

ABSTRACT

Title of thesis: MEMORY ALLOCATION FOR EMBEDDED
 SYSTEMS WITH A COMPILE-TIME-
 UNKNOWN SCRATCH-PAD SIZE

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This paper presents the first memory allocation scheme for embedded systems having a scratch-pad memory(SPM) whose size is unknown at compile-time. All existing memory allocation schemes for SPM require the SPM size to be known at compile-time; therefore tie the resulting executable to that size of SPM and not portable to other platforms having different SPM sizes. As size-portable code is valuable in systems supporting downloaded codes, our work presents a compiler method whose resulting executable is portable across SPMs of any size.

Our technique is to employ a customized installer software, which decides the SPM allocation just before the program's first run, then modifies the program executable accordingly to implement the decided SPM allocation. Results show that our benchmarks average a 41% speedup versus an all-DRAM allocation, with overheads of 1.5% in code-size, 2% in run-time, and 3% in compile-time for our benchmarks. Meanwhile, an unrealistic upper-bound is approximated only slightly faster at 45% better than all-DRAM.

MEMORY ALLOCATION FOR EMBEDDED SYSTEMS WITH A
COMPILE-TIME-UNKNOWN SCRATCH-PAD SIZE

by

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List of Abbreviations

CPU	Central Processing Unit
DRAM	Dynamic Random Access Memory
FPB	Frequency Per Byte
ILP	Integer Linear Programming
ISA	Instruction Set Architecture
LFPB	Latency Frequency Per Byte
PC	Program Counter
ROM	Read Only Memory
SPM	Scratch-Pad Memory
SRAM	Static Random Access Memory

Chapter 1

Introduction

In both desktop and embedded systems, SRAM and DRAM are the two most common writable memories used for program data and code. SRAM is fast but expensive while DRAM is slower (by a factor of 10 to 100) but less expensive (by a factor of 20 or more). To combine their advantages, a large amount of DRAM is often used to provide low-cost capacity, along with a small-size SRAM to reduce run-time by storing frequently used data. The proper use of SRAM in embedded systems can result in significant run-time and energy gains compared to using DRAM only. This gain is likely to increase in the future since the speed of SRAM is increasing by an average of 50% per year at a similar rate to processor speeds [1] versus only 7% a year for DRAM [2].

There are two common ways of adding SRAM: either as a hardware-cache or a Scratch Pad Memory (SPM). In desktop systems, caches are the most popular approach. A cache dynamically stores a subset of the frequently used data or instructions in SRAM. Caches have been a great success for desktops because they are flexible enough to be used by any executable; a trend that is likely to continue in the future. On the other hand, in most embedded systems where code is often tied to particular implementations, the overheads of cache are less justifiable. Cache incurs a significant penalty in area cost, energy, hit latency and real-time guarantees. A

detailed study [3] compares the tradeoffs of a cache as compared to a SPM. Their results show that a SPM has 34% smaller area and 40% lower power consumption than a cache memory of the same capacity. Further, the run-time with a SPM using a simple static knapsack-based [3] allocation algorithm was measured to be 18% better as compared to a cache. Thus, defying conventional wisdom, they found absolutely no advantage to using a cache, even in high-end embedded systems where performance is important. Given the power, cost, performance and real-time advantages of SPM, it is not surprising that SPM is the most common form of SRAM in embedded CPUs today.

Examples of embedded processor families having SPM include low-end chips such as the Motorola MPC500, Analog Devices ADSP-21XX, Philips LPC2290; mid-grade chips like the Analog Devices ADSP-21160m, Atmel AT91-C140, ARM 968E-S, Hitachi M32R-32192, Infineon XC166 and high-end chips such as Analog Devices ADSP-TS201S, Hitachi SuperH-SH7050, and Motorola Dragonball; there are many others. Recent trends indicate that the dominance of SPM will likely consolidate further in the future [4, 3], for regular as well as network processors.

A great variety of allocation schemes for SPM have been proposed in the last decade [5, 3, 6, 7, 8, 9]. They can be categorized into static methods, where the selection of objects in SPM does not change at run time; and dynamic methods, where the selection of memory objects in SPM can change during runtime to fit the program's dynamic behavior (although the changes may be decided at compile-time). All of these existing techniques have the same drawback of requiring the SPM size to be known at compile-time. This is because they establish their solutions by

reasoning about which data variables and code blocks will fit in SPM at compile-time, which inherently and unavoidably requires knowledge of the SPM size. This has not been a problem for traditional embedded systems where the code is typically fixed at the time of manufacture, usually by burning it onto ROM, and is not changed thereafter.

There is, however, an increasing class of embedded systems where SPM-size-portable code is desirable. These are systems where the code is updated after deployment through downloads or portable media, and there is a need for the same executable to run on different implementations of the same ISA. Such a situation is common in networked embedded infrastructure where the amount of SPM is increased every year, due to technology evolution, as expected by Moore's law. Code-updates that fix bugs, update security features or enhance functionality are common. Consequently, the downloaded code may not know the SPM size of the processor, and thus is unable to use the SPM properly. This leaves the designers with no choice but to use an all-DRAM allocation or a processor with caching mechanism, in which the well-known advantages of SPMs are lost.

To make code portable across platforms with varying SPM size, one theoretical approach is to recompile the source code separately using all the SPM sizes that exist in practice; download all the resulting executables to each embedded node; discover the node's SPM size at run-time; and finally discard all the executables for SPM sizes other than the one actually present. However this approach wastes network bandwidth, energy and storage; complicates the update system; and cannot handle un-anticipated sizes used at a future time. Further, un-anticipated sizes will

require a re-compile from source, which may not be available for intellectual property reasons, or because the software vendor is no longer in business since it could be years after initial deployment. This approach is our speculation – we have not found it being suggested in the literature, which is not surprising considering its drawbacks listed above. It would be vastly preferable to have a single executable that could run on a system with any SPM size.

Another alternative is to choose the smallest common SPM size for a particular platform. We can then make the SPM allocation decision based on this size, and the resulting executable can be run accurately on this family of system. This alternative approach sounds promising since it only requires the production of a single executable to be used for multiple systems. However, one obvious drawback is that this scheme will deliver poor performance for systems that have significant differences in SPM sizes. If an engineer were to use the binary optimized for 4KB SPM size on a system with 16KB SPM, then 12KB of the SPM would be idle and wasted. Moreover this alternative performs even worse if the base of the SPM address range is different in the two systems, which is often the case in practice.

1.1 Challenges

Without knowing the size of the SPM at compile-time, it is impossible to know which variables or code blocks should be placed in SPM at compile-time. This makes it hard to generate code to access variables. To illustrate, consider a variable A of size 4000 bytes. If the available size of SPM is less than 4000 bytes, this variable A

must remain allocated in DRAM at some address, say, 0x8000. Otherwise, A may be allocated to SPM to achieve speedup at some address, say, 0x100. Without the knowledge of the SPM size, the address of A could be either 0x8000 or 0x100, and thus remains unknowable at compile-time. Hence, it becomes difficult to generate an instruction at compile-time that accesses this variable since that requires knowledge of its assigned address. A level of indirection can be introduced in software for each memory access to discover its location first, but that would incur an unacceptably high overhead.

1.2 Method Features and Outline

Our method is able to allocate global variables, stack variables, and code regions, for both application code and library code, to SPMs of compile-time unknown size. Like almost all SPM allocation methods, heap data is not allocated to SPM by our method. Our method is implemented by both modifying the compiler and introducing a customized installer.

At compile-time, our method analyzes the program to identify all the locations in the code that contain *unknown variable addresses*. These are all occurrences of the addressing constants of variables in the code representation; they are unknown at compile-time since the allocations of variables to SPM or slow memory are decided only later at install-time. Unknown addressing constants are the stack pointer offsets in instruction sequences that access stack variables, and all locations that store the addresses of global variables. The compiler then stores the addresses of

these addressing constants as part of the executable, along with the profile-collected Latency-Frequency-Per-Byte (LFPB) of each variable. To avoid an excessive code-size increase from these lists of locations, they are maintained as *in-place linked lists*. In-place linked lists are a space-efficient way of storing the list of unknown addressing constants in the code. Rather than storing an external list of addresses of these addressing constants, in-place linked lists store the linked lists in bit fields of instructions that will be changed anyway at install-time, greatly reducing the code-size overhead.

The next step of our method is when the program is installed on a particular embedded device. Our customized installer discovers the SPM size, computes an allocation for this size, and then modifies the executable just before installing it to implement the decided allocation. The SPM size can be found either by making an OS call if available on that ISA, or by probing addresses in memory with a binary search pattern to observe the latency difference for finding the SPM's address ranges. Next, the SPM allocation is computed, giving preference to objects with higher LFPB, while also considering the differing gains of placing code and data in SPM because of the differing latencies of Flash and DRAM, respectively. At its end, the installer implements the allocation by traversing the locations in the code segment of the executable that have unknown variable addresses and replacing them with the SPM stack offsets and global addresses for the install-time-decided allocation. In the case of allocating code blocks to SPM, appropriate jump instructions are inserted before and after the code blocks in DRAM and SPM to maintain program correctness. The resulting executable is thus tailored for the SPM size on the

target device. The executable can be re-run indefinitely, as is common in embedded systems, with no further overhead.

1.3 Organization of Thesis

The rest of the paper is organized as follows. Chapter 2 overviews related works. Chapter 3 describes scenarios where our method is useful. Chapter 4 discusses the method in [5] whose allocation we aim to reproduce, but without knowing the SPM size. Chapter 5 discusses our method in detail stage-by-stage. Chapter 6 discusses the allocation policy used in the customized installer. Chapter 7 describes how program code is allocated into SPM. Chapter 8 discusses the real-world issues of handling library functions and separate compilation. Chapter 9 presents the experimental environment, benchmarks properties, and our method's results. Chapter 10 compares our method with systems having caches in various architectures. Chapter 11 concludes.

Chapter 2

Related Work

Among existing work, static methods to allocate data to SPM include [10, 11, 3, 6, 12, 13, 5]. Static methods are those whose SPM allocation does not change at run-time. Some of these methods [10, 3, 6] are restricted to allocating only global variables to SPM, while others [11, 12, 13, 5] can allocate both global and stack variables to SPM. These static allocation methods either use greedy strategies to find an efficient solution, or model the problem as a knapsack problem or an integer-linear programming problem (ILP) to find an optimal solution.

Some static allocation methods [14, 15] aim to allocate code to SPM rather than data. In the method presented by Verma et. al. in [15], the SPM is used for storing program code; a generic cache-aware SPM allocation algorithm is proposed for energy saving. The SPM in this work is similar to a preloaded loop cache, but with an improvement of energy saving. Other static methods [16, 17] can allocate both code and data to SPM. The goal of the work in [18] is yet another: to map the data in the scratch-pad among its different banks in multi-banked scratch-pads; and then to turn off (or send to a lower energy state) the banks that are not being actively accessed.

Another approach to SPM allocation are dynamic methods; in such methods the contents of the SPM can be changed during run-time [19, 7, 20, 9, 21, 22, 8,

23]. The method in [19] can place global and stack arrays accessed through affine functions of enclosing loop induction variables in SPM. No other variables are placed in SPM; further the optimization for each loop is local in that it does not consider other code in the program. The method in [7] allocates global and stack data to SPM dynamically, with explicit compiler-inserted copying code that copies data between slow memory and SPM when profitable. All dynamic data movement decisions are made at compile-time based on profile information. The method by Verma et. al. in [8] is a dynamic method that allocates code as well as global and stack data. It uses an ILP formulation for deriving an allocation. The work in [20] also allocates code and data, but using a polynomial-time heuristic. A more complete discussion of the above schemes can be found in [20]. The method in [9] is a dynamic method that is the first SPM allocation method to place a portion of the heap data in the SPM.

All the existing methods discussed above require the compiler to know the size of the SPM. Moreover, the resulting executable is meant only for processors with that size of SPM. Our method is the first to yield an executable that makes no assumptions about SPM size and thus is portable to any size. In future work, our method could be made dynamic; chapter 11 discusses this possibility.

Chapter 3

Scenarios where our proposed method is useful

Our method is useful in the situations when the application code is not burned into ROM at the time of manufacture, but is instead downloaded later; and moreover, when due to technological evolution, the code may be required to run on multiple processor implementations of the same ISA having differing amounts of SPM.

One application where this often occurs is in distributed networks such as a network of ATM machines at financial institutions. Such machines may be deployed in different years and therefore have different sizes of SPM. Code-updates are usually issued to these ATM machines over the network to update their functionality, fix bugs, or install new security features. Currently, such updated code cannot use the SPM since it does not know the SPM's size. Our method makes it possible for such code to run on any ATM machine with any SPM size.

Another application where our technology may be useful is in sensor networks. Examples of such networks are the sensors that detect traffic conditions on roads or the ones that monitor environmental conditions over various points in a terrain. In these long-lived sensor networks, nodes may be added over a period of several years. At the pace of technology evolution today, where a new processor implementation is released every few months, this may represent several generations of processors with

increasing sizes of SPM that are present simultaneously in the same network. Our method will allow code from a single remote code update to use the SPM regardless of its size. Such code updates are common in sensor networks.

A third example is downloadable programs for personal digital assistants (PDAs), mobile phones and other consumer electronics. These applications may be downloaded over a network or from portable media such as flash memory sticks. These programs are designed and provided independently from the configurations of SRAM sizes on the consumer products. Therefore, to efficiently utilize the SPM for such downloadable software, a memory allocation scheme for unknown size SPMs is much needed. There exists a variety of these downloadable programs on the market used for different purposes such as entertainment, education, business, health and fitness, and hobbies. Real-world examples include games such as Pocket DVD Studio [24], FreeCell, and Pack of Solitaires [25]; document managers such as PhatNotes [26] and its manual [27], PlanMaker [28] and its manual [29], and e-book readers; and other tools such as Pocket Slideshow [30] and Pocket Quicken [31] and its manual [32]. In all these applications our technology allows these codes to take full advantages of the SPM for the first time.

Furthermore, we expect that our technology may eventually even allow desktop systems to use SPM efficiently. One of the primary reasons that caches are popular in desktops is that they deliver good performance for any executable, without requiring it to be customized for any particular cache size. This is in contrast to SPMs, which so far have required customization to a particular SPM size. By freeing programs of this restriction, SPMs can overcome one hurdle to their use in desktops. However,

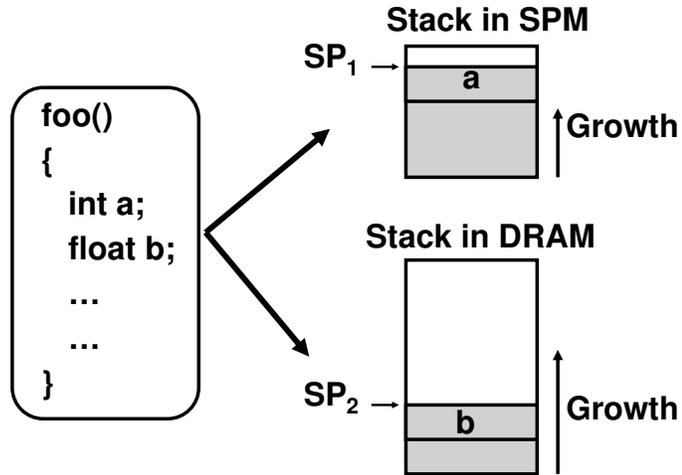


Figure 3.1: Example of stack split into two separate memory units. Variables a and b are placed on SRAM and DRAM respectively. A call to $foo()$ requires the stack pointers in both memories to be incremented.

there are still other hurdles to have SPMs become the norm in desktop systems, including that heap data, which our method does not handle, is more common in desktops than in embedded systems. In addition, the inherent advantages of SPM over cache are less important in desktop systems. For this reason we do not consider desktops further in this paper.

Chapter 4

Background

The allocation strategy used by the installer in our method aims to produce an allocation that is as similar as possible to the optimal static allocation method presented by Avissar et. al. in [5]. That paper only places global and stack variables in SPM; we extend that method to place code in SPM as well. Since the ability to establish an SPM allocation without knowing its size, instead of the allocation decision itself, is the central contribution of our method, we decided to build our allocation decision upon [5] since it is an optimal static allocation scheme. This chapter outlines that method to better understand the aims of our allocation policy.

The allocation in [5] is as follows. In effect, for global variables, the ones with highest Frequency-Per-Bytes (FPB) are placed in SPM. However, this is not easy to do for stack variables, since the stack is a sequentially growing abstraction addressed by a single stack pointer. To allow stack variables to be allocated to different memories (SPM vs DRAM), a *distributed stack* is used. Here the stack is partitioned into two stacks for the same application: one for SPM and the other for DRAM. Each stack frame is partitioned, and two stack pointers are maintained, one pointing to the top of the stack in each memory. A distributed stack example is shown in figure 1. The allocator places the frequently used stack variables in the SPM stack, and the rest are in the DRAM stack. In this way only frequently-used

stack variables (such as variable a in figure 1) appear in SPM.

The method in [5] formulates the problem of searching the space of possible allocations with an objective function and a set of constraints. The objective function to be minimized is the expected run-time with the allocation, expressed in terms of the proposed allocation and the profile-discovered FPBs of the variables. The constraints are that for each path through the call graph of the program, the size of the SPM stack fits within the SPM's size. This constraint automatically takes advantage of the limited life-time of stack variables: if $main()$ calls $f1()$ and $f2()$, then the variables in $f1()$ and $f2()$ share the same space in SPM, and the constraint correctly estimates the stack height in each path. As we shall see later, our method also takes advantage of the limited life-time of stack variables.

This search problem may be solved in two ways: using a greedy search and a provably optimal search based on Integer-Linear Programming (ILP). Our greedy approach chooses variables for SPM in decreasing order of their frequency-per-byte until the SPM is full. The ILP solver is the one in [5]. Results show that the ILP solver delivers only 1-2% better application run-time¹. Therefore the greedy solver is also near-optimal. For this reason and since the greedy solver is much more practical in real compilers, in our evaluation we use the greedy solver for both the

¹This is not the same as the result in [5] which finds that the run-time with greedy is 11.8% worse than ILP. The reason for the discrepancy is that the greedy method in [5] is less sophisticated than our greedy method. Their greedy method is a profile-independent heuristic that considers variables for SPM in increasing order of their size. However, for fairness, we use the better greedy formulation in this paper for [5] as well in our results chapter.

method in [5] and in the off-line component of our method, although either could be used.

4.1 Challenges in adapting a compile-time allocator to install-time

The first thought that we might have in devising an install-time allocator for SPM is to take an existing compile-time allocator and split into two parts: the first part which does size-independent tasks such as profiling, and a second part that computes the allocation at install-time using the same approach as an existing compile-time allocator. However this approach is not possible or desirable without solving at least three challenges, listed below. *Solving these challenges constitutes the contribution of the paper.* **First**, even in this approach we need a method to implement an SPM allocation by only changing the binary. Changing the allocation of a variable in the binary involves understanding and modifying each of the many addressing modes of variables, which is an important contribution of this paper. **Second**, using a simple minded split as above, most of the tasks other than profiling are size-dependent and must be done at install-time, including computing the allocation, the limited-lifetime stack bonus set, the memory layout in SPM, and the needed number and contents of literal tables in SPM. This will consume precious run-time and energy on the embedded device during deployment. Our method avoids these overheads by pre-computing all these quantities at compile-time for *each possible SPM size* (see cut-off points in chapter 6) and storing the resulting information in a customized highly compact form. Our **third** contribution

is the representation of all the accesses of a variable using in-place linked lists. If the list of accesses whose offsets need to be changed were stored in the binary in a naive way as an external list of addresses, the code size overhead would be large (at least equal to the percentage of memory instructions in the code, which is often 25% or more.) Our in-place representation reduces this overhead to 1-2%.

Chapter 5

Method

Although our resulting install-time allocation, including the use of a distributed stack, will be similar to that in Avissar et. al. [5], our mechanisms at compile-time and install-time will be significantly different and more complicated because of not knowing the SPM's size at compile-time. Our method introduces modifications to the profiling, compilation, linking and installing stages of code development to consider both code and data for SPM allocation. Our method can be understood as two main parts: first consists of the profiling, compilation, and linking stages which happen before deploying time, and the second part as the installing stage happening after deployment. Since the SPM size is not known before deployment, additional compiler techniques to the method in [5] are introduced to reduce the overheads, and handle new problems occurring due to the lack of SPM size. The part after deployment of our method is also very much different from the part where variables are assigned SPM addresses in the method in [5]. This is because the SPM size in our scheme is not known until the install-time, when limited information about the program are available; this makes the variable address assignment more complicated.

In this chapter we consider the allocation in SPM of *data variables only*. The allocation of code in SPM will be discussed in chapter 7. The tasks at each stage

are described below. Examples are from the 32-bit portion of the ARM instruction set (not including the 16-bit THUMB instructions.)

5.1 Profiling Stage

The application is run multiple times with different inputs to collect the number of accesses to each variable for each input; and an average is taken. Input data sets should represent typical real-world inputs. If the application has more than one typical behavior (for example, running only one part of the code for one kind of input, and running another part of the code for another kind of input) then at least one typical data set should be selected for each kind of input. The average number of times each variable is accessed across the different data sets is then computed. Next, this frequency of each variable is divided by its size in bytes to yield its FPB.

With this profiling information, the profiling stage also prepares a list of variables in decreasing order of their LFPB products. The LFPB for a code or data object is obtained by multiplying its FPB by the excess latency of the slow memory it resides in, compared to the latency of SPM ($Latency_{slow-memory} - Latency_{SPM}$). The slow memory for code objects is typically Flash memory, and for data objects is DRAM. This LFPB-sorted list is stored in the output executable for use by the installer in deciding an allocation. The installer will later consider variables for SPM allocation in the decreasing order of their LFPB. The reason this makes sense is that the LFPB value is roughly equal to the gain in cycles of placing that code or data object in SPM.

Address:	Assembly instruction:
83b4:	ldr r1, [fp, #-44] // Load stack var.
83b8:	...
83bc:	...
83c0:	str r4, [fp, #-44] // Store stack var.
83c4:	...
83c8:	mov r3, #-4160 // Addr computation 1
83cc:	add r3, fp, r3 // Addr computation 2
83d0:	ldr r0, [r3, #0] // Load stack var.

Figure 5.1: Example accesses to stack variables with instructions at 0x83b4 and 0x83c0 access one variable and 0x83c8 - 83d0 access another.

5.2 Compiling Stage

Since the SPM size is unknown, the allocation is not fixed at compile-time; instead it is done at install-time. Various types of pre-processing are done in the compiler to reduce the customized installer overhead. These are described as the follows.

As the first step, the compiler analyzes the program to identify all the code locations that contain variable addresses which are unknown due to not knowing the SPM's size at compile-time. These locations are identified by their actual addresses in the executable file.

To see how these locations are identified, let us first consider how stack accesses are handled. For the ARM architecture, on which we performed our experiments, the locations in the executable file that affect the stack offset assignments are the load and store instructions that access the stack variables, and all the arithmetic instructions that calculate their offsets. In the usual case, when the stack offset value is small enough to fit into the immediate field of the load/store instruction, these

Address:	Assembly instruction:	
821c:	<main>:	
...		
8234:	ldr r0,[pc,#16]	<i>// Load addr from literal table</i>
8238:	ldr r1,[r0]	<i>// Load global variable</i>
...		
8244:	b linkreg	<i>// Procedure return</i>
8248:	address of global variable 1	→ Literal table
824c:	address of global variable 2	

Figure 5.2: Example access to global variable with **ldr** instruction at 0x8234 accesses the literal table location at 0x8348, while the instruction at 0x8238 actually accesses the global.

load and store instructions are the only ones that affect the stack offset assignments. The first **ldr** and the subsequent **str** instructions in Figure 5.1 illustrate two accesses of this type, where the offset value of -44 from the frame pointer (*fp*) to the accessed stack variable fits in the 12-bit immediate field of the load/store instructions in ARM.

In some rare cases, when a variable's offset from the frame pointer is larger than the value that can fit in the immediate field of load/store instruction, additional arithmetic instructions are needed to calculate the correct offset. Such cases arise for procedures with frame sizes that are larger than the span of the immediate field of the load/store instructions. In ARM, this translates to stack offsets larger than $2^{12} = 4096$ bytes. In these rare cases, the offset is first moved to a register and then added to the frame pointer. An example is seen in the three-instruction sequence (**mov**, **add**, **ldr**) at the bottom of figure 5.1. Since the **mov** instruction in ARM can load a constant that is larger than 4096 bytes to a destination register, the **mov**

and **add** together are able to calculate the correct address of the stack variable for the load instruction. Here, only the **mov** instruction needs to be added to the linked list of locations with unknown addresses that we maintain, since only its field actually contains the stack variable's offset and needs to be changed by the customized installer.

After identifying the locations in the executable that need to be modified by the customized installer, the compiler creates a linked-list of such locations for each variable for use in the linking stage. This compiler linked-list is used later to establish the actual in-place linked-list at link-time, when the exact displacements of the to-be-modified locations are known.

For ARM global variables, the addresses are stored in the *literal tables*. These are tables stored as part of the code that store the full 32-bit addresses of global variables. Thereafter global variables are accessed by the code in two stages: first, the address of the global is loaded into a register by doing a PC-relative load from the literal table; and second, a load/store instruction that uses the register for its address accesses the global from memory. In ARM, literal tables reside just after each procedure's code in the executable file. An example of a literal table is presented in the figure 5.2. In some rare situations, the literal tables can also appear in the middle of the code segment of a function with a branch instruction jumping around it since the literal table is not meant to be executed. This situation occurs only when the code length of a procedure is larger than the range of the load immediates used in the code to access the literal tables.

In our method, each global-variable address in the literal tables needs to be

changed at install-time depending on whether that global is allocated to SPM or DRAM. Thus literal table locations are added to the linked lists of code addresses whose contents are unknown at compile-time; one linked list per global variable. Like the linked lists for stack variables, these will be traversed later by our installer to fill in install-time dependent addresses in code.

The compiler also analyzes the limited life-times of the stack variables to determine the additional sets of variables that can be allocated into SPM for better memory utilization. Details of the allocation policy and life-time analysis are presented in chapter 6. The final step in the compiling stage is to generate the customized installer code, and then either insert or broadcast it along with the executable. More details about the installer is presented below at the installing stage.

5.3 Linking Stage

At the end of the linking stage, to avoid significant code-size overhead, we store all the compiler-generated linked-lists of locations with unknown variable addresses in-place in those locations in the executable. This is possible since these locations will be overwritten with the correct immediate values only at the install-time, and until then, they can be used to store their displacements, expressed in words, to the next element in the list. The exact calculation of one location's displacement to the next is possible since at this stage the linker knows precisely the position of each location with an unknown address in the executable. The addresses of the first locations in the linked-lists are also stored elsewhere in a table in the executable to

Address:	Assembly instruction:	
83b4:	ldr r1, [fp, #-44]	↑ 12
83b8:	...	
83bc:	...	
83c0:	str r4, [fp, #-44]	↓ 8
83c4:	...	
83c8:	ldr r3, [fp, #-44]	

Figure 5.3: Before link-time.

Address:	Assembly instruction:	
83b4:	ldr r1, [fp, #12]	↘
83b8:	...	
83bc:	...	↘
83c0:	str r4, [fp, #8]	
83c4:	...	↘
83c8:	ldr r3, [fp, #0]	

// End of linked list

Figure 5.4: After link-time. In-place linked list is shown.

be used at install-time as the starting addresses of the linked-lists. Storing the head of the linked-lists in this way is necessary to traverse each list at install-time.

An example of the code conversion in the linker is shown in figures 5.3 and 5.4. Figure 5.3 shows the output code from the compiler stage with the stack offsets assuming an all-DRAM allocation. Figure 5.4 shows the same code after the linker converts the separately stored linked-lists to in-place linked-lists in the code. Each instruction now stores the displacement to the next address. The stack offset in DRAM (-44 in the example) is overwritten; this is no loss since our method changes the stack offsets of variables at install-time anyway.

The in-place linked list representation is efficient since in most cases the bit-width of the immediate fields is sufficient to store the displacement of two consecutive accesses, which are usually very close together. For example, in ARM architecture,

the offsets of stack variables are often carried in the 12-bit immediate fields of a **ldr**. This yields a displacement up to 4096 bytes, which is adequate for most offsets. In the rare case when the displacement to next access to the same variable from the current instruction is greater than the value that can fit in a 12-bit immediate, a new linked-list for the same variable is started at that point. This causes no difficulty since more than one linked-list of locations with unknown addresses can be efficiently maintained for the same variable, each with a different starting address.

For the case of global variables, the addresses are stored in the literal table whose entries are 32-bits wide. This is wide enough in a 32-bit address space to store all possible displacements. So a single linked list is always adequate for each global variable.

Application to other instruction sets Although our method above is illustrated with the example of the ARM ISA, it is applicable to most embedded ISAs. To apply our method to another ISA, all possible memory addressing modes for global and stack variables must be identified. Next, based on these modes, the locations in the program code that store the immediates for stack offsets and global addresses must be found and stored in the linked lists. The exact widths of the immediate fields may differ from ARM, leading to more or fewer linked lists than in ARM. However because accesses to the same variable are often close together in the code, the number of linked lists is expected to remain small.

5.4 Customized Installer

The customized installer is implemented in a set of compiler-inserted codes that are executed just after the program is loaded in memory. A part of the application executable, the code is invoked by a separate installer routine or by the application itself using a procedure call at the start of *main()*. In the later case, care must be taken that the procedure is executed only once before the first run of the program, and not before subsequent runs; these can be differentiated by a persistent *is-first-time* boolean variable in the installer routine. In this way, the overhead of the customized installer is encountered only once even if the program is re-run indefinitely, as is common in embedded systems.

The installer routines perform the following three tasks. First, they discover the size of SPM present on the device, either by making an OS system call (such calls are available on most ISAs), or by probing addresses in memory using a binary search and observing the latency to find the range of the SPM addresses. Second, the installer routines compute a suitable allocation to the SPM using its just-discovered size and the LFPB-sorted list of variables. The details of the allocation are described in chapter 6. Third, the installer implements the allocation by traversing the locations in the code that have unknown addresses and replacing them with the stack offsets and global addresses for the install-time-decided allocation. The resulting executable is now tailored for the SPM size on that device, and can be executed without any further overhead.

Chapter 6

Allocation policy in customized installer

The SPM-DRAM allocation is decided by the customized installer using the run-time discovered SPM size, the LFPB-sorted list of variables and information about the life-times of stack variables that the compiler provides. The greedy profile-driven cost allocation in the installer is as follows. The installer traverses the list of all global and stacks variables stored by the compiler in a decreasing order of their LFPBs, placing variables into SPM until their cumulative size exceeds the SPM size. This point in the list is called its *cut-off* point.

We observe, however, that the SPM may not actually be full on each call graph path at the cut-off point because of the limited life-times of stack variables. For example, if *main()* calls *f1()* and *f2()*, then the variables in *f1()* and *f2()* can share the same space in SPM since they have non-overlapping life-times, and simply cumulating their sizes over-estimates the maximum height of the SPM stack. Thus the greedy allocation under-utilizes the SPM.

Our method uses this opportunity to allocate an additional set of stack variables into SPM to utilize the remaining SPM space. We call this the *limited-lifetime-bonus-set* of variables to place in SPM. To avoid an expensive search at install-time, this set is computed off-line by the compiler and stored in the executable for each possible cut-off point in the LFPB-sorted list. Since the greedy search can cut-off at

Define:

A: is the list of all global and stack variables in decreasing LFPB order

Greedy_Set: is the set of variables allocated greedily to SPM

Limited_Lifetime_Bonus_Set: is the limited-lifetime-bonus-set of variables SPM

GREEDY_SIZE: is the cumulated size of greedily allocated variables to SPM at each cutoff point

BONUS_SIZE: is the cumulated size of variables in limited-lifetime-bonus-set

MAX_HEIGHT_SPM_STACK: the maximum height of the SPM stack during lifetime of current variable

```
void    Find_allocation(A) { /* Run at compile-time */
1.  for (i = beginning to end of LFPB list A) {
2.    GREEDY_SIZE ← 0; BONUS_SIZE ← 0;
3.    Greedy_Set ← NULL; Limited_Lifetime_Bonus_Set ← NULL;
4.    for (j = 0 to i) {
5.      GREEDY_SIZE ← GREEDY_SIZE + size of A[j]; /* jth variable in LFPB list */
6.      Add A[j] to the Greedy_Set;
7.    }
8.    Call Find_limited_lifetime_bonus_set(i, GREEDY_SIZE);
9.    Save Limited_Lifetime_Bonus_set for cut-off at variable A[i] in executable;
10. }
11. return;}

void    Find_limited_lifetime_bonus_set(cut-off-point, GREEDY_SIZE) {
12. for (k = cut-off-point to end of LFPB list A) {
13.   Add stack variables in Greedy_Set ∪ Limited_Lifetime_Bonus_Set to SPM stack;
14.   if (A[k] is a stack variable) {
15.     Find MAX_HEIGHT_SPM_STACK among all call-graph paths from main() to leaf
           procedures that go through procedure containing A[k];
16.   } else { /* A[k] is global variable */
17.     Find MAX_HEIGHT_SPM_STACK among all call-graph paths from main() to leaf procedures;
18.   }
19.   ACTUAL_SPM_FOOTPRINT ← (Size of globals in Greedy_Set ∪ Limited_Lifetime_Bonus_Set)
           + MAX_HEIGHT_SPM_STACK;
20.   if (GREEDY_SIZE - ACTUAL_SPM_FOOTPRINT ≥ size of A[k]) { /* L.H.S. is
           over-estimate amount */
21.     add A[k] into the Limited_Lifetime_Bonus_Set;
22.     BONUS_SIZE ← BONUS_SIZE + size of A[k];
23.   }
24. }
25. return;}

```

Figure 6.1: Compiler pre-processing pseudo-code that finds Limited_Lifetime_Bonus_Set at each cut-off

any variable, a bonus set must be pre-computed for each variable in the program. Once this list is available to our customized installer at the start of run-time or as we call the install-time, it implements its allocations in the same way as for other variables.

The compiler algorithm to compute the *limited-lifetime-bonus-set* of variables at each cut-off point in the LFPB list is presented in figure 6.1. Lines 1-11 show the main loop traversing the LFPB-sorted list in decreasing order of LFPB. Lines 4-7 find the greedy allocation for the cut-off point at variable i . Line 8 makes the call to a routine to find the *limited-lifetime-bonus-set* at this cut-off point; the routine is in lines 12-25. The un-utilized space in SPM is computed as the difference of the greedily-estimated size and the actual memory footprint (line 20), which may be lower because of limited life-times. Additional variables are then found to fill this space in decreasing order of LFPB among the remaining variables. This search of bonus variables considers the stack allocation only along paths through the current variable's procedure if it is a stack variable (line 15); therefore it does not itself over-estimate the memory footprint.

Figure 6.2 illustrates our allocation algorithm with an example. Figure 6.2(a) shows the call graph of an application. It also shows the program's global variables and the local variables for each procedure in a list next to each node. From this call graph, the stack of the application are derived in figure 6.2(b)¹. In this figure, each

¹For simplicity, only the program variables in the stack are shown; not the overhead words such as return address and old frame pointer, since the overhead words do not contend for SPM allocation.

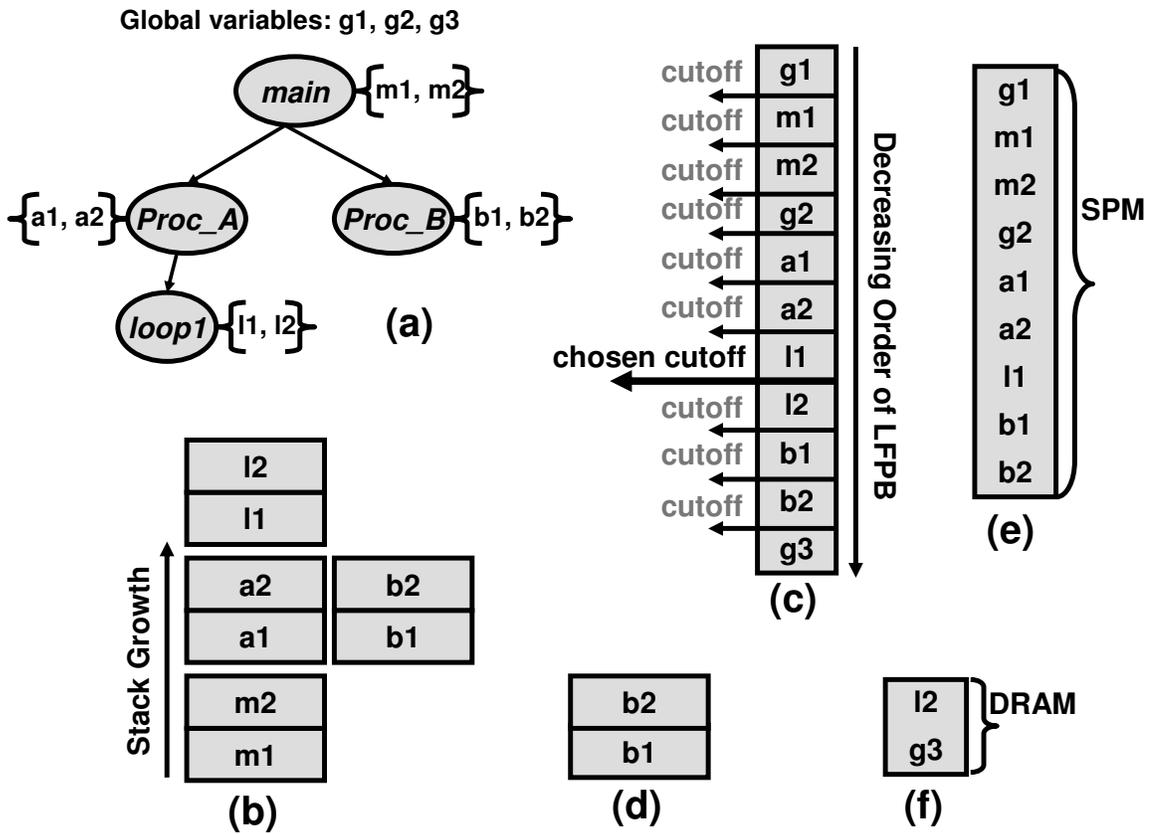


Figure 6.2: (a) program call graph with global and local variables lists (b) the stack of this application (c) sorted list of all program variables (d) bonus set (e) SPM variable list (f) DRAM variable list

stack frame, consisting the local variables of each procedure, is added and removed from the stack as it grows and sinks at runtime. This figure also shows that the local variables (b1,b2) have disjoint lifetimes both with local variables (a1,a2) and with (l1,l2). Figure 6.2(c) shows the sorted list of all variables of the application in the decreasing order of their LFPB. There is a cut-off point after each variable. For a particular SPM size assume that the first seven variables fit, and thus the chosen cut-off point at install-time is as shown. The seven variables are the original set of variables allocated to SPM. However, since (b1,b2) have disjoint lifetimes with already-allocated variables a1,a2 and l1, the former are contenders for the limited-lifetime bonus. Suppose $size(b1) + size(b2) \leq size(a1) + size(a2) + size(l1)$, then both b1 and b2 are in the bonus set as shown in figure 6.2(d). With this choice the set of SPM-allocated variables increases to the nine variables in figure 6.2(e). The remaining variables in figure 6.2(f) are placed in DRAM.

Two factors limit the code-size increase arising from storing the bonus sets at each cut-off. First, the bonus sets are stored in bit-vector representation on the set of variables, and so are extremely compact. Second, in a simple optimization, instead of defining cut-off at each variable, a cut-off is defined at a variable only if the cumulative size of variables from the previous cut-off exceeds `CUT_OFF_THRESHOLD`, a small pre-defined constant currently set to 10 words. This avoids defining a new cut-off for every single scalar variable; instead groups of adjacent scalars with similar LFPBs are considered together for purposes of computing a bonus set. This can reduce the code space increase by up to a factor of 10, with only a small cost in SPM utilization.

Chapter 7

Code Allocation

Our method is extended to allocate program code to SPM as follows. Code is considered for placement in SPM at the granularity of *regions*. The program code is partitioned into regions at compile-time. The allocator decides to place frequently accessed code regions in SPM. Thereafter at install-time, each such code blocks are copied from its current location in slow memory (usually Flash) to SPM at run-time. To preserve the intended control flow, a branch is inserted from the code's original location to its SPM location, and another branch at the end of the code in SPM back to the original code. These extra branches are called *patching* code and are detailed later.

Some criteria for a good choice of regions are (i) the regions should not too big thus allowing a fine-grained consideration of placement of code in SPM; (ii) the regions should not be too small to avoid a very large search problem and excessive patching of code; (iii) the regions should correspond to significant changes in frequency of access, so that regions are not forced to allocate infrequent code in them just to bring their frequent parts in SPM; and (iv) except in nested loops, the regions should contain entire loops in them so that the patching at the start and end of the region is not inside a loop, and therefore has low overhead.

With these considerations, we define a new region to begin at (i) the start of

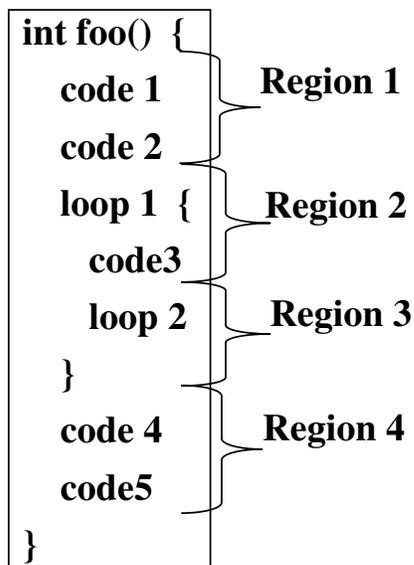


Figure 7.1: Program code is divided into code regions

each procedure; and (ii) just before the start, and at the end of every loop (even inner loops of nested loops). Other choices are possible, but we have found this heuristic choice to work well in practice. An example of how code is partitioned into regions is in Figure 7.1. As the following step, each region’s profiled data such as its size, LFPB, start and end addresses are collected at the profiling stage along with profiled data for program variables.

Since code regions and global variables have the same life-time characteristics, code-region allocation is decided at customized install-time using the same allocation policy as global variables. The greedy profile-driven cost allocation in the customized installer is modified to include code regions as follows. The customized installer traverses the list of all global variables, stacks variables, and code regions stored by the compiler in a decreasing order of their LFPBs, placing variables and transferring code regions into SPM, until the cumulative size of the variables and code allocated

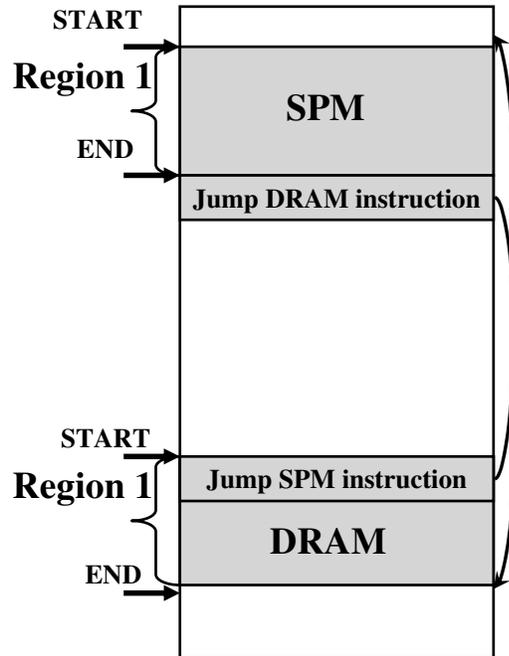


Figure 7.2: Jump instruction is inserted to redirect control flow between SPM and DRAM

so far exceeds the SPM size. At this cut-off point, an additional set of variables and code regions, which are established at compile-time by the limited-lifetime-bonus-set algorithm for both data variables and code regions, are also allocated to SPM. The limited-lifetime-bonus-set algorithm is modified to include code regions, which are treated as additional global variables.

Code-patching is needed at several places to ensure that the code with SPM allocation is functionally correct. Figure 7.2 shows the patching needed. At customized install-time, for each code region that is transferred to SPM, our method inserts a jump instruction at the original DRAM address of the start of this region. The copy of this region in DRAM becomes unused DRAM space¹. Upon reaching

¹We do not attempt to recover this space since it will require patching code even when it is not moved to SPM, unlike in our current scheme. Moreover, since the SPM is usually a small fraction

this install-time inserted instruction, execution will jump to the SPM address this region is assigned to, as intended.

Similarly, patching also inserts an instruction as the last instruction of the SPM-allocated code region, which redirects program flow back to DRAM. The distance from the original DRAM space to the newly allocated SPM space of the region usually fits into the immediate field of the jump instructions. In the ARM architecture, which we use for evaluation, jump instructions have a 24-bit offset which is large enough in most cases. In the rare cases that the offset is too large to fit in the space available in the jump instruction, a longer sequence of instructions is needed for the jump; this sequence first places the offset into a register and then jumps to the contents of the register.

Besides incoming and outgoing paths, side entries and exits in the middle of regions in SPM also need modification to ensure correct control flow. With our definition of regions, side entries are mostly caused by unstructured control flows from “goto” statements, which are rare in applications. Our method does not consider regions which are the target of unstructured control flow for SPM allocation; thus, no further modification is needed for side entries of SPM-allocated regions. However, side exits such as procedure calls from code regions are common. They are patched as follows. For each SPM-allocated code region, the branch offsets of all control transfer instructions that branch to outside of the region they belong to, are adjusted to new corrected offsets. These new branch offsets are calculated by adding the original branch offsets to the distance between DRAM and SPM starting of the DRAM space, the space recovered in DRAM will be insignificant.

addresses of the transferring regions. The returns from these procedure calls do not need any patching since their target address is automatically computed at run-time.

The final step in code patching is needed to modify the load-address instructions of global variables that are accessed in the SPM-allocated regions. In ARM, the load-address instruction of a global variable is a PC-relative load that loads the address of the global variables from the literal table, which is also in the code. Allocating code regions with such load-address instructions into SPM makes the original relative offsets invalid. Besides, for ARM, the relative offsets of the load-address instructions are 12-bit. Thus, it is likely that the distance from the load-address instructions in SPM to the literal tables in DRAM is too large to fit into those 12-bit relative offsets.

To solve this problem of addressing globals from SPM, our method generates a second set of literal tables which reside in SPM. During when code objects are being placed in SPM one-after-another, a literal table is generated at a point in the SPM layout if code about to be placed cannot refer to the previously generated literal table in SPM since it is out of range. This leads to (roughly) one literal table per $2^{12} = 4096$ bytes of code in SPM. These secondary SPM literal tables contain the addresses of only those global variables that refer to it. Afterward, the relative offsets of these load-address instructions are adjusted to the corrected offsets, which are calculated by the distance from the load-address instructions to the SPM literal table. Architectures other than ARM that do not use literal tables do not need this patching step.

The installer must account for the increase in size from patching code and

secondary literal tables. First, the installer adds the size of the jump instruction at the end of each code block to the size needed for allocating that block to SPM. This sum is used in both the calculation of code block's FPB number as well as in satisfying the space constraint. The size of the jump *to* the SPM block does not need to be added since it overwrites the unused space left in slow memory when that block is moved to SPM. Further, patching side exits does not increase code size either. Second, the SPM space required for secondary literal tables is also added. Since each cutoff point specifies an exact set of global variables and code blocks allocated, the size of the literal table for each cut-off point is known at compile-time, and stored with the cut-off point in the code. This size is added to the space required for the cut-off point by the customized installer.

Chapter 8

Real-world issues

8.1 Library Variables

Many downloadable applications make at least some use of functions from system libraries such as *libc*, *newlib*, and *libM* for the C and C++ languages. For this reason, we have extended our method to consider library variables for SPM allocation. This is accomplished by compiling all library functions used from their sources with our optimizing compiler¹. They can still be separately compiled as usual. The linker links the library codes, and thereafter in the linker, library stack and global variables are then treated in the same way as program variables. Profile information for library variables is collected per application, and supplements the static compile-time information gathered. The linked-lists of unknown addresses for library variables is stored with the library object-code for reuse across various applications.

The above method requires that library routines be statically linked with each application that uses them. For dynamically linked shared libraries, which can be

¹Libraries should be re-compiled for best performance. However, if the library sources are not available, then this handling of library variables can be omitted and our method can still be used for application variables. Results in figure 18 show that allocating library variables to SPM improves run-time by only a little for our benchmarks. Over 90% of the gain from SPM allocation is realized because of allocating application variables to SPM.

shared across applications, the same library allocation must be followed for every application that uses that library routine. For this reason, we expect that in a deployed system only a few frequently executed library routines would use static linking; the rest can use dynamic linking if desired with little loss in performance.

8.2 Separate Compilation

Our method does not use any whole-program analysis at compile-time; hence, separate compilation of the different files in an application can be maintained. However, during separate compilation, care must be taken for external and global variable references since the compiler produces a distinct linked-list of locations with unknown addresses for these variables in each file. This results in multiple linked-lists for the same global variable declaration across different object files. These linked lists must be correlated and marked as belonging to the same variable, since otherwise multiple allocations would result for the same variable, violating correctness². This simple step in the linking stage ensures that separate compilation remains possible.

²The multiple lists for a variable also can be merged into a single list if the displacements permit, but this is not essential for correctness.

Chapter 9

Results

This chapter presents our results by comparing the proposed allocation scheme for unknown size SPMs against an all-DRAM-allocation and against Avissar et. al’s method in [5]. We compare our method to the all-DRAM-allocation method since all other existing methods are inapplicable in our target systems where the SPM size is unknown at compile-time; thus, they have to force all data and code allocation to DRAM. We also compare our scheme to [5] to show that our scheme obtains a performance which is close to the un-achievable optimal upper-bound.

9.1 Experimental Environment

Our method is implemented on the GNU tool chain from CodeSourcery [33] that produces code for the ARM v5e embedded processor family [34]. The process of identifying variable accesses and addresses, analysis of variable limited life-times, and customized installer code generation are implemented in the GCC v3.4 cross-compiler. The modifications of the executable file are done in the linker of the same tool chain. An external DRAM with 20-cycle latency, Flash memory with 10-cycle latency, and an internal SRAM (SPM) with 1-cycle latency are simulated. The memory latencies assumed for Flash and DRAM are representative of those in modern embedded systems [35, 36]. Data is placed in DRAM and code in Flash

Application	Source	Description	Data Size (Bytes)	Lines of Code	# of asm instr.
CRC	MIBench	32 BIT ANSI X3.66 CRC checksum	1068	187	504
Dijkstra	MIBench	Shortest path Algorithm	4097	174	501
EdgeDetect	UTDSP	Edge Detection in an image	196848	297	701
FFT	UTDSP	Fast Fourier Transform	16568	189	478
KS	PtrDist	Minimum Spanning Tree for Graphs	27702	408	1327
MMULT	UTDSP	Matrix Multiplication	120204	164	416
Qsort	MIBench	Quick Sort Algorithm	7680000	45	116
PGP	MIBench	Public Key Encryption	64576	24870	55609
Rijndael	MIBench	AES Algorithm	22160	1142	9225
StringSearch	MIBench	A Pratt-Boyer-MooreStringSearch	12820	3037	4433
Susan	MIBench	Method for DigitallyProcessing Images	380212	2120	9886

Table 9.1: Application Characteristics

memory. Code is most commonly placed in Flash memory today when it needs to be downloaded. A set of most frequently used data and code is greedily allocated to SPM by our compiler. The SRAM size is configured to be 20% of the total data size in the program¹. This percentage is varied in additional experiments. The benchmarks' characteristics are shown in table 9.1.

The energy consumed by programs is estimated using the instruction-level power model proposed in [37]. In that model, the overall energy used is estimated as the sum total of energy consumed by each instruction, where the energy for each

¹We could have also chosen a different SRAM size – 20% of total code + data size – when evaluating the methods for both code and data. However, to make comparisons between methods for data only and for both code and data, we had to choose one consistent SRAM size (20% of data size alone) so that the comparison is fair.

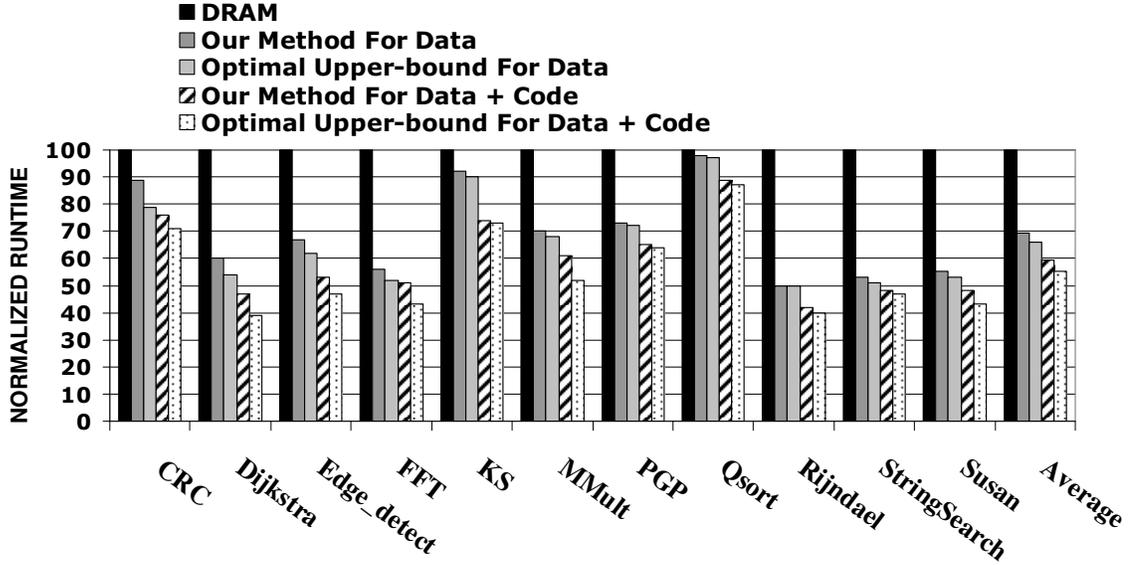


Figure 9.1: Runtime speedup compared to all-DRAM method and Static Optimal Method

instruction type (opcode) is estimated using synthetic workloads composing only of that instruction in an infinite loop and measuring the current drawn by the circuit. Thereafter no current measurements are needed per application; instead the pre-calculated energy per instruction type is used to calculate the total. Experimental results in that paper show that the approach is quite accurate and has a small percentage error in estimating the energy use. The energy numbers for each ARM instruction are extracted from [38]. We modified our simulator to add the energy of each instruction, based on its opcode, to a counter measuring the total energy.

9.2 Runtime Speedup

The run-time for each benchmark is presented in figure 9.1 for five configurations: all-DRAM, our method for data allocation only, optimal upper bound obtained by using [5] for data allocation only, our method enhanced for both code

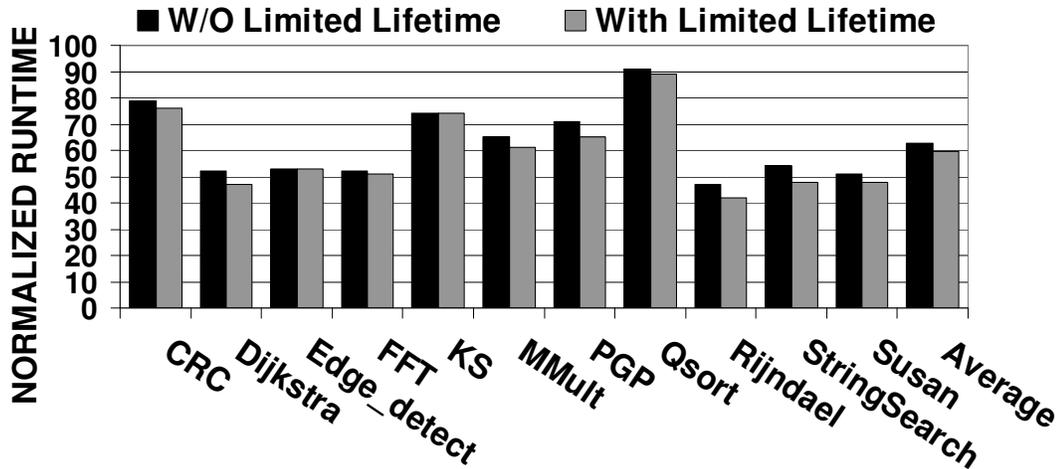


Figure 9.2: Runtime speedup of our method with and without limited lifetime compared to all-DRAM method

and data allocations, and the optimal upper bound obtained by using [5] for both code and data allocations. Averaging across the eleven benchmarks, our full method (the fourth bar) achieves a 41% speedup compared to all-DRAM allocation (the first bar). The provable optimal static allocation method [5], extended for code in addition to data (the fifth bar), achieves a speedup of 45% on the same set of benchmarks. This small difference indicates that we can obtain a performance that is close to that in [5] without requiring the knowledge of the SPM's size at compile-time.

The figure also shows that when only data is considered for allocation to SPM, a smaller run-time gain of 31% is observed versus an upper bound of 34% for the optimal static allocation. This shows that considering code for SPM placement rather than just data yields an additional $41\% - 31\% = 10\%$ improvement in run-time for the same size of SPM.

The performance of the limited-lifetime algorithm is showed in figure 9.2. The difference between the first and second bars gives the improvement using our limited

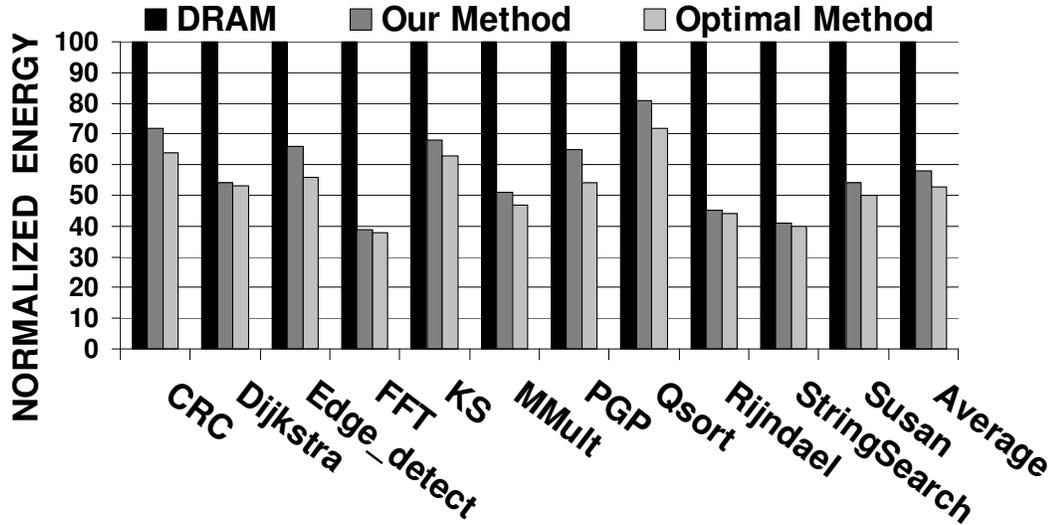


Figure 9.3: Energy consumption compared to all-DRAM method and Static Optimal Method

life-time analysis, as compared to a greedy allocation, for each benchmark. Although the average benefit is small (4% on average), for certain benchmarks (for example, PGP and Rijndael), the benefit is greater. This shows that the limited life-time enhancement is worth doing but not critical.

9.3 Energy Saving

Figure 9.3 compares the energy consumption of our method against an all-DRAM allocation method and against the optimal static allocation method [5] for the benchmarks in table 9.1. As the results of more frequently-accessed objects allocated in SRAM, our method is able to achieve a 42% gain in energy consumption compared to the all-DRAM allocation scheme. The optimal static scheme in [5] gets a slightly better result of 47% gain in energy for when the SPM size is provided at compile-time.

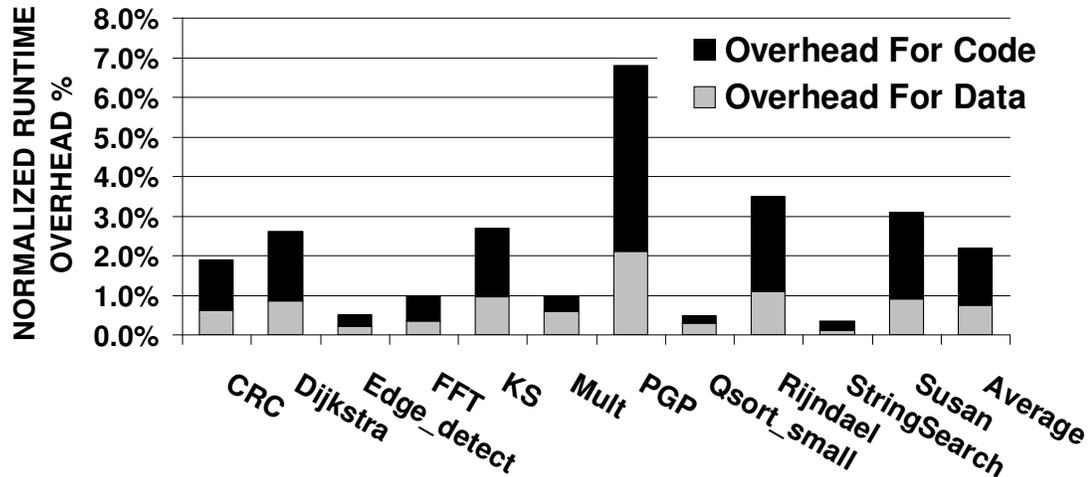


Figure 9.4: Runtime Overhead

9.4 Run Time Overhead

Figure 9.4 shows the increase in the run-time from the customized installer as a percentage of the run-time of one execution of the application. The figure shows that this run-time overhead averages only 2% across the benchmarks. A majority of the overhead is from code allocation including the latency of copying code from DRAM to SRAM at install-time. The overhead is an even smaller percentage when amortized over several runs of the application; re-runs are common in embedded systems. The reason why the run-time overhead is small can be understood as follows. The customized install-time is proportional to the total number of appearances in the executable file of load and store instructions that accesses the program stack, and the locations that store global variable addresses. These numbers are in-turn upper-bounded by the number of static instructions in the code. On the other hand, the run-time of the application is proportional to the number of dynamic instructions executed, which usually far exceeds the number of static instructions because of loops and repeated calls to procedures. Consequently the overhead of the installer

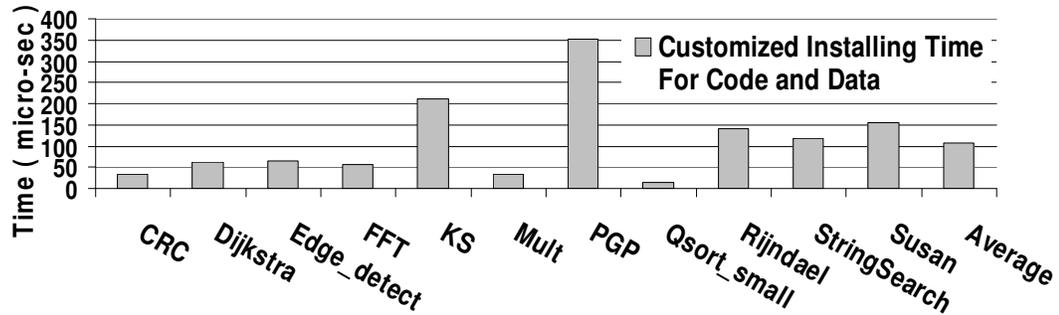


Figure 9.5: Variation of customized installing time across benchmarks

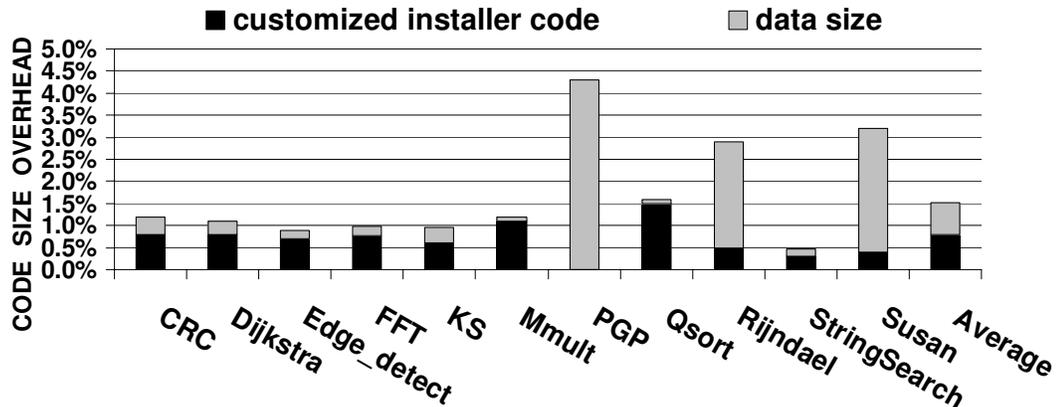


Figure 9.6: Variation of code size overhead across benchmarks

is small as a percentage of the run-time of the application.

Another metric is the absolute time taken by the customized installer. This is the waiting time between when the application has finished downloading and is ready to run after the installer has executed. For a good response time, this number should be low. Figure 9.5 shows that this waiting time is very low, averaging 100 micro-seconds across the eleven benchmarks. It will be larger for larger benchmarks, and is expected to grow roughly linearly in the size of the benchmark.

9.5 Code Size Overhead

Figure 9.6 shows the code size overhead of our method for each benchmark. The code-size increase from our method compared to the unmodified executable

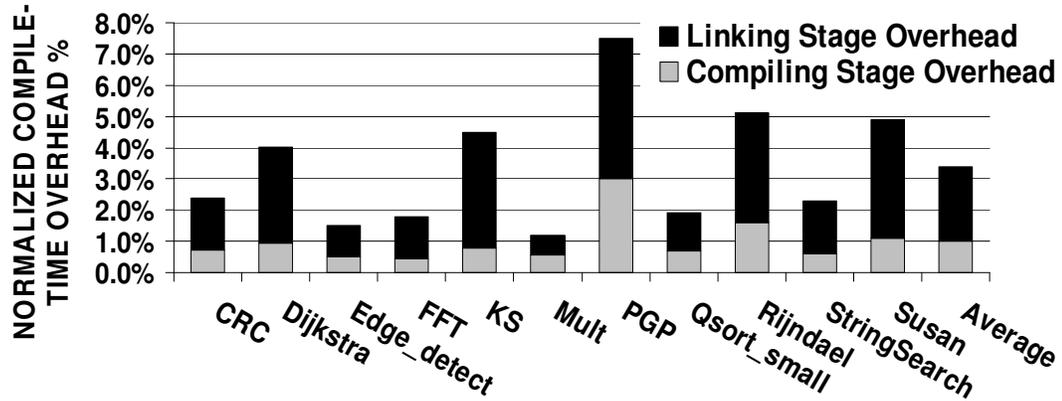


Figure 9.7: Compile Time Overhead

that does not use the SPM averages 1.5% across the benchmarks. The code-size overhead is small because of our technique of reusing the un-used bit-fields in the executable file to store the linked lists containing locations with unknown stack offsets and global addresses. In addition, the figure shows the code size overhead from its constituent two parts: the customized installer codes and the additional information about each variable and code region in the program which is stored until install-time. These additional information are the starting addresses of the location linked-lists, region sizes, region start and end addresses, variable sizes, original stack offsets and global variable addresses.

9.6 Compile Time Overhead

Figure 9.7 shows the compile-time overhead of our technique as a percentage of the time required to compile and link the applications using the unmodified GNU toolchain. Each bar in the figure is further made up of the overheads in the initial compilation and final linker stages. On an average across the benchmarks

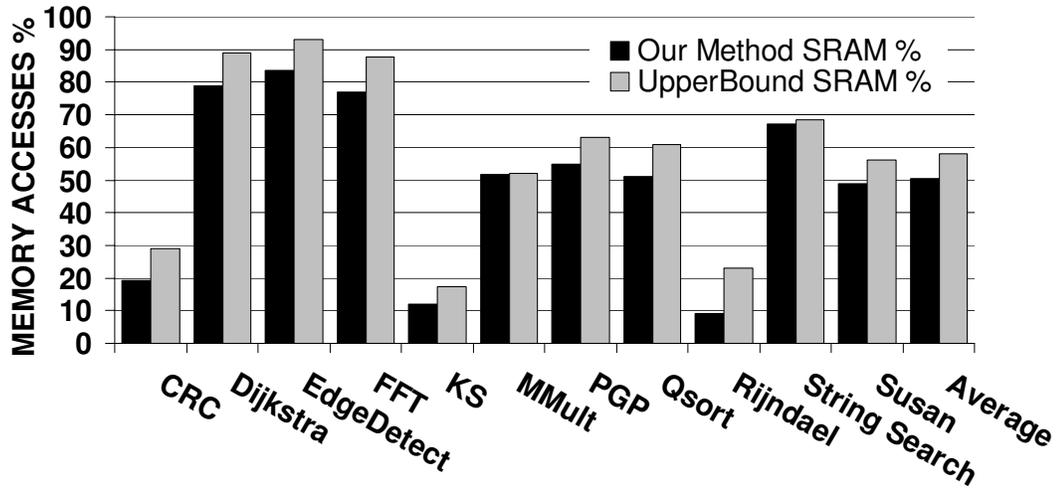


Figure 9.8: Percentage of memory accesses going to SRAM (remaining go to DRAM or FLASH).

the compile-time overhead for our method is only 3%. In this overhead, the ratio of extra time spent in the linking stage versus the compilation stage is approximately 2 to 1.

9.7 Memory Access Distribution

Figure 9.8 shows the percentage of application memory accesses going to SRAM (the remaining accesses go to DRAM or FLASH). It shows that on average, 50% of memory references access SRAM for our method; vs. 58% for Avissar et. al’s method in [5], explaining why our method is close in performance to the un-achievable upper bound in [5].

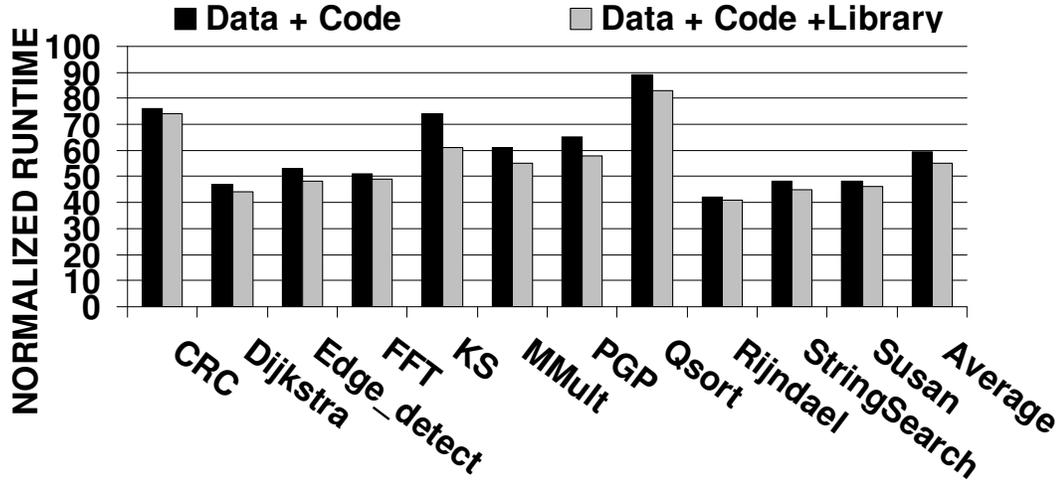


Figure 9.9: Normalized Runtime with Library Variables

9.8 Library Variables

In figure 9.9, we show the normalized run-time (all-DRAM = 100) of our method for two different SPM allocation strategies which target (i) program code, stack variables and global variables, or (ii) library variables and program code, stack variables and global variables. We see that by considering library variables for SPM allocation, an additional 4% run-time speedup is obtained, taking the overall speedup from our method vs. all-DRAM from 41% to 45%. To be fair to [5], a comparison of our method placing library data in SPM versus the optimal static method in [5] is not made since that method does not place library functions in SPM.

9.9 Runtime vs. SPM size

Figure 9.10 shows the variation of run-time for the Dijkstra benchmark with different SPM size configurations ranging from 5% to 35% of the data size. When

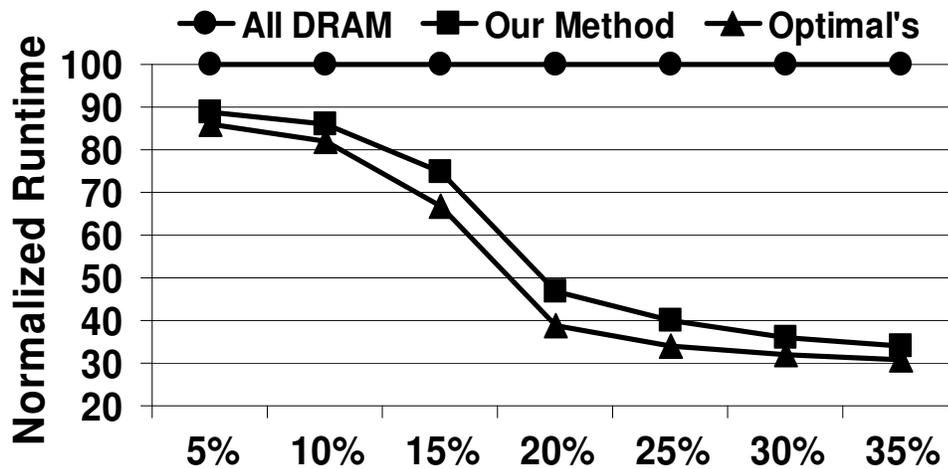


Figure 9.10: Runtime Speedup with varying SPM Sizes for Dijkstra Benchmark

the SPM size is set to lower than 15% of the data size, both our method and the optimal solution in [5] do not gain much speedup for this particular benchmark. Our method starts achieving good performance when the SPM size is more than 15% of the data size since at that point more significant data structures in the benchmark start to fit in the SPM. When the SPM size exceeds 30% of the data set, a point of diminishing returns is reached in that the variables that do not fit are not frequently used. The point of this example is not so much to illustrate the absolute performance of the methods. Rather it is to demonstrate that our method is able to closely track the performance of the optimal static allocation in a robust manner across the different sizes by using the exact same executable. In contrast the optimal static allocation uses different executables for each size.

Chapter 10

Comparison with Caches

The key advantage of our method over all existing SPM allocation schemes is that we are able to deliver near-optimal performance while not requiring the knowledge of the SRAM size at compile-time. In cache-based embedded systems, frequently used data and code are moved in and out of SRAM dynamically by hardware at run-time; therefore caches are also able to deliver good results without the compile-time knowledge of SRAM sizes. For this reason, it is insightful to evaluate our performance versus cache-based systems. In this chapter, we discuss several comparisons in term of performance of our method for SPM versus alternative architectures, using either cache alone or cache and SPM together.

It is, however, important to note that our method is useful regardless of the results of the comparisons with caches. This is because there are a great number of embedded architectures which have a SPM and DRAM directly accessed by the CPU, but have no cache. Examples of such architectures include low-end chips such as the Motorola MPC500 [39], Analog Devices ADSP-21XX [40], Motorola Coldfire 5206E [40]; mid-grade chips such as the Analog Devices ADSP-21160m [41], Atmel AT91-C140 [42], ARM 968E-S [34], Hitachi M32R-32192 [43], Infineon XC166 [44] and high-end chips such as Analog Devices ADSP-TS201S [45], Hitachi SuperHSH7050 [46], and Motorola Dragonball [47].

We have found at least 80 such embedded processors with no caches but with SRAM and external memory (usually DRAM) in our search but have listed only the above eleven for the lack of space. These architectures are popular because SPMs provide better real-time guarantees [48], power consumption, access time and area cost [14, 17, 15, 3] compared to caches.

Nevertheless, it is interesting to see how our method compares against processors containing caches. We compare three architectures (i) an SPM-only architecture; (ii) a cache-only architecture; and (iii) an architecture with both SPM and cache of equal area. To ensure a fair comparison the total silicon area of fast memory (SPM or cache) is equal in all three architectures. For an SPM and cache of equal area, the cache has lower data capacity because of the area overhead of tags and other control circuitry. Area and energy estimates for cache and SPM are obtained from Cacti [49, 50]. The cache area available is split in the ratio of 1:2 among the I-cache and D-cache. This ratio is selected since it yielded the best performance in our setup compared to other ratios we tried. The caches simulated are direct-mapped (this is varied later), have a line size of 8 bytes, and are in 0.5 micron technology. This technology may seem obsolete in high-end desktop system, but with its affordable price, it is a common choice in embedded systems nowadays, where competitive performance, power, and productivity are required at an effective price point. The SPM is of the same technology except we remove the tag memory array, tag column multiplexers, tag sense amplifiers and tag output drivers in Cacti since they are not needed for SPM. The Dinero cache simulator [51] is used to obtain run-time results; it is combined with Cacti's energy estimates per access to yield the energy results.

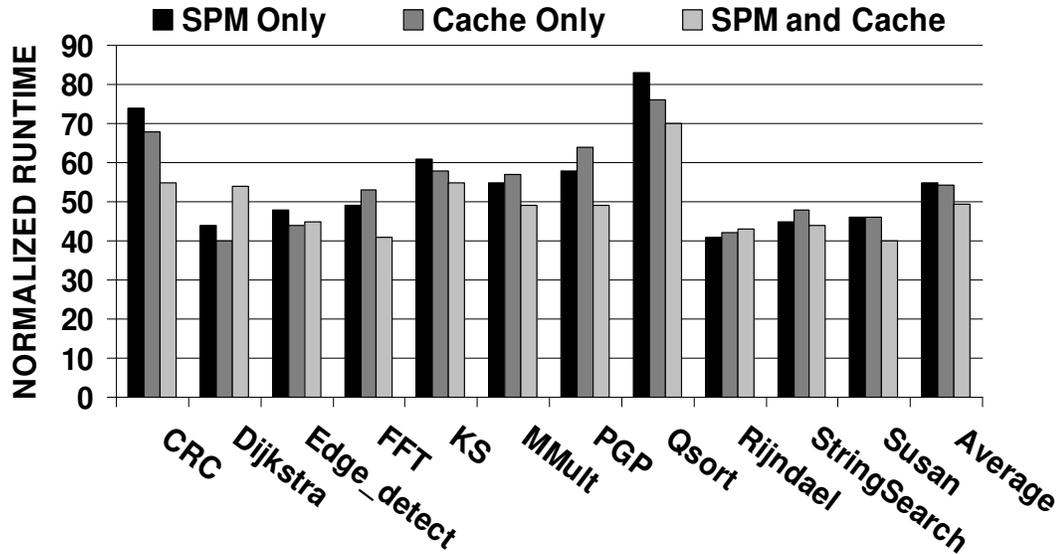


Figure 10.1: Comparisons of Normalized Runtime with Different Configurations

Figure 10.1 compares the run-times of different architectures, normalized with respect to an all-DRAM allocation (=100). The first bar shows the run-time with our method in a SPM-only system allocating variables, code and library data. The second bar shows the run-time of a pure cache-based system. The third bar shows the run-time of our method in a cache and SPM design. In this cache and SPM design, all less-frequently accessed data and code that our method presumes to be in DRAM is placed in cached-DRAM address space instead; thus the slow memory transfers are accelerated¹. By comparing the first and second bar, we see that our method for SPM-based systems has slightly lower, but nearly equal, run-time to that of cache-based systems. The third bar shows that applying our method in a cache and SPM system delivers the best run-time.

Figure 10.2 shows the normalized energy consumption, normalized with re-

¹This may not be the best way to use a memory system with both cache and SPM [15]. Future work could consider how to make an install-time allocator such as ours cache-aware.

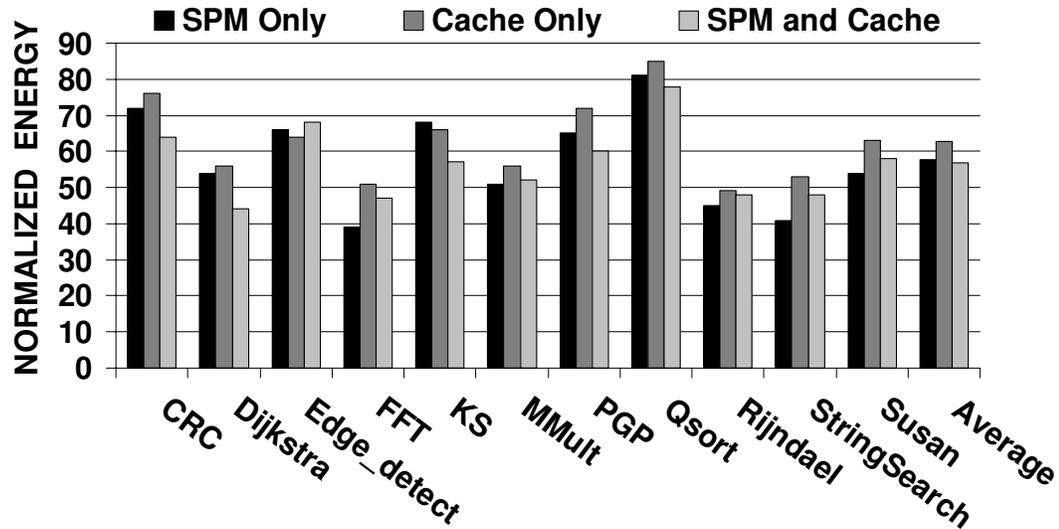


Figure 10.2: Comparisons of Normalized Energy Consumption with Different Configurations

spect to the all-DRAM allocation scheme (=100) for the same configurations as in figure 10.1. Our energy results are collected as the total system-wide energy consumption of application programs. It consists of the energy usage of DRAM, SRAM, FLASH, and the main processor. By comparing the first and second bar of figure 10.2, we see that SPM-only consumes less energy than cache-only architecture. Further, the SPM + Cache combination delivers the lowest energy use.

Figures 10.3 and 10.4 measure the impact of varying cache associativity on the run-time and energy usage, respectively, on the cache-only architecture. The two figures show that with increasing associativity the run-time is relatively unchanged although the energy gets worse; for this reason a direct-mapped cache is used in the earlier experiments in this chapter.

In conclusion, the results show that our method for SPM is comparable to a cache-only architecture and that a SPM + cache architecture provides the best

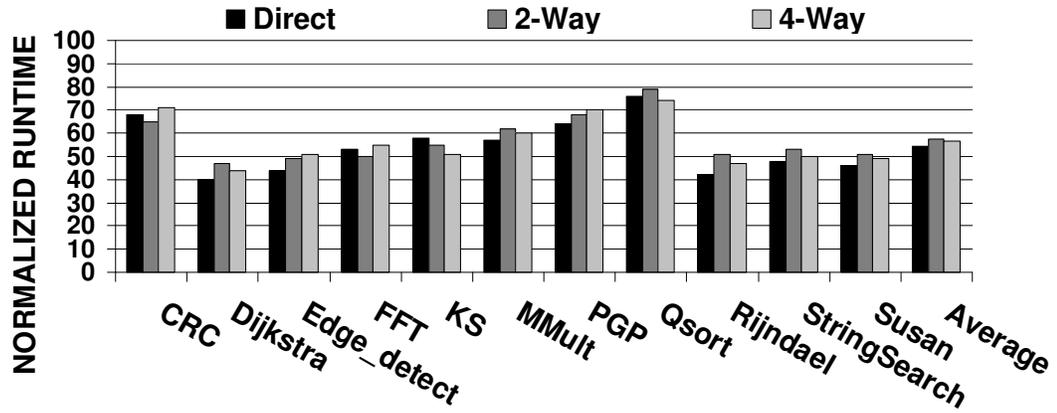


Figure 10.3: Normalized run-time for different set associativities for a cache-only configuration

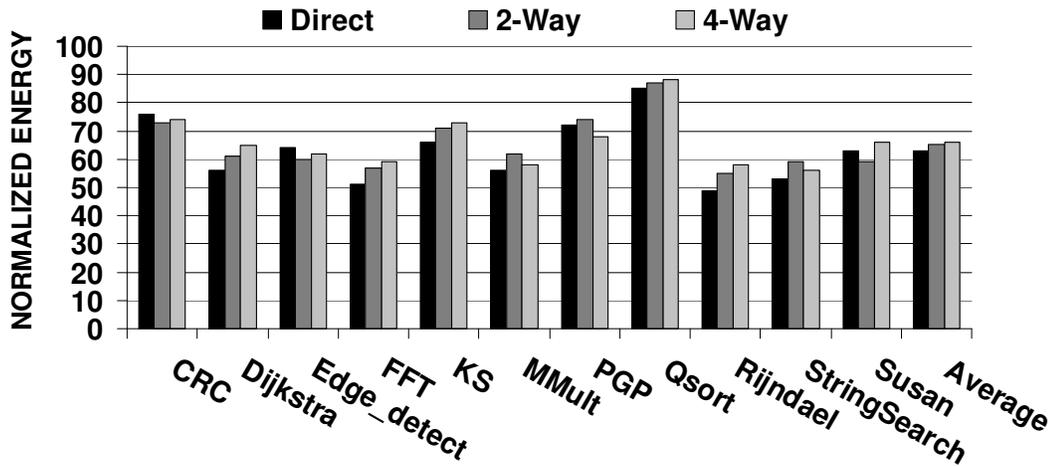


Figure 10.4: Normalized energy usage for different set associativities for a cache-only configuration

energy and run-time. Since the differences are not great, we can only conclude that the overall performance of our method and a cache-based method is comparable. Despite the similar performance vs. caches, our method still has merit because of two other advantages of SPMs over caches not apparent from the results above. First, it is widely known that for global and stack data, SPMs have significantly better real-time guarantees than caches [48, 17, 5]. Second, as described above, there are a great number of important embedded architectures which have SPM and DRAM but no caches of any type.

Chapter 11

Conclusion

In this paper, we introduce a compiler technique that, for the first time, is able to generate code that is portable across different SPM sizes. With technological evolution every year leading to different SPM sizes for the same ISA’s processor implementations, there is a need for a method that can generate such portable code. Our method is also able to share memory between stack variables that have mutually disjoint life-times. Our results indicate that on average, the proposed method achieves a 41% speedup compared to an all-DRAM allocation without knowing the size of the SPM at compile-time. The speedup is only slightly higher (45% vs all-DRAM) with an unattainable optimal upper-bound allocation that requires knowing the SPM size [5].

A possible direction of future work is to devise a dynamic allocation scheme for SPM for unknown-size SPMs. Dynamic schemes can better match the allocation to the program’s access patterns at each instant, and may lead to better performance. Having said that, install-time dynamic SPM allocation schemes suffer increased install-time, run-time and energy overhead on the target device. These device overheads are not seen in off-line compile-time dynamic schemes [7, 20, 8]. These overheads include (i) the dynamic allocator itself, which is inherently more complex than static allocators; (ii) the overheads of binary rewriting to change the

offsets of branches whose displacements are changed as a result of new copying code inherent to dynamic schemes; and (iii) the SPM memory layout will need to be computed at run-time, further complicating the addressing mechanisms and install-time algorithms. Finally, our earlier work on compile-time dynamic schemes [7, 20] shows that the gain for dynamic schemes over static are low or zero for SPM sizes beyond a few Kbytes. In the future, when large SPM sizes will be common (are already common?) the benefit of dynamic schemes will be limited.

Another direction of future work is extending our install-time method to allocate heap data to SPM. The compile-time allocator for heaps in SPM in [9] will likely be a good starting point.

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