

# Improving Bandwidth Utilization using Eager Writeback

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## Abstract

Cache memories have been incorporated into almost all modern, general-purpose microprocessors. To maintain data consistency between cache structures and the rest of the memory systems, most of these caches employ either a *writeback* or a *write-through* strategy to deal with store operations. Write-through caches propagate data to more distant memory levels at the time each store occurs, producing a significant bus traffic overhead to maintain consistency between the memory hierarchy levels. Writeback caches can significantly reduce the bandwidth requirements between caches and memory by marking cache lines as *dirty* when stores are processed and writing those lines to the memory system only when that dirty line is evicted. Writeback caches work well for many applications; however, for applications that experience significant numbers of cache misses over a very short interval due to streaming data, writeback cache designs can degrade overall system performance by clustering bus activity when dirty lines contend with data being fetched into the cache.

In this paper we present a technique called *Eager Writeback*, which avoids performance loss due to clustered memory traffic patterns found in streaming and graphics applications by speculatively "cleaning" dirty cache lines prior to their eviction. Eager Writeback can be viewed as a compromise between write-through and writeback policies, in which dirty lines are written later than write-through, but prior to writeback. We will show that this approach can effectively avoid the performance degradation caused by clustering bus traffic in a writeback approach, while incurring very minimal additional memory traffic.

## 1 Introduction

Caches are very effective at reducing memory bus traffic by intercepting most of the read requests generated by the processor and servicing the requests with data retained in the cache. To accomplish this, caches must maintain a consistent storage state in the presence of both reads and writes to memory. Generally, support for writes (or stores) tends to be simple – on a store the data item is either written into the cache and through the cache hierarchy to the memory (referred to as a *write-through* policy), or it is written into the cache exclusively and the data item is written out to memory only when the cache line is evicted (known as a *writeback* policy).

Caches employing a write-through policy generate memory traffic to the next level of the cache hierarchy or memory system for each store request. Since it largely defeats the purpose of having a cache if the processor has to block on each store until the write completes, write-through caches use a structure known as a *store buffer* or *write buffer* [13] to buffer writes to memory. Whenever a write occurs, the data item is written into both the cache and this structure (assuming the store buffer is

not full), allowing the processor to continue executing without blocking. The store buffer will send its contents to memory as soon as the bus is idle.

Writeback caches, on the other hand, generate memory traffic much less frequently. When a store occurs in a writeback cache the data value is written into the corresponding line in the cache, which is then marked *dirty*. Writes to memory occur only when a line marked *dirty* is evicted from the cache (usually due to a cache miss for some other memory block mapped to the same cache line) in order to make room for the incoming data item.

Whenever there are clustered misses (caused by context switches, or working set changes in an application) the writeback cache can find itself blocked waiting for a dirty line to be written back to memory. This is similar to the problem faced by the write-through cache, and can be dealt with in much the same manner by adding a writeback buffer. However, there are certain classes of programs which suffer from memory delay penalties that even a large writeback buffer cannot eliminate. For example, many emerging applications (e.g. 3D graphics or multimedia) have enormous incoming data streams. Due to the finite capacity of the data caches, in these programs, the stream of incoming data items can cause many conflict cache misses over a very short time frame and trigger the eviction of many dirty lines. This dirty writeback traffic must compete for available memory bandwidth with the arriving data, and often impedes the delivery of the data to the processor. For programs where overall performance is bound by memory bandwidth, this competition for bandwidth can have a substantial negative impact.

In this paper we propose a modification to the writeback policy which spreads out memory activity by speculatively writing certain dirty lines to memory whenever the bus is free, instead of waiting until that line in the cache is replaced. This early writing of dirty lines to the memory system reduces the potential impact of bursty reference streams, and can effectively re-distribute and balance the memory bandwidth and thereby improve system performance. In Section 2 we describe this problem in a concurrent system architecture, in Section 3 we describe our “Eager Writeback” strategy, in Section 4 we present the simulation infrastructure and benchmarks used in this study, in Section 5 we discuss the results, and Section 6 presents our conclusions.

## 2 Memory Subsystem

Modern high-performance and embedded processors provide several different memory attributes so that programmers and compilers can manipulate application data in a more flexible manner. These memory attributes generally contain the following basic types — *uncacheable*, *cacheable*, *transient*, *write-combinable* and *cache locking*. *Uncacheable* and *cacheable* are the most commonly used memory attributes, with the *cacheable* memory type typically implemented using a writeback or a write-through policy. The *transient* memory type is primarily used for data that demonstrates spatial but not temporal locality. Thus, transient type data will have a shorter life expectancy. The *write-combinable* type is a specially tailored memory type for multimedia applications, used to write out graphics or image data more efficiently. It collapses byte or word writes into an internal buffer (in cache line size) and evicts them later when the buffer is filled or displaced. Finally, the *cache locking* memory type is essentially the opposite of the transient memory type. Data declared as this type will be locked in the cache during execution. The operating system can designate a memory attribute to a specified memory region at the page granularity through system flags and registers.

As discussed in Section 1, caches that employ a writeback policy reduce memory traffic by delaying the transfer of data to memory as long as possible. Many modern microprocessors using a writeback cache policy incorporate a writeback (or cast-out) buffer, which is used as temporary storage space for holding dirty cache lines while the data request that caused the eviction is serviced. Upon eviction, the displaced dirty cache line is deposited into the writeback buffer, which usually has the highest bus scheduling priority among all types of non-read bus transactions. Once the writeback buffer fills

up, subsequent dirty line replacements cannot take place, and their corresponding data demand fetch operations cannot be committed into the cache before the writeback buffer is drained. This causes the processor pipeline to stall waiting for the dependent data.

It is possible to alleviate this problem somewhat by using existing cache hardware. *Non-blocking caches* have been proposed by Kroft [8] which use a set of *miss status holding registers* (MSHRs) to manage several outstanding cache misses. When a cache miss occurs in a non-blocking cache, it is allocated an empty MSHR entry. Once the MSHR entry is allocated, processor execution can continue. If none of the MSHRs are available (i.e. a structural hazard [5] exists due to resource conflicts), the processor will have to block until an MSHR entry becomes free.

By adding data fields to the MSHRs, it would be possible to use them to temporarily store returning cache lines. This would allow fetched data to be immediately forwarded to the appropriate destination registers, and help overcome the situation where the cache cannot be written to because the writeback buffer is full. However, this scenario delays MSHR deallocation and can lead to processor stalls on a cache miss if there are no free MSHRs. Figure 1 illustrates a non-blocking cache organization.

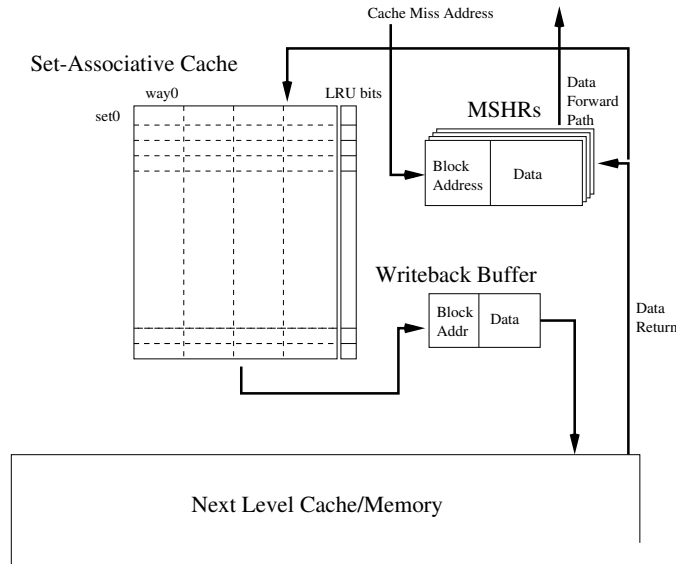


Figure 1: Block Diagram of Non-blocking Caches.

In addition, in a modern computer system memory bandwidth is not exclusively dedicated to the host processor. There are often multiple agents on the bus (such as graphics accelerators or multiple processors) issuing requests to memory over a short period of time. In a contemporary multimedia PC platform with an Accelerated Graphics Port (AGP) interface running a graphics-centric application, for example, the graphics accelerator shares system memory bandwidth with the host processor in order to retrieve graphics commands and texture maps from the system memory. A typical system architecture of a contemporary multimedia PC system is illustrated in Figure 2.

In a common 3D graphics application, the processor reads instructions and triangle vertices, performs the specified computations, and then stores them with rendering state commands back into AGP memory space, usually configured as an uncacheable region. The graphics accelerator then reads these commands out of AGP memory for rasterization that draws filled polygons on the display. In addition to the command traffic, the graphics accelerator also reads a large amount of texture data (which constitutes the major portion of AGP traffic on the bus). These textures are mapped onto polygon surfaces to increase the visual realism of computer-generated images. With richer content 3D graphics applications/games or graphics accelerators with enhanced quality features such as bi-linear/tri-linear interpolation, AGP command and data bandwidth demands for graphics accelerators will undoubtedly

be even greater than they are now.

Current cache designs have difficulty in efficiently managing the flow of data in and out of the cache hierarchy in these data intensive applications. Buffering techniques such as write buffers and MSHRs can help, but do not alleviate the problems of clustering bus traffic caused by writeback data. In the next section we introduce a new technique designed to distribute the writes of dirty blocks to times when the bus is idle.

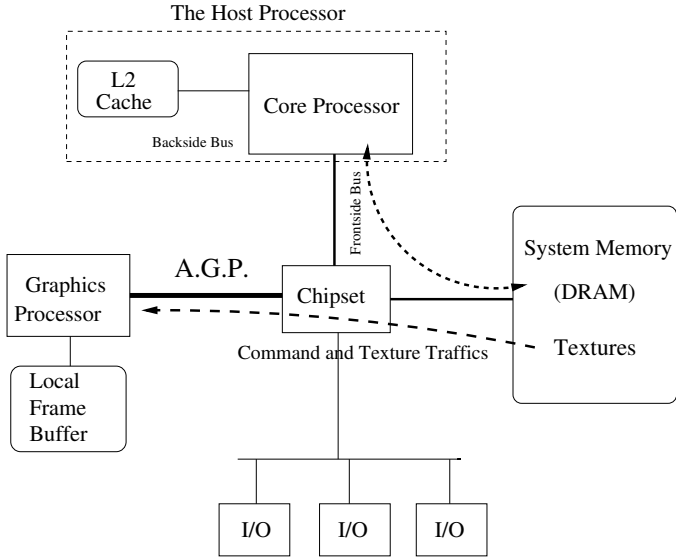


Figure 2: A Multimedia PC Architecture.

### 3 Eager Writeback

#### 3.1 Overview

To address the performance drawbacks of a conventional writeback policy, we have developed a novel technique called *Eager Writeback*. The fundamental idea behind Eager Writeback is to write selected dirty cache lines to the next level of the memory hierarchy and clear their dirty bits earlier than would happen in a conventional writeback cache design, in order to better distribute bandwidth utilization and alleviate memory bus congestion. If dirty cache lines are written to memory when the bus is less congested, there will be fewer dirty lines that require eviction during peak memory activity.

In essence, we are speculating that once dirty lines enter a certain state, they will not be written to again before eviction. Thus, there is no need to wait until eviction time to perform the cache line write. Note that an Eager Writeback will never impact the correctness of the architectural state even if the operation that triggers it was wrongly speculated - if the speculation was incorrect and writes occur too often, we approach the limiting case of write-through cache behavior. If we do not speculate often enough, we approach the behavior of a writeback cache. However, in either case no correctness constraints will be violated.

This work is similar in spirit to that of Lai and Falsafi [9], in which they identify cache lines in a shared memory system that can be speculatively self-invalidated in order to hide the invalidation time and reduce the coherence overhead. However, we are applying the idea to uniprocessor caches instead of DSM machines, which enables us to use a far simpler mechanism to identify which lines should be speculatively written out.

In order to identify which lines are the best candidates for being speculatively cleaned, we examined the probability of rewriting a dirty line in a set-associative cache when it was in a given state (MRU

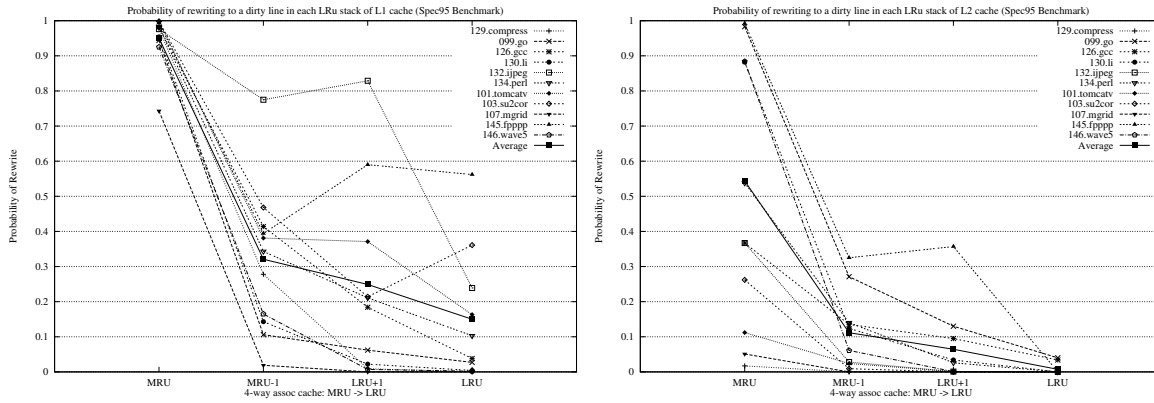


Figure 3: Probability of writing to a dirty line in each LRU stack of L1 and L2 caches (SPEC95)

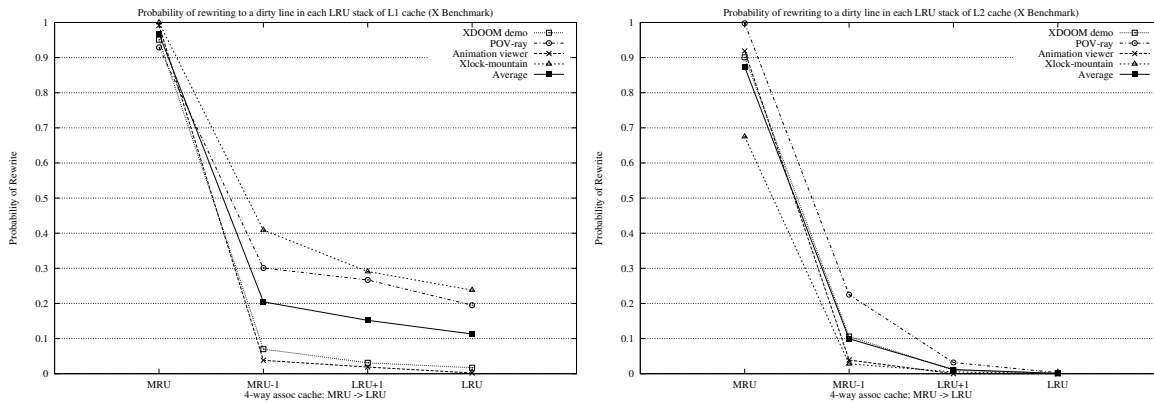


Figure 4: Probability of writing to a dirty line in each LRU stack of L1 and L2 caches (X benchmark)

through LRU) for the well-known SPEC95 benchmarks [4] and four applications from the lesser-known X benchmark suite [12]. The X benchmark suite consists of four applications representing different graphics algorithms based on X Windows. *X-DOOM*, a popular video game, uses a polygon-based rendering algorithm. *POV-ray* is a public domain ray tracing package developed for generating photo-realistic images on a computer. *xlock*, a popular screen saver, renders a 3D polygonal object on the screen. The final application is an animation viewer which processes an MPEG-1 data stream to display an animated sequence.

Our results indicate that cache lines that have been marked dirty and reach the LRU (Least Recently Used) state in a 4-way set-associative data cache are rarely written to again before they are evicted. In Figure 3 and Figure 4, we show the probability of a line that was marked dirty being written to again as it moves from the MRU (Most Recently Used) state to the LRU state for both L1 and L2 caches. The cache configurations are described in Table 1. The graph on the left in Figure 3, for example, shows that in the L1 cache the average probability (the solid line) of a dirty line in the LRU state being re-written is 0.15, while the similar probability for a dirty line in the MRU state is 0.95. The probabilities of re-dirtying lines in the LRU state are even lower in the L2 cache - in fact, close to 0 as shown in the graphs on the right of Figure 3 and Figure 4.

These figures indicate there are some programs (such as *103.su2cor* and *145.fpppp*) that do have a fairly high probability of writing to dirty lines after they have entered the LRU state, however. In order to further evaluate these cases, we looked at the ratio of the number of times a dirty line in the

LRU state is written to, normalized to the number of times a dirty line in the MRU state is written to. The results are presented in Figure 5, which shows that while the probabilities may be high, the actual number of these occurrences is negligible compared to the rewriting that occurs when a line is in other states (MRU, MRU-1, etc.). Two bars are shown for each state of each benchmark in this figure. Symbol *# enter dirty* accounts for the total number of writes including the first time writes while *# re-dirty* shows only the number of writes to "dirty" lines. These trends held across a wide range of cache configurations, and imply that once a line enters the LRU state it becomes a prime candidate for Eager Writeback, since it has a very low occurrence of being written to (and marked dirty) again.

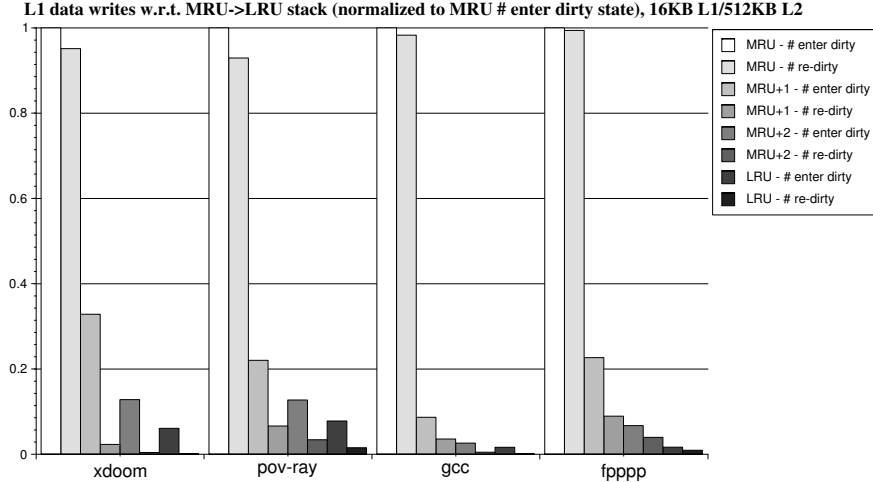


Figure 5: Normalized number of writes and rewrites to a dirty line in each MRU-LRU state

### 3.2 Design Issues in Eager Writeback Caches

There can be many different approaches to deciding when to trigger an Eager Writeback. As was shown in the previous section, one obvious candidate is to use the transition of a dirty line into the LRU state as a trigger point for an Eager Writeback. For example, when a cache set is being accessed and its corresponding LRU bit is being updated, the line can be checked to see if it is marked dirty. If it is, then a dirty writeback can be scheduled, and the dirty bit can be reset. Unless a later write updates this particular cache line again during its lifetime in the cache, otherwise, no additional dirty writeback will be generated for this line.

If the writeback buffer is full at this point, two approaches can be considered; (a) simply abort the Eager Writeback - the actual dirty writeback will take place later when the line is evicted, or (b) perform the eager writeback when an entry in the writeback buffer becomes free. This provides the ability to perform eager writeback anytime between when a line is marked LRU and when it is evicted.

To provide this capability using a minimum of hardware, we chose to implement an *Eager Queue* which holds attempted eager writebacks which were unable to acquire writeback buffer entries. Whenever an entry in the writeback buffer becomes available, the Eager Queue checks the cache set on the top of the queue to see if the dirty bit of the LRU line in the indexed set is set. If it is, the line is moved into the writeback buffer. Figure 6 illustrates a writeback cache organization with an Eager Queue.

An alternate implementation considered during this research was *Autonomous Eager Writeback*. This implementation used a small independent state machine which autonomously polled each cache set in round-robin fashion and checked the dirty bit of its LRU line, initiating eager writeback on those lines when the writeback buffer was not full. Whether Eager Queues or the autonomous state machine is more feasible is highly dependent on the processor and cache organization. For this study we present results for the more conservative approach which used Eager Queues.

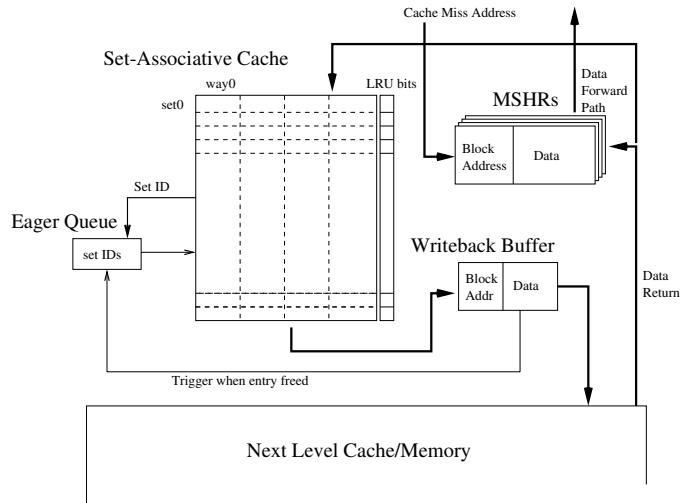


Figure 6: Eager Writeback Cache with an Eager Queue.

## 4 Simulation Framework

Processor Architectural Parameters	Specifications
Core frequency	1 GHz
1st Level I-Cache	2-way 16KB, virtual-index physical-tag
1st Level D-Cache	4-way 16KB, virtual-index physical-tag
2nd Level Cache	Unified, 4-way 512KB, physical-index physical-tag
Cache line size	32 bytes
I- and D-TLBs	2-way 8KB, 128 entries each
Backside bus	500 MHz (half-speed), 8B wide
Frontside bus	200 MHz, 8B wide
Memory model	Rambus DRAM (peak: 1.6 GB/s)
Branch predictor	2-level adaptive, 10-bit gshare
Instr. fetch/decode/issue/commit width	8 / 8 / 8 / 8
1st Level Cache MSHR entries	8
2nd Level Cache MSHR entries	8
Load/Store Queue size	32
Register update unit size	64
Memory port size	2
INT/FP ALU size	4 / 4
INT/FP MULT/DIV size	1 / 1

Table 1: Summary of the Baseline Processor Model.

Our simulation environment was based on the SimpleScalar tool set [1], a re-targetable execution-driven simulator which models speculative and out-of-order execution. The machine employs a Register Update Unit (RUU), which combines the functions of the reservation stations and the re-order buffer necessary for supporting out-of-order execution [14]. Functional unit binding, instruction dispatch and retirement all occur in the RUU. Separate address and data buses were implemented and their contentions were all modelled appropriately. Writeback buffers were implemented between cache hierarchies. All the binaries used were compiled using the SimpleScalar GCC compiler that generates code in the Portable ISA (PISA) format, whose encoding and addressing modes are almost identical to the MIPS ISA [7].

The microarchitectural parameters used in our baseline processor model are shown in Table 1. Table 2 lists the latencies of each functional unit modelled in the simulation. A non-blocking cache structure, writeback buffer and eager queue associated with each cache level were added to the simulator

for this study. The number of entries in each buffer was re-configurable from 1 to 256, and varied from simulation to simulation. Note that the actual RDRAM memory access latency simulated is also determined by other side-effects, e.g. bank conflict, in addition to memory page access timings shown in the table. For example, the minimal memory access latency is 70 core clocks if the RDRAM access causes a memory page miss without any contention.

Processor Parameters	Cycles in Processor Clocks
1st Level I- and D-Cache	3 clks, thruput = 1 clk
2nd Level Cache	18 clks, thruput = 10 clks
I- and D-TLBs	2 clks, thruput = 1 clk
Backside bus arbitration	4 clks
Frontside bus arbitration	10 clks
RDRAM Trcd, RAS-to-CAS	20 clks
RDRAM Tcac, CAS-to-data return	20 clks
RDRAM Trp, Row Precharge	20 clks
RDRAM page hit timing	30 clks
RDRAM row miss timing	50 clks
RDRAM page miss timing	70 clks
INT ALU latency/thruput	1 / 1
INT multiplier latency/thruput	3 / 1
INT divider latency/thruput	20 / 19
FP ALU latency/thruput	2 / 1
FP multiplier latency/thruput	4 / 1
FP divider latency/thruput	12 / 12

Table 2: Latency Table (in core clocks) of Functional Units in the Baseline Processor.

A pseudo-Rambus DRAM model was used in the external memory system. This single-channel RDRAM with a 64 split bank architecture can address up to 2GB of system memory. In the model, 32 independent open banks can be accessed simultaneously<sup>1</sup>. Row control packets, column control packets and data packets can be pipelined and use separate busses. RDRAM address re-mapping [10] was modelled to reduce the rate of bank interference. The theoretical peak bandwidth that can be reached in our RDRAM model is 1.6GB/sec.

A simplified uncacheable write-combining (or write-coalescing) memory [2][3] was implemented as well for the purpose of correctly simulating our benchmark behavior. Whenever a data write to an uncacheable yet marked as write-combinable memory region results in an L1 cache miss, the write operation will immediately request access to the bus and drive data out to the system memory directly (skipping a next-level cache look-up). Only complete cache line writes are modelled - any partial cache line update will be treated as a full cache line write in the simulator.

For modelling multiple agents on the memory bus, a memory traffic injector was also implemented. This injector allowed us to imitate the extra bandwidth consumed by other bus agents by configurable periodic injections of data streams onto the memory bus.

## 4.1 Benchmarks

In order to evaluate the effectiveness of the Eager Writeback technique, we ran extensive simulations on the SPEC95 benchmark suite and two programs representative of future graphics and streaming applications. Concentrating the analysis on these small, representative kernels enables us to illustrate the potential benefits of our scheme in far greater detail than can be achieved running an entire application.

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<sup>1</sup>Note that a bank conflict occurs while simultaneously accessing adjacent banks that share the same sense amplifier for driving data out of the RAM cells



```

3D-GEOMETRY()
while ( frames )
for ( objects in each frame )
for ( every 4 vertices )
/* Transformation */
tx = m11 * InV[]x + m21 * InV[]y + m31 * InV[]z + m41;
ty = m12 * InV[]x + m22 * InV[]y + m32 * InV[]z + m42;
tz = m13 * InV[]x + m23 * InV[]y + m33 * InV[]z + m43;
w = m14 * InV[]x + m24 * InV[]y + m34 * InV[]z + m44;
OutV[]rw = 1/w;
OutV[]tx = X_offset + tx * OutV[]rw;
OutV[]ty = Y_offset + ty * OutV[]rw;
OutV[]tz = tz * OutV[]rw;
/* Texture coordinates copying */
OutV[]tu = InV[]u;
OutV[]tv = InV[]v;
/* Lighting Loop */
IDr = IDg = IDb = 0.0;
for ( every light source )
dot = LDir[]x * InV[]nx + LDir[]y * InV[]ny + LDir[]z * InV[]nz;
IDr = IDr + Ambientr + Diffuse_r * dot;
IDg = IDg + Ambientg + Diffuse_g * dot;
IDb = IDb + Ambientb + Diffuse_b * dot;
OutV[]cd = ((int)IDr << 24)|((int)IDg << 16)|((int)IDb << 8)|alpha;

/* Device driver loop */
for ( each transformed and lit vertex )
/* Assume Tri-Strip triangles */
/* Copy entire OutV records to graphics AGP memory */
GfxCommand[vertex - 2] = OutV[vertex - 2];
if (even - numberedvertex)
GfxCommand[vertex] = OutV[vertex];
GfxCommand[vertex - 1] = OutV[vertex - 1];
else
GfxCommand[vertex - 1] = OutV[vertex - 1];
GfxCommand[vertex] = OutV[vertex];

```

Figure 7: Algorithm of the 3D geometry pipeline

The first of these kernels is a 3D geometry processing pipeline (*3D geometry*), which is present in most triangle-based rendering algorithms [17]. Two different geometric rendering configurations were simulated, one which was very simple (i.e. ambient light with no external light sources), and one which included multiple diffuse light sources. The ambient light configuration reduces the computational requirements of the algorithm in order to maximize frame rate at the expense of image realism<sup>2</sup>. The multiple light source configuration increases the computational demands, reducing the relative impact of bus utilization as the processor spends more time processing between each data element request.

The second kernel represents the class of streaming data algorithms which work with large data sets. This kernel processes a large array of data, performing both reads and writes, generating frequent cache misses as well as many dirty writebacks (behavior common to many current streaming applications). In a real program the data would have some computation performed upon it – however, in order to highlight the behavior of the memory system in this kernel no actual computations are performed.

#### 4.1.1 3D Geometry Pipeline

Today’s polygon-based 3D graphics engine is composed of two major components, a 3D geometry processing pipeline and a rasterization pipeline. The 3D geometry processing pipeline as shown in Figure 7 is representative of a very frequently used algorithm in most polygon-based rendering engines. 3D geometry processing, consisting of intensive floating-point operations on a large quantity of vertex data from memory, handles vertices from the object model database. It maps the vertices from the world coordinate space to viewer’s space, i.e. the display. It also computes the interaction between the light sources and their effect on each vertex for generating the shading intensity to be used during rasterization. These functions are typically done on the host processor. Once the vertices are transformed and lit, they are sent to the rasterization pipeline for scan-converting them into pixels on the display. The

<sup>2</sup>This would be preferred in real-time 3D applications when processor performance is lacking.

rasterization pipeline interpolates color values for each interior pixel within a polygon. Rasterization is typically performed by a dedicated graphics accelerator.

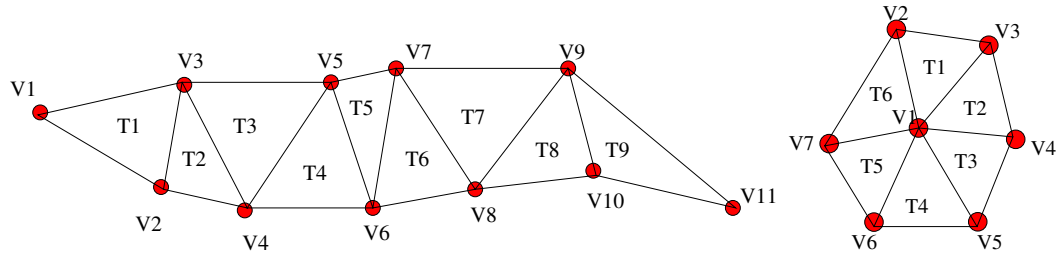


Figure 8: Geometry Representations — Triangle Strip and Fan

This pipeline consists of three nested loops wrapped by two outer loops which iterate through frames and 3D objects in the world space. The first innermost loop processes vertices for each 3D object assuming the entire object is modelled by a single triangle strip or a triangle fan. A triangle strip or fan as illustrated in Figure 8 is an object representation using a pre-ordered triangulation scheme. It is supported by popular graphics libraries such as OpenGL [16]. Using such an ordering, only one incremental vertex needs to be specified to describe a new triangle. For example, T1 is constructed by V1, V2 and V3. T2 is constructed by V2, V3 and V4. It is a more compact representation, therefore it requires fewer number of operations performed in the geometry pipeline. The basic functions performed inside this loop are *transformation*, *lighting*, and *rendering command output*. A triangle strip or fan as illustrated in Figure 8 is an object representation using a pre-ordered triangulation scheme. It is supported by popular graphics libraries such as OpenGL [16]. Using such an ordering, only one incremental vertex needs to be specified to describe a new triangle. For example, T1 is constructed by V1, V2 and V3. T2 is constructed by V2, V3 and V4. It is a more compact representation, therefore it requires fewer number of operations performed in the geometry pipeline.

The *transformation* function projects the new location of each vertex on screen through a 4x4 matrix multiplication and a viewport transformation. The *lighting* function calculates the interaction of each vertex with light sources and generates the color intensity for each vertex. This calculation involves a dot product between the light direction vector and the vertex normal vector using a Phong illumination model [15]. A single parallel light source with diffuse only components is assumed in the lighting model. For a parallel light source, per-vertex normal transformations can be replaced by an inverse transformation of the light source location on a per-scene basis, thus eliminating a large number of computations for generating the light direction vector of each vertex. A color packing conversion then packs four single-precision floating-point RGBA color intensities into a packed 4-byte integer. The instruction set architecture of interest is assumed to support four-wide SIMD floating-point computation, similar to Intel’s Pentium III Processor [11] and Pentium 4 Processor [6].

After finishing with all the vertices in one object, a loop imitating the functionality of an ”online device driver” [17] is invoked (the *command output* function). This driver loop breaks one triangle strip into individual triangles and copies these transformed and lit vertices to the uncacheable AGP memory. Figure 9 illustrates the complete control flow diagram of each functional block in the geometry pipeline.

#### 4.1.2 Streaming Kernel

The *Streaming* kernel is presented in Figure 10, and consists of three inner loops that exercise the L2 cache. The first loop writes data into *array<sub>A</sub>*. The second loop reads data from *array<sub>A</sub>*, performs some floating-point computation and passes the results to inner loop invariant array elements. Finally the third loop accesses a new array (*array<sub>B</sub>*), displacing elements of *array<sub>A</sub>* from the cache.

This program is designed to represent the typical behavior of many streaming applications from

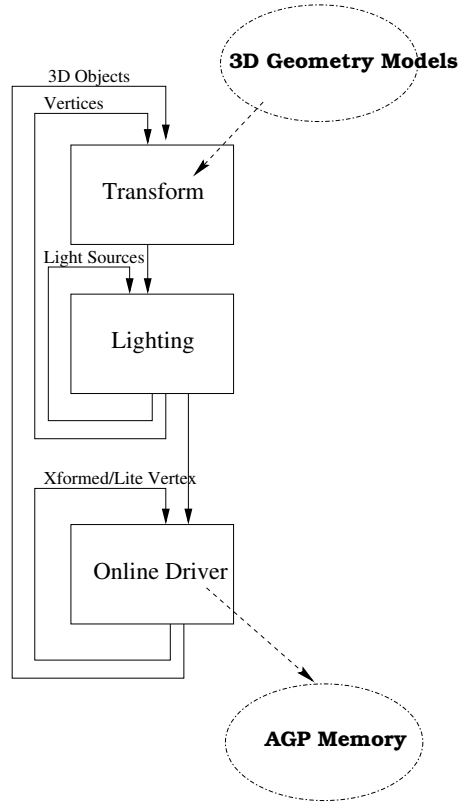


Figure 9: Geometry Processing Loop Structure

multimedia and digital signal processing algorithms. However, as pointed out previously, in order to highlight the interaction of Eager Writeback and the memory system in this uniprocessor, no actual computational work is performed per data read. In other words, it represents a memory bandwidth bound program.

## 5 Simulation Results and Analysis

The simulation results are presented and analyzed in this section. For each kernel studied, we present two different data sets, one with no memory contention from other potential bus agents, and one with artificially injected memory traffic.

<i>sim cycle</i>	<i>Baseline</i>	<i>Eager</i>		<i>Free Writeback</i>	
benchmark	cycles	cycles	speedup	cycles	speedup
go	4106741898	4106316586	1.000	4105050891	1.000
gcc	1425690611	1423578223	1.001	1419981686	1.004
li	401639628	401635232	1.000	401481752	1.000
jpeg	2125521487	2123322070	1.001	2117908634	1.004
perl	3705579465	3701065936	1.001	3683430936	1.006
tomcatv	5436594306	5436670500	1.000	5436456381	1.000
su2cor	4625207540	4625248569	1.000	4625117247	1.000
mgrid	2138832527	2132120132	1.003	2061823555	1.037
fpppp	8404705112	8410760399	0.999	8404047239	1.000
wave5	2221747518	2208702430	1.006	2179225372	1.020

Table 3: Performance of SPEC95 Benchmarks. (WB buffer = 1, EQ = 4)

```

STREAMING()
float arrayA[MAX], arrayB[MAX];
for (m = 0; m < loop; m++)
  for ( arrayA[i] ∈ each set of L2 cache )
    write arrayA[i] to way 0;
    write arrayA[i + 1 * 8 * set_size] to way 1;
    write arrayA[i + 2 * 8 * set_size] to way 2;
    write arrayA[i + 3 * 8 * set_size] to way 3;
  for ( arrayA[j] ∈ each cache line in L2 cache )
    read arrayA[j];
    compute arrayA[j];
    write arrayA[j];
  for ( arrayB[k] ∈ each set of L2 cache )
    read arrayB[k] into way 0;
    read arrayB[k + 1 * 8 * set_size] into way 1;
    read arrayB[k + 2 * 8 * set_size] into way 2;
    read arrayB[k + 3 * 8 * set_size] into way 3;
    write arrayA[m];

```

Figure 10: Algorithm of the Streaming Kernel

## 5.1 Spec95 Benchmarks

Table 3 shows the simulation results for the SPEC95 benchmark suite using 3 configurations - Baseline, Eager and Free Writeback. The Baseline case uses a single entry writeback buffer, while Free Writeback models a system in which dirty writebacks do not generate any memory traffic on the bus (thus serving as an upper bound on performance.) SPEC95 *reference* input data set were used for all the benchmarks simulated.

Looking at the table it is apparent that there is little performance gain possible for the programs in this suite, since the difference in the cycle count between the baseline case and the upper bound is negligible. This is not surprising, since it is well-known that the SPEC95 benchmark suite does not exercise the memory system aggressively. The SPEC95 suite is not a good candidate for memory system performance studies primarily due to its small working set size. Even though the eager writeback mechanism does not show any overall performance improvement, Table 4 shows that eager writeback can effectively reduce stall cycles of the major machine resource RUU in several benchmarks.

For the rest of this study we will focus on the benchmarks that more aggressively exercise the memory system, and are arguably more representative of future workloads.

Resource Hazard	RUU stall cycles		
	Baseline	Eager	impr%
go	1536683	1533417	0.21%
gcc	2014449	1821639	9.57%
li	23688	23612	0.32%
ijpeg	13210350	9918520	24.92%
perl	67143494	61436723	8.50%
tomcatv	8469	8469	0.00%
su2cor	61514	61410	0.17%
mgrid	164720709	149735310	9.10%
fp PPP	219377	219209	0.08%
wave5	134599566	121701473	9.58%

Table 4: Resource Hazard Improvement of SPEC95 Benchmark (WB Buffer =1, EQ = 4)

## 5.2 Analysis of 3D Geometry Pipeline

### 5.2.1 Without Injected Memory Traffic

Table 5 contains the number of 3D geometry pipeline execution cycles for a variety of memory configurations. In this table, each row represents a different combination of writeback buffer size and lighting conditions, while the columns contain different writeback strategy cycle counts. The first column, *Baseline*, shows the cycle count using a conventional writeback policy. There are six different *Baseline* cases shown in the table, each with its own writeback buffer size and lighting condition. The next 6 columns contain the results for 3 different variations of the *Eager Writeback* scheme and the speed-up of each scheme over their corresponding *Baseline* case, with each scheme identified by the size of its *Eager Queue* (EQ). The simplest design choice is EQ=0, in which Eager Writebacks are dismissed if the writeback buffer is full. The other two cases can queue up attempted eager writebacks within Eager Queues of specified sizes. The rightmost column contains the Free Writeback case, which as stated earlier is the upper bound to available performance.

write buffer size	<i>Baseline</i>		<i>Eager (EQ=0)</i>		<i>Eager (EQ=4)</i>		<i>Eager (EQ=256)</i>		<i>Free Writeback</i>	
	cycles	cycles	speedup	cycles	speedup	cycles	speedup	cycles	speedup	
No light, WB Buf=1	25364637	23876911	1.062	21838002	1.162	21837952	1.162	21798206	1.164	
No light, WB Buf=4	25320139	21820627	1.160	21820566	1.160	21820566	1.160	21798206	1.162	
No light, WB Buf=256	25320139	21820566	1.160	21820566	1.160	21820566	1.160	21798206	1.162	
3 diff. lights, WB Buf=1	30643341	29200004	1.049	27176616	1.128	27176333	1.128	27134147	1.129	
3 diff. lights, WB Buf=4	30643153	27158044	1.128	27158049	1.128	27158044	1.128	27134147	1.129	
3 diff. lights, WB Buf=256	30643153	27158044	1.128	27158044	1.128	27158044	1.128	27134147	1.129	

Table 5: Simulated cycles of 3D Geometry Pipeline.

There are several things of interest to note in this table. Perhaps most significantly, it can be seen that increasing the depth of the writeback buffer has virtually no impact on the performance of the Baseline case. In fact, going from 1 to 256 entries in the writeback buffer only improves performance by 0.17%. This is because a large number of dirty writebacks are competing for bandwidth with the demand fetches, and the bus congestion can not be alleviated by a deeper writeback buffer.

On the other hand, adding the Eager Writeback scheme increases the performance of the system by 4.9% to 16.2% (depending on the light sources and the depth of the Eager Queue). For the simplest case of no Eager Queue and a single entry writeback buffer, the speedup ranges from 6.2% (for no light source) to 4.9% (with 3 light sources). This speedup is smaller than for the other cases, because many eager writebacks are dropped due to the lack of space in the writeback buffer. When the Eager Queue size is increased (or the number of writeback buffer entries is increased), the speedup achieved approaches the upper bound.

The “bandwidth shifting” effect is quite apparent in Figure 11 and Figure 12. These two figures present the utilization profile of memory bandwidth requested by the processor using the Baseline (Figure 11) and Eager Writeback (Figure 12) configurations, running the 3D geometry pipeline. The y-axis plots the instantaneous bandwidth versus the execution timeline on the x-axis, which was calculated by sampling the data phase on the memory bus every 2000 core clocks (e.g. if 1600 bytes are seen on the bus in 2000 core cycle period, its instantaneous bandwidth is 800MB/sec for a 1GHz processor).

The 12 broad spikes that saturate the peak RDRAM bandwidth<sup>3</sup>(1.6GB/sec) in Figure 12 occur within the driver loop, where rendering command output is being written into the write-combining graphics memory while eager writebacks of dirty lines are concurrently taking place. Since within the

<sup>3</sup>According to Table 1, it is assumed that the maximum front-side bus bandwidth is equivalent to the maximum memory bandwidth for our study. Without external devices competing for memory bandwidth, front-side bus bandwidth and memory bandwidth are interchangeably used in this paper hereafter.

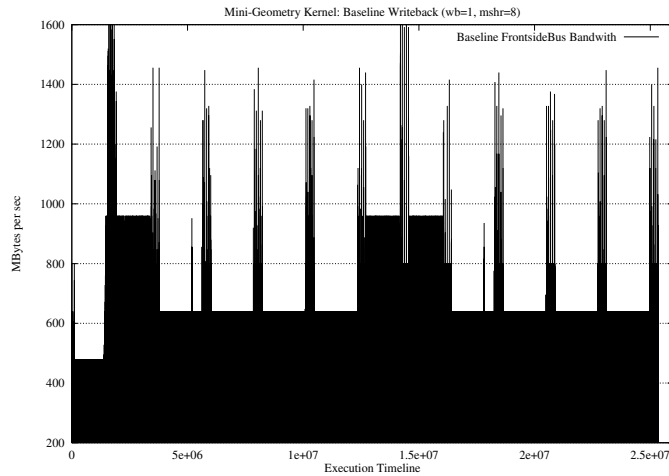


Figure 11: Memory Bandwidth Profile by *Baseline Writeback* for 3D Geometry Pipeline (No light)

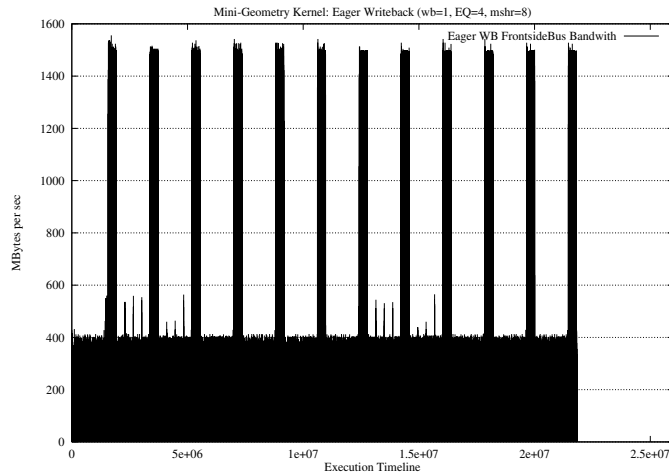


Figure 12: Memory Bandwidth Profile by *Eager Writeback* for 3D Geometry Pipeline (No light)

driver loop there is still some computation occurring, the bandwidth is not fully utilized, and eager writeback writes can use the available idle slots and maximize bandwidth. Conversely, in the baseline case, the same writebacks occur within the geometry computation loop. This means these requests compete for the bus with the return of the data requested by vertex loads, and thus slow down the processing. This maximization of the utilization of the bandwidth during the driver loop leads to a lower and sparser average memory bandwidth in Eager Writeback than in the Baseline case outside the driver loop<sup>4</sup>.

The overall performance improvement is obviously gained from the shifting of dirty writeback traffic to where this traffic does not impede the return of any data on the critical path of performance. This can be seen in Figure 13, which presents an execution profile of the benchmark. In this figure the sequence of vertex data load requests appears on the y-axis, and the cycle upon which the corresponding data item returns is plotted on the x-axis. As execution begins, the profiles of Baseline and Eager Writeback are completely overlapped, because data is returning at the same time for both schemes. Beginning at around 2.6 million cycles, these two curves start to deviate from one other, and continue

<sup>4</sup>It should be emphasized that the total bandwidth required by a system using Eager Writeback is not reduced; rather, it is re-distributed by the early eviction of dirty cache lines. In some certain cases, the total bandwidth could be slightly increased due to re-writes to the LRU lines.

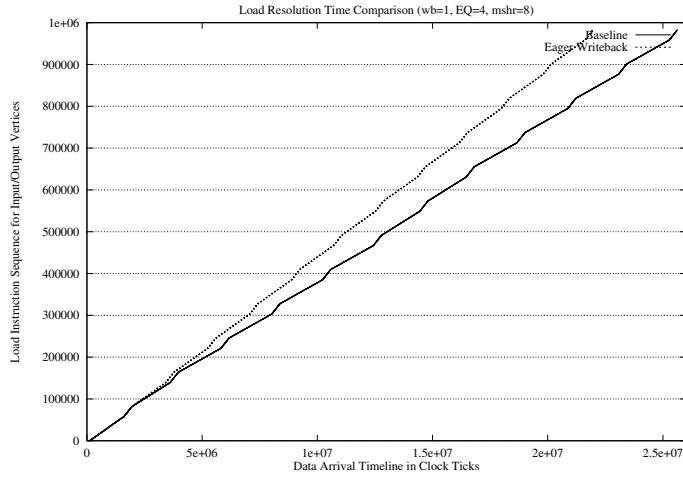


Figure 13: Load Response Time for Input Vertex in 3D Geometry Pipeline

to diverge as execution time increases. The first deviation indicates the location where data returns of the Baseline model were getting delayed because of dirty writeback contentions. The speedup due to Eager Writeback as measured is around 16%.

By looking carefully at this figure it is possible to distinguish the geometry computation loop from the device driver loop on each curve. The segments with shorter but steeper slopes are where the driver loop is executing. The steepness of the slope occurs because the requested data was returned faster (since the loop read the output vertices generated in the transformation and lighting stages from the L2 cache directly, rather than from memory).

	<i>baseline</i>		<i>Eager (EQ=0)</i>		<i>Eager (EQ=4)</i>		<i>Eager (EQ=256)</i>		<i>Free Writeback</i>	
RUU Full cycles	cycles		cycles	improved	cycles	improved	cycles	improved	cycles	improved
No light, WB Buf = 1	8404023		6678659	20.5%	4452469	47.0%	4452265	47.0%	4409553	47.5%
No light, WB Buf = 4	8375679		4439397	47.00%	4439226	47.00%	4439226	47.00%	4409553	47.35%
3 diffuse lights, WB Buf=1	8045791		6541028	18.7%	4361651	45.8%	4361259	45.8%	4313710	46.4%
3 diffuse lights, WB Buf=4	8033850		4344799	45.92%	4344670	45.92%	4344653	45.92%	4313710	46.31%

Table 6: Resource Hazard Improvement of 3D Geometry Pipeline.

Table 6 shows how Eager Writeback affects the performance bottleneck in the Register Update Unit (RUU) of the processor. The layout of this table is similar to Table 5, and contains the number of cycles the processor is stalled due to the RUU being full. As the table shows, Eager Writeback is able to remove a substantial number of stall cycles due to a full RUU and keep the execution pipeline running smoother. These stalls are reduced because in conventional writeback schemes dirty writebacks are competing with demand fetches for available bandwidth, causing delays in data arrival and the filling of the reservation stations in the RUU. The eager writebacks shift the dirty writes to an earlier time, freeing up the bandwidth to handle just data reads and reducing the pressure on the RUU.

### 5.2.2 With Injected Memory Traffic

In order to evaluate the effectiveness of Eager Writeback in a real system, we implemented a memory traffic injector which we used to model other bus agents requesting the memory bus and consuming as well competing for memory bandwidth. For this study, we injected three different external bandwidths using two different injection frequencies onto the bus during the simulations. The external bandwidths chosen were 400MB/sec, 800MB/sec and 1.2GB/sec. For each bandwidth configuration, data was

injected at a high frequency (every 400 processor clock cycles) and a low frequency (every 3200 processor clock cycles). Data was injected onto the bus in blocks - for example, in the 800MB high frequency case, every 400 cycles the injector took over the bus and held it until it had completed transferring 320 bytes of data. The injections are uniformly distributed throughout the simulation.

<i>bandwidth injection (no light)</i>	<i>sim cycles</i>			<i>RUU Full cycles</i>		
	Baseline	Eager	speed-up	Baseline	Eager	improved
0 GB/sec	25364637	21838002	1.16	8404023	4452469	47.0%
0.4GB/sec (160B/400clks)	27323771	25434535	1.07	10529817	8448695	19.76%
0.8GB/sec (320B/400clks)	33567580	33775835	0.99	16760998	17024045	-1.6%
1.2GB/sec (480B/400clks)	60699573	59162773	1.03	44206642	42864369	3.0%
0.4GB/sec (1280B/3200clks)	32539684	28636072	1.14	15604083	11364679	27.2%
0.8GB/sec (2560B/3200clks)	47365936	42559653	1.11	30356564	25269290	16.8%
1.2GB/sec (3840B/3200clks)	87400980	83426435	1.05	70248220	66015191	6.0%

Table 7: Memory Traffic Injection to 3D Geometry Pipeline. (EQ = 4)

The results for simulations of the 3D geometry pipeline using no light sources are shown in Table 7. The top line of the table is the base case with no injected memory traffic, while the other entries are for the different injected bandwidths at the different frequencies. In this table we can see that (as expected) memory traffic injection causes extra stall cycles in the RUU. In addition, as the amount of injected bus traffic increases, the opportunity to do Eager Writeback decreases and the RUU stalls climb dramatically.

The table also shows that Eager Writeback provides virtually no speedup when a bandwidth of 0.8GB/sec is injected at the higher frequency, while the same bandwidth injected at a lower frequency allows a speedup of 11%. By examining the dirty writeback bandwidth utilization profile of this scenario ( Figure 14 and Figure 15), one can see that many eager writebacks (i.e. the spikes) are prevented from occurring by the higher frequency injection. The advantages of Eager Writeback are lost and it performs almost on par with the baseline scenario, due to more frequent bus contention.

### 5.3 Streaming Kernel

The 3D geometry pipeline highlighted the problem of implicit dirty writebacks causing loss of performance due to delays in receiving data. Finite memory peak bandwidth is another serious performance issue, which is exposed by the Streaming kernel.

#### 5.3.1 Without Injected Memory Traffic

Table 8 contains the results of simulation runs of the Streaming kernel, presented in the same format used in Table 5. For this benchmark, an eager queue of length 0 (EQ=0) is enough to approximate the optimal case of no dirty writeback traffic at all. Further size increases of the EQ provide only marginal performance gains.

<i>sim cycle write buffer size</i>	<i>Baseline cycles</i>	<i>Eager (EQ=0) cycles</i>	<i>speedup</i>	<i>Eager (EQ=4) cycles</i>	<i>speedup</i>	<i>Eager (EQ=256) cycles</i>	<i>speedup</i>	<i>Free Writeback cycles</i>	<i>speedup</i>
WB buf = 1	10230328	9054559	1.130	9053851	1.130	9053851	1.130	9045154	1.131
WB buf = 4	10067331	9052957	1.112	9052957	1.112	9052957	1.112	9045154	1.113
WB buf = 256	10223055	9052821	1.129	9052821	1.129	9052821	1.129	9045154	1.113

Table 8: Simulated cycles of Streaming Kernel.

Looking at the memory bandwidth utilization profiles for this kernel (Figure 16 and Figure 17), we see three spikes that appear repeatedly in both writeback schemes (because the outer loop contains



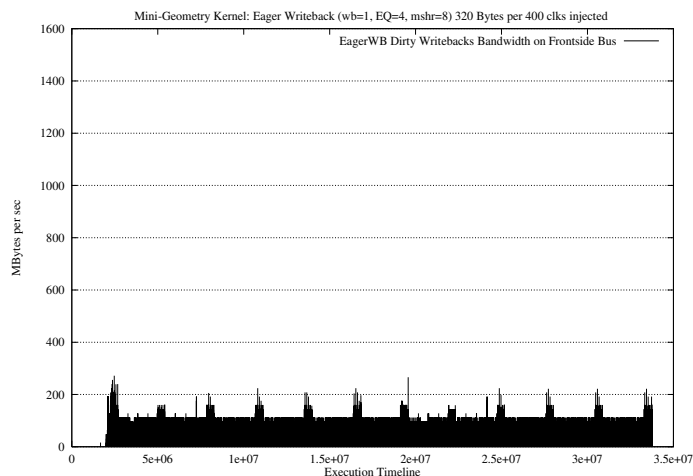


Figure 14: Dirty WB L2-to-Mem Bandwidth with 320B/400clks Injection (*Eager*) for Geometry

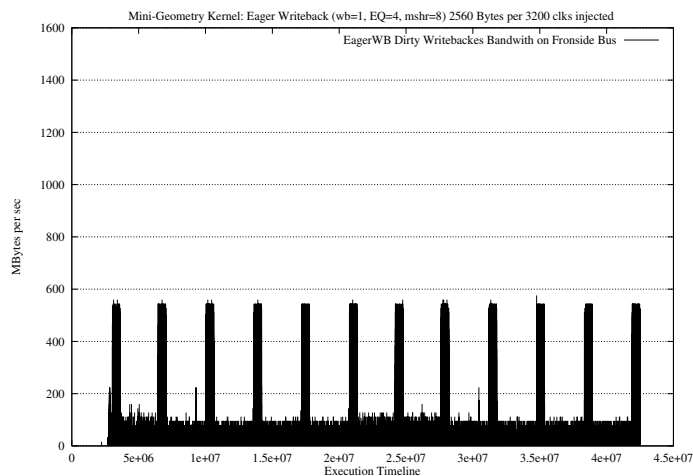


Figure 15: Dirty WB L2-to-Mem Bandwidth with 2560B/3200clks Injection (*Eager*) for Geometry

three iterations). The spikes are much wider in the Baseline case, however, indicating the program is spending more execution cycles in these phases. Examining the algorithm, it is clear these spikes are related to the time during the third inner loop where incoming  $array_B$  data items collide and share memory bandwidth with the induced dirty writebacks of  $array_A$ . Because the finite memory bandwidth (1.6 GB/sec in this study) must be shared between both memory accesses<sup>5</sup>, the rate of demand fetches for  $array_B$  in the third inner loop is (theoretically) cut in half and thus the overall performance degrades.

Figure 16 also shows three bandwidth grooves where memory bus bandwidth has dropped to zero. These correspond to the second inner loops, where all data references hit in the cache. To take advantage of this available resource, Eager Writeback fills these bus idle states with early evictions of dirty data cache lines (as shown in Figure 17). By shifting these bandwidth requests to idle cycles, i.e. “bandwidth filling,” the memory bandwidth during the course of the third inner loop can be fully dedicated to the demand fetches of  $array_B$ , speeding up the cache fill requests.

As was done for the 3D geometry pipeline, we examined how Eager Writeback interacted with internal processor resources when running this benchmark. Table 9 shows that the Load/Store Queue

<sup>5</sup>Read and write turnarounds between demand fetch and dirty writeback streams also prevent peak memory bandwidth from being achieved.

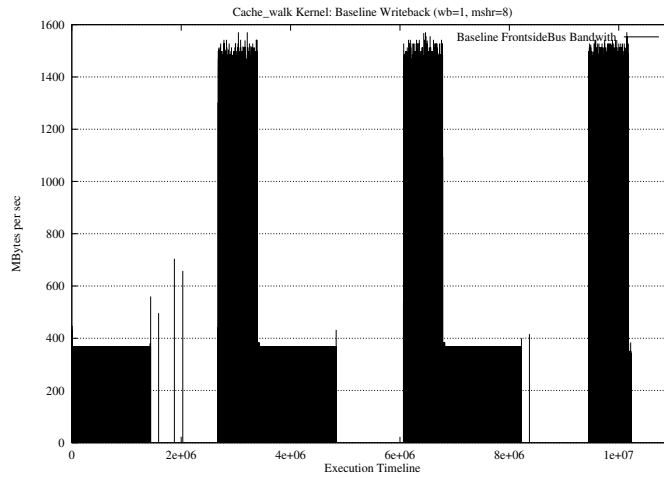


Figure 16: Memory Bandwidth Distribution by *Baseline Writeback* for Streaming Kernel

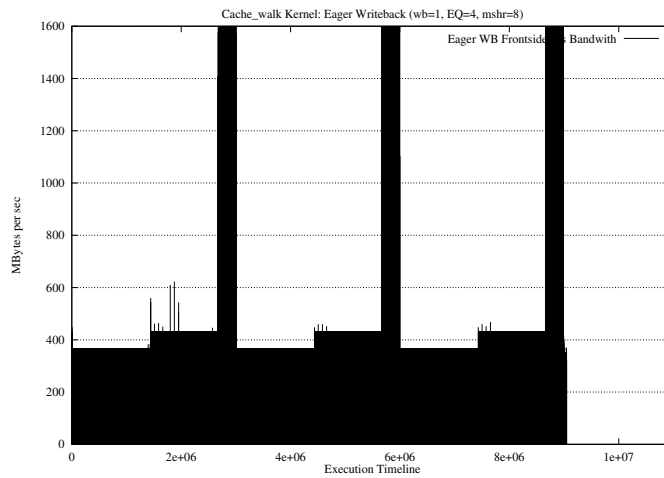


Figure 17: Memory Bandwidth Distribution by *Eager Writeback* for Streaming Kernel

is used heavily by this benchmark, and that Eager Writeback can remove more than half of the stalls due to a full Load/Store Queue. As the LSQ is kept less full, instructions are able to leave the Instruction Fetching Queue (IFQ) faster and as a result cycles lost due to a full IFQ are reduced substantially.

Bottlenecks	baseline cycles	<i>Eager</i> ( $EQ=0$ ) cycles	improved	<i>Eager</i> ( $EQ=4$ ) cycles	improved	<i>Eager</i> ( $EQ=256$ ) cycles	improved	<i>Free Writeback</i> cycles	improved
IFQ Full cycles	5770175	4594401	20.38%	4594631	20.37%	4594631	20.37%	4587638	20.49%
RUU Full cycles	4274868	4260784	0.33%	4260703	0.33%	4260703	0.33%	4258811	0.38%
LSQ Full cycles	1978596	864867	56.29%	866341	56.21%	866341	56.21%	862880	56.39%

Table 9: Resource Constraint Improvement of Streaming Kernel. (Writeback buffer = 1)

### 5.3.2 With Injected Memory Traffic

We also repeated the experiments involving injecting memory traffic onto the bus for this benchmark program. The results are shown in Table 10, and indicate that higher frequency injection seems to have a greater impact on the Baseline case than on the Eager Writeback case. The number of simulated

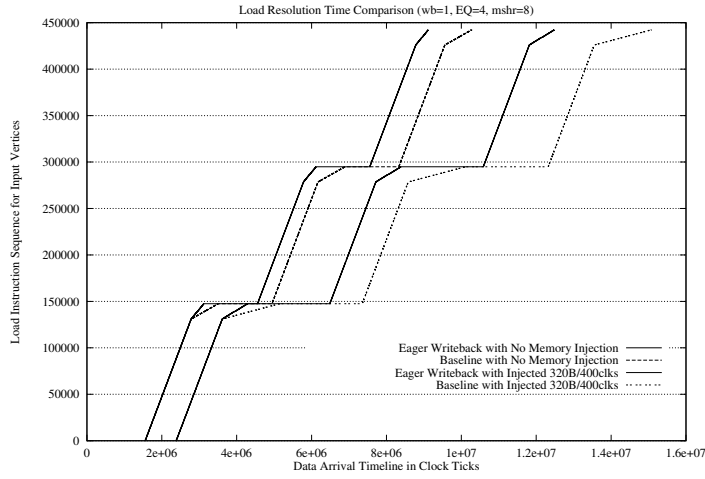


Figure 18: Load Response Time for Data Reads in Streaming Kernel (Higher Frequency Injection)

cycles for the Baseline case using high frequency injection increases faster than for the Eager Writeback case, while the increase stays roughly the same for both schemes while injecting lower frequency traffic.

<i>bandwidth injection</i>	<i>simulated cycles</i>			<i>IFQ Full cycles</i>			<i>LSQ Full cycles</i>		
	Baseline	Eager	speed-up	Baseline	Eager	improved	Baseline	Eager	improved
0 MB/sec	10230328	9053851	1.13	5770175	4594631	20.4%	1978596	866341	56.2%
0.4GB/sec (160B/400clks)	11807448	10039848	1.18	7340618	5576536	24.0%	2903145	1205358	58.5%
0.8GB/sec (320B/400clks)	15025957	12389159	1.21	10540877	7908077	25.0%	4428473	1882587	57.5%
1.2GB/sec (480B/400clks)	24250335	21412735	1.13	19717746	16880309	14.4%	8309036	5480188	34.05%
0.4GB/sec (1280B/3200clks)	12379290	10991058	1.13	7908538	6521201	17.5%	2030932	1417595	30.2%
0.8GB/sec (2560B/3200clks)	16593748	15115348	1.10	12101456	10622058	12.2%	4264295	2818313	33.9%
1.2GB/sec (3840B/3200clks)	29048835	27135235	1.07	24495295	22585042	7.8%	8903039	7007451	21.3%

Table 10: Memory Traffic Injection to Streaming Kernel. (Eager Queue = 4)

The reason the cycle count climbs faster for the Baseline case than for the Eager Writeback case can be understood by analyzing Figure 18. This figure contains an execution profile of the Streaming benchmark, plotting the arrival time for each load instruction. Each curve can be divided into 3 repeated patterns, which bear the following three piecewise line segments: flat (zero increment), steep rise, and slowdown knee. These 3 line segments correspond to the three inner loops in the benchmark.

The first loop contains only data stores, so the load instruction count stays flat as execution time continues. The steep vertical climb corresponds to the second inner loop, which has a high number of cache hits (a large number of loads completing in a short period of time). Finally, the third segment represents the behavior of the third loop, which loads another array that misses in both the caches.

This third segment, shown as a knee in the curve, reveals the reason for the performance deviation between Baseline and Eager Writeback. Figure 19 shows a close-up view of part of Figure 18, focusing on the knees of the curve. The slopes ( $\tan\theta$ ) of these knees are the key - the flatter the slope, the longer the completion time. Comparing the slope changes between Baseline and Eager Writeback, it is obvious that the slope of the Baseline segment is much shallower than that of the Eager Writeback segment. This means that for the same number of loads in the third loop, the execution time of the Baseline case was more sensitive to and severely delayed by other transactions, which in this case are composed of the dirty writebacks induced by the loads and the periodic injection of memory traffic. For the Eager Writeback case, the dirty writebacks were mostly completed in the second loop, so the slope of the knee is steeper and the third loop can be completed more swiftly than its Baseline counterpart.

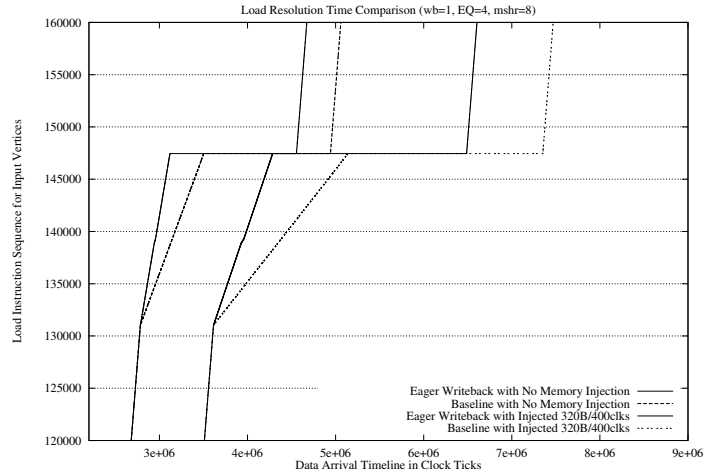


Figure 19: Details of the Load Response Time for Data Reads (Higher Frequency Injection)

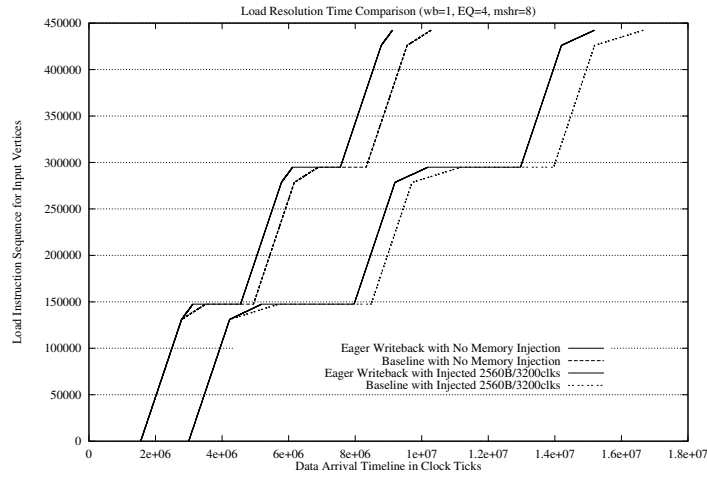


Figure 20: Load Response Time for Data Reads in Streaming Kernel (Lower Frequency Injection)

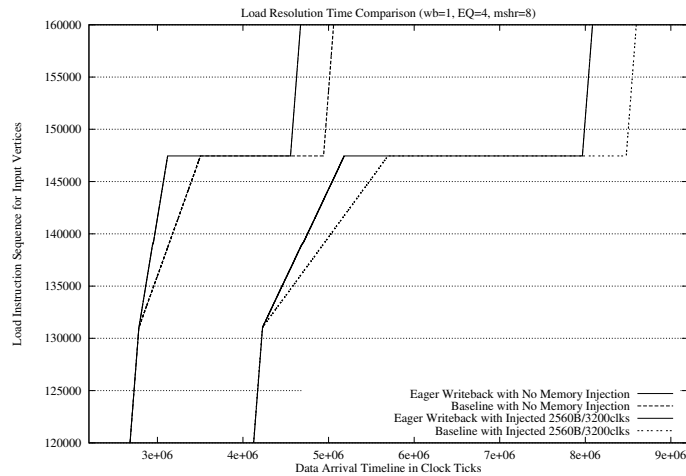


Figure 21: Details of the Load Response Time for Data Reads (Lower Frequency Injection)

Repeating the same experiment using lower frequency injection (as plotted in Figure 20 and Figure 21) reveals that the slope of the knees of the curve are more similar to one another. As a result, roughly the same number of penalty cycles were added to both Baseline and Eager Writeback, and the speedups due to Eager Writebacks are smaller in Table 10. These results suggest higher frequency interference can deteriorate performance in the baseline case more in a bandwidth-limited code.

## 6 Conclusions

Systems employing write-back caches have to contend with the following two issues: (1) Dirty writebacks contend with demand fetches for bandwidth and can impede the delivery of data, and (2) Finite memory bandwidth shared among demand fetches, implicit dirty writebacks, and other bus agents limit the performance of memory bound programs. These performance issues are important to a large and growing class of programs – those that consume large amounts of memory bandwidth and generate many data stores.

In this paper we have presented a new technique for dealing with these issues, called Eager Writeback, which can effectively improve overall system performance by shifting the writing of dirty cache lines from on-demand to times when the memory bus is idle. This time-shifting is accomplished by identifying and speculatively writing (“cleaning”) dirty lines whenever the bus is free. We have shown that for a wide variety of programs, once a dirty cache line has entered the LRU state it is rarely written to again. We use this fact to identify the lines that can be speculatively written (although this information could be of interest to many other intelligent cache management techniques as well).

We have shown that applying this technique can alleviate bandwidth congestion and improve performance for two kernels that are representative of these classes of applications. We have shown that when conventional writebacks compete with memory loads and defer the delivery of data, the Eager Writeback technique is able to remove the competition by evicting dirty data earlier. We have also shown that when “finite” memory or front-side bus bandwidth limits overall performance, eager writeback can alleviate this situation by utilizing earlier idle bus cycles. Eager Writeback can be implemented in a number of ways - for example, one way would be as an additional programmable memory attribute on top of the existing memory types, as described in Section 2, provided by a processor to speed up bandwidth-hungry applications, e.g. 3D games or content-rich multimedia applications.

Further investigation of this Eager Writeback mechanism will include the effect this approach has on other system performance issues. For example, Eager writeback can be expected to reduce context switching time overhead by flushing dirty lines in advance of a context switch. In addition, Eager Writeback can be used to push modified data closer to the globally observable memory level earlier, in order to reduce coherence miss latency. Similarly, the same analysis performed in this paper can be applied to write-update and write-invalidate protocols in a shared memory system to reduce coherence traffic in a way similar to (but perhaps simpler to implement than) that presented in [9].

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