Understanding and Improving Operating System Effects in Control Flow Prediction

Abstract

Many modern applications exercise the operating system kernel significantly, resulting in several implications including affecting the control flow transfer in the execution environment. This paper focuses on understanding the operating system effects on control flow transfer and prediction, and designing architectural support to alleviate the bottlenecks.

We characterize the control flow transfer of several emerging applications on a commercial operating system. We find that the exception-driven, intermittent invocation of OS code and the user/OS branch history interference increase the misprediction in both user and kernel code.

We propose two simple OS-aware control flow prediction philosophies to alleviate the destructive impact of user/OS branch interference. The first one consists of capturing separate branch correlation information for user and kernel code. The second philosophy involves using separate branch prediction tables for user and kernel code. We study the improvement contributed by the OS-aware philosophies to various branch prediction strategies ranging from simple Gshare to more elegant Agree, Multi-Hybrid and Bi-Mode predictors. On 32K entry predictors, incorporating OS-aware strategies yields up to 34%, 23%, 27% and 9% prediction accuracy improvement in Gshare, Multi-Hybrid, Agree and Bi-Mode predictors, resulting in up to 8% execution speedup.

1. Introduction

Every once in a while in systems research, we come to the point where we need to evaluate how well is the hardware suited for a given application, how well is it accommodating the operating system (OS), and how well is the application exploiting an OS's capabilities. Such issues are extremely important in order to fine tune system performance since we find that the three subsystems - application (workload), OS and hardware - are constantly evolving, and many times quite independently.

The nature and diversity of workloads has seen substantial changes that we need to go back to understanding the interplay of the three subsystems based on these new workloads. In particular, we note the growing importance of the OS in emerging application environments, with OS services being invoked much more often. Several recent studies [Rose95, Red00] have reported high operating system contribution to the overall execution time. In commercial applications such as databases and web services, the OS component has been observed to reach as high as 55% of the execution time in some of the workloads considered in this paper. Li et al. [Li00] also report higher OS involvement in Java applications compared to traditional SPEC workloads. The reason for the higher OS involvement in all these emerging environments is because the applications are in general multi-threaded and exercise the I/O subsystem much more extensively. This trend is likely to continue in the near future and it is very important to consider the operating system not only for complete system evaluations as other studies have pointed out [Rose95], but also when attempting to optimize the hardware and/or the application [Zhang02]. A detailed characterization of the interactions between the application-OS-hardware can have considerable ramifications in the design of each system component, and this paper takes a step in that direction.

We focus on one specific issue that has long been considered an important issue for performance optimization of state-of-the-art processors - control flow prediction. Current high performance processors provision aggressive support for Instruction Level Parallelism (ILP) and have deep pipelines to keep cycle times low. The delivered ILP and pipelining performance is critically dependent on being able to accurately predict the control (branch) flow in the program, so that we can execute more useful instructions and avoid stalling/squashing the pipeline.

Branch predictors for control flow prediction have been studied extensively with different programs [Yeh91, Young95, Sech96 and Mich97] and also with OS effects [Gloy96]. The OS affects control flow predictability by introducing the additional user/OS branch aliasing in branch predictor tables. The negative impact of kernel branches on branch prediction has been reported in [Gloy96]. We also find that user/OS execution can significantly increase the mispredictions in each part (Figure 1). For example, as shown in Figure 1a, kernel code nearly doubles the misprediction rates in 7 out of 13 of our benchmarks in a Gshare predictor.



Figure 1. Impact of User/OS Execution on Branch Prediction

Branch aliasing characterization shows that user/OS aliasing contributes to up to 24% of all misprediction and 46% of aliasing misprediction in the benchmarks studied in this paper. While eliminating all branch prediction aliasing is not trivial, it is our belief that the destructive user/OS

part can be alleviated with appropriate architectural support. There are numerous branch predictors that have been proposed to address different situations [Yeh93, Far93, Ever96, Spra97, Lee97, Chang96, Mish97 and Eden98]. These prediction mechanisms have paid less attention to the OS requirements and no particular scheme was proposed on tuning control flow prediction hardware for the OS. Our intention in this paper, however, is not to propose a new predictor to add to this list. Rather, it is to understand the fundamental nature of the user/OS interference and suggest simple and cost-effective optimizations that one can incorporate into any predictor to alleviate the impact of the OS activity on the control flow prediction.

Adhering to this philosophy, we investigate what causes the execution of a spectrum of applications with significant OS involvement to give worse branch prediction in the user and kernel modes by characterizing their execution using complete system simulation. This investigation shows that the interference between the branches in the user and kernel modes is leading to this poor performance. User and kernel branches have different characteristics (such as the direction bias) that cause the history information used by the predictors - and shared by both the user and kernel - to become polluted. Such pollution would not have happened if we had a separate predictor for each mode.

These observations lead us to advocate separating out branch prediction logic for user and kernel modes. By doing this, we are reducing the interference between the two. This approach can be easily integrated into existing prediction schemes without significantly complicating the logic. Separating resources for OS to reduce the user/OS interference exists for other resources (e.g., TLB, memory, etc.). However, to our knowledge, this paper presents the first study and quantitative analysis on orchestrating control flow prediction resources for OS.

The rest of this paper is organized as follows. In section 2, we characterize kernel branch behavior in different applications. We also quantify the effect of user/OS branch aliasing or interference¹. Section 3 introduces OS-aware prediction designed specifically to reduce user/OS branch aliasing and evaluates the improvement contributed by the OS-aware philosophies to various branch prediction strategies. Finally, conclusions are provided in Section 4.

2. Characterizing OS Branch Behavior

In this section, we use simulation of complete system activity to characterize OS branch execution and evaluate its impact on branch predictability. We use the SimOS [Rose95] full system simulation environment and select thirteen benchmarks with significant OS involvement. We provide a brief overview of the selected applications. A set of six Java applications from the

SPECjvm98 suite (s1 dataset) is executed using a commercial JDK from Sun Microsystems simulated on top of IRIX 5.3. *Vortex* is a database manipulation code and *gcc* is a compiler code from the SPECint95 benchmark suite. *Pmake* is a program development workload that is a variant of the compiled phase of the modified Andrew benchmark employed in [Rose95a]. The *sendmail* benchmark forwards email messages to local accounts on the system using the Simple Mail Transport Protocol (SMTP). The sizes of the messages vary from 1K to 1.5M. We also use three benchmarks that use PostgreSQL [Psql1, Psql2], a relational database management system (DBMS). The database is populated with relational tables for the TPC-C benchmark [TPC-C]. We evaluate the execution of three specific queries on this data set: a sequential table scan of a table with 1 million rows and a selectivity of 3% (*postgres.select*), an update to a field of a 300,000 row table (*postgres.update*) and a nested loop join involving two tables of sizes 11MB and 24KB (*postgres.join*). The SPECjvm98 applications, *pmake* and *gcc* are all executed to completion. All other applications are simulated for 2 billion cycles. Table 1 summarizes the complete system branch execution statistics of each studied benchmark.

	ns	ц	# =f C = = t == t	Conditional Branch Statistics				
	1) Ctio	Switch	# of Context Switches	Switches U		OS		
Benchmarks	Instruc (N	, (Exect	between User/OS	Static Sites	Dynamic Instances	Static Sites	Dynamic Instances	
db ¹	201	31	935,783	33,957	13,147,512	6,016	19,742,706	
jess ²	467	30	4,852,221	38,654	35,986,299	6,037	28,266,026	
javac ³	366	19	2,039,387	38,815	34,766,245	6,070	20,807,714	
jack ⁴	1,782	17	23,530,133	40,640	210,722,195	6,142	40,451,532	
mtrt ⁵	1,431	7	5,949,357	36,629	195,674,102	6,099	23,343,298	
compress ⁶	2,428	6	11,819,663	33,907	406,427,219	6,081	26,101,839	
gcc ⁷	1,036	8	4,975,087	13,570	138,915,436	4,696	13,845,466	
vortex ⁸	1,811	8	21,486,430	4,108	133,545,812	1,189	11,976,141	
pmake ⁹	1,117	17	1,018,543	11,651	122,460,692	5,273	33,821,182	
sendmail ¹⁰	1,494	54	1,438,961	4,516	139,259,991	5,553	75,069,918	
postgres.select11	1,516	38	5,632,788	8,417	107,228,678	6,201	93,551,585	
postgres.update ¹²	1,438	55	6,385,224	8,144	83,362,599	6,325	149,084,522	
postgres.join ¹³	1,849	15	5,858,258	8,606	220,730,099	6,099	72,657,859	

Table 1. Complete System Branch Execution Statistics

1. Performs multiple database functions on a memory resident database

2. Java expert shell system based on NASA's CLIPS expert system

3. The JDK 1.0.2 Java compiler compiling 225,000 lines of code 4. Parser generator with lexical analysis, early version of what is now JavaCC

5. Dual-threaded raytracer

6. Modified Lempel-Ziv method (LZW) to compress and decompress large file

7.Compiles pre-processed source into optimized SPARC assembly code

8.A full object oriented database

9. Two parallel compilation processes compile the Modified Andrew Benchmark[Oust90]

10.UNIX electronic mail transport agent

12.Object-Relational DBMS PostgreSQL executes an update query

13.Object-Relational DBMS PostgreSQL executes a join query

^{11.}Object-Relational DBMS PostgreSQL [psql] executes a select query

¹ We use the terms branch aliasing and branch interference interchangeably in this paper.

As illustrated in Table 1, the OS activity in the selected benchmarks ranges from 6% in *compress* to as high as 55% in *postgres.update*. The kernel portion of dynamic branch instances can be found to constitute a significant part in these applications. On the average, kernel branches, which include loops, error/bound checking, and other routine conditionals, constitute 27% of branch sites and 30% of dynamic branch instances in our benchmark executions. Branches are more frequent in OS (than in user mode) [Red00] because it has to be designed to handle all possible situations (i.e., abundant error and bound checking). Further, the OS functions are performed not just for one process/application but also for the system as a whole (other daemons, periodic book-keeping duties etc.).

2.1. Context Switch Profile and Branch Distribution

During the execution, branch instructions from user and OS code get interspersed. OS is activated either voluntarily by a system call from the application, or from a call by some other application, or implicitly by some underlying periodic/asynchronous (timer/device interrupt) mechanism. The inter-mingling of user and kernel branches can affect their behavior, compared to the execution when they were isolated from each other. Figure 2 shows the average number of executed branches in each mode per context invocation on the studied benchmarks. In all benchmarks except *db* and *postgres.update*², OS exercises fewer branches than user code in each visit to that mode.



Figure 2. Average Number of Executed Branches per Visit in User and Kernel Modes

We tracked the distribution of the number of executed branches for each context switch and the profiling results for a 5,000 context switch sample of benchmark *jack* are shown in Figure 3 for user and kernel code separately. While the user part shows random and sparse distribution, most of the OS contexts are caused by exception driven OS routines (e.g. TLB miss and page fault) and

 $^{^{2}}$ For benchmarks *db* and *postgres.update*, OS service *read* and *write*, which consists of far more branch instructions, dominates OS execution, causing higher average number of executed branches in OS.

execute very few branches. The distributions in Figure 3 for the kernel are a cause for concern since it indicates the possibility that the branch history may be not accurate for correct predictions (with user mode branches interference). On the other hand, the user branch distribution suggests that this problem may not be as severe for the user mode. Kernel invocations are more short lived, while user execution has reasonable time quanta to work with and build history.



2.2. OS Branch Execution Profile

We next examine what are the dominant kernel branches, and how their performance can be affected by the user code executing between OS operations. The pie chart of Figure 4 shows the percentage of OS branches (in benchmark *jack*) executed in the different services. Additional **benchmark results can be found in Appendix A1.** The top five components include: TLB miss (*TLB miss, 39%*); OS scheduling (*scheduling, 37%*); performing file and I/O services (*I/O & file system, 5%*), idle looping (*idle, 4%*); and kernel synchronization

(synchronization, 4%).



Figure 4. Where do the OS Dynamic Branches Come from?

These results show that we really need to focus mainly on the TLB handler (it is done in software on the given MIPS platform to facilitate the use of flexible page table structure and simplify the handling of sparse address spaces.) and the scheduler. Further, it should be noted that other services such as file system, synchronization etc., are directly invoked by the user code. Hence, their behavior (including that for branches) is influenced by the current state of the invoking application and the parameters of the call. So one would not like to associate the term "interference" for such services. On the other hand, TLB handling and scheduler invocations are not necessarily voluntary. It is useful to understand how the branches in these OS subsystems are invoked and whether history would have any bearing on their behavior for predictability – so that we can better understand if the predictability of these branches would be affected by the user code getting in-between.

Table 2 gives further evidence of the significance of the TLB handling and scheduler subsystems on the overall branches within the OS. The *utlb* routine that handles TLB misses and the *checkRunq* routine that does scheduling (picking the next process to run) account for over 70% of the OS dynamic branches. Though *utlb* and *checkRunq* both have high dynamic branch instances, the number of actual branch sites is quite small. We briefly go over these routines below identifying the branches in these routines and their anticipated behavior qualitatively.

OS Poutina	% Dynamic	Active Branch	
OS Koutine	Branches	Sites	
utlb	38.7	1	
checkRunq	34.2	6	
idle	3.89	3	
syscall	2.8	14	
io_splock	2.38	5	
exception_ip12	2.08	6	
bcopy	1.5	6	
mrlock	1.17	8	
vsema	0.65	5	
uiomove	0.6	10	
findchunk	0.48	8	
blkclr	0.48	1	
ufget	0.48	8	
mrunlock	0.45	3	
copyout	0.42	3	
getff	0.42	7	
psema	0.42	6	

Table 2. OS Routine Branch Characterization

The *utlb* handler has only 1 branch, and the reason for its high dynamic instance is because this routine is invoked frequently. The *utlb* routine is invoked directly by the hardware which is the only entity that can invoke this operation. On the other hand, the scheduler (*checkRunq*) is

invoked from several places. First, this operation is needed for scheduling decisions (by consulting the ready queue) whenever the time quantum expires (triggered by timer interrupt), when I/O device activity completes (there are usually priority boosts and rescheduling may be needed) and idle looping, or even voluntarily during blocking (making semaphore, I/O requests etc.) or other process state change activities (such as termination). Consequently, it is to be noted that, while *utlb* invocations are only the consequence of application behavior, the scheduler actions are invoked from all over the OS and are invoked either asynchronously (by hardware events) or voluntarily due to system load/behavior. In all, we found there are more than 23 events that can cause *checkRunq* to be invoked.

2.3. Characteristics of OS Branches

We investigate specific properties of these OS branches and their architectural implications in this subsection.

2.3.1. Branch Directions and Weakly Biased Branches in Kernel

It is well known that branches often have biased behavior and many branches are either usually "taken" or usually "not taken". The conventional branch history table (BHT) counters exploit this behavior to predict future outcomes of that branch. However, when branches showing different biases are mapped into the same entry of the predictor table, aliased branches update BHT counters with different directions, leading to aliasing mispredictions.

We measure branch direction distribution in order to gain more insight on bias behavior of the user and OS branches. Figure 5 shows the result on benchmark *jack*. Additional **benchmark results can be found in Appendix A2.** The branch sites are categorized into 100% "taken" (always-taken), 0% "taken" (always-not-taken) and groups between them. For example, the marker "70%-79%" on X-axis implies that branch sites that fall into this category have a possibility of 70% to 79% to be "taken".



Figure 5. User and OS Branch Directions

Figure 5 shows that user and OS branches behave differently in terms of the bias or direction distribution. For example, 46% of dynamic branches in kernel are "always taken" while their counterparts in user code are only 15%. On the other hand, 18% of dynamic branches in kernel are "always not taken" and that number in user mode can be as high as 42%. This implies that even when the strongly biased user and kernel branches are mapped into the same BHT counter, it is likely that they will lead to aliasing misprediction.

Another interesting observation in Figure 5 is that while the dominant portion of branch sites is strongly biased (i.e. always taken or always not taken) in user code, a significant number of branches are weakly biased in OS code. More precisely, we observed that 8.8% of dynamic branches that contribute to the weakly biased (with the category of 40%-49%) branches shown in Figure 5, come from a wide range of 22 kernel service routines. The weakly biased OS branches showing interleaved directions are also found on other benchmarks. Among these is the *checkRunq* routine that is frequently invoked. This routine checks through queues to find out if a rescheduling decision needs to be made. Intuitively, it can be hypothesized that the execution characteristics of such a routine are more a function of the load on the system more than anything else. Even when the load does not change very much during the course of this execution, there are bursts of I/O, synchronization activity and other events that can exercise the *checkRunq* differently, causing its branch to vary direction. Weakly biased branches can be a problem to many branch predictors, which rely on the persistent history and saturated 2-bit counter for accurate branch prediction.

2.3.2. How Correlated are Kernel Branches?

We observe that many OS branches are very correlated and hence benefit from two-level predictors that exploit global history correlation. It should be noted that the *utlb* routine has a single branch that is nearly always taken. While static predictors would suffice for this branch, previous history is also a very good indicator for this particular branch that accounts for a large portion of the kernel's dynamic branches. Further, OS exception handlers frequently use binary decision trees to classify and dispatch vectored interrupts from the trap entry point to the specific fault handler. Figure 6a shows an example use of such a structure in the general exception handler *exception_ip12* OS code. This handler dispatches an exception to the corresponding kernel processing routine based on the value of the exception vector. The binary decision tree based branch sequence of this handler is given in Figure 6b. It can be observed that the branches in the OS routine *inttrap* will be correlated with a NNT branching sequence while the branches in

systrap will be correlated with a NNNT branching sequence. Hence Gshare [McFar93] and GAg [Yeh91] predictors work extremely well with these branches.



2.3.3. Impact of Intermittent Kernel Execution on Strongly Biased Kernel Branches

Even strongly biased OS branches can experience mispredictions due to the user code interference. An example for this can be obtained from the *utlb* routine from the OS. Since the *utlb* handler needs to be very efficient, this code is usually written in assembly and is hand-optimized. There are exactly 13 instructions in this routine, with the bulk of the instructions used to read the page table entry from the memory system and load it into the TLB. There is exactly 1 branch within this code that is strongly taken. But intervening user code interference can result in mispredictions in even such strongly biased branches. Consider a correlation based branch predictor, and two scenarios of branch history shift register (BHSR) contents in Figure 7. In the absence of user code intervention, the correlation shift register may look like (a), and leads to correct prediction, whereas the intervening user code may result in the correlation information to look like (b) and result in aliasing misprediction.



2.3.4. Characterization of User/OS Aliasing

	II Dentrice Osnarc,	11110 11110	prediction	(Matc)
Danahmarka	Matria	OS/OS	User/User	User/OS
Denchinarks	Metric	Aliasing	Aliasing	Aliasing
db	% of Misprediction	6.2	28.2	19.4
(MR=4.8%)	% of Correct Prediction	1.4	2.7	2.1
jess	% of Misprediction	3.3	37.3	20.1
(MR=8.8%)	% of Correct Prediction	1.4	6.5	3.9
javac	% of Misprediction	3.1	34.7	16.4
(MR=7.1%)	% of Correct Prediction	0.7	5.2	2.4
jack	% of Misprediction	1.3	35.7	18.8
(MR=8%)	% of Correct Prediction	0.6	7.9	4.7
mtrt	% of Misprediction	1.3	23.5	10.2
(MR=4%)	% of Correct Prediction	0.2	3.8	1.1
compress	% of Misprediction	0.7	12.0	2.5
(MR=3.1%)	% of Correct Prediction	0.1	4.7	0.2
gcc	% of Misprediction	0.3	41.5	6.2
(MR=10.2%)	% of Correct Prediction	0.1	10.5	1.9
vortex	% of Misprediction	0.1	39.4	11.7
(MR=7.8%)	% of Correct Prediction	0	11.8	3.8
pmake	% of Misprediction	3.6	25.1	9.4
(MR=6.6%)	% of Correct Prediction	0.5	4.6	1
sendmail	% of Misprediction	22.2	9	23.7
(MR=9.3%)	% of Correct Prediction	3.8	1.7	2.9
postgres.select	% of Misprediction	7.4	16	19.7
(MR=3.1%)	% of Correct Prediction	0.9	2.4	2.2
postgres.update	% of Misprediction	7.8	18.4	22.4
(MR=5.7%)	% of Correct Prediction	1.7	3.3	3.8
postgres.join	% of Misprediction	1.1	15	4.5
(MR=5.6%)	% of Correct Prediction	0.2	5.3	1.1

 Table 3. Characterization of Branch Aliasing
 (8K BHT Fentries Cshare MR: Misprediction Rate)

It is well known that branch aliasing, namely, several branches mapping to the same entry in the prediction tables, impacts the branch prediction accuracy. Although some of the aliasing can be neutral or constructive, a large part of the aliasing is often destructive. We performed a branch aliasing characterization to understand the impact of user/OS aliasing. In order to do that, we instrument our branch prediction simulators to record the histograms of mappings between branch instructions and BHT entries. We capture BHT access references, and we mark each as a member of one of the following categories: (1) if a BHT access is mapped to the same branch site from user (or kernel) space, we record an instance of a user/user (or a OS/OS) hit; (2) if a BHT access is mapped to different branch sites from user (or kernel) space, we record an instance of a user/user (or a OS/OS) hit; (2) if a BHT access is mapped to different branch sites from user (or kernel) space, we record an instance of a user/user (or a OS/OS) hit; (2) if a BHT access is mapped to different branch sites from user (or kernel) space, we record an instance of a user/user (or a OS/OS) hit; (2) if a BHT access is mapped to different branch sites from user (or kernel) space, we record user/US aliasing; (3) if a BHT entry or access is mapped to branches from different spaces, we record user/OS aliasing; (4) Finally, a cold miss is counted when a BHT entry is first mapped to a branch. To reduce the effect of capacity misses, we examine the mapping behavior on a BHT with 8K entries.

Branch aliasing is attributed to user (User/User Aliasing), kernel (OS/OS Aliasing) and the interaction between them (User/OS Aliasing). The percentages of misprediction and correct prediction caused by different aliasing are shown in Table 3. In experiments with a Gshare predictor of size 8K BHT entries, user/OS aliasing on the average contributes to the 14.2% and 2.5% of misprediction and correct prediction respectively, implying most of the user/OS aliasing are negative. The percentage of misprediction caused by user/OS aliasing does not change significantly when the predictor size is increased from 8K entries to 64K entries. This indicates that just increasing the capacity of the branch predictor will not effectively solve the user/OS aliasing problem. The user/user aliasing that many previous studies have evaluated is still important as the results observed from Table 3 indicate. However, user/OS aliasing is also a big source for mispredictions. Table 4 characterizes the impact of branch aliasing on misprediction in user and OS component. With an 8K BHT entry Gshare, approximately 22-62% of mispredictions in OS code are found to be from user/OS aliasing, suggesting that it is essential to protect kernel branch predictors from interference from user code.

Table 4. Characterization of Mispredictiondue to Branch Aliasing in User and OS Component(8K BHT Entries Gshare, MR: Misprediction Rate)

Benchmarks		OS/OS	/OS User/User User/OS			
		Aliasing	Aliasing	Aliasing	MR%	
JL User			39.0	13.5	8.6	
db	OS	22.3		34.9	2.3	
inco	User		47.3	12.8	12.3	
Jess	OS	15.5		47.7	4.3	
101/00	User		42.0	10.0	9.3	
Javac	OS	17.9		47.0	3.5	
iaalt	User		43.9	11.6	7.8	
јаск	OS	6.9		50.4	9.4	
matut	User		26.6	5.8	3.9	
murt	OS	11.5		44.0	4.7	
00mmr044	User		12.5	1.3	3.1	
compress	OS	16.8		32.0	2.1	
gcc	User		43.6	3.3	10.6	
	OS	6.7		62	5.8	
vontor	User		44.7	6.6	7.5	
vonex	OS	1		49.5	11.3	
nmaka	User		28.8	5.4	7.2	
pillake	OS	28		36.2	4.3	
con denoil	User		19.9	26.2	6.3	
senuman	OS	40.5		21.6	14.9	
postgras salast	User		26.7	16.5	3.5	
posigies.select	OS	18.4		24.5	2.6	
postares undate	User		29.3	17.9	9.6	
posigres.update	OS	21		29.9	3.5	
nostaras ioin	User		16.1	2.4	7	
posigies.join	OS	16.2		33.5	1.6	

In summary, we observe that user/OS branch aliasing can significantly deteriorate branch prediction accuracy. This is primarily attributed to the exception-driven and intermittent kernel branch execution that causes inaccurate branch history information in BHSR. Moreover, user and kernel branches have different bias distribution, which in turn spreads user-kernel branch aliasing references across a wide range of BHT entries. The above observations motivate the need for OS-aware branch prediction techniques.

3. Alleviating Impact of User/OS Interference

It is clear from the prior sections that user and kernel code possess different branch behavior, often resulting in conflicts in unified structures that capture branch history. In subsections 3.1 and 3.2, we present two philosophies that aim to alleviate the destructive impact of OS branch execution on branch predictability.

During the initial period of a context switch, both user and kernel history patterns coexist in history capturing structures. In Gshare and any correlation based predictor, this can happen in shift registers (BHSRs) that capture correlation between branches and/or branch history tables (BHTs). One solution is to use separate shift registers to individually keep track of branch correlation and another solution is to use separate BHTs.

3.1. OS-aware Philosophy 1 - Splitting Correlation Shift Registers

We illustrate our OS-aware philosophy in the context of a Gshare predictor, but it can be applied to other correlation-based predictors as well. A Gshare predictor with split correlation history shift registers is illustrated in Figure 8. This predictor functions exactly the same as a conventional Gshare predictor except that two dedicated BHSRs (i.e., U-BHSR for user and K-BHSR for kernel) are used to gather branch correlation patterns and to generate BHT indexing. By using K-BHSR for kernel branches, OS-aware Gshare-1 overcomes the loss of branch history patterns in kernel mode. Meanwhile, OS-aware Gshare-1 dynamically switches between BHSRs when a context switch occurs, preventing the BHT indexing ambiguity during the initial stages of a context switch.



Figure 8. OS-aware Philosophy 1 applied to Gshare - Employs Split Correlation History

3.2. OS-aware Philosophy 2 - Splitting BHTs for User and Kernel

The proposed OS-aware philosophy-1 aims to preserve accurate BHT counter indexing during a context switch. However, user/OS aliasing can still occur when user and kernel branches have the same XORed global history pattern, but opposite biases. Due to their different branch bias distribution, user and kernel branches can update BHT counters in different manners. To reduce the destructive user/OS branch aliasing in BHT, we propose the use of split BHT for each, which yields OS-aware Gshare-2 predictor, as shown in Figure 9. This predictor eliminates the destructive user/OS aliasing by using separate correlation and history information for user mode and kernel mode. It is also observed that when branch history tables are split into user and kernel parts, the kernel BHT can be smaller than the user BHT because of the fewer active branch sites in kernel (as shown in Table 1). Due to the difficulty in creating a 7:1 or 3:1 split (due to the user BHT becoming not power of 2), we kept the user BHT at half the size of the original Gshare and allocate kernel BHT with a fixed size of 2K entry in our experiment.



Figure 9. OS-aware Philosophy 2 applied to Gshare- Employs Split BHTs

Filtering out kernel branches can easily be done at run time by using the Processor Status Register (PSR). Typically, in a microprocessor a set of PSR bits is used to record and identify kernel-user execution mode or privilege level. For example, MIPS R10000 [Yeag96] uses KSU field in PSR to identify current execution mode and Intel's next generation IA-64 Itanium (Merced) [MPR0400] uses PSR.cpl to determine one of 4 privilege levels (level 0-3). The corresponding field in PSR can be used to select the appropriate predictor. At runtime, instructions from a fetch unit are filtered into an active part of prediction resource (user or kernel, depending on execution mode).

In summary, the OS-aware Gshare-1 (split correlation-Gshare) and OS-aware Gshare-2 (split history table-Gshare) predictors are designed specifically to reduce user/OS branch aliasing without adding extra hardware for branch de-aliasing. They consume equivalent or less resource than a conventional Gshare.

3.3. Integrating OS-aware Prediction Philosophy with Other Predictors

OS-aware prediction is a philosophy suggested by our characterization study, not necessarily a particular predictor. We surveyed literature to identify branch predictors, which may be poised to handle branches with the characteristics unveiled in the earlier sections. Although not targeted for OS-user branch interference, Multi-Hybrid [Ever96], Agree [Spra97] and Bi-Mode schemes [Lee97] do contain mechanisms tailored for branches with heterogeneous characteristics and/or de-aliasing. Table 5 summarizes these schemes, and the additional cost used for branch de-aliasing. The sizes of all the predictors are normalized to Gshare to give an indication of the associated area cost.

Predictor	Description of feature to exploit heterogeneous branches or De-aliasing	Additional Branch De-aliasing Hardware	Predictor Size Normalized to Gshare(8k- 256k)
Gshare [McFar93]	Consists of one correlation shift register (BHSR) and one BHT. BHSR is XORed with branch address bits of a branch address to index BHT entry. The XORing helps to reduce aliasing effects.	0	1
Multi-Hybrid ^{1, 2} [Ever97]	Consists of multiple single-scheme components: simple 2-bit (2bc), GAs, Gshare, Pshare and always taken predictor. Use of simple 2-bit predictors (2bc) and static predictors as components of the multi-hybrid predictor provides quick warm up after a context switch.	5×2K predictor selection counters in BTB	1.04-2.25
Agree [Spra97]	Converts instances of destructive aliasing into either constructive or neutral aliasing by attaching each branch with a biasing bit that predicts the most likely outcome of that branch.	2K biasing bits in BTB	1-1.13
Bi-Mode [Lee97]	Uses separate history tables for taken and not-taken branches, and a selection branch history table. This classification helps to alleviate destructive aliasing while keeping the harmless aliasing together.	the third BHT for dynamic bias selection	1.5
OS-aware philosophy-1 [this paper]	OS-aware Gshare predictor uses separate shift registers (U-BHSR and K-BHSR) for capturing path history patterns.	1 shift register	1
OS aware philosophy-2- [this paper]	OS-aware Gshare predictor that uses separate branch history tables for user and kernels. Kernel-BHT is 2K and User-BHT is 50% of Gshare.	consumes less BHT resource than Gshare	0.51-1

 Table 5. A Comparison of Several Branch De-aliasing Schemes

1. Our simulated Multi-Hybrid does not include AVG predictor [Chang95] because it needs source recompilation which often is difficult for commercial and complicated software like OS and JVM.

2. As indicated by [Ever96], we allocate half of the total budget for Gshare, a quarter of the total budget for Pshare, and 1/8 for 2bc and Gas respectively. The priority ordering of the component predictors is 2bc, GAs, Gshare, Pshare and always taken scheme.

All these predictors contain a Gshare predictor or a Gshare indexing [Ever96, Spra97 and Lee97]. To integrate the proposed techniques, we simply replace the conventional Gshare component used in the above predictors with the proposed OS-aware split-correlation-Gshare and split history-Gshare.

Table 6a shows the average (of the 13 studied benchmarks) misprediction rates of each baseline predictor and the percentage of misprediction reduction by incorporating the OS-aware philosophies proposed in this paper. Table 6b further illustrates the breakdown of the misprediction reduction in user and OS parts, for each individual benchmark.

		Size (Number of BHT entries, not including de-						
Schemes	Metric	aliasing overhead)						
		8k	16k	32k	64k	128k	256k	
Gshare	Misprediction(in %)	14.03	12.35	10.89	9.64	8.66	8	
Gshare+OS-aware Phil. 1	% of Misprediction Reduction	31%	33%	34%	32%	31%	29%	
Gshare+OS-aware Phil. 2	% of Misprediction Reduction	20%	24%	22%	20%	17%	15%	
Multi-Hybrid	Misprediction(in %)	10.87	9.53	8.58	7.66	6.96	6.3	
Multi-Hybrid+OS-aware Phil. 1	% of Misprediction Reduction	21%	22%	23%	23%	22%	22%	
Multi-Hybrid+OS-aware Phil. 2	% of Misprediction Reduction	13%	12%	13%	11%	10%	8%	
Agree	Misprediction(in %)	12.59	11.41	10.46	9.66	9.13	8.78	
Agree+OS-aware Phil. 1	% of Misprediction Reduction	27%	27%	27%	26%	25%	24%	
Agree+OS-aware Phil. 2	% of Misprediction Reduction	19%	22%	22%	20%	20%	19%	
Bi-Mode	Misprediction(in %)	7.7	6.95	6.42	6.07	5.79	5.57	
Bi-Mode+OS-aware Phil. 1	% of Misprediction Reduction	10%	9%	9%	9%	9%	9%	
Bi-Mode+OS-aware Phil. 2	% of Misprediction Reduction	4%	2%	1%	1%	0%	0%	

Table 6a Misprediction Reduction by Introducing OS-aware Philosophies

As described in subsection 3.1, philosophy 1 only separates the branch history shift registers. The partitioning of the BHT for user or OS happens dynamically. The resource available for the code is not less than that in the baseline. Hence, Philosophy 1 is never inferior to the baseline. Philosophy 2 is at times worse than the baseline. In Philosophy 2, the partitioning of the BHT between user and kernel code is done statically. Both the user and kernel BHTs are smaller than the unified BHT in the baseline configuration. In the configurations studied in this paper, the user BHT is only 50% of the baseline BHT, and the K-BHT is fixed at 2K in all cases. Hence, the overall size of the philosophy 2 BHT is not much greater than 50% of the BHT in the baseline. A 2K K-BHT is seen to be sufficient to capture all history patterns in the OS code and except in *postgres.update*, the mispredictions in OS code goes down. For the user part, the small size of the U-BHT (4K BHT entries) can detrimentally affect the performance on benchmarks *compress*, *gcc*, *pmake*, *postgres.select* and *postgres.join*.

On the average, with a 32K BHT entry Gshare, incorporating OS-aware Phil.1 and Phil.2 strategies reduces 34% and 22% of the misprediction. OS-aware philosophies also reduce the misprediction of Multi-Hybrid, Agree and Bi-Mode predictors. For instance, compared with the 32K BHT entry baseline predictors, OS-aware Multi-Hybrid, Agree and Bi-Mode predictors yield up to 23%, 27% and 9% prediction accuracy improvement respectively, implying that OS-aware predictions still provide significant improvements on some of the most powerful predictors.

Benchmarks		Schemes and % of Misprediction Reduction							
		Gshare		Multi-Hybrid		Agree		Bi-Mode	
		+ OS-aware		+ OS-aware		+OS-aware		+ OS-aware	
		Phil. 1	Phil. 2	Phil. 1	Phil. 2	Phil. 1	Phil. 2	Phil. 1	Phil. 2
	User	28%	23%	20%	15%	21%	17%	9%	8%
db	OS	28%	8%	7%	11%	15%	7%	7%	10%
	Full-System	28%	19%	16%	14%	20%	16%	8%	8%
	User	39%	31%	31%	25%	34%	27%	13%	8%
jess	OS	52%	42%	12%	15%	44%	36%	13%	20%
	Full-System	42%	34%	28%	23%	36%	29%	13%	10%
	User	28%	19%	20%	13%	24%	17%	8%	4%
javac	OS	40%	36%	10%	20%	42%	41%	9%	18%
-	Full-System	30%	22%	18%	14%	27%	21%	8%	6%
	User	57%	47%	47%	39%	51%	42%	21%	13%
jack	OS	79%	82%	29%	49%	64%	70%	43%	53%
	Full-System	61%	53%	46%	40%	53%	46%	23%	17%
	User	27%	15%	27%	19%	20%	11%	7%	4%
mtrt	OS	60%	59%	15%	23%	49%	48%	19%	27%
	Full-System	31%	20%	25%	19%	22%	15%	8%	6%
	User	11%	-27%	10%	-3%	7%	-30%	3%	2%
compress OS Full-Sys	OS	43%	29%	7%	11%	19%	12%	8%	13%
	Full-System	12%	-25%	10%	1%	7%	-29%	3%	3%
gcc User	User	16%	2%	10%	-1%	12%	2%	10%	-1%
	OS	46%	55%	3%	26%	62%	68%	14%	31%
-	Full-System	18%	5%	10%	0%	15%	7%	10%	1%
	User	76%	63%	71%	48%	73%	65%	35%	28%
vortex C	OS	96%	97%	30%	54%	98%	99%	67%	77%
	Full-System	78%	68%	70%	48%	78%	72%	37%	31%
	User	8%	-6%	4%	-11%	6%	-7%	4%	-6%
pmake	OS	11%	2%	2%	8%	7%	13%	3%	8%
	Full-System	8%	-4%	4%	-8%	6%	-5%	4%	-4%
	User	5%	3%	1%	0%	3%	2%	2%	1%
sendmail	OS	5%	0%	3%	1%	3%	2%	2%	2%
	Full-System	5%	1%	2%	0%	3%	2%	2%	2%
	User	56%	45%	47%	12%	50%	48%	36%	-34%
postgres.select	OS	27%	8%	17%	22%	26%	29%	14%	13%
-	Full-System	45%	30%	35%	16%	40%	40%	26%	-14%
	User	35%	30%	25%	24%	25%	25%	23%	21%
postgres.update	OS	14%	-10%	6%	6%	9%	17%	5%	5%
	Full-System	27%	14%	17%	17%	19%	22%	16%	15%
	User	12%	-6%	8%	-1%	10%	-6%	3%	-6%
postgres.join	OS	42%	32%	15%	26%	35%	44%	26%	34%
1 U - J	Full-System	14%	-4%	9%	0%	12%	-3%	4%	-5%

Table 6b. Breakdown of Misprediction Reduction by Introducing OS-aware Philosophies (8K BHT Entries)

As shown in Table 6a and Table 6b, philosophy 1 outperforms philosophy 2 on most of the dealiasing predictors we examined. Considering overall performance, in more than half the cases, the performance gain due to the elimination of user/OS aliasing by philosophy 2 outweighs the performance loss due to individually using smaller prediction tables for each part. More precisely, for example, the OS-aware Phil.2 reduces 22% of misprediction on a conventional Agree predictor of 32K BHT entries, using only 18K entries BHT consisting of a 16K entries U-BHT and a 2k entries K-BHT.

3.4. Performance Evaluation

We evaluate the benefits of integrating the above predictors with OS-aware predictions on a dynamically scheduled superscalar processor using a full-system simulator that captures OS behavior as well. Table 7 summarizes the configuration of the simulated machine architecture. We use SimOS MXS model [Benn95], which simulates a superscalar microprocessor with multiple instruction issue, register renaming, dynamic scheduling, and speculative execution with precise exceptions. The simulated architectural model is an 8-issue superscalar processor with instruction latencies as in the MIPS R10000 [Yeag96]. By default, the branch prediction algorithm allows fetch unit to fetch through up to 4 unresolved branches. In our model, a misprediction will cause a 10-cycle penalty. BHSR is speculatively updated and later corrected after a misprediction. BHT counter update takes place in order at instruction commit time.

Processor Core				
Fetch/Decode/Issue/Retire Width	8			
Instruction Window Size	128			
Reorder Buffer Size	128			
Number of Function Units	2×Issue Width			
Latency of Function Units	MIPS R10000 Like			
Branch Target Buffer (BTB)	2048-entry, 4-way			
Return Address Stack	32-entry w/ misprediction repair			
Misprediction Penalty	10 cycles			
Load Store Queue Size	64			
Memor	y Hierarchy			
MMI	Fully associative TLB, 48-entries, 4KB			
WWO	page size			
I 1 L Cache	32KB, 2-way(LRU), 64B blocks,			
El I-Cacile	4MSHRs, 2 ports, 1 cycle latency			
L1D-Cache	32KB, 2-way(LRU), 32B blocks,			
El D-Cache	4MSHRs, 2 ports, 1 cycle latency			
L2 Cache	1MB, 2-way(LRU), 128B blocks,			
L2 Cache	4MSHRs, 2 ports, 10 cycle latency			
Memory	256MB, 60-cycle access			

Table 7. Simulated Machine Architecture

Figure 13 shows the IPC performance for this scenario. Since instruction counts are the same, IPC improvement is indicative of execution cycle improvement. Results are depicted for the 13

evaluated programs. Comparison of predictors integrating OS-aware philosophy with Gshare, Multi-Hybrid, Agree and Bi-Mode predictors is presented. The scale of Y-axis is varied for each benchmark due to their differences in IPC. Philosophy 1 improves IPC performance on all of the benchmarks for all of the four types of base predictors. This benefit is particularly substantial in those programs where user/OS aliasing is significant, such as *jess*, *jack*, *vortex*, and *postgres.update* (as was illustrated in Figure 1). The same trend can be observed in programs such as *javac* and *db*. For those programs where the impact of user/OS aliasing on misprediction is less significant (for instance, *compress* and *pmake*), the integration of OS-aware philosophies show only limited improvement.





Figure 13. IPC of OS-aware Predictors

Integration of philosophy 2 results in improvement in many cases, even though the predictor size is not much more than 50% of the baseline predictor. In most of the cases in Gshare, Multi-Hybrid and Agree predictors, despite the small size, philosophy 2 still results in improvement. In the case of the Bi-Mode predictors, philosophy 2-integrated case is inferior to the baseline for 5 of 13 benchmarks. However, if one compares them to a baseline that is comparable in size (i.e., 16K BHT entries), philosophy 2 with 18K BHT entries (16K U-BHT + 2K K-BHT) outperforms 16K BHT entries baseline predictor in all cases, resulting up to 10% of IPC speedup (**See**

Appendix 3 for details).

Compared with a Gshare predictor, the two proposed philosophies - Phil.1 and Phil.2 yield up to 8% and 7% of IPC improvement respectively. This improvement is a result of the removal of aliasing mispredictions. The integration of OS-aware philosophies into Multi-Hybrid predictor yields up to 5% of IPC gain. As described earlier, Multi-Hybrid allocates the largest prediction resource to its Gshare component and its overall prediction accuracy is more impacted by Gshare than any other predictor. Hence, the replacement of the conventional Gshare with the proposed OS-aware Gshare predictors improves performance.

By introducing OS-aware philosophies on the Agree predictor, up to 7% of IPC improvement can be achieved. The performance of Agree predictor is largely dependent on branch biases and possibility of identifying the biased behavior the first time the branch is introduced into the BTB. If the branch does not show strongly biased behavior, there is still frequent aliasing between instances of a branch that do not comply with the biasing bit and instances which do comply with the biasing bit. Once we incorporate OS-aware policies into the Agree predictor, the filtering out of the visible portion of weakly biased kernel branches leads more U-BHT entries to reach "agree" status.

The IPC improvement of OS-aware Bi-Mode is marginal (1%), but it should be noted that the OS-aware Bi-Mode consumes only equivalent or less resource to achieve this performance enhancement. Thus, OS-aware prediction leads to the same performance with less hardware.

The results shown in Figure 13 also indicate that the combination of the OS-aware prediction and a simple predictor (for instance, Gshare) can outperform sophisticated predictors (e.g., Multi-Hybrid and Agree) with larger size configuration.

In summary, architectural support for specific OS branch behavior can enhance prediction performance without increasing predictor size or complexity. Current and next generation microprocessors are becoming increasingly sensitive to branch prediction accuracy due to the use of deeper pipelines and wider issue microarchitecture. The proposed techniques are expected to yield more ILP performance benefit on aggressive implementations with higher misprediction penalties.

3.5. Discussion

We motivated the research in this paper using Figure 1, which showed that kernel interference increases user misprediction from 1.1x to 6x (with an average of 2.1x). Similarly, we observed that user interference increases OS misprediction from 1.3x to 129x (with an average of 13x). In this subsection, we revisit this characterization in the presence of the OS-aware philosophies.



Figure 14. Impact of User/OS Execution on OS-Phil. 1 Branch Predictors

Figure 14 illustrates the impact of user/OS execution on branch prediction after OS-aware philosophy 1 is integrated with Gshare. Compared with Figure 1, OS-aware philosophy 1 significantly reduces the negative impact of user/OS interference on branch prediction, resulting in the drop of mispredictions from 2.1x to 1.2x and from 13x to 2x in user and OS space respectively.

As described in subsection 3.1, philosophy 1 reduces the misprediction by providing interference free branch history for both user and kernel sides. In order to investigate the impact of the initial state of the K-BHSR when switching to the OS mode from the user mode, an experiment is performed with a variant of Philosophy 1 where the K-BHSR is cleared out upon an OS call is made. We compare the misprediction rates of a baseline Phil. 1 with this variant of Phil. 1 scheme. As shown in Table 8, the Phil. 1 with K-BHSR zeroing out on an OS call provides similar performance result with that of a baseline Phil. 1, implying that removing the interference in the predictor state between user and kernel modes is more important than really figuring out what state to leave the predictor in when entering the kernel.

ostgres. postgres. postgres. db jess javac jack mtrt compress gcc vortex pmake sendmail select update ioin Baseline 3.49 5.09 4.98 3.15 2.75 2.68 8.42 1.69 6.01 8.8 1.7 4.16 4.86 with K-BHSR Zeroing out 3.40 5.12 5.01 3.13 2.73 2.74 8.43 1.69 6.03 9.02 1.69 4.15 4.85

Table 8. Effect of Zeroing out K-BHSR in Phil. 1

Similarly, Figure 15 revisits the impact of user/OS on branch misprediction after OS-aware philosophy 2 is integrated. Compared with Figure 1, OS-aware philosophy 2 cost-effectively reduces the negative impact of kernel code on branch misprediction in user part. The misprediction reduction by OS interference removal outweighs the extra misprediction caused by using less (50%) BHT resource on all benchmarks except *compress* and *pmake*. In the OS part, the fixed size 2K K-BHT still outperforms the performance of a unified 16K BHT on benchmarks *jess*, *javac*, *jack*, *mtrt*, *gcc*, *vortex* and *postgres.join*.





Figure 15. Impact of User/OS Execution on OS-Phil. 2 Branch Predictors

4. Summary and Conclusion

Control flow prediction is one of the key issues in the design of high performance processors. It is extremely important that processor hardware, software and the operating system collaborate with each other to deliver high performance. The operating system affects control flow predictability by introducing the additional user/OS branch aliasing in predictor hardware. Compared to the branches in user code, the OS branches are usually invoked by the exception-driven and intermittently executed kernel routines and may have different biased behavior caused by performing operations not common in user mode. Thus, when interacted with user branches, the OS branches increase misprediction significantly. Current branch predictors have paid less attention to the OS requirements and therefore, do not contain mechanisms to specifically alleviate the user/OS aliasing.

This paper focuses on understanding and improving the control flow predictability in the light of the OS code. Having characterized kernel and user branch execution behavior and quantified their impact on misprediction rates of a commercial OS, we propose OS-aware branch prediction designed to reduce user/OS branch aliasing without adding extra hardware for branch de-aliasing.

The proposed OS-aware prediction is a philosophy that advocates orchestrating branch correlation information and/or branch history information for user and kernel branches individually. The proposed OS-aware philosophy can be incorporated into any other predictor, ranging from a naïve Gshare to the more sophisticated Multi-Hybrid, Agree and Bi-Mode predictors, to further improve prediction accuracy. More precisely, on the 32K BHT entry predictors, incorporating OS-aware strategies into previously proposed Gshare, Multi-Hybrid, Agree and Bi-Mode predictors yields up to 34%, 23%, 27% and 9% prediction accuracy improvement and up to 8%, 5%, 7% and 1% execution speedup respectively. Simulation results also show that the combination of the OS-aware prediction and a simple predictor (for instance,

Gshare) can outperform sophisticated predictors (e.g., Multi-Hybrid and Agree) with larger size configuration.

OS-aware philosophy improves prediction performance by cost-effectively alleviating the user/kernel branch aliasing. Moreover, it provides opportunities for catering user and kernel branches with differently tuned structures. For example, compared with a conventional design, the OS-aware philosophy 2 requires access to only one of the smaller prediction tables for a given branch instruction mode (kernel or user), which can result in energy savings and low-latency access. These advantages are valuable in the light of power and clock frequency constraints in emerging processor and branch predictor designs [Jim00][Pari02]. In our future work, we plan to model the energy consumption and the access latency of the OS-aware branch predictors.

Historically, privilege bits are used to protect system critical resources such as page tables and process control blocks for security purposes. The results of our study show that on a fine-grained resource sharing superscalar microprocessor, the protection of performance-critical microarchitecture hardware, such as branch prediction tables, also is important. It is likely that future high-end microprocessors will aggressively support more predictions (e.g. value prediction). Further research is needed to investigate the benefit of OS-aware microarchitecture on these predictors.

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Appendix A1. Breakdown of Dynamic OS Branches based on Service







A2. Illustration of Weakly Biased Branches in OS











Bi-





