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Virtual-Time Round-Robin: An $O(1)$ Proportional Share Scheduler

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Abstract

Proportional share resource management provides a flexible and useful abstraction for multiplexing time-shared resources. However, previous proportional share mechanisms have either weak proportional sharing accuracy or high scheduling overhead. We present Virtual-Time Round-Robin (VTRR), a proportional share scheduler that can provide good proportional sharing accuracy with $O(1)$ scheduling overhead. VTRR achieves this by combining the benefits of fair queueing algorithms with a round-robin scheduling mechanism. Unlike many other schedulers, VTRR is simple to implement. We have implemented a VTRR CPU scheduler in Linux in less than 100 lines of code. Our performance results demonstrate that VTRR provides accurate proportional share allocation with constant, sub-microsecond scheduling overhead. The scheduling overhead using VTRR is two orders of magnitude less than the standard Linux scheduler for large numbers of clients.

1 Introduction

Proportional share resource management provides a flexible and useful abstraction for multiplexing scarce resources among users and applications. The basic idea is that each client has an associated weight, and resources are allocated to the clients in proportion to their respective weights. Because of its usefulness, many proportional share scheduling mechanisms have been developed [3, 8, 10, 14, 16, 18, 25, 28, 30, 33, 34]. In addition, higher-level abstractions have been developed on top of these proportional share mechanisms to support flexible, modular resource management policies [30, 33].

Proportional share scheduling mechanisms were first developed decades ago with the introduction of weighted round-robin scheduling [29]. Later, fair-share algorithms based on controlling priority values were developed and incorporated into some UNIX operating systems [10, 16, 18]. These earlier mechanisms were typi-

cally fast, requiring only constant time to select a client for execution. However, they were limited in the accuracy with which they could achieve proportional sharing. As a result starting in the late 1980s, fair queueing algorithms were developed [3, 8, 14, 25, 28, 30, 33, 34], first for network packet scheduling and later for CPU scheduling. These algorithms provided better proportional sharing accuracy. However, the time to select a client for execution using these algorithms grows with the number of clients. Most implementations require linear time to select a client for execution. For server systems which may service large numbers of clients, the scheduling overhead of linear time algorithms can waste more than 20 percent of system resources [5] for large numbers of clients. Hierarchical data structures can be used to reduce the selection time complexity, but they are not generally used as they are often less efficient in practice. This is because they add implementation complexity and their performance depends on being able to balance the data structures efficiently.

In this paper, we introduce VTRR, a Virtual-Time Round-Robin scheduler for proportional share resource management. VTRR combines the benefits of low overhead round-robin execution with high accuracy virtual-time allocations. It provides accurate control over client computation rates, and it can schedule clients for execution in $O(1)$ time. The constant scheduling overhead makes VTRR particularly suitable for server systems that must manage large numbers of clients. VTRR is simple to implement and can be easily incorporated into existing scheduling frameworks in commercial operating systems. We have implemented a prototype VTRR CPU scheduler in Linux in less than 100 lines of code. We have compared our VTRR Linux prototype against schedulers commonly used in practice and research, including the standard Linux scheduler [2] and fair queueing. Our performance results on micro-benchmarks and real applications demonstrate that VTRR delivers excellent proportional share control with lower scheduling overhead than other approaches.

This paper is organized as follows: Section 2 dis-

cusses background and related work. Section 3 presents the VTRR scheduling algorithm. Section 4 describes our prototype Linux implementation. Section 5 presents performance results from both simulation studies and real kernel measurements that compare VTRR against weighted round-robin, fair queueing, and standard Linux scheduling. Finally, we present some concluding remarks and directions for future work.

2 Background

Previous proportional sharing mechanisms can be classified into four categories: those that are fast but have weaker proportional fairness guarantees, those that map well to existing scheduler frameworks in current commercial operating systems but have no well-defined proportional fairness guarantees, those that have strong proportional fairness guarantees and higher scheduling overhead, and those that have weaker proportional fairness guarantees but have higher scheduling overhead. The four categories correspond to round-robin, fair-share, fair queueing, and lottery mechanisms.

To discuss these different approaches, we first present in Section 2.1 a simple proportional share model for scheduling a time-multiplexed resource and more precisely define the notion of proportional fairness. In Sections 2.2 to 2.4, we use this background to explain the round-robin, fair-share, fair queueing, and lottery sharing mechanisms in further detail. We briefly mention other related work in Section 2.6.

2.1 Proportional Fairness

Proportional share scheduling has a clear colloquial meaning: given a set of clients with associated weights, a proportional share scheduler should allocate resources to each client in proportion to its respective weight. In this paper, we use the term share and weight interchangeably. Without loss of generality, we can model the process of scheduling a time-multiplexed resource among a set of clients in two steps: 1) the scheduler orders the clients in a queue, 2) the scheduler runs the first client in the queue for its *time quantum*, which is the maximum time interval the client is allowed to run before another scheduling decision is made. Note that the time quantum is typically expressed in time units of constant size determined by the hardware. As a result, we refer to the units of time quanta as time units (tu) in this paper rather than an absolute time measure such as seconds.

Based on the above scheduler model, a scheduler can achieve proportional sharing in one of two ways. One way is to adjust the frequency that a client is selected to

run by adjusting the position of the client in the queue so that it ends up at the front of the queue more or less often. The other way is to adjust the size of the time quantum of a client so that it runs longer for a given allocation. The manner in which a scheduler determines how often a client runs and how long a client runs directly affects the accuracy and scheduling overhead of the scheduler.

A proportional share scheduler is more accurate if it allocates resources in a manner that is more proportionally fair. We can formalize this notion of proportional fairness in more technical terms. The definition we use is a simple one that suffices for our discussion; more extended definitions are presented in [12, 15, 26, 32]. Our definition draws heavily from the ideal sharing mechanism GPS [19]. To simplify the discussion, we assume that clients do not sleep or block and can consume whatever resources they are allocated.

We first define *perfect fairness*, an ideal state in which each client has received service exactly proportional to its share. We denote the proportional share of client A as S_A , and the amount of service received by client A during the time interval (t_1, t_2) as $W_A(t_1, t_2)$. Formally, a proportional sharing algorithm achieves perfect fairness for time interval (t_1, t_2) if, for any client A ,

$$W_A(t_1, t_2) = (t_2 - t_1) \frac{S_A}{\sum_i S_i} \quad (1)$$

If we had an ideal system in which all clients could consume their resource allocations simultaneously, then an ideal proportional share scheduler could maintain the above relationship for all time intervals. However, in scheduling a time-multiplexed resource in time units of finite size, it is not possible for a scheduler to be perfectly proportionally fair as defined by Equation 1 for all intervals.

Although no real-world scheduling algorithm can maintain perfect fairness, some algorithms stay closer to perfect fairness than others. To evaluate the fairness performance of a proportional sharing mechanism, we must quantify how close an algorithm gets to perfect fairness. We can use a variation of Equation 1 to define the *service time error* $E_A(t_1, t_2)$ for client A over interval (t_1, t_2) . The error is the difference between the amount time allocated to the client during interval (t_1, t_2) under the given algorithm, and the amount of time that would have been allocated under an ideal scheme that maintains perfect fairness for all clients over all intervals. Service time error is computed as:

$$E_A(t_1, t_2) = W_A(t_1, t_2) - (t_2 - t_1) \frac{S_A}{\sum_i S_i} \quad (2)$$

A positive service time error indicates that a client has received more than its ideal share over an interval; a negative error indicates that a client has received less. To be

precise, the error E_A measures how much time client A has received beyond its ideal allocation.

The goal of a proportional share scheduler should be to minimize the allocation error between clients. In this context, we now consider how effectively different classes of proportional share algorithms are in minimizing this allocation error.

2.2 Round-Robin

One of the oldest, simplest and most widely used proportional share scheduling algorithms is round-robin. Clients are placed in a queue and allowed to execute in turn. When all client shares are equal, each client is assigned the same size time quantum. In the weighted round-robin case, each client is assigned a time quantum equal to its share. A client with a larger share, then, effectively gets a larger quantum than a client with a small share. Weighted round-robin (WRR) provides proportional sharing by running all clients with the same frequency but adjusting the size of their time quanta. A more recent variant called deficit round-robin [28] has been developed for network packet scheduling with similar behavior to a weighted round-robin CPU scheduler.

WRR is simple to implement and schedules clients in $O(1)$ time. However, it has a relatively weak proportional fairness guarantee as its service ratio error can be quite large. Consider an example in which 3 clients A , B , and C , have shares 3, 2, and 1, respectively. WRR will execute these clients in the following order of time units: A, A, A, B, B, C . The error in this example gets as low as -1 tu and as high as $+1.5$ tu. The real trouble comes with large share values: if the shares in the previous example are changed to 3000, 2000, and 1000, the error ranges instead from -1000 to $+1500$ tu. A large error range like this illustrates the major drawback of round-robin scheduling: each client gets all service due to it all at once, while other clients get no service. After a client has received all its service, it is well ahead of its ideal allocation (it has a high positive error), and all other clients are behind their allocations (they have low negative errors).

2.3 Fair-Share

Fair-share schedulers [10, 16, 18] arose as a result of a need to provide proportional sharing among users in a way compatible with a UNIX-style time-sharing framework. In UNIX time-sharing, scheduling is done based on multi-level feedback with a set of priority queues. Each client has a priority which is adjusted as it executes. The scheduler executes the client with the highest priority. The idea of fair-share was to provide proportional sharing among users by adjusting the priorities of a

user's clients in a suitable way. Fair-share provides proportional sharing by effectively running clients at different frequencies, as opposed to WRR which only adjusts the size of the clients' time quanta. Fair-share schedulers were compatible with UNIX scheduling frameworks and relatively easy to deploy in existing UNIX environments. Unlike round-robin scheduling, the focus was on providing proportional sharing to groups of users as opposed to individual clients. However, the approaches were often ad-hoc and it is difficult to formalize the proportional fairness guarantees they provided. Empirical measurements of show that these approaches only provide reasonable proportional fairness over relatively large time intervals [10]. It is almost certainly the case that the allocation errors in these approaches can be very large.

The priority adjustments done by fair-share schedulers can generally be computed quickly in $O(1)$ time. In some cases, the schedulers need to do an expensive periodic re-adjustment of all client priorities, which required $O(N)$ time, where N is the number of clients.

2.4 Fair Queueing

Fair queueing was first proposed by Demers et. al. for network packet scheduling as Weighted Fair Queueing (WFQ) [8], with a more extensive analysis provided by Parekh and Gallager [25], and later applied by Waldspurger and Wehl to CPU scheduling as stride scheduling [33]. WFQ introduced the idea of a virtual finishing time (VFT) to do proportional sharing scheduling. To explain what a VFT is, we first explain the notion of virtual time. The *virtual time* of a client is a measure of the degree to which a client has received its proportional allocation relative to other clients. When a client executes, its virtual time advances at a rate inversely proportional to the client's share. In other words, the virtual time of a client A at time t is the ratio of $W_A(t)$ to S_A :

$$VT_A(t) = \frac{W_A(t)}{S_A} \quad (3)$$

Given a client's virtual time, the client's *virtual finishing time* (VFT) is defined as the virtual time the client would have after executing for one time quantum. WFQ then schedules clients by selecting the client with the smallest VFT. This is implemented by keeping an ordered queue of clients sorted from smallest to largest VFT, and then selecting the first client in the queue. After a client executes, its VFT is updated and the client is inserted back into the queue. Its position in the queue is determined by its updated VFT. Fair queueing provides proportional sharing by running clients at different frequencies by adjusting the position in at which each client is inserted back into the queue; the same size time

quantum is used for all clients.

To illustrate how this works, consider again the example in which 3 clients A, B, and C, have shares 3, 2, and 1, respectively. Their initial VFTs are then 1/3, 1/2, and 1, respectively. WFQ would then execute the clients in the following order of time units: A, B, A, B, C, A. In contrast to WRR, WFQ’s service time error ranges from $-5/6$ to $+1$ tu in this example, which is less than the allocation error of -1 to $+1.5$ tu for WRR. The difference between WFQ and WRR is greatly exaggerated if larger share values are chosen: if we make the shares 3000, 2000, and 1000 instead of 3, 2, and 1, WFQ has the same service time error range while WRR’s error range balloons to -1000 to $+1500$ tu.

It has been shown that WFQ guarantees that the service time error for any client never falls below -1 , which means that a client can never fall behind its ideal allocation by more than a single time quantum [25]. More recent fair queueing algorithms [3, 30] provide more accurate proportional sharing (by also guaranteeing an upper bound on error) at the expense of additional scheduling overhead. Fair queueing provides stronger proportional fairness guarantees than round-robin or fair-share scheduling. Unfortunately, fair queueing is more difficult to implement, and the time it takes to select a client to execute is $O(N)$ time for most implementations, where N is the number of clients. With more complex data structures, it is possible to implement fair queueing such that selection of a client requires $O(\log N)$ time. However, the added difficulty of managing complex data structures in kernel space causes most implementers of fair queueing to choose the more straightforward $O(N)$ implementation.

2.5 Lottery

Lottery scheduling was proposed by Waldspurger and Weihl [33] after WFQ was first developed. In lottery scheduling, each client is given a number of tickets proportional to its share. A ticket is then randomly selected by the scheduler and the client that owns the selected ticket is scheduled to run for a time quantum. Like fair queueing, lottery scheduling provides proportional sharing by running clients at different frequencies by adjusting the position in at which each client is inserted back into the queue; the same size time quantum is typically used for all clients.

Lottery scheduling is somewhat simpler to implement than fair queueing, but has the same high scheduling overhead as fair queueing, $O(N)$ for most implementations or $O(\log N)$ for more complex data structures. However, because lottery scheduling relies on the law of large numbers for providing proportional fairness, its accuracy is much worse than WFQ [33], and is also worse

than WRR for smaller share values.

2.6 Other Related Work

Higher-level resource management abstractions have also been developed [1, 33], and a number of these abstractions can be used with proportional share scheduling mechanisms. This work is complementary to our focus here on the underlying scheduling mechanisms. Other scheduling work has also been done in supporting clients with real-time requirements [4, 6, 13, 17, 20, 21, 22, 23, 24] and improving the response time of interactive clients [9, 11, 24]. Considering these issues in depth is beyond the scope of this paper.

3 VTRR Scheduling

VTRR is an accurate, low-overhead proportional share scheduler for multiplexing time-shared resources among a set of clients. VTRR combines the benefit of low overhead round-robin scheduling with the high accuracy mechanisms of virtual time and virtual finishing time used in fair queueing algorithms. At a high-level, the VTRR scheduling algorithm can be briefly described in three steps:

1. Order the clients in the run queue from largest to smallest share. Unlike fair queueing, a client’s position on the run queue only changes when its share changes, an infrequent event, not on each scheduling decision.
2. Starting from the beginning of the run queue, run each client for one time quantum in a round-robin manner. VTRR uses the fixed ordering property of round-robin in order to choose in constant time which client to run. Unlike round-robin, the time quantum is the same size for all clients.
3. In step 2, if a client has received more than its proportional allocation, skip the remaining clients in the run queue and start running clients from the beginning of the run queue again. Since the clients with larger share values are placed first in the queue, this allows them to get more service than the lower-share clients at the end of the queue.

To provide a more in depth description of VTRR, we first define the state VTRR associates with each client, then describe precisely how VTRR uses that state to schedule clients. In VTRR, a client has five values associated with its execution state: share, virtual finishing time, time counter, id number, and run state. A client’s *share* defines its resource rights. Each client receives

a resource allocation that is directly proportional to its share. A client’s *virtual finishing time* (VFT) is defined in the same way as in Section 2.4. Since a client has a VFT, it also has an implicit virtual time. A client’s VFT advances at a rate proportional to its resource consumption divided by its share. The VFT is used to decide when VTRR should reset to the first client in the queue. This is described in greater detail in Section 3.1, below. A client’s *time counter* ensures that the pattern of allocations is periodic, and that perfect fairness is achieved at the end of each period. Specifically, the time counter tracks the number of quanta the client must receive before the period is over and perfect fairness is reached. A client’s *id number* is a unique client identifier that is assigned when the client is created. A client’s *run state* is an indication of whether or not the client can be executed. A client is *runnable* if it can be executed, and not runnable if it cannot. For example for a CPU scheduler, a client would not be runnable if it is blocked waiting for I/O and cannot execute.

3.1 Basic VTRR Algorithm

We will initially only consider runnable clients in our discussion of the basic VTRR scheduling algorithm. We will discuss dynamic changes in a client’s run state in Section 3.2. VTRR maintains the following scheduler state: time quantum, run queue, total shares, and queue virtual time. As discussed in Section 2.1, the *time quantum* is the duration of a standard time slice assigned to a client to execute. The *run queue* is a sorted queue of all runnable clients ordered from largest to smallest share client. Ties can be broken either arbitrarily or using the client id numbers, which are unique. The *total shares* is the sum of the shares of all runnable clients. The *queue virtual time* (QVT) is a measure of what a client’s VFT should be if it has received exactly its proportional share allocation.

Previous work in the domain of packet scheduling provides the theoretical basis for the QVT [8, 25]. The QVT advances whenever a client executes at a rate inversely proportional to the total shares. If we denote the system time quantum as Q and the share of client i as S_i , then the QVT is updated as follows:

$$QVT(t+Q) = QVT(t) + \frac{Q}{\sum_i S_i} \quad (4)$$

The difference between the QVT and a client’s virtual time is a measure of whether the respective client has consumed its proportional allocation of resources. If a client’s virtual time is equal to the queue virtual time, it is considered to have received its proportional allocation of resources. An earlier virtual time indicates that the client has used less than its proportional share. Sim-

ilarly, a later virtual time indicates that it has used more than its proportional share. Since the QVT advances at the same rate for all clients on the run queue, the relative magnitudes of the virtual times provide a relative measure of the degree to which each client has received its proportional share of resources.

First, we explain the role of the time counters in VTRR. In relation to this, we define a *scheduling cycle* as a sequence of allocations whose length is equal to the sum of all client shares. For example, for a queue of three clients with shares 3, 2, and 1, a scheduling cycle is a sequence of 6 allocations. The time counter for each client is reset at the beginning of each scheduling cycle to the client’s share value, and is decremented every time a client receives a time quantum. VTRR uses the time counters to ensure that perfect fairness is attained at the end of every scheduling cycle. At the end of the cycle, every counter is zero, meaning that for each client A , the number of quanta received during the cycle is exactly S_A , the client’s share value. Clearly, then, each client has received service proportional to its share. In order to guarantee that all counters are zero at the end of the cycle, we enforce an invariant on the queue, called the *time counter invariant*: we require that, for any two consecutive clients in the queue A and B , the counter value for B must always be no greater than the counter value for A .

The VTRR scheduling algorithm starts at the beginning of the run queue and executes the first client for one time quantum. We refer to the client selected for execution as the *current client*. Once the current client has completed its time quantum, its time counter is decremented by one and its VFT is incremented by the time quantum divided by its share. If we denote the system time quantum as Q , the current client’s share as S_C , and the current client’s VFT as $VFT_C(t)$, $VFT_C(t)$ is updated as follows:

$$VFT_C(t+Q) = VFT_C(t) + \frac{Q}{S_C} \quad (5)$$

The scheduler then moves on to the next client in the run queue. First, the scheduler checks for violation of the time counter invariant: if the counter value of the next client is greater than the counter of the current client, the scheduler makes the next client the current client and executes it for a quantum, without question. This causes its counter to be decremented, preserving the invariant. If the next client’s counter is not greater than the current client’s counter, the time counter invariant cannot be violated whether the next client is run or not, so the scheduler makes a decision using virtual time: the scheduler compares the VFT of the next client with the QVT the system would have after the next time quantum a client executes. We call this comparison the *VFT*

inequality. If we denote the system time quantum as Q , the current client’s VFT as $VFT_C(t)$, and its share as S_C , the VFT inequality is true if:

$$VFT_C(t) - QVT(t + Q) < \frac{Q}{S_C} \quad (6)$$

If the VFT inequality is true, the scheduler selects and executes the next client in the run queue for one time quantum and the process repeats with the subsequent clients in the run queue. If the scheduler reaches a point in the run queue when the VFT inequality is not true, the scheduler returns to the beginning of the run queue and selects the first client to execute. At the end of the scheduling cycle, when the time counters of all clients reach zero, the time counters are all reset to their initial values corresponding to the respective client’s share, and the scheduler starts from the beginning of the run queue again to select a client to execute. Note that throughout this scheduling process, the ordering of clients on the run queue does not change.

To illustrate how this works, consider again the example in which 3 clients A, B, and C, have shares 3, 2, and 1, respectively. Their initial VFTs are then 1/3, 1/2, and 1, respectively. VTRR would then execute the clients in the following repeating order of time units: A, B, C, A, B, A. In contrast to WRR and WFQ, VTRR has a maximum allocation error between A and B of 1/3 tu in this example. This allocation error is much better than WRR and comparable to WFQ.

Since VTRR simply selects each client in turn to execute, selecting a client for execution can be done in $O(1)$ time. We defer a more detailed discussion of the complexity of VTRR to Section 3.3.

3.2 VTRR Dynamic Considerations

In the previous section, we presented the basic VTRR scheduling algorithm, but we did not discuss how VTRR deals with dynamic considerations that are a necessary part of any on-line scheduling algorithm. We now discuss how VTRR allows clients to be dynamically created, terminated, change run state, and change their share assignments.

We distinguish between clients that are runnable and not runnable. As mentioned earlier, clients that are runnable can be selected for execution by the scheduler, while clients that are not runnable cannot. Only runnable clients are placed in the run queue. With no loss of generality, we assume that a client is created before it can become runnable, and a client becomes not runnable before it is terminated. As a result, client creation and termination have no affect on the VTRR run queue.

When a client becomes runnable, it is inserted into the run queue so that the run queue remains sorted from

largest to smallest share client. Ties can be broken either arbitrarily or using the unique client id numbers. One issue remains, which is how to determine the new client’s initial VFT. When a client is created and becomes runnable, it has not yet consumed any resources, so it is neither below or above its proportional share in terms of resource consumption. As a result, we set the client’s implicit virtual time to be the same as the QVT. We can then calculate the VFT of a new client A with share S_A as:

$$VFT_A(t) = QVT_A(t) + \frac{Q}{S_A} \quad (7)$$

After a client is executed, it may become not runnable. If the client is the current client and becomes not runnable, it is preempted and another client is selected by the scheduler using the basic algorithm described in Section 3.1. The client that is not runnable is removed from the run queue. If the client becomes not runnable and is not the current client, the client is simply removed from the run queue. While the client is not runnable, its VFT is not updated. When the client is removed from the run queue, it records the client that was before it on the run queue, and the client that was after it on the run queue. We refer to these clients as the *last-previous* client and *last-next* client, respectively.

When a client that is not runnable becomes runnable again, VTRR inserts the now runnable client back into the run queue. If the client’s references to its last-previous and last-next client are still valid, it can use those references to determine its position in the run queue. If either the last-previous or the last-next reference is not valid, VTRR then simply traverses the run queue to find the insertion point for the now runnable client.

Determining whether the last-previous and last-next references are valid can be done efficiently as follows. The last-previous and last-next client references are valid if both clients have not exited and are runnable, if there are no clients between them on the run queue, and if the share of the newly-runnable client is no more than the last-previous client and no less than the last-next client. Care must be taken, however, to ensure that the last-previous and last-next references are still valid before dereferencing them: if either client has exited and been deallocated, last-previous and last-next may no longer refer to valid memory regions. To deal with this, a hash table can be kept that stores identifiers of valid clients. Hash function collisions can be resolved by simple replacement, so the table can be implemented as an array of identifiers. A client’s identifier is put into the table when it is created, and deleted when the client exits. The last-previous and last-next pointers are not dereferenced, then, unless the identifier of the last-previous and

last-next clients exist in the hash table. As described in Section 4, the use of a hash table was not necessary in our Linux VTRR implementation.

Once the now runnable client has been inserted in the run queue, the client’s VFT must be updated. The update is analogous to the VFT initialization used when a new client becomes runnable. The difference is that we also account for the client’s original VFT in updating the VFT. If we denote the original VFT of a client A as $VFT_A(t')$, then the client’s VFT is updated as follows:

$$VFT_A(t) = \text{MAX}\left\{QVT_A(t) + \frac{Q}{S_A}, VFT_A(t')\right\} \quad (8)$$

This treats a client that has been not runnable for a while like a new client that has not yet executed. At the same time, the system keeps track of the client’s VFT so that if it that has recently used more than its proportional allocation, it cannot somehow game the system by making itself not runnable and becoming runnable again.

We use an analogous policy to set the initial value of a client’s time counter. A client’s time counter tracks the number of quanta due to the client before the end of the current scheduling cycle, and is reset at the beginning of each new cycle. We set the time counter of a newly-inserted client to a value which will give it the correct proportion of remaining quanta in this cycle. The counter C_A for the new client A is computed:

$$C_A = \frac{S_A}{\sum_i S_i} \sum_i C_i \quad (9)$$

Note that this is computed *before* client A is inserted, so S_A is not included in the $\sum_i S_i$ summation.

This value is modified by a rule similar to the rule enacted for the VFT: we require that a client cannot come back in the same cycle and receive a larger time count than it had previously. Therefore, if the client is being inserted during the same cycle in which it was removed, the counter is set to the minimum of C_A and the previous counter value. Finally, to preserve the time counter invariant (as described in Section 3.1), the counter value must be restricted to be between the time counter values of the clients before and after the inserted client.

If a client’s share changes, there are two cases to consider based on the run state of the client. If the client is not runnable, no run queue modifications are needed. If the client is runnable and its share changes, the client’s position in the run queue may need to be changed. This operation can be simplified by removing the client from the run queue, changing the share, and then reinserting it. Removal and insertion can then be performed just as described above.

3.3 Complexity

The primary function of a scheduler is to select a client to execute when the resource is available. A key benefit of VTRR is that it can select a client to execute in $O(1)$ time. To do this, VTRR simply has to maintain a sorted run queue of clients and keep track of its current position in the run queue. Updating the current run queue position and updating a client’s VFT are both $O(1)$ time operations. While the run queue needs to be sorted by client shares, the ordering of clients on the run queue does not change in the normal process of selecting clients to execute. This is an important advantage over fair queueing algorithms, in which a client needs to be reinserted into a sorted run queue after each time it executes. As a result, fair queueing has much higher complexity than VTRR, requiring $O(N)$ time to select a client to execute, or $O(\log N)$ time if more complex data structures are used (but this is rarely implemented in practice).

When all clients on the run queue have zero counter values, VTRR resets the counter values of all clients on the run queue. The complete counter reset takes $O(N)$ time, where N is the number of clients. However, this reset is done at most once every N times the scheduler selects a client to execute (and much less frequently in practice). As a result, the reset of the time counters is amortized over many client selections so that the effective running time of VTRR is still $O(1)$ time. In addition, the counter resets can be done incrementally on the first pass through the run queue with the new counter values.

In addition to selecting a client to execute, a scheduler must also allow clients to be dynamically created and terminated, change run state, and change scheduling parameters such as a client’s share. These scheduling operations typically occur much less frequently than client selection. In VTRR, operations such as client creation and termination can be done in $O(1)$ time since they do not directly affect the run queue. Changing a client’s run state from runnable to not runnable can also be done in $O(1)$ time for any reasonable run queue implementation since all it involves is removing the respective client from the run queue. The scheduling operations with the highest complexity are those that involve changing a client’s share assignment and changing a client’s run state to runnable. In particular, a client typically becomes runnable after it is created or after an I/O operation that it was waiting for completes. If a client’s share changes, the client’s position in the run queue may have change as well. If a client becomes runnable, the client will have to be inserted into the run queue in the proper position based on its share. Using a doubly linked list run queue implementation, insertion into the sorted queue can require $O(N)$ time, where N is the number of

runnable clients. A priority queue implementation could be used for the run queue to reduce the insertion cost to $O(\log N)$, but probably does not have better overall performance than a simple sorted list in practice.

Because queue insertion is required much less frequently than client selection in practice, the queue insertion cost is not likely to dominate the scheduling cost. In particular, if only a constant number of queue insertions are required for every N times a client selection is done, then the effective cost of the queue insertions is still only $O(1)$ time. Furthermore, the most common scheduling operation that would require queue insertion is when a client becomes runnable again after it was blocked waiting on a resource. In this case, the insertion overhead can be $O(1)$ time if the last-previous client and last-next client references remain valid at queue insertion time. If the references are valid, then the position of the client is already known on the run queue so the scheduler does not have to find the insertion point.

An alternative implementation can be done that allows all queue insertions to be done in $O(1)$ time, if the range of share values is fixed in advance. The idea is similar to priority schedulers which have a fixed range of priority values and have separate run queue for each priority. Instead of using priorities, we can have a separate run queue for each share value and keep track of the run queues using an array. We can then find the queue corresponding to a client's share and insert the client at the end of the corresponding queue in $O(1)$ time. Such an implementation maps well to scheduling frameworks in a number of commercial operating systems, including Solaris [31] and Windows NT [7].

4 Implementation

We have implemented a prototype VTRR CPU scheduler in the Linux operating system. For this work, we used the Red Hat Linux version 6.1 distribution and the Linux version 2.2.12-20 kernel. We had to add or modify less than 100 lines of kernel code to complete the VTRR scheduler implementation. We describe our Linux VTRR implementation in further detail to illustrate how easy VTRR is to implement. These scheduling frameworks are commonly found in commercial operating systems. While VTRR can be used in a multiprocessor scheduling context, we only discuss the single CPU implementation here.

The Linux scheduling framework for a single CPU is based on a run queue implemented as a single doubly linked list. We first describe how the standard Linux scheduler works, and then discuss the changes we made to implement VTRR in Linux.

The standard Linux scheduler multiplexes a set of

clients that can be assigned different priorities. The priorities are used to compute a per client measure called *goodness* to schedule the set of clients. Each time the scheduler is called, the goodness value for each client in the run queue is calculated. The client with the highest goodness value is then selected as the next client to execute. In the case of ties, the first client with the highest goodness value is selected. Because the goodness of each client is calculated each time the scheduler is called, the scheduling overhead of the Linux scheduler is $O(N)$, where N is the number of runnable clients.

The standard way Linux calculates the goodness for all clients is based on a client's priority and counter. The counter is not the same as the time counter value used by VTRR, but is instead a measure of the remaining time left in a client's time quantum. The standard time unit used in Linux for the counter and time quantum is called a jiffy, which is 10 ms by default. The basic idea is that the goodness of a client is its priority plus its counter value. The client's counter is initially set equal to the client's priority, which has a value of 20 by default. Each time a client is executed for a jiffy, the client's counter is decremented. A client's counter is decremented until it drops below zero, at which point the client cannot be selected to execute. As a result, the default time quantum for each client is 21 jiffies, or 210 ms. When the counters of all runnable clients drop below zero, the scheduler resets all the counters to their initial value. There is some additional logic to support static priority real-time clients and clients that become not runnable, but an overview of the basic way in which the Linux scheduler works is sufficient for our discussion here. Further details are available elsewhere [2].

To implement VTRR in Linux, we reused much of the existing scheduling infrastructure. We used the same doubly linked list run queue structure as the standard Linux scheduler. The primary change to the run queue was sorting the clients from largest to smallest share. Rather than scanning all the clients when a scheduling decision needs to be made, our VTRR Linux implementation simply picks the next client in the run queue based on the VTRR scheduling algorithm.

One idiosyncrasy of the Linux scheduler that is relevant to this work is that the smallest counter value that may be assigned to a client is 1. This means that the smallest time quantum a client can have is 2 jiffies. To provide a comparable implementation of VTRR, the default time quantum used in our VTRR implementation is also 2 jiffies, or 20 ms.

In addition to the VTRR client state, two fields that were added to the standard client data structure in Linux were last-previous and last-next pointers which were used to optimize run queue insertion efficiency. In the Linux 2.2 kernel, memory for the client data structures

is statically allocated, and never reclaimed for anything other than new client data structures. Therefore, in our implementation, we were free to reference the last-next and last-previous pointers to check their validity, as they always refer to some client’s data; the hash table method described in Section 3.2 was unnecessary.

5 Measurements and Results

To demonstrate the effectiveness of VTRR, we have quantitatively measured and compared its performance against other leading approaches from both industrial practice and research. We have conducted both extensive simulation studies and detailed measurements of real kernel scheduler performance on real applications.

We conducted simulation studies to compare the proportional sharing accuracy of VTRR against both WRR and WFQ. We used a simulator for these studies for two reasons. First, our simulator enabled us to isolate impact of the scheduling algorithms themselves and purposefully do not include the effects of other activity present in an actual kernel implementation. Second, our simulator enabled us to examine the scheduling behavior of these different algorithms across hundreds of thousands of different combinations of clients with different share values. It would have been much more difficult to obtain this volume of data in a repeatable fashion from just measurements of a kernel scheduler implementation. Our simulation results are presented in Section 5.1.

We also conducted detailed measurements of real kernel scheduler performance by comparing our prototype VTRR Linux implementation against both the standard Linux scheduler and a WFQ scheduler. In particular, comparing against the standard Linux scheduler and measuring its performance is important because of its growing popularity as a platform for server as well as desktop systems. The experiments we have done quantify the scheduling overhead and proportional share allocation accuracy of these schedulers in a real operating system environment under a number of different workloads. Our measurements of kernel scheduler performance are presented in Sections 5.2 to 5.4.

All of our kernel scheduler measurements were performed on a Gateway 2000 E1400 system with a 433 MHz Intel Celeron CPU, 128 MB RAM, and 10 GB hard drive. The system was installed with the Red Hat Linux 6.1 distribution running the Linux version 2.2.12-20 kernel. The measurements were done by using a minimally intrusive tracing facility that logs events at significant points in the application and the operating system code. This is done via a light-weight mechanism that writes timestamped event identifiers into a memory log. The mechanism takes advantage of the high-

resolution clock cycle counter available with the Intel CPU to provide measurement resolution at the granularity of a few nanoseconds. Getting a timestamp simply involved reading the hardware cycle counter register, which could be read from user-level or kernel-level code. We measured the cost of the mechanism on the system to be roughly 70 ns per event.

The kernel scheduler measurements were performed on a fully functional system to represent a realistic system environment. By fully functional, we mean that all experiments were performed with all system functions running and the system connected to the network. At the same time, an effort was made to eliminate variations in the test environment to make the experiments repeatable.

5.1 Simulation Studies

We built a scheduling simulator that we used to evaluate the proportional fairness of VTRR in comparison to two other schedulers, WRR and WFQ. The simulator is a user-space program that measures the service time error, described in Section 2.1, of a scheduler on a set of clients. The simulator takes four inputs, the scheduling algorithm, the number of clients N , the total number of shares S , and the number of client-share combinations. The simulator randomly assigns shares to clients and scales the share values to ensure that they sum to S . It then schedules the clients using the specified algorithm as a real scheduler would, and tracks the resulting service time error. The simulator runs the scheduler until the resulting schedule repeats, then computes the maximum (most positive) and minimum (most negative) service time error across the nonrepeating portion of the schedule for the given set of clients and share assignments. The simulator assumes that all clients are runnable at all times. This process of random share allocation and scheduler simulation is repeated for the specified number of client-share combinations. We then compute an average highest service time error and average lowest service time error for the specified number of client-share combinations to obtain an “average-case” error range.

To measure proportional fairness accuracy, we ran simulations for each scheduling algorithm considered on 40 different combinations of N and S . For each set of (N, S) , we ran 10,000 client-share combinations and determined the resulting average error ranges. The average service time error ranges for VTRR, WRR, and WFQ are shown in Figures 1 and 2.

Figure 1 shows a comparison of the error ranges for VTRR versus WRR, one graph showing the error ranges for VTRR and the other showing the error ranges for WRR. Each graph shows two surfaces plotted on axes of the same scale, representing the maximum and mini-

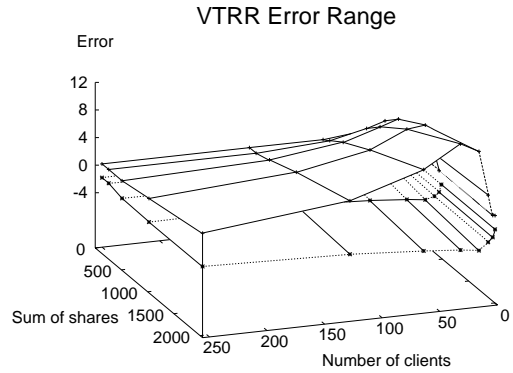
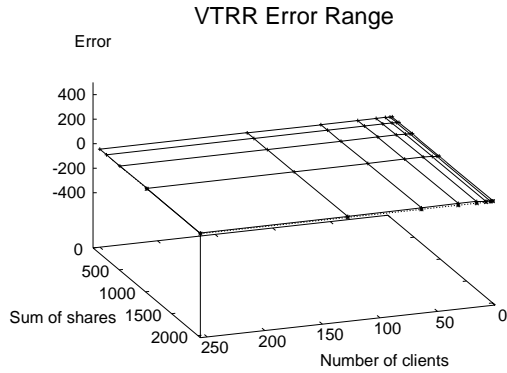
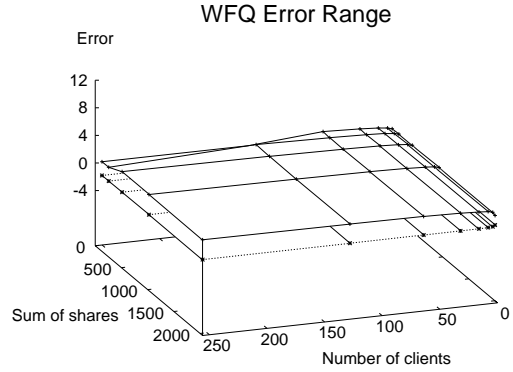
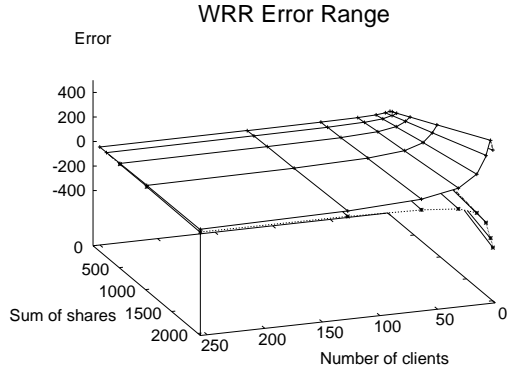


Figure 1: VTRR vs. WRR service time error

Figure 2: VTRR vs. WFQ service time error

num service time error as a function of N and S . Within the range of values of N and S shown, WRR's error range reaches as low as -398 tu and as high as 479 tu. With the time units expressed in 10 ms jiffies as in Linux, a client under WRR can on average get ahead of its correct CPU time allocation by 4.79 seconds, or behind by 3.98 seconds, which is a substantial amount of service time error. In contrast, Figure 1 shows that VTRR has a much smaller error range than WRR and is much more accurate. Because the error axis is scaled to display the wide range of WRR's error values, it is difficult to even distinguish the two surfaces for VTRR in Figure 1. VTRR's service time error only ranges from -3.8 to 10.6 tu; this can be seen more clearly in Figure 2.

Figure 2 shows a comparison of the error ranges for VTRR versus WFQ, one graph showing the error ranges for VTRR and the other showing the error ranges for WFQ. As in the case in Figure 2, each graph shows two surfaces plotted on axes of the same scale, representing the maximum and minimum service time error as a function of N and S . The VTRR graph in Figure 2 includes the same data as the VTRR graph in Figure 1, but the error axis is scaled more naturally. Within the range of values of N and S shown, WFQ's average error range reaches as low as -1 tu and as high as 2 tu, as opposed

to VTRR's error range from -3.8 to 10.6 tu. The error ranges for WFQ are smaller than VTRR, but the difference between WFQ and VTRR is much smaller than the difference between VTRR and WRR. With the time units expressed in 10 ms jiffies as in Linux, a client under WFQ can on average get ahead of its correct CPU time allocation by 10 ms, or behind by 20 ms, while a client under VTRR can get ahead by 38 ms or behind by 106 ms. In both cases, the service time errors are small. In fact, the service time errors are even below the threshold of delay noticeable by most human beings for response time on interactive applications [27]. Note that another fair queueing algorithm WF^2Q was not simulated, but its error is mathematically bounded [3] between -1 and $+1$ tu, and so would be very similar to WFQ in practice.

The data produced by our simulations confirm that VTRR has fairness properties that are much better than WRR, and nearly as good as WFQ. For the domain of values simulated, the service time error for VTRR falls into an average range almost two orders of magnitude smaller than WRR's error range. While VTRR's error range is not quite as good as WFQ, even the largest error measured, 10.6 tu, would likely be unnoticeable in most applications, given the size of time unit used by most schedulers. Furthermore, we show in Section 5.2 that

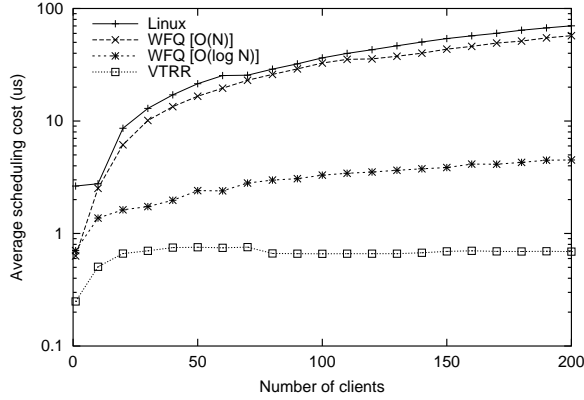


Figure 3: Average scheduling overhead

VTRR provides this degree of accuracy at much lower overhead than WFQ.

5.2 Scheduling Overhead

To evaluate the scheduling overhead of VTRR, we implemented VTRR in the Linux operating system and compared the overhead of our prototype VTRR implementation against the overhead of both the Linux scheduler and a WFQ scheduler. We conducted a series of experiments to quantify how the scheduling overhead for each scheduler varies as the number of clients increases. For this experiment, each client executed a simple micro-benchmark which performed a few operations in a while loop. A control program was used to fork a specified number of clients. Once all clients were runnable, we measured the execution time of each scheduling operation that occurred during a fixed time duration of 30 seconds. This was done by inserting a counter and timestamped event identifiers in the Linux scheduling framework. The measurements required two timestamps for each scheduling decision, so variations of 140 ns are possible due to measurement overhead. We performed these experiments on the standard Linux scheduler, WFQ, and VTRR for 1 client up to 200 clients.

Figure 3 shows the average execution time required by each scheduler to select a client to execute. For this experiment, the particular implementation details of the WFQ scheduler affect the overhead, so we include results from two different implementations of WFQ. In the first, labeled “WFQ [$O(N)$]” the run queue is implemented as a simple linked list which must be searched on every scheduling decision. The second, labeled “WFQ [$O(\log N)$]” uses a heap-based priority queue with $O(\log N)$ insertion time. Most fair queueing-based schedulers are implemented in the first fashion, due to the difficulty of maintaining complex data structures in the kernel. In our implementation, for example, a sep-

arate, fixed-length array was necessary to maintain the heap-based priority queue. If the number of clients ever exceeds the length of the array, a costly array reallocation must be performed. We chose an initial array size large enough to contain more than 200 clients, so this additional cost is not reflected in our measurements.

As shown in Figure 3, the increase in scheduling overhead as the number of clients increases varies a great deal between different schedulers. VTRR has the smallest scheduling overhead. It requires less than 800 ns to select a client to execute and the scheduling overhead is essentially constant for all numbers of clients. In contrast, the overhead for Linux and for $O(N)$ WFQ scheduling grows linearly with the number of clients. The Linux scheduler imposes 100 times more overhead than VTRR when scheduling a mix of 200 clients. In fact, the Linux scheduler still spends almost 10 times as long scheduling a single micro-benchmark client as VTRR does scheduling 200 clients. VTRR outperforms Linux and WFQ even for small numbers of clients because the VTRR scheduling code is simpler and hence runs significantly faster. VTRR performs even better compared to Linux and WFQ for large numbers of clients because it has constant time overhead as opposed to the linear time overhead of the other schedulers.

While $O(\log N)$ WFQ has much smaller overhead than Linux or $O(N)$ WFQ, it still imposes significantly more overhead than VTRR, particularly with large numbers of clients. With 200 clients, $O(\log N)$ WFQ has an overhead more than 6 times that of VTRR. WFQ’s more complex data structures require more time to maintain, and the time required to make a scheduling decision is still dependent on the number of clients, so the overhead would only continue to grow worse as more clients are added. VTRR’s scheduling decisions always take the same amount of time, regardless of the number of clients.

5.3 Microscopic View of Scheduling

Using our prototype VTRR implementation, we conducted a number of experiments to measure the scheduling behavior of the standard Linux scheduler, WFQ, and VTRR at fine time resolutions. We discuss the results of one of the studies in which we ran a 30 second workload of five micro-benchmarks with different proportional sharing parameters. Using VTRR and WFQ, we ran the five micro-benchmarks with shares 1, 2, 3, 4, and 5, respectively. To provide similar proportional sharing behavior using the Linux scheduler, we ran the five micro-benchmarks with user priorities 19, 17, 15, 13, and 11, respectively. This translates to internal priorities used by the scheduler of 1, 3, 5, 7, and 9, respectively. This then translates into the clients running for 20

ms, 40 ms, 60 ms, 80 ms, and 100 ms time quanta, respectively. The smallest time quantum used is the same for all three schedulers. At the very least, the mapping between proportional sharing and user input priorities is non-intuitive in Linux. The scheduling behavior for this workload appears similar across all of the schedulers when viewed at a coarse granularity. The relative resource consumption rates of the micro-benchmarks are virtually identical to their respective shares at a coarse granularity.

We can see more interesting behavior when we view the measurements over a shorter time scale of one second. We show the actual scheduling sequences on each scheduler over this time interval in Figures 4, 5, and 6. These measurements were made by sampling a client’s execution from within the client by recording multiple high resolution timestamps each time that a client was executed. We can see that the Linux scheduler does the poorest job of scheduling the clients evenly and predictably. Both WFQ and VTRR do a much better job of scheduling the clients proportionally at a fine granularity. In both cases, there is a clear repeating scheduling pattern every 300 ms.

Linux does not have a perfect repeating pattern because the order in which it schedules clients changes depending on exactly when the scheduler function is called. This is because once Linux selects a client to execute, it does not preempt the client even if its goodness drops below that of other clients. Instead, it runs the client until its counter drops below zero or an interrupt or other scheduling event occurs. If a scheduling event occurs, then Linux will again consider the goodness of all clients, otherwise it does not. Since interrupts can cause a scheduling event and can occur at arbitrary times, the resulting order in which clients are scheduled does not have a repeating pattern. As a result, applications being scheduled using WFQ and VTRR will receive a more even level of CPU service than if they are scheduled using the Linux scheduler.

5.4 Application Workloads

To demonstrate VTRR’s efficient proportional sharing of resources on real applications, we briefly describe two of our experiments, one running multimedia applications and the other running virtual machines. We contrast the performance of VTRR versus the standard Linux scheduler and WFQ.

One experiment we performed was to run multiple MPEG audio encoders with different shares on each of the three schedulers. The encoder test was implemented by running five copies of an MPEG audio encoder. The encoder clients were allotted shares of 1, 2, 3, 4, and 5, and were instrumented with time stamp event recorders

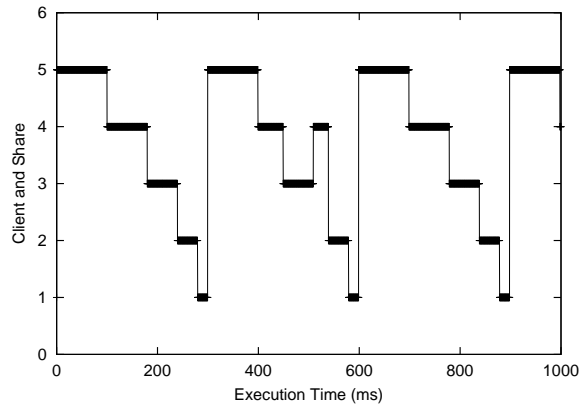


Figure 4: Linux scheduling behavior

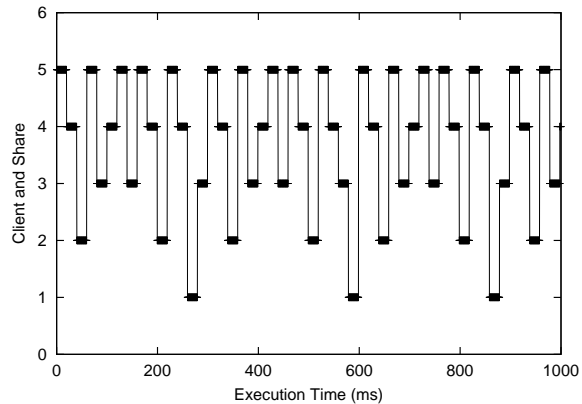


Figure 5: WFQ scheduling behavior

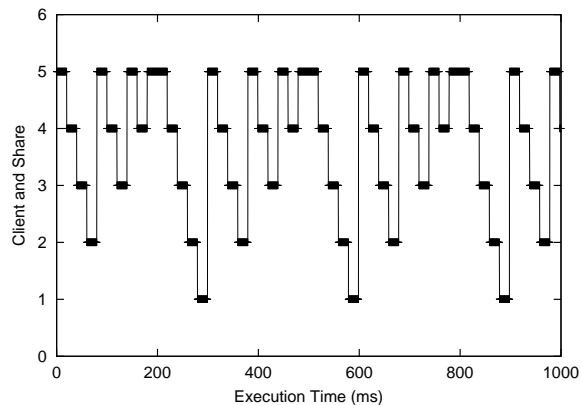


Figure 6: VTRR scheduling behavior

in a manner similar to how we recorded time in our micro-benchmark programs. Each encoder took its input from the same file, but wrote output to its own file. MPEG audio is encoded in chunks called frames, so our instrumented encoder records a timestamp after each frame is encoded, allowing us to easily observe the effect of resource share on single-frame encoding time.

Figures 7, 8, and 9 show the number of frames encoded over time for the Linux default scheduler, WFQ, and VTRR. The Linux scheduler clearly does not provide sharing as fairly as WFQ or VTRR when viewed over a short time interval. The “staircase” effect indicates that CPU resources are provided in bursts, which, for a time-critical task like audio streaming, can mean extra jitter, resulting in delays and dropouts. It can be inferred from the smoother curves of the WFQ and VTRR graphs that WFQ and VTRR scheduling provide fair resource allocation at a much smaller granularity. When analyzed at a fine resolution, we can detect some differences in the proportional sharing behavior of the applications when running under WFQ versus VTRR, but the difference is far smaller than the difference compared with Linux, which is clearly visible. VTRR trades some precision in instantaneous proportional fairness for much lower scheduling overhead.

Schedulers that explicitly support time constraints can do a more effective job than just proportional share schedulers of ensuring that real-time applications can meet their deadlines [24]. However, these real-time schedulers typically require modifying an application in order for the application to make use of scheduler-supported time constraints. For applications that have soft timing constraints but can adapt to the availability of resources, accurate proportional sharing may provide sufficient benefit in some cases without the cost of having to modify the applications.

Another experiment we performed was to run several VMware virtual machines on top a Linux operating system, and then compare the performance of applications within the virtual machines when the virtual machines were scheduled using different schedulers. For this experiment, we ran three virtual machines simultaneously with respective shares of 1, 2, and 3. We then executed a simple timing benchmark within each virtual machine to measure the relative performance of the virtual machine. We were careful to make use of the hardware clock cycle counters in doing these measurements as the standard operating system timing mechanisms within a virtual machine are a poor measure of elapsed time. We conducted the experiment using the standard Linux scheduler, WFQ, and VTRR. The results were similar to the previous experiments, with Linux doing the worst job in terms of evenly distributing CPU cycles, and VTRR and WFQ scheduling providing more comparable schedul-

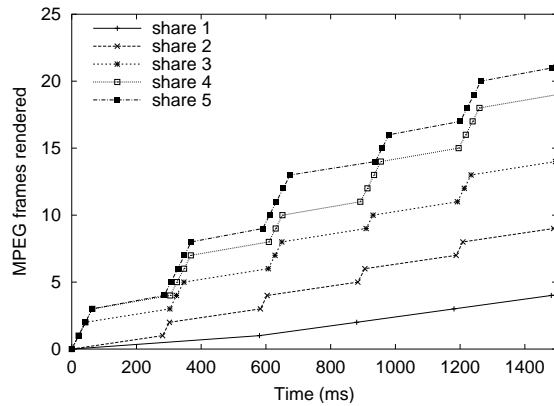


Figure 7: MPEG encoding with Linux

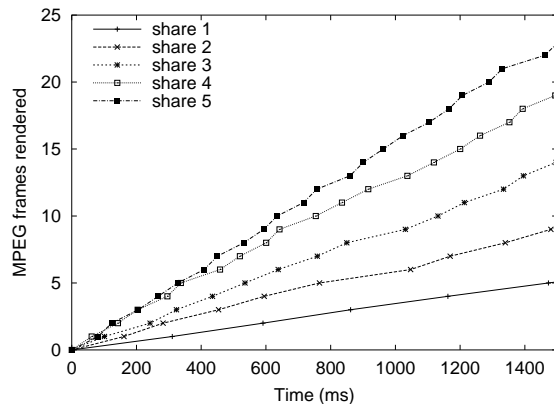


Figure 8: MPEG encoding with WFQ

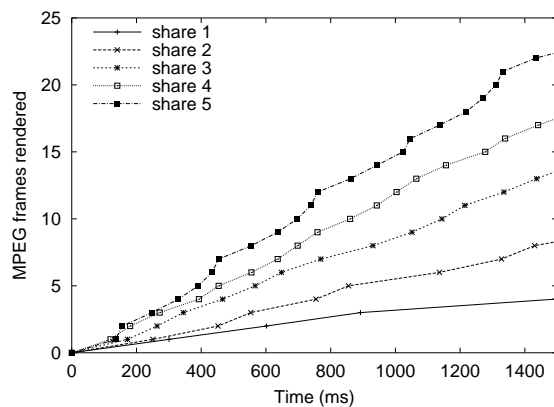


Figure 9: MPEG encoding with VTRR

ing accuracy in proportionally allocating resources.

6 Conclusions and Future Work

We have designed, implemented, and evaluated Virtual-Time Round-Robin scheduling in the Linux operating system. Our experiences with VTRR show that it is simple to implement and easy to integrate into existing commercial operating systems. We have measured the performance of our Linux implementation and demonstrated that VTRR combines the benefits of accurate proportional share resource management with very low overhead. Our results show that VTRR scheduling overhead is constant, even for large numbers of clients. Despite the popularity of the Linux operating system, our results also show that the standard Linux scheduler suffers from $O(N)$ scheduling overhead and performs much worse than VTRR, especially for larger workloads.

VTRR's ability to provide low overhead proportional share resource allocation make it a particularly promising solution for managing resources in large-scale server systems. Since these systems are typically multiprocessor machines, we are continuing our evaluation of VTRR in a multiprocessor context to demonstrate its effectiveness in supporting large numbers of users and applications in these systems.

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