15% during the last 10 seconds. The next section presents the fairness measurement of the simulations using Jain's Fairness index, which visualizes system's fairness on individual flows in a normalized number.

4.3.2 Jain's Fairness Measurement

Figure 4.3.5 shows the simulated systems' fairness on individual flows using Jain's Fairness Index, where the periodic (0-10, 10-20 and 20-30) average throughput of each individual flow was given as input to the original Jain's equation.



<Figure 4.3.5> Jain's Fairness Comparison

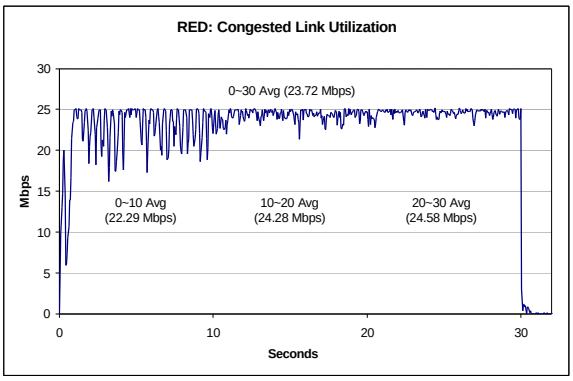
The Jain's fairness measurement shows that the simulated system that use RED queue management fails to fairly assign bandwidth to each individual flow from 10 seconds when the unresponsive flows join in the system. The low Jain's index value for the RED system indicates that some flows are experiencing severe starvation during 10-20 seconds and even more severe during 20-30 seconds when 30 extra TCP flows join.

The system that uses the CBT queue management mechanism was fair overall in distributing bandwidth to each flow. However, during 20-30 seconds, the system's fairness was degraded because the 10 tagged (multimedia) flows got about twice as much bandwidth as the other flows. One thing to note is that the CBT's fairness was engineered. In a circumstance where traffic mixes change a lot, CBT might show more degraded fairness.

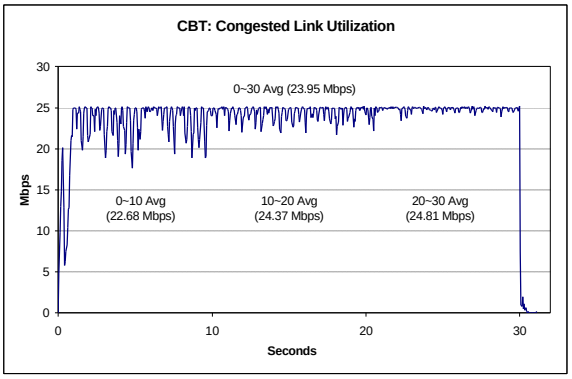
On the other hand, the systems that use D-CBT or D-CBT with ChIPS were dynamically adjusting to changing flow mixes, and were very fair not only to the classes of flows but also to individual flows as Jain's index numbers indicate. The Jain's fairness measurement results on D-CBT and D-CBT with ChIPS also reconfirmed that ChIPS did not degrade the system's fairness. The next section shows the congested link utilization, and the section after presents the improved multimedia jitter performance of the system that used ChIPS.

4.3.3 Congested Link Utilization

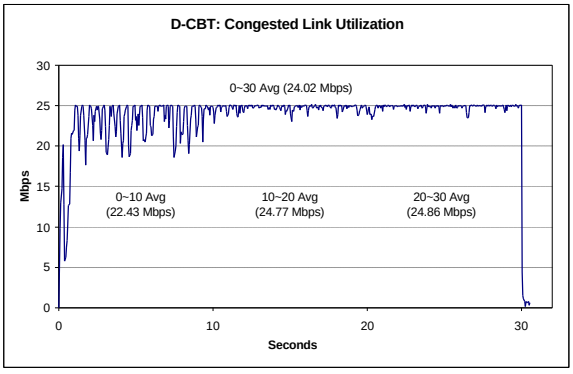
In the previous section, we showed that D-CBT and D-CBT with ChIPS outperforms RED and CBT in term of fairness in managing bandwidth. However, the ability of a system to exercise fairness without efficiently utilizing available resources is not always acceptable. Therefore, to compare efficiency of D-CBT and D-CBT with ChIPS with that of RED and CBT, we present the congested link throughput for the simulated systems.



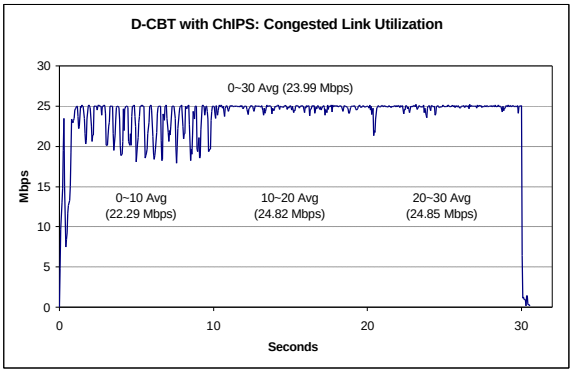
(a) RED



(b) CBT



(c) D-CBT



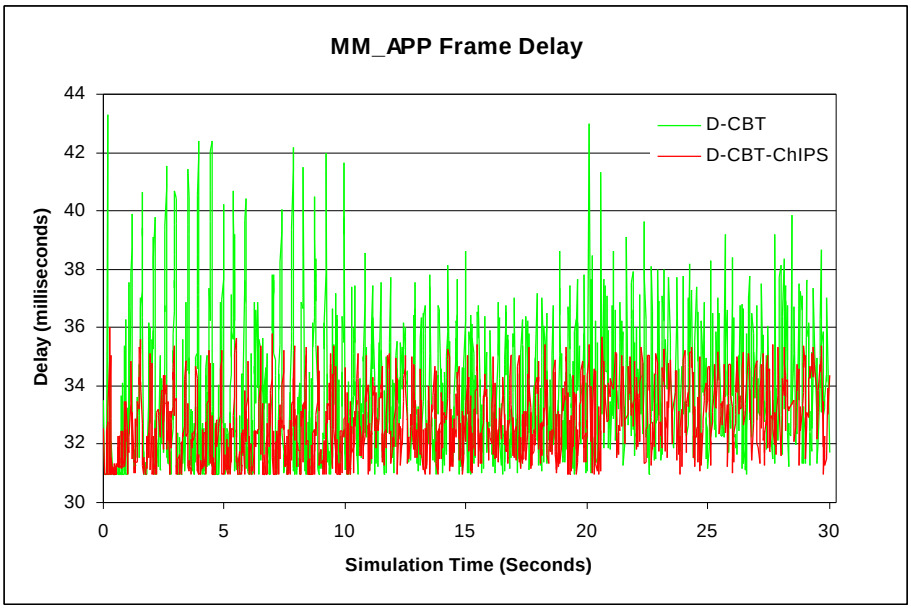
(d) D-CBT with ChIPS

<Figure 4.3.6> Congested Link Utilization (Throughput)

Figure 4.3.6 shows that the efficiency of the systems that use D-CBT or D-CBT with ChIPS was comparable with that of the systems that use RED or CBT. All the systems get to a stable state within few seconds after the start of the simulation, and generally used the link bandwidth up close to the maximum. During 10-30 seconds when the UDP flows joined, D-CBT and D-CBT with ChIPS were even a bit more efficient than RED and CBT. This was mainly because UDP packets were admitted without its threshold test and without forcing packets from responsive connections (i.e. TCP or Tagged UDP) to be dropped, when the RED unit indicates no congestion.

4.4 Analysis of ChIPS

Now that we have shown that D-CBT outperforms RED and CBT in managing bandwidth, and that the use of ChIPS does not degrade the performance of D-CBT, this section presents the performance of ChIPS on multimedia jitter and TCP throughput. Figure 4.4.1 shows tagged UDP (or multimedia) jitter by comparing a MM_APP application's end-to-end frame delay under D-CBT and D-CBT with ChIPS.



<Figure 4.4.1> Improved Multimedia Jitter – ChIPS Effect on Jitter

The result indicates that ChIPS does improve tagged stream jitter by inserting tagged packets into the line of the queue to which the RED average points on transient congestion. Under ChIPS, the maximum tagged-UDP jitter was about 5*ms* (36*ms* – 31*ms*) while it was about 12*ms* (43*ms* – 31*ms*) under normal D-CBT. Noting that the maximum threshold of RED was set to 15 packets, which gives about 3*ms* (15pkts *

8Kbits / 25Mbps) of queuing delay, ChIPS was able to regulate the maximum tagged stream jitter around the queuing delay of the RED's maximum threshold.

We believe that ChIPS effect on improved multimedia jitter could be very significant because of the following two reasons. First, what we show in Figure 4.4.1 is the jitter gain due to a single router. Assuming that a multimedia connection that is configured similar to our multimedia traffic generator has to go through 10 routers that use D-CBT with ChIPS and set the maximum threshold of RED to 15. In this environment, the application can almost completely remove the jitter effect on perceptual quality by using a delay buffer of only 50*ms*, while it would be extremely difficult to determine the optimal amount of the playout delay under the traditional Internet environment. Second, in the simulation, we used multimedia frames that are the same size as that of a network packet, meaning that no frame fragmentation occurs in the IP layer. Assuming that a multimedia application uses frames that are larger than a network packet and are chopped into multiple packets in the network, the jitter improvement due to ChIPS could be even more significant, since the multimedia packets have better chance to be transmitted close to each other at routers. Thus, we believe that the potential of ChIPS to improve multimedia jitter is larger than shown in our experiments.

In summary, we present TCP packet accounting and tagged packet accounting for the simulation in Table 4.4.1 and Table 4.4.2. Table 4.4.1 shows the TCP packet drop rate and throughput under ChIPS is very compatible with those of basic D-CBT. The TCP throughput under ChIPS was about 99.6% of the throughput under basic D-CBT. This

indicates that ChIPS, when used along with D-CBT, does not significantly affect the TCP throughput. Comparing the TCP throughput loss with the multimedia jitter gain, ChIPS compensates 14.3% ((42*ms* - 36*ms*) / 42*ms* * 100) of multimedia jitter gain for 0.4% of TCP throughput loss for the simulation.

	TCP Packets Delivered	TCP Packet Drop Rate	TCP Throughput
D-CBT	66,648 pkts	4.46 %	17,773 Kbps
D-CBT with ChIPS	66,386 pkts	4.44 %	17,703 Kbps

<Table 4.4.1> TCP Packet Accounting (0 ~ 30 Seconds)

	Tagged Packets Delivered	Tagged Packet Drop Rate
D-CBT	21,126 pkts	11.85 %
D-CBT with ChIPS	21,519 pkts	12.95 %

<Table 4.4.2> Tagged (Multimedia) Packet Accounting (0 ~ 30 Seconds)

Table 4.4.2 shows the multimedia packet drop rate of the system that used ChIPS is very compatible with that of the system that used basic D-CBT. This result shows that ChIPS has a high potential to improve end-user multimedia performance (perceptual quality) on the Internet by improving jitter without increasing the multimedia packet drop rate, which is another important factor in multimedia perceptual quality and for congestion control and system utilization.

5. Future Work

As shown in Chapter 4, our queue management approaches (D-CBT and ChIPS) were mostly examined from the viewpoint of the network, concentrating on fairness and multimedia jitter. However, there still remain many performance aspects to be evaluated. The first research would be to extend this study to evaluate D-CBT and ChIPS with TCP Vegas, especially on fairness and TCP throughput. The second research would be to evaluate D-CBT and ChIPS under the environment where fragile and robust TCP connections as well as multimedia connections with different end-to-end delay coexist in the system.

The third essential research would be to extend this study to evaluate the limitation of ChIPS on the fairness and the link utilization offered by D-CBT. As denoted earlier in Chapter 3.3.2, ChIPS introduces an additional delay to other traffic which may affect TCP throughput. Therefore, in order for the use of ChIPS to be more practical, future work suggests an extended study to determine the maximum average ChIPS enqueue ratio between tagged and the other classes of flows without degrading fairness or link utilization. Another study that we could not do due to the lack of time but suggest as a future work is to compare the performance of the D-CBT with that of FRED. We expect that CBT could give better throughput performance for tagged UDP flows than FRED, since it frees flow-controlled multimedia flows from the strict per-flow punishment.

Another area for future work is measure of D-CBT and ChIPS on jitter, packet and/or frame drop rates and the drop patterns for current Internet multimedia applications in order to evaluate the effect on perceptual quality. This work would involve a user study as well as a network performance study, and could be carried out in two ways. The first approach is through simulations using NS, which may require implementing application behavior of real world such as MPEG-2 or H.261 into NS. The second approach is through experiments. This requires implementing D-CBT and ChIPS into an operating system's kernel such as Linux, which is currently a standalone project. Andrew Chillar, Nicolas Leazard, Miguel Maldonado and Edwin Mercado are implementing the mechanisms into a Linux kernel as their Major Qualifying Project (MQP), which is a senior project at Worcester Polytechnic Institute (WPI) [CLM+00].

Another area for future work is combining simulation techniques with experiments. Currently, NS is starting to support emulation, in which a machine running the simulator gets a stream of packets from a host through a network interface, simulates a network adding its effects to the stream and passes it to another host. Although it may sound attractive, emulation has a limitation that the size and capacity of a network that NS can emulate is strictly determined by the processing power of the machine. For example, simulating the network shown in Chapter 4 took much more than the simulated time under the Linux powered Intel Pentium 200Mhz machine with 64M memory and a relatively slow hard disk drive. However, it appears that emulating a small size network is more than possible even with low performance machines such as the one we used, and offers lots of options to researchers. For example, in experiments, it is hard to generate a

relatively large link delay. But using emulation this job can be simply done by setting up a virtual network of two nodes that is connected by a link with desired delay.

The last future work that may be interesting is to apply the dynamic three-class bandwidth management concept of D-CBT into Class-Based Queuing (CBQ) [FJ95, Fl]. This would be similar to variety of Fair Queuing (FQ) [DH90] in that the output bandwidth is dynamically managed. However, the mechanism, say D-CBQ, is different from common FQ mechanisms in that bandwidth management is done in class level, and thus reduces the overhead of per-flow accounting. It appears that CBQ is a mechanism that could possibly offer a significant improvement for multimedia jitter due to its packet or link scheduling. However, the downside of CBQ is the complexity of the scheduling that supports hierarchical queues, its potential unfairness in bandwidth management and lack of a congestion avoidance mechanism. Therefore, reinforcing its fairness with a simple three-class dynamic scheduling and adding an appropriate congestion avoidance mechanism to the queues for responsive flows could be a very interesting combination.

6. Conclusion

In this thesis, we have presented the design and evaluation of our proposed router queue mechanisms, Dynamic Class-Based Threshold (D-CBT) and Cut-In Packet Scheduling (ChIPS), by comparing their performance with that of RED and CBT. D-CBT is a new active queue management mechanism that addresses the problem of fairness by grouping flows into TCP, tagged (flow-controlled multimedia) UDP and untagged (other) UDP classes and regulating the average queue usage of the UDP classes to their fair shares. ChIPS is a multimedia-favored lightweight packet scheduling mechanism that can substitute the FCFS enqueue style packet scheduling part of a RED-managed queue for D-CBT and possibly for other RED-like queue mechanisms.

Prior to the evaluation, we built a general purpose and a trace driven MPEG-1 multimedia traffic generator that use a rate-based flow control mechanism since NS did not support such traffic generators, while tagged multimedia flows were essential to evaluate D-CBT and ChIPS. While designing and testing the multimedia traffic generators trying to make their congestion response behavior similar to TCP, we found out that multimedia applications that use a rate-based flow control mechanism can be greedier than TCP that uses a window-based mechanism. This is because a rate-based multimedia application will inject packets into the network at the rate lastly determined while TCP could be waiting for ACK packets after transmit up to the congestion window size, or even get timed out. However, we found a strong reason why multimedia applications today do not want to use TCP as the underlying transport agent. That is,

TCP's bursty transmission policy introduces high jitter, and the fact that it does not separate flow control from loss recovery gives possibly unbounded delays.

After developing the traffic generators, we implemented CBT, D-CBT and ChIPS on NS, and validated CBT by simulating previous CBT experiments. Then, we compared the performance of RED, CBT, D-CBT and D-CBT with ChIPS under heterogeneous traffic mixes in which tagged flows were generated with the general-purpose multimedia application. In the comparison, we primarily focused on fairness and congested link utilization in order to evaluate the improved performance of D-CBT and to measure ChIPS effect on fairness and link utilization. Also, we measured and compared tagged multimedia jitter and packet drop rate for the systems that use D-CBT and D-CBT with ChIPS.

As expected, RED, that has shown to be fair among TCP, showed an extreme unfairness with mixed traffic. The result showed that under RED, high bandwidth unresponsive flows get most of the bandwidth up to close to their transmission rates while TCP flows suffer from severe starvation. CBT that uses a fixed threshold on UDP classes was able to avoid extreme unfairness. However, during the analysis, we found that CBT has a potential defect in its design, which we refer to as "unsynchronized weighted-average updates". That is, the ratio between independently updated UDP class averages and RED average does not correctly indicate the actual class bandwidth utilization ratio, since whichever flows updates the average more frequently will have higher weighted-average than the others will, although they all use the same amount of bandwidth. Therefore, in

CBT, bandwidth utilization ratio between each class was not only determined by the UDP thresholds but also considerably affected by the number of incoming packets that belonged to each class.

D-CBT fixes CBT's problem by dynamically determining the UDP thresholds to cooperate with RED by fairly assigning the output bandwidth to each class for all traffic mixes. The simulation result shows that ChIPS, when used with D-CBT, does not degrade the fairness when the packet enqueue ratio between tagged and other flows are about 30%, and improves tagged (flow-controlled multimedia) jitter significantly. Table 6.1.1 shows summary of our results.

	Fair to TCP	Fair to Mixed (Fixed)	Fair to Mixed (Variable)	Improve MM Jitter
RED	✓			
CBT	✓	✓		
D-CBT	✓	✓	✓	
D-CBT w/ ChIPS	✓	✓	✓	✓

<**Table 6.1.1**> Comparison of the Queue Mechanisms (✓ = yes)

Thus, we have shown that D-CBT, through the class-based accounting, protects TCP from the effect of UDP flows and also fairly protects tagged UDP flows from untagged flows. We also have shown that ChIPS can improve multimedia jitter without degrading the fairness. Further contributions we made through this thesis are the design and

implementation of two flow-controlled multimedia traffic generators in NS as well as D-CBT and ChIPS, and the first study of multimedia jitter due to network queue management policy.

7. References

[AL97] Ahonen, J. and Laine, A., "Realtime speech and voice transmission on the Internet", In *Proceedings of the HUT Internetworking Seminar*, Department of Computer Science, Helsinki University of Technology, Espoo, Finland, May 1997, URL: http://www.tcm.hut.fi/Opinnot/Tik-110.551/1997/seminar_paper.html

[AP97] Anttila, I. And Paakkunainen, M., "Transferring real-time video on the Internet", In *Proceedings of the HUT Internetworking Seminar*, Department of Computer Science, Helsinki University of Technology, Espoo, Finland, May 1997, URL: http://www.tcm.hut.fi/Opinnot/Tik-110.551/1997/iwsem.html

[BRS99] Balakrishnan, H., Rahul, H. S. and Seshan, S., "An Integrated Congestion Management Architecture for Internet Hosts", In *Proceedings of SIGCOMM '99*, Cambridge, MA, September 1999

[CLM+00] Chiller, A., Leazard, N., Maldonado, M. and Mercado E., "Implementation of Dynamic CBT", *MQP-MLC-DC99*, Spring 2000

[CT99] Claypool, M. and Tanner, J., "The Effects of Jitter on the Perceptual Quality of Video", *ACM Multimedia Conference*, Volume 2, Orlando, FL, October 30 - November 5, 1999

[DH90] Davin, J. R. and Heybey, A. T., "A Simulation Study of Fair Queueing and Policy Enforcement", *Computer Communication Review*, Vol. 20, No. 5, pp. 23-29, October 1990

[DHH+93] Delgrossi, L., Halstrick, C., Hehmann, D., Herrtwich, R. G., Krone, O., Sandvoss, J. and Vogt, C., "Media Scaling for Audiovisual Communication with the Heidelberg Transport System", In *Proceedings of the conference on Multimedia '93*, Anaheim, CA, August 1993

[FF96] Floyd, S. and Fall, K., "Simulation-based Comparisons of Tahoe, Reno, and SACK TCP", *Computer Communication Review, V.26 N.3*, July 1996, pp. 5-21

[FF97] Floyd, S. and Fall, K., "NS Simulator Tests for Random Early Detection (RED) Queue Management", *Lawrence Berkeley Laboratory, One Cyclotron Road, Berkeley, CA 94704*, April 29, 1997

[FF98] Floyd, S. and Fall, K., "Promoting the Use of End-to-End Congestion Control in the Internet", *IEEE/ACM Transactions on Networking*, February 1998

[FJ93] Floyd, S. and Jacobson, V., "Random Early Detection Gateways for Congestion Avoidance", *IEEE/ACM Transactions on Networking*, August 1993

[FJ95] Floyd, S. and Jacobson, V., "Link-sharing and Resource management Models for Packet Networks", *IEEE/ACM Transactions on Networking*, Vol. 3 No. 4, August 1995

[Fl] Floyd, S., "References on CBQ (Class-Based Queueing)", URL http://www-nrg.ee.lbl.gov/floyd/cbq.html

[Fl94] Floyd, S., "TCP and Explicit Congestion Notification", *Computer Communication Review*, October 1994

[GBC98] Gerek, J., Buchanan, W. and Claypool, M., "MMLIB – a library for end-to-end simulation of multimedia over a WAN", Technical Report MQP MLC-ML98, Worcester Polytechnic Institute, May 1998

[GP97] Guo, X. and Pattinson, C., "Quality of Service Requirements for Multimedia Communications", In *Proceedings of Time and the Web Workshop*, Staffordshire University, Staffordshire, UK, June 1997, appeared in SIGCHI Bulletin, January 1998, URL: http://www.soc.staffs.ac.uk/seminars/web97/papers/guo.html

[Ha97] Handley, M., "An Examination of Mbone Performance", USC Information of Science Institute Research Report: ISI/RR-97-450, 1997

[HSK98] Hardman, V., Sasse, M. A. and Kouvelas, I., "Successful Multiparty Audio Communication", *Communications of the ACM*, May 1998, vol. 41, No. 5

[Ja91] Jain, R., "The Art of Computer Systems Performance Analysis: Techniques for Experimental Design, Measurement, Simulation, and Modeling", John Wiley & Sons, Inc., New York, NY, 1991

[Jv88] Jacobson, V., "Congestion Avoidance and Control", In *Proceedings of SIGCOMM '88*, Stanford, CA, August 1988

[LC00] Liu, Y. and Claypool, M., "Using Redundancy to Repair Video Damaged by Network Data Loss", In *Proceedings of IS&T/ACM/SPIE Multimedia Computing and Networking 2000 (MMCN00)* January 25-27, 2000. San Jose, California, USA

[lc95] Multimedia Communications Forum, Inc. "Multimedia Communications Quality of Service", *MMCF/95-010, Approved Rev 1.0*, 1995, URL: http://www.luxcom.com/library/2000/mm_qos/qos.htm

[LM97] Lin, D. and Morris R., "Dynamics of Random Early Detection", In *Proceedings of SIGCOMM '97*, Cannes, France, September 1997

[MPFG97] Mitchell, J.L., Pennebaker, W.B., Fogg, C.E. and LeGall, D.J., "MPEG Video Compression Standard", *In Digital Multimedia Standards Series*, Chapman & Hall, New York, NY, 1997

[NJZ97] Nichols, K., Jacobson, V., Zhang, L., "Two-bit Differentiated Services Architecture for the Internet", Internet draft, work in progress, 1997

[ns2] VINT, "Virtual InterNetwork Testbed, A Collaboration among USC/ISI, Xerox PARC, LBNL, and UCB", URL: http://netweb.usc.edu/vint/

[PJS99] Parris, M., Jeffay, K. and Smith, F. D., "Lightweight Active Router-Queue Management for Multimedia Networking", *Multimedia Computing and Networking*, SPIE Proceedings Series, Vol. 3020, San Jose, CA, January 1999

[rfc2001] Stevens, W., "TCP Slow Start, Congestion Avoidance, Fast Retransmit, and Fast Recovery Algorithms", Internet Request for Comment 2001, Network Working Group, January 1997, URL: http://www.cis.ohio-state.edu/htbin/rfc/rfc2001.html

[RK99] Raghavendra, A. M. and Kinicki, R. E., " A Simulation Performance Study of TCP Vegas and Random Early Detection", *IPCCC99*, February 1999

[WKC+97] Walpole, J., Koster R., Cen, S., Cowan, C., Maier, D., McNamee, D., Pu, C., Steere, D. and Yu, L., "A player for adaptive MPEG video streaming over the Internet", In *Proceedings 26th Applied Imagery Pattern Recognition Workshop AIPR-97*, SPIE, Washington DC, October 1997