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Software Estimation from Executable Specifications

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Abstract

Previous work in software/hardware codesign has addressed issues in system modeling, partitioning, and mixed module simulation and integration. Software estimation, which provides software metrics to assist the software/hardware partitioning, has not been studied. In order to rapidly explore large design space encountered in software/hardware systems, automatic software estimation is indispensable in software/hardware partitioning in which designers or partitioning tools must trade off a hardware with a software implementation for the whole or a part of the system under design. In this report we present a software estimator that provides three software metrics — execution time, program-memory size and data-memory size for a specification executing on a given processor. Experiments have shown that our estimator has less than 20% estimation errors on different designs spanning from straight line code to code with branches and loops and even to hierarchical specifications. Experiments also show that our estimator is fast and can provide rapid feedback to the designers or partitioning tools to quickly evaluate different design alternatives.

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1 Introduction

System design is a set of tasks which convert the system-level specification into a set of completely specified interconnected modules implementing the specification. Each module could be implemented in hardware or software on a processor. One of the system design tasks is partitioning of specification into software and hardware parts. A hardware implementation has better performance whereas a software implementation has lower cost, shorter development time and allows changes late in the design cycle. Thus, the most efficient implementation has a minimal amount of costly application-specific hardware while still meeting the required performance constraints.

Several researchers have described frameworks for modeling, simulation and integration of software/hardware designs [SB92][KL92][BV92]. Software/hardware partitioning techniques have been addressed in [GM92a][EH92]. Gupta and De Micheli [GM92a] propose a partitioning algorithm that starts with an initial partition where all operations, except for the unbounded delay operations, are assigned to hardware. The partition is refined by migrating operations from hardware to software in the search for a lower cost feasible partition. The approach used by Ernst and Henkel [EH92] starts with a complete software implementation from which those portions that violate the timing constraints are extracted for hardware implementation. These two approaches start from different directions but work towards the same goal of minimizing the amount of application-specific hardware required. This software/hardware partitioning requires a software estimator that will predict the execution time of the software implementation in order to identify which portion in the specification can be migrated from hardware to software while not violating the constraints or which portion needs to be implemented in hardware to satisfy the timing constraints. To our knowledge, no previous work has addressed the issue of software estimation assisting in the software/hardware partitioning. In the absence of an automatic estimator, the effect of each partitioning can be evaluated only through actual implementation which prevents a designer from considering other design alternatives.

In this report we present our software estimator which provides three software metrics — execution time, program-memory size and data-memory size for a given specification and a given target processor. The generic model used in our estimator does not require different estimators for different target processors. Instead a single estimator and a set of technology files for different target processors are used. This makes our estimator fast and easy to extend for different target processors. Also, the probability-based flow analysis technique used for the performance estimation in our estimator provides further advantages over dynamic simulation [PK90].

The input to our software estimator is a system-level description specified with the executable specification language — SpecChart. A SpecChart consists of hierarchical concurrent/sequential behaviors with leaf behaviors specified using the VHDL language. Details on the language and its constructs may be found in [NVG91].

In the next section, we present the underlying model used for software estimation. Performance and memory size estimation for system-level specifications are discussed in Section 3 and Section 4 respectively. The results of our experiments are presented in Section 5 followed by the conclusion in Section 6.

2 Model for Estimation

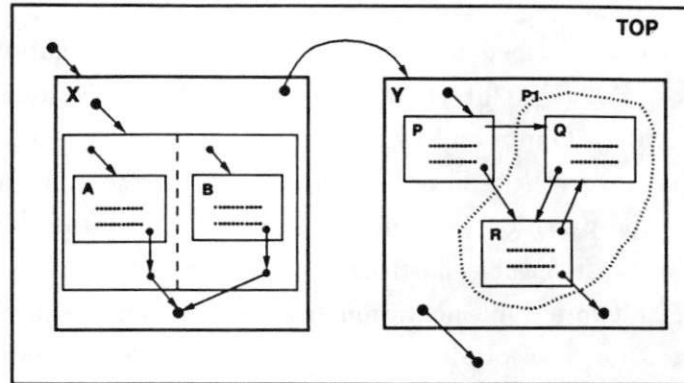


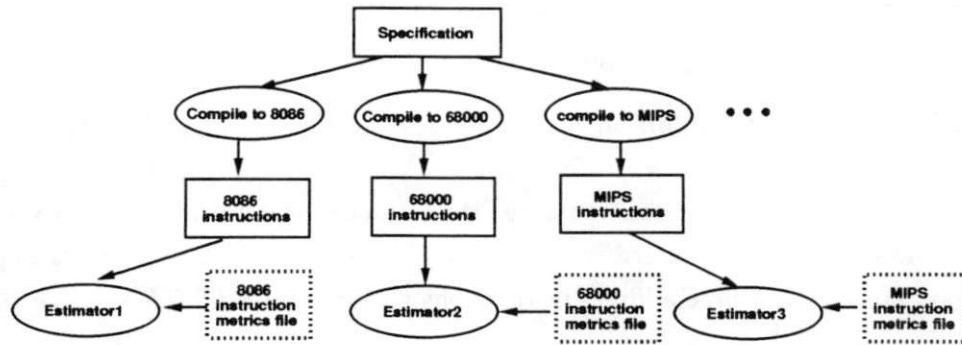
Figure 1: A sample system-level specification

The estimation model we propose is targeted to the system-level specification which consist of hierarchical concurrent/sequential behaviors. A behavior which is a set of actions and a set of conditions describing when each action is to occur, can in turn contain sequential or concurrent sub-behaviors. For example, in Figure 1, behavior TOP consists of two sequential sub-behaviors X and Y. Behavior X in turn contains two concurrent sub-behaviors A and B. Behavior Y contains three sequential sub-behaviors P, Q, R. Concurrency is represented by the dashed line like the one between behavior A and B whereas sequencing is represented by those transition arcs. The behavior which does not contain any sub-behaviors is called leaf behavior. In SpecChart, each leaf behavior consists of a set of VHDL sequential statements. Behavior A, B, P, Q and R in Figure 1 are leaf behaviors. The dots in the specification indicate the starting or completion of the behaviors. Our estimators are intended to estimate the software metrics for any given leaf/non-leaf behavior of the specification as well as any given partition (a set of behaviors) in the specification. P1 in Figure 1 is a partition which contains two behaviors Q and R.

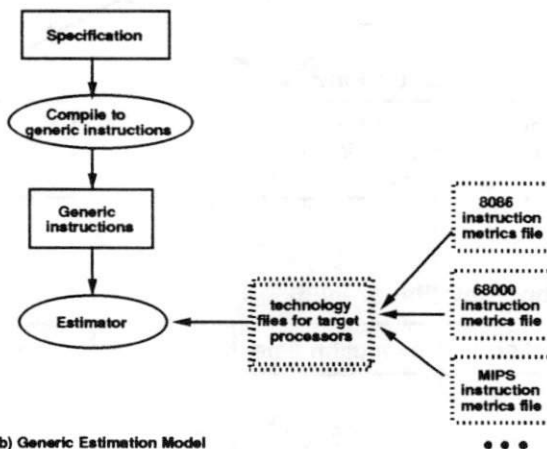
2.1 Estimation Model for Leaf Behaviors

In order to obtain the estimates for leaf behaviors, we need to compile the code in the leaf behaviors into machine instructions of the target processor. For example, if a leaf behavior will be implemented on an Intel 8086 processor, it needs to be compiled into the 8086 instruction set. Using the timing

and size information associated with each type of instruction such as how many clock cycles the 8086 instruction executes and how many bytes it takes, we can obtain the performance and program size of the behavior. Similarly, if the leaf behavior is going to be implemented on a Motorola 68000 processor, it needs to be compiled into 68000 machine instructions. Based on the 68000 instruction timing and size information, the estimator can obtain the software metrics for the behavior. This model, in which the estimator is targeted to one specific processor, is called processor-specific model (shown in Figure 2(a)).



(a) Processor-specific Estimation Model



(b) Generic Estimation Model

Figure 2: Two different estimation models

Instead of using different compilers and estimators for different target processors in the processor-specific model (Figure 2(a)), we propose a generic estimation model (Figure 2(b)) in which the leaf behavior specification is converted into a set of generic three-address instructions. After that the estimator will compute the software metrics for the leaf behavior based on the generic instructions and the technology files for the target processors. For example, if the leaf behavior is going to be implemented on an Intel 80286 processor, then the technology file for the 80286 processor will be used during the estimation. The technology file for a target processor supplies information about how many clock cycles each type of generic instruction needs and how many bytes it takes if the generic

instruction is executed on that target processor. The technology file for a target processor is derived from the timing and size information of the processor's instruction set.

The generic three-address instructions used in our estimation model have the following formats:

1. Arithmetic/logic/relational operation: $des \leftarrow src1 \text{ op } src2$; For unary operations, $src1$ is empty;
2. Move/load/store operation: $des \leftarrow src$;
3. Conditional jump operation: **if** *cond* **goto** *label*;
4. Unconditional jump operation: **goto** *label*;
5. Procedure call operation: **call** *label*;

Here *des*, *src* and *cond* are either constants, registers or memory locations. Memory locations could be directly addressed like *A*, *B* or addressed with offset like *A[I]* or *B[J]*. *label* refers to procedure names or instruction labels. The three-address instructions also include *RETURN* and *NULL* instructions.

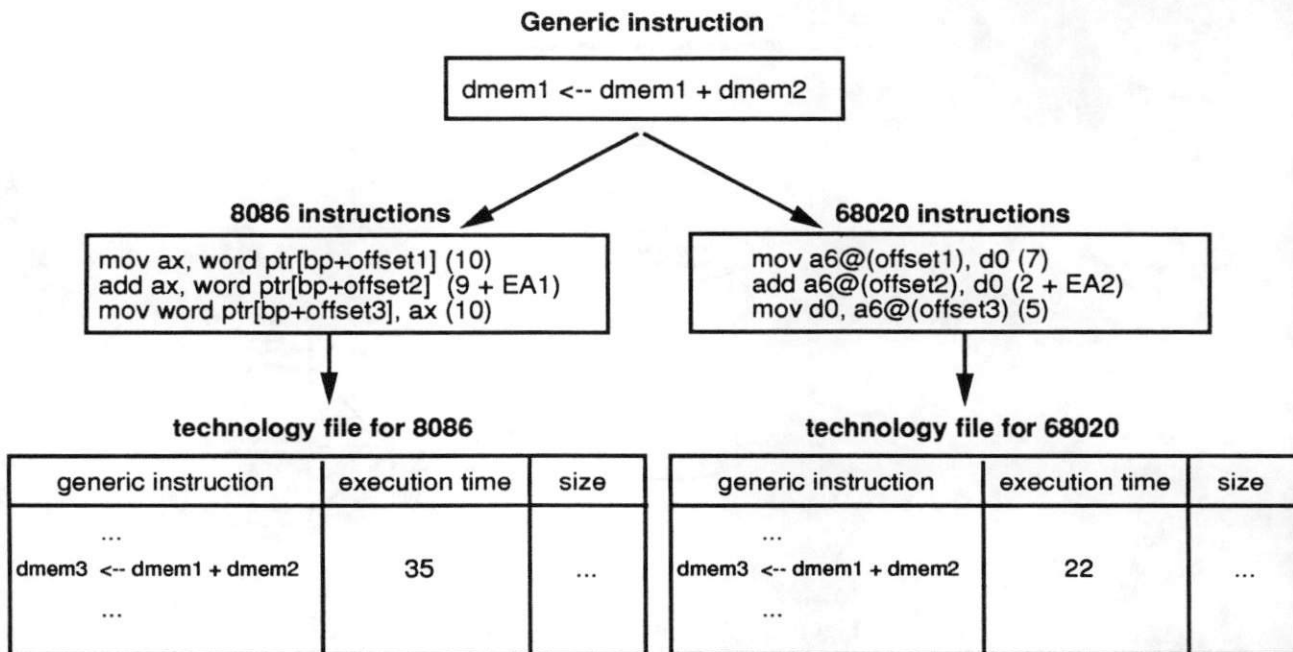


Figure 3: Execution time of the generic instruction for different processors

To provide the technology file for a given processor, we need to find out how many clock cycle each type of generic instruction needs and how many bytes it takes on that processor. Figure 3 shows the computation for the number of clock cycles for the generic instruction with type of $dmem3 \leftarrow dmem1 + dmem2$. Here, *dmem* indicates direct memory addressing mode. The generic instruction is first mapped to a sequence of target processor instructions followed which the total number of clock cycles of the generic instruction is obtained by summing the clock cycles of each individual

instruction in the sequence. *EA1* and *EA2* in Figure 3 are the effective address calculation times used for displacement memory addressing mode, which are 6 and 8 clock cycles in the 8086 and 68020 respectively. The generic instruction thus will take 35 and 22 clock cycles on the 8086 and 68020 processors respectively. Using a similar approach we can derive how many bytes each type of generic instruction will take if it is executed on the 8086 or 68000 processor. Presently the technology files for 8086, 80286, 68000 and 68020 processors are supported in our estimator. The 8086, 80286, 68000 and 68020 technology files are derived from the timing and size information of their corresponding instruction sets given in [I8086][I80286][M68000][M68020]. The technology file for Intel 8086 processor is shown in the Appendix.

Compared with the processor-specific model, our generic model has the following advantages:

1. In the generic model, we do not need to use different compilers and different estimators for different target processors. Instead, only a single compiler, estimator and a set of technology files are required for the estimation.
2. The generic model makes it much easier to apply the estimator to other target processors. The estimation can be carried out as long as the technology file for the target processor is supplied.
3. The peculiarities of each type of processor is reflected in the technology file for the processor. The generic three-address instructions are free of instruction idiosyncrasies. Thus it is much easier and faster to compile the specification into the generic instruction set than those associated with specific processors.

2.2 Estimation Model for Non-leaf Behaviors and Partitions

Non-leaf behaviors possess hierarchical or concurrent constructs. To evaluate the software implementation of a given non-leaf behavior or a partition on a specific microprocessor, we must first flatten the hierarchy and sequentialize/schedule the specification to diminish the concurrency [GM92b] since our target machine is a uni-processor. In other words, the specification needs to be mapped (flattened/sequentialized) into a program written in a language which can be directly compiled to the machine instructions of the given processor. Based on the machine instructions generated, the software metrics such as performance and memory size for the specification can thus be computed. The software metrics obtained in such a way are accurate since they are computed from the actual implementation of the specification on the given processor. However, due to that automatic partitioning tools will evaluate hundreds or thousands of partitions, this approach is too costly and time consuming since we would have to actually implement each partition on the given processor through flattening, sequentializing and compiling process in order to get the software estimates for that partition. To get fast estimates while not sacrificing too much accuracy, the estimation model we propose combines two different approaches: an accurate approach for estimating leaf behaviors and a fast approach for estimating non-leaf behaviors and partitions. Prior to the partitioning process, each leaf behavior is compiled

and estimated using the approach described in the previous section. During the partitioning process, the software estimates for each partition are constructively computed bottom up from the estimates of the leaf behaviors. Such a combined approach may be less accurate than the approach based on the actual implementation of a partition. However it is much faster because it does not involve flattening, scheduling, compiling for each partition during the design process. It only requires some computation based on the pre-obtained estimates for the leaf behavior specifications. Therefore this model allows rapid evaluation of different design alternatives.

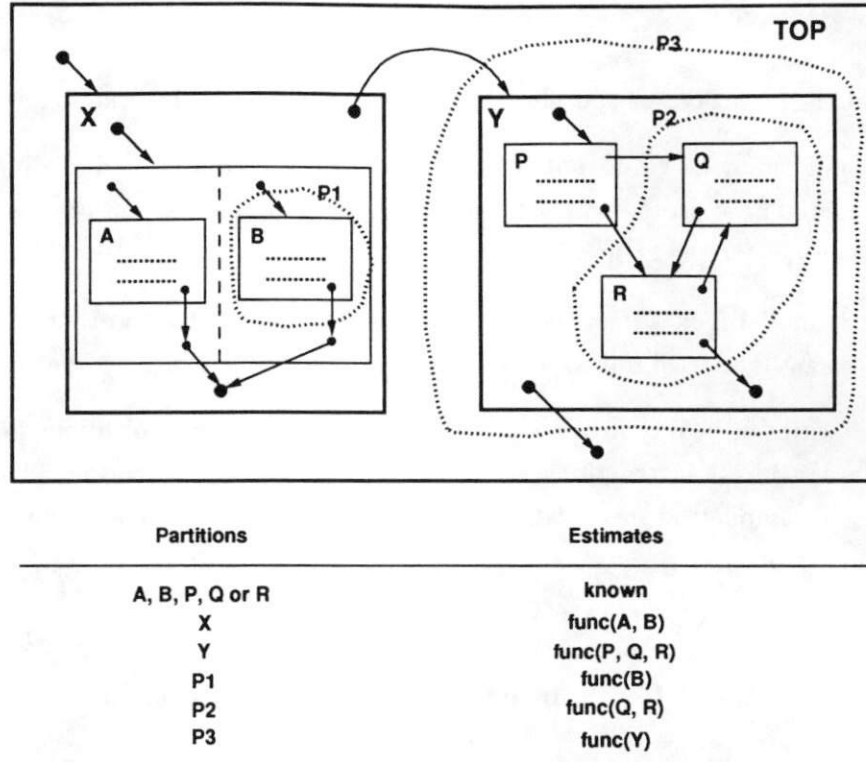


Figure 4: A sample specChart and its estimates

Figure 4 illustrates our estimation model with a sample SpecChart. Before partitioning process, the software metrics for leaf behaviors A, B, P, Q, and R are estimated. During partitioning process, the software metrics for each partition are computed based on the already known estimates of its constituent behaviors. For example, the estimates of partition P2 is a function of the estimates of behavior Q and R.

3 Performance Estimation

There are two possible ways to find out how many clock cycles a program takes to execute — dynamic simulation and static estimation. Given a set of input data, dynamic simulation actually executes

the program and records the clock cycles used in each execution. Given different sets of input data, dynamic simulation may obtain different number of clock cycles for that program due to the data dependent conditional branches and loops. Static estimation, on the other hand, is insensitive to input data. It just computes the average number of clock cycles needed to execute the program. Static estimation can yield good results if the number of loop iterations is known and the conditional branching probability can be predicated correctly. Besides, static estimation has a number of advantages: (1) It takes much less time and space than dynamic simulation. (2) It does not need input data (test vectors). The probability-based technique and its application to the performance estimation for system-level specifications are presented in this section.

3.1 Flow Analysis

Flow analysis is a technique used in the static estimation for design with conditional branching (including loops). Given a control flow graph $G = (V, E)$ representing a portion of the design, where V is the set of vertices v_i , and E is the set of *directed* edges e_{ij} connecting vertex v_i to v_j and indicating sequencing between v_i and v_j , we wish to determine the execution frequencies of each of its nodes based on the branching probabilities. By determining the execution frequencies of the nodes, we can obtain useful information about the design by associating with each node in the graph, *weight* representing some design parameters. For example, if each node weight represents the execution time of a basic block derived from the code of a leaf behavior, then we can use the execution frequency of the nodes and the weights of the nodes to determine the total execution time for the leaf behavior by taking a weighted sum of these two quantities.

3.1.1 Determining the Branching Probabilities

Branching probabilities are associated with the edges in the control flow graph. They could be determined in the following ways:

1. **Equal Probabilities** : In case of branching, we assign equal probabilities to all the edges emerging from the node. Thus, if there are n edges emerging from a node, all of them are assigned a probability of $1/n$.
2. **Loop Related Probabilities** : When the number of loop iterations is known, say n , the exit edge has a probability of $1/n$ while the back edge has a probability of $(n - 1)/n$.
3. **User Defined Probabilities** : The user may specify the branch probabilities in the SpecChart description using annotations.
4. **Simulation Based Probabilities** : For every input data, a record can be kept of the branches taken during the simulation. From this observed behavior, the probabilities of the branches can be derived.

Currently the first two approaches have been implemented in our estimator.

3.1.2 Determining Node Execution Frequencies

The execution frequency of a node is defined as the number of times on the average that the node will be executed in a single execution of the graph. We use the branching probabilities between the nodes to determine the execution frequencies. This is given in the following procedure.

Determining Node Execution Frequencies:

1. Determine the branch probabilities using one of the methods outlined above.
2. A start node, S, preceding the first node in the graph, is added. Its execution frequency, $F(S)$ is set to 1 since this node is executed exactly once whenever the control flow graph is executed.
3. The execution frequency $F(N_j)$ for any node N_j depends on the weighted execution frequencies of all its immediate predecessor nodes. The execution frequency for each predecessor node N_i is multiplied by the branch probability of the edge between N_i and N_j , $P(e_{ij})$. For each node in the graph, we first formulate the equation for the node execution frequency.

$$F(N_j) = \sum_{\text{all predecessor nodes } N_i \text{ of } N_j} F(N_i) \times P(e_{ij}) \quad (1)$$

4. We then solve the set of equations formulated in step 3 to obtain the individual execution frequencies. There are a variety of methods such as Gaussian Elimination, LU decomposition, Chomsky's method which can be used. We have selected the Gaussian Elimination method in our estimator.

The procedure is illustrated by an example shown in Figure 5. A control flow graph with the branch probabilities is shown in Figure 5(a). Figure 5(b) shows the same graph with the added dummy node, S. Figure 5(c) shows the equations for the execution frequency of each node. As an example, consider node N_2 . It has two predecessor nodes N_1 and N_3 . The probabilities of the edges e_{12} and e_{32} are 1.0 and 0.8 respectively. Thus the equation for the execution frequency for N_2 is :

$$F(N_2) = 1.0 \times F(N_1) + 0.8 \times F(N_3) \quad (2)$$

Solving the equations in Figure 5(c) yields the following values for execution frequency :

$$F(N_1) = 1, \quad F(N_2) = 5, \quad F(N_3) = 5, \quad F(N_4) = 1$$

These are the expected execution frequencies returned after probability based analysis of the control flow graph.

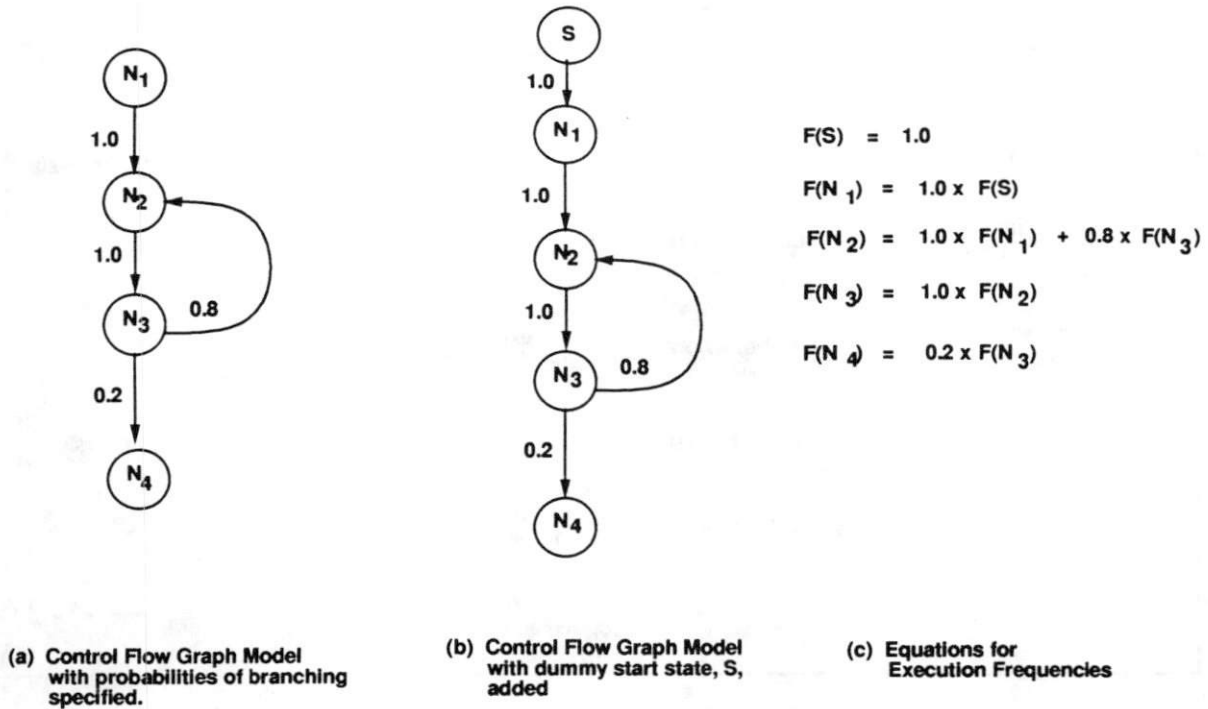


Figure 5: Obtaining a set of linear equations from the control flow graph model

In situations where the edges are also associated with some weights, we may need to know the execution frequency of each edge $F(e_{ij})$. It is obvious that the execution frequency of an edge is the same as that of its target node. Therefore we have $F(e_{ij}) = F(N_j)$.

3.1.3 Determine the Performance of Control Flow Graph

For a control flow graph $G = (V, E)$ with each node $v \in V$ associated with a weight $W(v)$ and each edge $e \in E$ associated with a weight $W(e)$, if we know the execution frequency of each node $F(v)$ and the execution frequency of each edge $F(e)$ in the graph, then the performance $P(G)$ of the graph G can be calculated as follows.

$$P(G) = \sum_{\text{for all } v \in V} (W(v) \times F(v)) + \sum_{\text{for all } e \in E} (W(e) \times F(e)) \quad (3)$$

3.2 Applying Flow Analysis to Performance Estimation

Figure 6. summarizes the performance estimation for different entities such as *partition*, *behavior* of the specification, *basic block* of the leaf behavior and *generic instruction* of the basic block. The performance of a partition depends on that of its containing behaviors. The performance of a sequential non-leaf behavior is determined using flow analysis on the performance of its containing sub-behaviors. The performance of a concurrent non-leaf behavior is the sum of the performance of

Entity to be estimated	Components of entity	Weights of components	Technique to be applied
partition	behaviors	execution times of behaviors	sum/flow analysis
sequential non-leaf behavior	sub-behaviors	execution times of sub-behaviors	flow analysis
concurrent non-leaf behavior	sub-behaviors	execution times of sub-behaviors	sum
leaf behavior	basic blocks	execution times of basic blocks	flow analysis
basic block	generic instructions	execution times of generic instructions	sum
generic instruction	——	execution times specified in the technology file	——

Figure 6: Performance estimation for different entities

its containing sub-behaviors. The performance of a leaf behavior is determined using flow analysis on the performance of its containing basic blocks. The performance of each basic block is computed by summing the performance of its containing generic instructions. The performance of each generic instruction is taken from the technology files supplied to the estimator.

3.2.1 Performance Estimation for Leaf Behavior

Constructing Basic Blocks from VHDL statements

We divide the VHDL code segment in each leaf behavior into *basic blocks*. A *basic block* [ASU88] is a sequence of consecutive VHDL statements in which flow of control enters at the beginning and leaves at the end without halting or the possibility of branching, except at the end. To determine the basic blocks, we first determine the set of *leaders*, the first statement of a basic block. The rules we use to determine leader statements are: (1) The first statement of the code segment is a leader. (2) All wait statements and procedure calls are leaders. (3) Any statement which is the target of a conditional statement (if, loop, case) is a leader. The target of a conditional statement is any statement to which control could possibly be transferred on evaluating the condition. (4) Any statement that immediately follows a conditional statement, wait statement, or a procedure call is a leader.

For each leader determined above, its basic block consists of the leader and all statements up to but not including the next leader (or the end of the given VHDL code segment). A basic block will then contain one of the following: (1) a set of assignment statements, or (2) a single wait statement

or (3) a single procedure call.

Figure 7 shows how the basic blocks can be constructed from a given set of sequential VHDL statements. In Figure 7(a), the leaders are denoted by horizontal arrows along with the corresponding rule number used to determine the leader. Figure 7(b) shows the basic block structure of the VHDL code. The conditions to be evaluated for the conditional branching are associated with the edges between basic blocks.

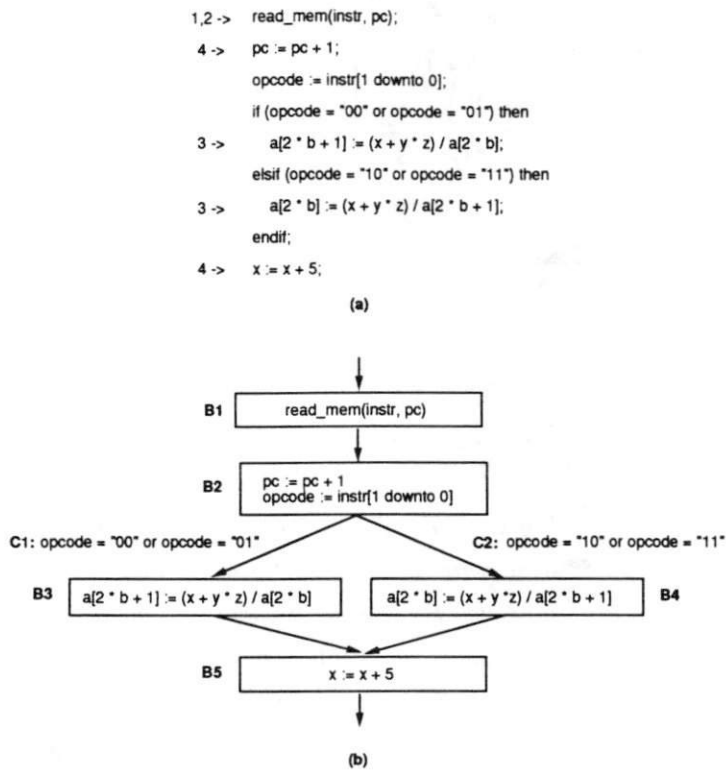


Figure 7: Building basic blocks from VHDL statements

Obtaining Weights for Basic Blocks and Conditions

Once basic blocks have been determined (Figure 7(b)), we need to find the weight (i.e. execution time) of each basic block and weight of each condition. The VHDL assignment statements in each basic block and condition are compiled into our generic three-address instructions. Figure 8(a) shows the basic block structure of Figure 7(b) after the VHDL code is compiled into generic instructions. The execution time of each generic instruction can be obtained from the technology file based on its type. For example, the instruction $pc \leftarrow pc + 1$ has the type of $dmem \leftarrow dmem + constant$ ($dmem$ means directly memory addressing mode). In Figure 8(a), $T1$, $T2$ and $T3$ are temporary variables which can be considered as type of register or $dmem$ depending on how many general-purpose registers the target processor has. Suppose after the compilation, m temporary variables $T1, T2, \dots, Tm$ are used and the target processor has n general-purpose registers. If $n \geq m$, then $T1, T2, \dots, Tm$ will

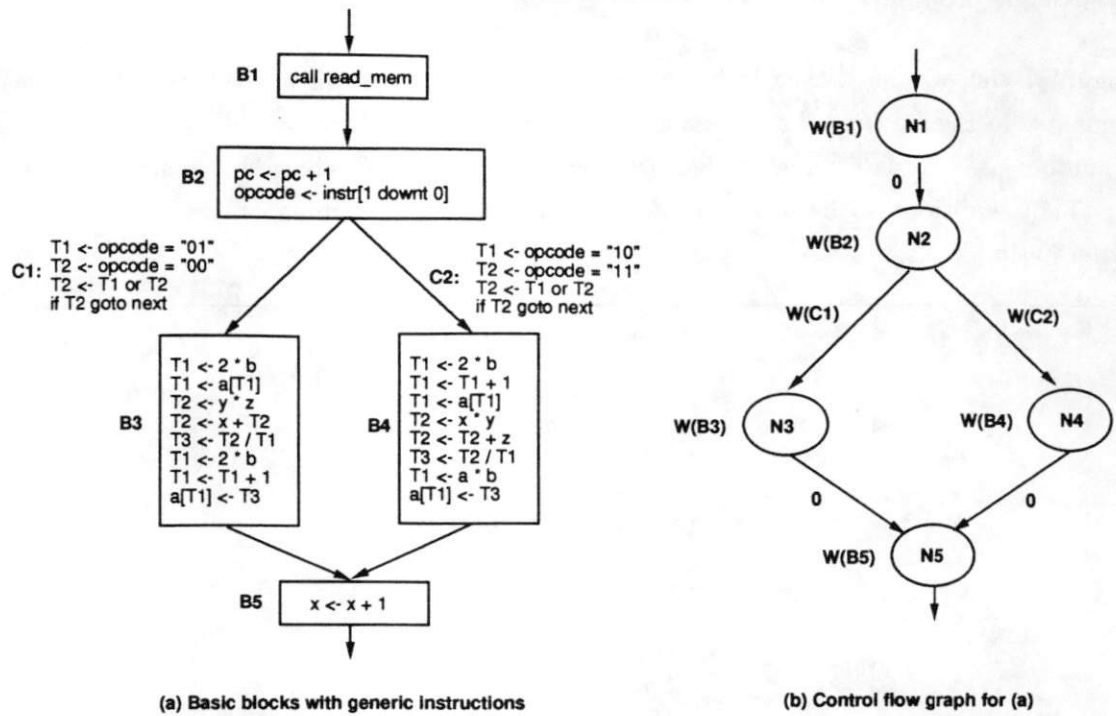


Figure 8: A basic block structure and its corresponding control flow graph

be considered as register type, otherwise, T_1, T_2, \dots, T_n will be considered as register type whereas T_{n+1}, \dots, T_m will be considered as $dmem$ type. Temporary variable with smaller index (e.g. T_1) will be used more heavily than temporary variable with larger index (e.g. T_9) in the compilation. Therefore temporary variable with smaller index should have higher priority to be assigned with a register. In the example shown in Figure 8(a), suppose the target processor has two general-purpose registers. So T_1 and T_2 will be considered as register type whereas T_3 will be considered as a memory type. Thus instruction $T_1 \leftarrow T_1 + 1$ has the type of $register \leftarrow register + constant$. Instruction $T_3 \leftarrow T_1/T_2$ has the type of $dmem \leftarrow register/register$. The weight $W(B)$ of a basic block B can be computed by summing the execution time of each generic instruction in that basic block. The weight $W(C)$ of a condition C can be obtained in a similar way.

Obtaining Weight of Basic Block With a Wait Statement

The weight of a basic block with a wait statement is decided as follows:

1. **Wait** : In the absence of a timeout clause, the execution time is set to a very large number like MAXINT, the largest integer supported by the host machine.
2. **Wait on S_1, \dots, S_n until C_1, \dots, C_2 for T ns** : The execution time for a wait statement with only a *timeout clause* (i.e. *for* clause) is equivalent to the smallest multiple of the clock period higher than T . The effect of conditions (e.g. C_1) and an event on signals in the sensitivity list (e.g. S_1) are ignored.

3. **Wait on S_1, \dots, S_n until C_1, \dots, C_2** : i.e wait statement without a timeout clause. If the wait statement has a sensitivity list or condition clause but no timeout clause then the wait statement requires at least one control step to be executed. The effect of conditions (e.g. $C1$) and an event on signals in the sensitivity list (e.g. $S1$) are ignored.

Currently the estimation used for the wait statement is very primitive. Further work is needed to obtain better estimation for the wait statement by considering the input rate of the signals in the sensitivity list.

Obtaining Weight of Basic Block with a Procedure call

The third type of basic block is one which has a single procedure call. The execution time for the procedure can be determined by treating the body of the procedure declaration as a leaf state. This time is then used as the execution time of the basic block which contains the procedure call.

Computing Performance for Leaf Behavior

The basic block structure of a leaf behavior is mapped to an equivalent control flow graph G . Figure 8(b) shows the corresponding control flow graph of Figure 8(a). Each basic block B_i is mapped to a node N_i in G . Each edge connecting two basic blocks B_i and B_j is mapped to an edge connecting node N_i and N_j in the graph. Each node N_i in G has a weight which is the same as $W(B_i)$. Each edge in G has a weight which is the same as the weight of the condition associated with the corresponding edge in the basic block structure. Applying the flow analysis and using the weights obtained for the nodes and edges, the average execution time for the leaf behavior can be computed using equation 3 in section 3.1.3.

3.2.2 Performance Estimation for Non-leaf Behavior

Once execution times have been estimated for each of the leaf behaviors as shown above, we can merge the performance estimates of the leaf behaviors to yield the the performance estimate of the next higher behavior in the hierarchy. The approach adopted is similar to that of merging the performance estimates of basic blocks to obtain performance estimates for the leaf behaviors. At any level of the hierarchy, we will first determine the performance estimates of the child behaviors and then combine these estimates to determine the performance estimate of the parent behavior.

To estimate the performance for a non-leaf behavior with sequential sub-behaviors, B_{parent} , we create a control flow graph $G = (V, E)$ for its child behaviors whose performance estimates are already known. For each of the sub-behaviors, B_i , of B_{parent} , there exists a corresponding vertex v_i in the graph G . For every transition arc between the two sub-behaviors B_i and B_j , the set E has a directed edge e_{ij} from vertex v_i to vertex v_j in G . After the control flow graph model has been constructed for the sub-behaviors, we can apply the flow analysis (section 3.1.3) to obtain the performance of the parent behavior B_{parent} .

In case a behavior at any level of the hierarchy has concurrent sub-behaviors, the execution time of that behavior is computed as the sum of that of its child behaviors. It must be mentioned here that a non-leaf behavior may have a descendant sub-behavior which does not have a *stop dot* in SpecChart. In this case the behavior will never finish executing and consequently the execution time returned for that behavior is an arbitrarily large number.

4 Memory Size Estimation

Given a behavior, memory size estimation is to determine how much program-memory (i.e. bytes used to store the compiled program representing the behavior) and how much data-memory (i.e. bytes used to store the data manipulated by the behavior) are needed.

4.1 Program-memory Size Estimation

The size of each type of generic instruction is specified in the technology file for target processor. For example, the size of the generic instruction with type of $dmem \leftarrow dmem + dmem$ is 9 bytes in the technology file for the 8086 processor and 6 bytes in the technology file for the 68020 processor. Based on the size of each generic instruction, the program-memory size of each basic block is the sum of that of all generic instructions in that basic block. The program-memory size of a leaf behavior in turn is the sum of that of all its basic blocks. Analogously, the program-memory size of a non-leaf behavior is the sum of that of all its sub-behaviors.

4.2 Data-memory Size Estimation

The data-memory size is determined based on the data declaration parts in the specification. The data-memory size $DMS(D)$ of a declaration D is determined by the size of D 's base type and the number of base type elements in D . For example, the base type of the declaration 'variable a: integer' is *integer* and the number of base type elements is 1. For declaration 'variable b: bitvector [9 downto 0]', the base type is *bitvector* and the number of base type elements is 10. The data-memory size of a declaration D is computed as follows:

$$DMS(D) = DMS(BT(D)) \times N \quad (4)$$

where, $BT(D)$ is the base type of the declaration D . N is the number of base type elements in D .

The data-memory size of each base type is specified in a configuration file. The information currently used in the configuration file is shown in Figure 9.

Base Type	Data Memory Size (bytes)
Bit	1
Bitvector	$\lceil n/8 \rceil$ where n is the number of bits in the vector.
Boolean	1
Character	1
Integer	4
Natural	4
Positive	4
Real	8
String	4
Time	4

Figure 9: Size of the base type

After obtaining the data-memory size of each declaration, The data-memory size of a leaf behavior can be computed by summing that of each declaration in its declaration part. The data-memory size $DMS(B)$ of a non-leaf behavior B is given as follows:

$$DMS(B) = \sum_{\text{all sub-behaviors } B_i \text{ of } B} DMS(B_i) + \sum_{\text{all declaration } D_i \text{ in } Decl(B)} DMS(D_i) \quad (5)$$

5 Results

Given a SpecChart description and a leaf behavior name in the SpecChart, our estimator will output the estimates for each basic block in the leaf behavior and the execution frequency of each basic block as well as the total estimates for the leaf behavior. In this way, designers who are not only interested in the estimates of the whole behavior but also interested in the estimates of a loop body or some basic blocks can easily find out estimates of any part of the specification. Similarly, given a SpecChart description and a non-leaf behavior name, our estimator will output the estimates and execution frequency for each sub-behavior in the non-leaf behavior as well as the total estimates of the non-leaf behavior.

We have compared the estimation results for different target processors including 8086, 80286, 68000, 68020 with those results obtained by flattening (this step is not needed in the first two designs) and compiling the designs directly into the instruction set of the target processors. Since there is no VHDL compiler available for those target processors, we first manually convert the VHDL code to its equivalent C code. Following that we compiled the C code into the instruction set of 8086, 80286, 68000 and 68020 processors. Based on the machine instructions generated from the C compilers and the instruction timing and size information provided in [I8086][I80286][M68000][M68020], we have

manually computed the actual performance and program-memory size for the designs. Those actual results were then compared with the estimates based on the generic three-address instruction compiler supplemented by corresponding technology files for the target processors.

The first two descriptions we choose were the fifth order elliptical filter and the differential equation example adopted from [DR92]. The behavior of the elliptical filter consists of an infinite loop. We estimated only the loop body. In the differential equation example, there are three basic blocks and one of them is a loop. The loop body can be executed any number of times depending on the external parameters. In our experiments, we have assumed that the loop body would be executed 10 times. The performance results are shown in Figure 10 while program-memory sizes are shown in Figure 11. The data-memory size estimated for the elliptical filter example is 90 bytes. The data-memory size estimated for the differential equation example is 21 bytes.

Application	Target Machine	Actual Performance (in clock cycles)	Estimated Performance (in clock cycles)	Estimation Error
Elliptic filter	8086	2569	2488	-3.2%
Elliptic filter	80286	712	662	-7.0%
Elliptic filter	68000	1692	1632	-3.5%
Elliptic filter	68020	924	888	-3.9%
Differential equation	8086	9446	10586	12.1%
Differential equation	80286	2244	2304	2.7%
Differential equation	68000	6112	6452	5.6%
Differential equation	68020	3416	3676	7.6%

Figure 10: Performance estimation

Generally, compilers optimize the object code by using different optimization techniques such as global optimization, loop optimization, register allocation, optimization for speed or space. Users can invoke those optimizations by passing special flags to the compiler. In previous experiments we have disabled those optimizations during the C compilation since our generic compiler does not use optimization heuristics. Therefore our estimates are for those non-optimized code. In order to estimate for the optimized code, we need to know the optimization ratio of the compiler to be used by the designer to generate the machine instructions. The performance-optimization ratio α is defined as performance of the optimized code over the performance of the non-optimized code. The size optimization ratio, β , is defined similarly. To obtain the optimization ratio for each compiler, we have performed several experiments and obtained average optimization ratios for the four compilers used in our experiments (Figure 12).

Application	Target Machine	Actual instruction memory size (in bytes)	Estimated instruction memory size (in bytes)	Estimation Error
Elliptic filter	8086	336	319	-5.1%
Elliptic filter	80286	336	319	-5.1%
Elliptic filter	68000	230	214	-7.0%
Elliptic filter	68020	229	213	-7.0%
Differential equation	8086	151	143	-5.3%
Differential equation	80286	147	141	-4.1%
Differential equation	68000	106	98	-7.5%
Differential equation	68020	105	97	-7.6%

Figure 11: Program-memory size estimation

Compiler	Optimization ratio for performance	Optimization ratio for size
Cto8086	0.74	0.58
Cto80286	0.68	0.56
Cto68000	0.54	0.54
Cto68020	0.49	0.51

Figure 12: Optimization ratio of the compilers

After knowing the performance and size optimization ratio α and β , the estimates for the optimized code can be obtained by multiplying the estimates of the non-optimized code (obtained from our estimator) with α or β . Figure 13 compares our estimates with the actual performance of the optimized code compiled from the designs.

Application	Target Machine	Actual Performance (In clock cycles)	Estimated Performance (In clock cycles)	Estimation Error
Elliptic filter	8086	1984	1841	-7.2%
Elliptic filter	80286	483	450	-6.8%
Elliptic filter	68000	1020	881	-13.6%
Elliptic filter	68020	449	435	-12.9%
Differential equation	8086	9282	7834	-15.6%
Differential equation	80286	1690	1567	-7.3%
Differential equation	68000	3986	3484	-12.6%
Differential equation	68020	1970	1801	-8.6%

Figure 13: Performance estimation of optimized code

The next design we experiment is the real-time medical system used to measure a patient's bladder volume described in [Wu85]. The SpecChart description of this medical system is shown in Figure 14.

There are two timing constraints imposed on this medical system. One is associated with the behavior DATA_ACQUISITION, which requires that acquisition and conversion of 1000 data points take place in less than 1 ms. The other is associated with behavior ONE_SCAN, which requires that the maximum time between two scans, i.e. the time used to execute MOTOR_CONTROL2, DATA_ACQUISITION, VOLUME_COMPUTATION and DATA_STORAGE, is 1 second. We have estimated behavior DATA_ACQUISITION and behavior ONE_SCAN using our estimator. The estimates are compared with the actual results obtained from the (non-optimized) target machine instructions (Figure 15).

Since ONE_SCAN is a non-leaf behavior, we need to flatten it into a leaf behavior first. Following that the VHDL code is manually translated into C and compiled into the target machine instructions. For DATA_ACQUISITION, the flattening step is not needed since it is a leaf behavior by itself. During the translation from VHDL to C, wait statements were substituted with dummy procedure calls. When we computed the actual performance from the machine instructions, we substituted the performance of those dummy procedure call portions with the performance of the corresponding wait statement obtained from our estimator.

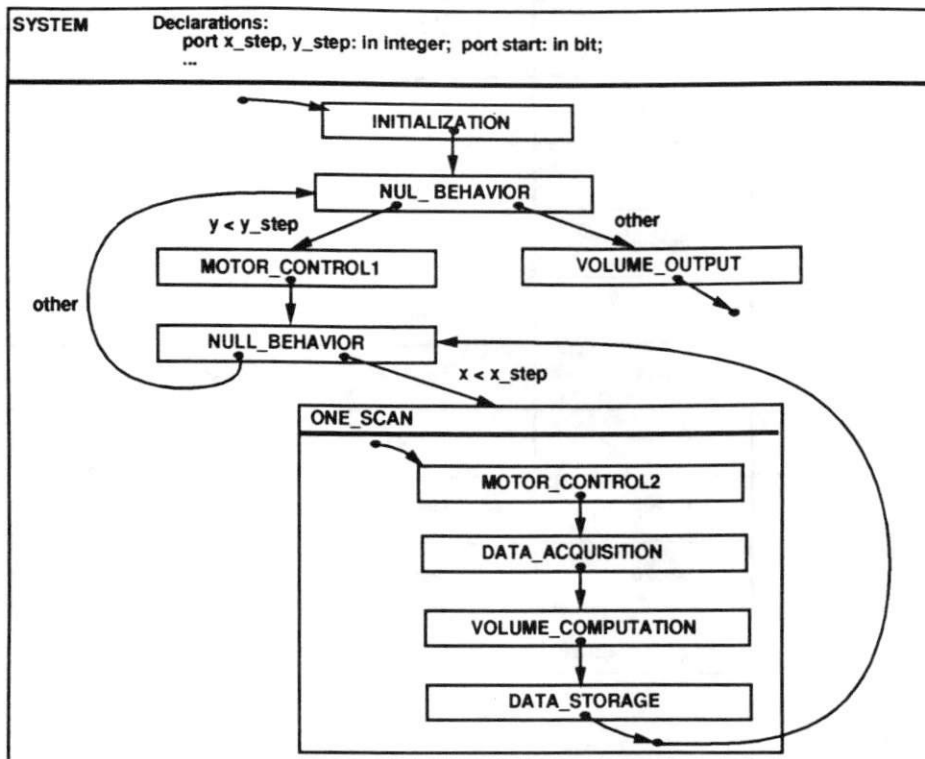


Figure 14: SpecChart of a medical system

In this experiment, equal branching probability are assumed for all the branches in the behaviors. The number of loop iteration in DATA_ACQUISITION were 1000 which is the number of data needed to be sampled in each scan. The loop iteration numbers in the VOLUME_COMPUTATION were set to 200 for computing the flat level of the bladder, 400 each for computing the anterior wall and the posterior wall of the bladder. The number of loop iteration in DATA_STORAGE were the same as number of data fetched, which is 1000.

If the behavior DATA_ACQUISITION is going to run on an 8086 microprocessor with 12 MHZ clock rate, the estimator predicated 71004 clock cycles which is equivalent to 5.9 ms in this case. Therefore the timing constraint (1 ms) imposed on DATA_ACQUISITION were violated. And thus custom hardware must be designed for this behavior. The timing constraint of 1 second imposed on the behavior of ONE_SCAN has not been violated since it only requires 12.8 ms (153641 clock cycles) to execute all behaviors in ONE_SCAN on the chosen microprocessor. If behavior DATA_ACQUISITION is extracted out and implemented by some faster design, the execution time for behavior ONE_SCAN can be expected to be less than 12.8 ms. Therefore all behaviors in ONE_SCAN except DATA_ACQUISITION can be implemented using the code running on the microprocessor.

Application	Target Machine	Actual Performance (In clock cycles)	Estimated Performance (In clock cycles)	Estimation Error
DATA ACQUISITION	8086	82015	71004	-13.4%
DATA ACQUISITION	80286	27056	24002	-11.3%
DATA ACQUISITION	68000	92089	88012	-4.4%
DATA ACQUISITION	68020	44044	41006	-6.9%
ONE_SCAN	8086	189914	153641	-19.1%
ONE_SCAN	80286	65931	55580	-15.7%
ONE_SCAN	68000	198665	175620	-11.6%
ONE_SCAN	68020	98982	84531	-14.6%

Figure 15: Performance estimation

6 Conclusion

In this report we have presented techniques for estimating from executable specification the performance and memory size of software code running on a given processor. The experiments has shown that our estimator has an average error of 7.4% and has a maximum error of 19.1% on designs spanning from straight line code (elliptic filter) to code with branches and loops (differential equation and DATA_ACQUISITION module) and even to hierarchical specification (ONE_SCAN module).

Since the generic three-address instructions and the technology files can only characterize the target machine instructions to some extent, there is always difference between the estimation obtained from our estimator and the results obtained directly by compiling to target machine instructions. Currently our technology files are primitive. We expect smaller estimation errors with more accurate technology files.

Another thing is that our generic instruction set has limited formats, especially in terms of memory addressing modes. If we can incorporate more memory addressing modes in our compiler to close the gap between the generic instructions and the target machine instructions, we can expect better estimation results. However this may increase the complexity of the generic instructions. Increasing complexity of the generic instructions may increase the compiling time and hence increase the whole estimation time. Therefore more studies are needed to investigate what makes a suitable generic instruction set.

In conclusion, our software estimator provides rapid feedback to the designers or partitioning too

and enable them to evaluate different design alternatives quickly. It takes 0.64, 0.33, 1.97 and 7.09 seconds on a Sun4 system to estimate the performance of the elliptic filter, differential equation, DATA_ACQUISITION and ONE_SCAN modules respectively for the 8086 technology file. On the contrary, it takes several days to manually compute the same information. Such instant feedback enables the designers or partitioning tools to rapidly explore larger design space, which may lead to faster and/or cheaper designs.

7 Acknowledgements

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A Technology Files

```
## Anything after '#' are comments.
## This is the technology file for 8086 processor.
## DirectMem means direct memory addressing.
## IndirectMem means indirect memory addressing.
```

#	OP	DESTINATION	SOURCE1	SOURCE2	time(clock cycles)	size(bytes)
	ALU	Register	Constant	Constant	4	3
	ALU	Register	Constant	Register	7	5
	ALU	Register	Register	Constant	6	6
	ALU	Register	Register	Register	5	4
	ALU	Register	DirectMem	Constant	17	7
	ALU	Register	Constant	DirectMem	18	6
	ALU	Register	DirectMem	Register	16	5
	ALU	Register	Register	DirectMem	16	5
	ALU	Register	DirectMem	DirectMem	27	6
	ALU	Register	IndirectMem	Constant	20	8
	ALU	Register	Constant	IndirectMem	21	7
	ALU	Register	IndirectMem	Register	19	6
	ALU	Register	Register	IndirectMem	19	6
	ALU	Register	IndirectMem	DirectMem	30	7
	ALU	Register	DirectMem	IndirectMem	30	7
	ALU	Register	IndirectMem	IndirectMem	33	8
	ALU	DirectMem	Constant	Constant	15	5
	ALU	DirectMem	Constant	Register	21	8
	ALU	DirectMem	Register	Constant	20	9
	ALU	DirectMem	Register	Register	19	7
	ALU	DirectMem	DirectMem	Constant	31	10
	ALU	DirectMem	Constant	DirectMem	32	9
	ALU	DirectMem	DirectMem	Register	30	8
	ALU	DirectMem	Register	DirectMem	30	8
	ALU	DirectMem	DirectMem	DirectMem	41	9
	ALU	DirectMem	IndirectMem	Constant	34	11
	ALU	DirectMem	Constant	IndirectMem	35	10
	ALU	DirectMem	IndirectMem	Register	33	9
	ALU	DirectMem	Register	IndirectMem	33	9
	ALU	DirectMem	IndirectMem	DirectMem	44	10
	ALU	DirectMem	DirectMem	IndirectMem	44	10
	ALU	DirectMem	IndirectMem	IndirectMem	47	11
	ALU	Register	Empty	Constant	4	3
	ALU	Register	Empty	Register	5	4
	ALU	Register	Empty	DirectMem	16	5
	ALU	Register	Empty	IndirectMem	19	6
	ALU	DirectMem	Empty	Constant	21	8
	ALU	DirectMem	Empty	Register	19	7
	ALU	DirectMem	Empty	DirectMem	30	8
	ALU	DirectMem	Empty	IndirectMem	33	9

MUL	Register	Constant	Constant	4	3
MUL	Register	Constant	Register	74	7
MUL	Register	Register	Constant	72	6
MUL	Register	Register	Register	135	4
MUL	Register	DirectMem	Constant	83	7
MUL	Register	Constant	DirectMem	79	7
MUL	Register	DirectMem	Register	146	5
MUL	Register	Register	DirectMem	146	5
MUL	Register	DirectMem	DirectMem	157	6
MUL	Register	IndirectMem	Constant	86	8
MUL	Register	Constant	IndirectMem	82	7
MUL	Register	IndirectMem	Register	149	6
MUL	Register	Register	IndirectMem	149	6
MUL	Register	IndirectMem	DirectMem	160	7
MUL	Register	DirectMem	IndirectMem	160	7
MUL	Register	IndirectMem	IndirectMem	163	8
MUL	DirectMem	Constant	Constant	15	5
MUL	DirectMem	Constant	Register	88	10
MUL	DirectMem	Register	Constant	86	9
MUL	DirectMem	Register	Register	149	7
MUL	DirectMem	DirectMem	Constant	97	10
MUL	DirectMem	Constant	DirectMem	93	10
MUL	DirectMem	DirectMem	Register	160	8
MUL	DirectMem	Register	DirectMem	160	8
MUL	DirectMem	DirectMem	DirectMem	171	9
MUL	DirectMem	IndirectMem	Constant	100	11
MUL	DirectMem	Constant	IndirectMem	96	10
MUL	DirectMem	IndirectMem	Register	163	9
MUL	DirectMem	Register	IndirectMem	163	9
MUL	DirectMem	IndirectMem	DirectMem	174	10
MUL	DirectMem	DirectMem	IndirectMem	174	10
MUL	DirectMem	IndirectMem	IndirectMem	177	11
DIV	Register	Constant	Constant	4	3
DIV	Register	Constant	Register	89	7
DIV	Register	Register	Constant	92	6
DIV	Register	Register	Register	164	4
DIV	Register	DirectMem	Constant	103	7
DIV	Register	Constant	DirectMem	94	7
DIV	Register	DirectMem	Register	175	5
DIV	Register	Register	DirectMem	175	5
DIV	Register	DirectMem	DirectMem	186	6
DIV	Register	IndirectMem	Constant	106	8
DIV	Register	Constant	IndirectMem	97	7
DIV	Register	IndirectMem	Register	178	6
DIV	Register	Register	IndirectMem	178	6
DIV	Register	IndirectMem	DirectMem	189	7
DIV	Register	DirectMem	IndirectMem	189	7
DIV	Register	IndirectMem	IndirectMem	192	8

DIV	DirectMem	Constant	Constant	15	5
DIV	DirectMem	Constant	Register	103	10
DIV	DirectMem	Register	Constant	106	9
DIV	DirectMem	Register	Register	178	7
DIV	DirectMem	DirectMem	Constant	117	10
DIV	DirectMem	Constant	DirectMem	108	10
DIV	DirectMem	DirectMem	Register	189	8
DIV	DirectMem	Register	DirectMem	189	8
DIV	DirectMem	DirectMem	DirectMem	200	9
DIV	DirectMem	IndirectMem	Constant	120	11
DIV	DirectMem	Constant	IndirectMem	111	10
DIV	DirectMem	IndirectMem	Register	192	9
DIV	DirectMem	Register	IndirectMem	192	9
DIV	DirectMem	IndirectMem	DirectMem	203	10
DIV	DirectMem	DirectMem	IndirectMem	203	10
DIV	DirectMem	IndirectMem	IndirectMem	206	11
CMP	Register	Constant	Constant	4	3
CMP	Register	Constant	Register	7	5
CMP	Register	Register	Constant	6	6
CMP	Register	Register	Register	5	4
CMP	Register	DirectMem	Constant	17	7
CMP	Register	Constant	DirectMem	18	6
CMP	Register	DirectMem	Register	16	5
CMP	Register	Register	DirectMem	16	5
CMP	Register	DirectMem	DirectMem	27	6
CMP	Register	IndirectMem	Constant	20	8
CMP	Register	Constant	IndirectMem	21	7
CMP	Register	IndirectMem	Register	19	6
CMP	Register	Register	IndirectMem	19	6
CMP	Register	IndirectMem	DirectMem	30	7
CMP	Register	DirectMem	IndirectMem	30	7
CMP	Register	IndirectMem	IndirectMem	33	8
CMP	DirectMem	Constant	Constant	15	5
CMP	DirectMem	Constant	Register	21	8
CMP	DirectMem	Register	Constant	20	9
CMP	DirectMem	Register	Register	19	7
CMP	DirectMem	DirectMem	Constant	31	10
CMP	DirectMem	Constant	DirectMem	32	9
CMP	DirectMem	DirectMem	Register	30	8
CMP	DirectMem	Register	DirectMem	30	8
CMP	DirectMem	DirectMem	DirectMem	41	9
CMP	DirectMem	IndirectMem	Constant	34	11
CMP	DirectMem	Constant	IndirectMem	35	10
CMP	DirectMem	IndirectMem	Register	33	9
CMP	DirectMem	Register	IndirectMem	33	9
CMP	DirectMem	IndirectMem	DirectMem	44	10
CMP	DirectMem	DirectMem	IndirectMem	44	10
CMP	DirectMem	IndirectMem	IndirectMem	47	11

MOV	Register	Empty	Constant	4	3
MOV	Register	Empty	Register	2	2
MOV	Register	Empty	DirectMem	13	3
MOV	Register	Empty	IndirectMem	16	4
MOV	DirectMem	Empty	Constant	15	5
MOV	DirectMem	Empty	Register	14	3
MOV	DirectMem	Empty	DirectMem	27	6
MOV	DirectMem	Empty	IndirectMem	27	6
MOV	IndirectMem	Empty	Constant	18	6
MOV	IndirectMem	Empty	Register	17	4
MOV	IndirectMem	Empty	DirectMem	30	7
MOV	IndirectMem	Empty	IndirectMem	33	8
NOP	Empty	Empty	Empty	3	1
CJUMP	Empty	Empty	Empty	16	2
JUMP	Empty	Empty	Empty	24	4
RET	Empty	Empty	Empty	17	3
CALL	Empty	Empty	Empty	37	4