Experiences Using the Parascope Editor: an Interactive Parallel Programming Tool

Mary W. Hall     Timothy J. Harvey
Ken Kennedy      Nathaniel McIntosh
Kathryn S. McKinley  Jeffery D. Oldham
Michael Paleczny   Gerald Roth

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Center for Research on Parallel Computation
Rice University
P.O. Box 1892
Houston, TX 77251-1892

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Experiences Using the ParaScope Editor:
an Interactive Parallel Programming Tool*†

Mary W. Hall‡ Timothy J. Harvey§ Ken Kennedy§ Nathaniel McIntosh§
Kathryn S. McKinley§ Jeffrey D. Oldham‡ Michael H. Paleczny§ Gerald Roth§

Abstract

The ParaScope Editor is an interactive parallel programming tool that assists knowledgeable users in developing scientific Fortran programs. It displays the results of sophisticated program analyses, provides a set of powerful interactive transformations, and supports program editing. This paper summarizes experiences of scientific programmers and tool designers using the ParaScope Editor. We evaluate existing features and describe enhancements in three key areas: user interface, analysis, and transformation. Many existing features prove crucial to successful program parallelization. They include interprocedural array side-effect analysis and program dependence view filtering. Desirable functionality includes improved program navigation based on performance estimation, incorporating user assertions in analysis and more guidance in selecting transformations. These results offer insights for the authors of a variety of programming tools and parallelizing compilers.

1 Introduction

The complexity of constructing parallel programs is a significant obstacle to widespread use of parallel computers. In the process of writing a parallel program, programmers must consider the implications of concurrency on the correctness of their algorithms. Ideally, an advanced compiler could free the programmer from this concern by automatically converting a sequential program into an equivalent shared-memory parallel program. Although a substantial amount of research has been devoted to automatic parallelization, such systems are not consistently successful [1, 9, 6, 30, 33]. When automatic systems fail, the ParaScope Editor (PED) assists in parallelizing Fortran applications by combining programmer expertise with extensive analysis and program transformations [3, 27, 28].

This paper reports on the utility of PED as an interactive parallel programming tool in the ParaScope parallel programming environment [10]. Through user evaluation of PED, we assess existing functionality and suggest enhancements. The critiques confirm the usefulness of tools like PED and offer new insights for compiler writers and parallel programming tool designers in three key areas.

User Interface. We describe a new user interface that was designed and implemented based on previous evaluations [17, 25]. This interface enhances the original [28], unifying improved navigation and viewing of the program, its dependences, and information about its variables. User evaluation confirms the appropriateness of the interface but exposes the need for new features such as navigation assistance based on performance estimation and a facility enabling users to assert information about variable values.

Program Analysis. Dependence analysis specifies a partial order of a program's statements that preserves its meaning [5, 15]. For safety, the compiler must assume a dependence exists if it cannot prove otherwise. In ParaScope, analysis of interprocedural and intraprocedural constants, symbolic array sections, and, if the precision of its dependence analysis [19, 27]. (We assume the reader is familiar with these analyses.) User evaluation confirms they are indispensable for discovering parallelism. In addition, more advanced symbolic analysis and exploiting user assertions in all analyses are essential when parallelizing many programs.

Transformations. Program transformations introduce, discover, and exploit parallelism without changing the meaning of the program. PED provides users with the same source-to-source transformations and accompanying analysis employed by ParaScope's compilers [8, 22], allowing users to apply transformations di-

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‡Center for Integrated Systems, Stanford University, Palo Alto, CA 94305. §Department of Computer Science, Rice University, Houston, TX 77251-1892. ¶ENSMP (CR1), 35, rue Saint Honoré, F-75305 Fontainebleau Cedex, France.
Table 1: Analyzed and Parallelized Programs.

<table>
<thead>
<tr>
<th>name</th>
<th>description &amp; contributor</th>
<th>lines</th>
<th>procedures</th>
</tr>
</thead>
<tbody>
<tr>
<td>spec77</td>
<td>weather simulation code Steve Poole, IBM Kingston &amp; Lo Hsih, IBM Palo Alto</td>
<td>5600</td>
<td>67</td>
</tr>
<tr>
<td>neoss</td>
<td>thermodynamics code Mary Zosel, Lawrence Livermore National Laboratory</td>
<td>350</td>
<td>5</td>
</tr>
<tr>
<td>nxsns</td>
<td>quantum mechanics code John Engle, Lawrence Livermore National Laboratory</td>
<td>1400</td>
<td>11</td>
</tr>
<tr>
<td>dpmin</td>
<td>molecular mechanics and dynamics program Marcia Pottile, Cornell Theory Center</td>
<td>5000</td>
<td>52</td>
</tr>
<tr>
<td>slab2d</td>
<td>2-D severe storm fluid flow prototype Roy Heimbach, National Center for Supercomputing Applications</td>
<td>350</td>
<td>9</td>
</tr>
<tr>
<td>slalom</td>
<td>benchmark program Roy Heimbach, National Center for Supercomputing Applications</td>
<td>1200</td>
<td>13</td>
</tr>
<tr>
<td>pueb3d</td>
<td>hydrodynamics benchmark program Ralph Brickner, Los Alamos National Laboratory</td>
<td>4000</td>
<td>30</td>
</tr>
<tr>
<td>arc3d</td>
<td>3-D hydrodynamics code Doreen Cheng, NASA Ames Research Center</td>
<td>3600</td>
<td>25</td>
</tr>
</tbody>
</table>

The contents of this paper are derived from the assistants' accounts and the discussions as well as talks by Stein and Fletcher on their experiences.

3 User Interface
3.1 Background

The user interface communicates PED's capabilities to programmers. Repeated evaluation of PED [17] and its predecessor Ptool [2, 25] by compiler writers, application programmers and human interface experts has led to its current design. Although many of the features in the new PED interface are found in the original [28], it improves upon the original by providing more consistent and powerful features.

PED employs three user interface techniques to manage and present the detailed information computed by the system: (1) a book metaphor with progressive disclosure, (2) user-controlled view filtering, and (3) power steering for complex actions. The book metaphor portrays a Fortran program as an electronic book with analysis results annotating the source, analogous to a book’s marginal notes, footnotes, appendices, and indices [37]. Progressive disclosure presents details incrementally as they become relevant [34]. View filtering emphasizes or conceals parts of the book as specified by a user [14]. Power steering automates repetitive or error-prone tasks, providing unobtrusive assistance while leaving the user in control.

The layout of a PED window is shown in Figure 1. The large area at the top is the source pane displaying the Fortran text. Two footnotes beneath it, the dependence pane and the variable pane, display dependence and variable information. All three panes operate similarly. Scroll bars can be used to bring other portions of the display into view, predefined or user-defined view filters may be applied to customize the display, the mouse may be used to make a selection, and menu choices or keyboard input may be used to edit the displayed infor-
<table>
<thead>
<tr>
<th>file</th>
<th>edit</th>
<th>view</th>
<th>search</th>
<th>dependence</th>
<th>variable</th>
<th>transform</th>
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</table>

![Figure 1: The ParaScope Editor.](image)

The Fortran text, the dependence information, and the variable information are each described below in terms of the information displayed, view filtering, and editing. We also discuss the work model best supported by PED and used during the workshop.

### Fortran Source Code

Fortran source code is displayed in pretty-printed form in the source pane. At the left are marginal annotations showing the ordinal number of each source line and asterisks (*) which indicate the start of each loop. The ordinal line numbers for the statements in the *current loop* are highlighted with color (not visible in Figure 1). Source view filtering can be used to highlight or conceal source lines, based on either the text or the underlying semantics of each line. For example, source view filter predicates can test attributes of a line such as if it contains certain text, if it is a loop header, or if it is erroneous.

The source pane allows arbitrary editing of the program using mixed text and structure editing techniques [36]. The user has the full freedom to edit character by character or to use the power steering afforded by template-based structure editing. Incremental parsing occurs in response to edits, and the user is immediately informed of any syntactic or semantic errors.

### Dependence Information

The editor reveals program dependences in a tabular list in the dependence
The display shows each dependence’s source and sink variable references and characteristics such as dependency distance. Ped uses progressive disclosure of dependencies based on the current loop. When the user expresses interest in a particular loop by selecting its marker in the source pane, the selected loop’s dependencies immediately appear in the dependence pane. The user may select a range of dependencies in the dependence pane to examine that are then highlighted. Simultaneously, the source pane graphically displays selected dependencies using red arrows from source variable references to sink variables. In Figure 1, the user has selected a single dependence to be displayed, which is reflected in the source and dependence panes.

Dependence view filter predicates can test the computed and user-controlled attributes of a dependence, such as its source and sink variable references and line numbers, its type, loop nesting level, mark and reason. The filtered dependence list may then be displayed graphically all at once, one dependence at a time, or according to line numbers in the source.

The dependence pane also permits an important form of editing known as dependence marking. The system marks each dependence as either proved, pending, accepted or rejected. If Ped proves a dependence exists with an exact dependence test [19], the dependence is marked as proved; otherwise it is marked pending. Users may sharpen Ped’s dependence marking by marking a pending dependence as accepted or rejected. Rejected dependencies are disregarded when Ped considers the safety of a parallelizing transformation, but they remain in the system so the user can reconsider them at a later time. If desired, the user can edit the dependence’s reason attribute to attach a comment explaining the classification decision. A Mark Dependencies dialog box provides power steering for dependence marking by allowing the user to classify in one step an entire set of dependences that satisfy a chosen predicate.

Variable Information. The variable pane displays a list of variables with progressive disclosure similarly based on the current loop. The attributes of a variable include its name, definitions and uses outside the current loop, dimension, common block if any, and shared or private status. View filter predicates can test any of these attributes.

The variable pane also supports editing in the form of variable classification. Locating variables that may legally be made private to a loop body can greatly reduce the number of spurious dependences in a loop (see Section 4.1). Analysis automatically locates many such variables, but it is conservative and may classify variables as shared that may legally be made private. With variable classification, the user may edit a variable’s shared/private attribute to correct overly conservative classifications for arrays and scalars. As with dependence editing the user may also edit a variable’s reason field, attaching a comment regarding the variable’s classification. A Classify Variables dialog box (analogous to Mark Dependences) provides power steering for variable classification.

Work Model. The following work model was used by the workshop participants. Users select a loop for consideration and examine any parallelism inhibiting dependences. If they are the result of overly conservative assumptions, the user employs dependence deletion and variable classification to increase the precision of analysis. If necessary, they perform transformations to expose parallelism. Ped supports this model by displaying the dependences and variables associated with a selected loop and relating them back to the source code. During the workshop, the attendees augmented this work model with program execution profiles to help them focus on the most computationally intensive loops in their program.

3.2 Evaluation

This section describes the evaluation of Ped’s user interface, as summarized in Table 2. The table lists a number of features in its rows and responses in the columns. Each of the five workshop groups, along with Fletcher and Stein, is represented by an asterisk, for a total of seven possible asterisks. Because the implementation of the interface described above was not robust at the time of the workshop, it was demonstrated but not used by the participants. Here, we address only issues that apply to both interfaces and are not addressed in the current implementation. This structure allows us to focus on user suggestions for improvement.

User Interaction. Users were comfortable with selecting a loop and then evaluating its dependences and variable annotations. Specifically, one user commented that having all information available in one window annotating the source code was preferable to the functional-based approach of Forge [31], which requires the programmer to bring up a series of windows to access the results of program analysis.
To correct overly conservative analysis, users performed both dependence deletion and variable reclassification to reflect their perception of the true program state. Variable reclassification proved to be the right level of abstraction. When users were able to determine a variable could be made private to a loop, changing the variable's classification and seeing the resulting decrease in dependences was effective. However, deleting individual dependences was too tedious for most users. One user disliked dependences as a paradigm and desired a higher level of abstraction based on variables. We discussed simplifying dependence deletion by introducing variable-based user assertions that the system would then use to automatically eliminate dependences (see Section 3.3).

Three users requested more access to all the analysis computed by the system in order to enhance their understanding of the program and provide insight into the reasons the system must assume dependences exist. For instance, one user felt that displaying array sections that summarize the portion of an array accessed by a loop, similar to those used in interprocedural analysis (see Section 4.1), would provide an intuitive representation of the accesses within the loop. Two other users requested that the system display constant values or ranges for loop bounds and subscript expressions. In general, they wanted access to any information the system uses to compute dependences in order to correct overly conservative analysis.

**Program Navigation.** Existing features in PED assist in navigating the source code. They include view filtering and annotations in the display. Source view filtering was not widely used during the workshop primarily because it was unfamiliar to the assistants. However, one group defined filters based on labels to help them understand a procedure's control flow (see Section 5.3). One user commented that they would have liked to apply view filtering to examine the procedure's loop structure; this feature is one of the predefined view filters. Another requested that the display annotations include a loop nesting depth indicator.

All users requested more assistance in locating the most computation-intensive procedures and loops in the program in order to better target their parallelization efforts. For this purpose, the users relied on external tools to profile their codes. They used the Unix utility `aprof` from Sun and Cray executions to derive procedure-level profiles. One user brought loop-level profiling available from Forge [31], which shows the loop structure of a procedure, with the number of iterations executed and its percentage of execution time. The users requested that similar profiling or static performance estimation be integrated into PED to help focus user attention on the loops where effective parallelization would have the highest payoff. ParaScope now includes a static performance estimator used to predict the relative execution time of loops and subroutines in parallel programs [26].

Several users wanted a graphical representation of the call graph, rather than the current textual presentation [21]. A visual program representation provides a much needed "big picture" when working with a large or unfamiliar program. Short cuts to accessing other procedures via this representation were also requested.

**Dependence Navigation.** In navigating dependences, the users requested a number of features subsumed by the current implementation. First, they needed to visit dependence endpoints quickly rather than having to scroll through the source. They wanted view filtering capabilities to examine in the source and dependence panes all dependences associated with a particular definition. They were comfortable with use-definition chains and requested a filter mechanism for viewing them.

The users also wanted additional navigational features to examine dependences with at least one endpoint outside the current loop, in another loop or even another procedure. Displaying dependences with endpoints outside the selected loop requires straightforward extensions to PED. For example, if a variable is used or defined outside the current loop, the variable display reveals that information. To find the reference(s) at the workshop, the users were forced to search the program. Now the user need only select the line number of interest in the USE or DEF field (see Figure 1), and the appropriate reference appears in the source pane.

Dependence endpoints spanning multiple procedures pose a more difficult problem. Because interprocedural side-effect information summarizes accesses, one endpoint may correspond to several reads or writes of an array in multiple procedure bodies. To fully support dependence navigation, PED must be able to display other procedures while iterating over all the endpoints corresponding to a dependence.

**Other.** Another ParaScope tool, the Composition Editor, compares a procedure definition to calls invoking it, ensuring the parameter lists agree in number and type [11, 21]. These types of errors exist in production codes because most compilers do not perform cross-procedure comparisons. Several matched parameters between a procedure call and its declaration as well as type errors were detected and subsequently corrected using this analysis. One user requested further analysis to ensure that common block declarations have the same shape in every procedure in which they appear and to perform static array bounds checking.

Two participants were eager to use PED as a teaching tool for parallel programmers at their institutions. They felt PED's tutorial nature could help explain parallel programming to novices. Two users found the interactive help facility useful, but one wanted it to provide more information. One user wanted the ability to print the program, dependences, and variable information.
3.3 Obstacles to Parallelization

User Assertions. Explicit deletion of dependences fails to capture the user’s reason for eliminating a dependence. A user deletes a dependence or a group of dependences because of some additional knowledge about the program that the system is unable to detect statically. This information is usually at a higher level than a specific dependence. Users would prefer to specify a high-level assertion and then have the system respond by deleting associated dependences. They proposed adding a user assertion language.

To support user assertions, we first need to define an assertion language based on three important concepts. (1) Assertions should express program properties that are natural to a user. (2) Assertions should provide information to the system that is useful in eliminating dependences. (3) It should be possible for the system to verify the correctness of the assertions at run time.

For example, the assertion language could enable users to specify the value or range of values of a variable using familiar Fortran syntax. Two simple but very useful assertions encountered during evaluation were specifying relationships between two symbolic variables and the properties of index arrays, arrays used in subscript expressions of another array. Consider the following fragment from _pueblo3d_.

```fortran
DO M = 1, 3
   MCN = ICM(M, IR)
   DO I = ISTRT(IR), IENDV(IR)
   ... = UF(I + MCN, 3)
   UF(I, M) = ... 
ENDDO
```

This construction appears in 10 loop nests in _pueblo3d_ and several of these consume the majority of the total execution time. _MCN_ represents “my current neighbor” and is used to index linearized three-dimensional arrays. Relative to the above loop structure, this program ensures that _MCN > IENDV(IR) - ISTRT(IR)_ and therefore, there are no loop-carried dependences on _UF_. As in this case, arriving at an appropriate and useful assertion may require some careful thought, but the system will be offering some assistance as well (see Section 4.3).

4 Program Analysis

4.1 Background

Discovering and evaluating potential parallelism in a program requires extensive program analysis. In particular, dependence analysis provides the fundamental paradigm for detecting parallelism. A loop that contains no loop-carried dependences may execute its iterations in parallel. Further, dependences are used to prove the safety of program transformations. This section briefly describes the dependence analysis and supporting analyses that are currently available in _PED_ [27].

Dependence Analysis. _PED_ detects data and control dependences. _Data dependences_ are located by testing pairs of references in a loop. A hierarchical suite of tests is used, starting with inexpensive tests, to prove or disprove that a dependence exists [19]. _Control dependences_ explicitly represent how control decisions affect statement execution [15].

Supporting Analysis. Scalar data-flow analysis, including def-use chains, constant propagation and symbolic analysis, provides additional information about the values and relationship of variables. They can vastly improve the precision of dependence analysis [19, 20]. Def-use chains expose dependences among scalar variables as well as linking all accesses to each array for dependence testing. A critical contribution of scalar data-flow analysis is recognizing scalars that are killed, or redefined, on every iteration of a loop and may be made private, thus eliminating dependences. Constant propagation can locate constant-valued loop bounds, step sizes and subscript expressions. Symbolic analysis locates auxiliary induction variables, loop invariant expressions and equivalent expressions. It also performs expression simplification on demand.

Interprocedural Data-flow Analysis. One of the distinguishing features of _Ped’s_ dependence information is the incorporation of an extensive suite of interprocedural analysis techniques that determine the effects of procedure calls on variables. Interprocedural constants are inherited from a procedure’s callers and directly incorporated into the intraprocedural constants. _Flow-insensitive side-effect analysis_, including _MOD_ and _REF_ analysis, describes the variables that may be accessed on some control flow path through the procedure [4]. _Flow-sensitive side-effect analysis_, such as _KILL_ analysis, describes accesses that occur on every possible control flow path [7]. _Regular section analysis_ is also used to describe more precisely, when possible, the side-effects to portions of arrays [24].

4.2 Evaluation

Table 3 demonstrates the importance of existing analysis and the need for additional analysis for the programs described in Table 1.

Dependence Analysis. The _dependence_ entry in Table 3 indicates whether dependence analysis locates parallel loops in each program. For all of the programs, the system is able to automatically detect many parallel loops. Small, inner parallel loops are almost always detected. However, outer loop parallelism is essential to achieving measurable performance improvements for applications programs on many parallel architectures, and it too often goes undetected. We discuss impediments and solutions for parallel loop detection in Section 4.3.

Scalar _Kill_ Analysis. As illustrated by the _scalar kills_ entry, almost all of the programs contain a loop...
<table>
<thead>
<tr>
<th>dependence</th>
<th>spec77</th>
<th>nessa</th>
<th>nxns</th>
<th>dpmin</th>
<th>slab2d</th>
<th>slalom</th>
<th>pueblo3d</th>
<th>arc3d</th>
</tr>
</thead>
<tbody>
<tr>
<td>scalar kills sections</td>
<td>U</td>
<td>U</td>
<td>U</td>
<td>U</td>
<td>U</td>
<td>U</td>
<td>U</td>
<td>U</td>
</tr>
<tr>
<td>array kills reductions</td>
<td>N</td>
<td>N</td>
<td>N</td>
<td>N</td>
<td>N</td>
<td>N</td>
<td>N</td>
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</tr>
<tr>
<td>index arrays</td>
<td>N</td>
<td>N</td>
<td>N</td>
<td>N</td>
<td>N</td>
<td>N</td>
<td>N</td>
<td>N</td>
</tr>
</tbody>
</table>

*U*: existing analysis was used. *N*: additional analysis was needed.

Table 3: Analysis Used or Needed During Workshop.

that becomes parallelizable following scalar privatization. In the program nxns, interprocedural scalar Kill analysis reveals a scalar variable is killed in a procedure invoked inside a loop. Experience using Prool, Ped's predecessor, also suggests interprocedural scalar Kill analysis is useful in eliminating spurious dependencies [25].

**Interprocedural Side-effect Analysis.** The sections entry indicates that scalar side-effect analysis or regular section analysis reduces the number of dependences on a loop containing a procedure call in six of the programs. Of the two remaining programs, one does not contain loops with procedure calls and analysis failed on the other. In spec77 and nxns, interprocedural side-effect analysis reveals that loops containing procedure calls can safely execute in parallel.

### 4.3 Obstacles to Parallelization

There are several areas where existing analysis in Ped is not sufficient to detect parallelism, but the users together with the workshop assistants were able to discover it. In many cases, more precise analysis can detect and eliminate overly conservative dependences. In others, static analysis will probably never be sufficient and user assertions are needed.

**Array Kill Analysis.** For loops in seven of the programs, array kill analysis would eliminate important dependences, revealing parallelism. In slab2d and arc3d, automatic privatization of one or more killed arrays is sufficient to prove that loops may be safely executed in parallel. Frequently, a temporary array is assigned and used in an inner loop and its value does not carry across iterations of the outer loop. In arc3d, an array is killed inside a procedure invoked in a loop, so interprocedural array kill analysis is required. To perform array privatization in slab2d, kill analysis must be combined with loop transformations. Because the need for array kill analysis has been discussed previously [6, 33], we do not elaborate further here.

**Reductions.** Five of the programs contain sum reductions which go unrecognized by Ped. For example, computing the sum of all the elements of an array. Because addition is associative, the additions do not need to be performed in order and so the loop can be parallelized after restructuring the accumulation of the sum. The need and methods for recognizing reductions are well known and we do not elaborate here. However, transforming reductions in an interactive setting is complicated by the property that efficient execution requires an architecture-specific approach.

**Symbolic Expressions.** Static analysis cannot derive information about certain symbolic expressions, such as variables read from an input file or index arrays used in subscript expressions. Symbolic terms in subscript expressions are a key limiting factor in precise dependence analysis. One study found that over 50% of the array references in some numerical packages contained at least one unknown symbolic term [20]. The index arrays entry in Table 3 demonstrates that three programs contained index arrays in subscript expressions that prevented parallelization.

We are using a three-pronged approach to improving the precision of dependence information in the presence of symbols: (1) sophisticated symbolic analysis; (2) partial evaluation, or compiling the program with all or part of an input data set [18]; and (3) incorporating user assertions to eliminate dependences (described from the user interface perspective in Section 3.3).

The following program fragment from the routine filter3d in arc3d demonstrates the type of advanced interprocedural symbolic analysis that would improve program parallelization.

```plaintext
DO 15 K = 1, N
    DO J = 1, JM
        WR1(J,K) = Q(JPL,K,L,N)-Q(J,K,L,N)
    CONTINUE
    DO 76 K = 2, KM
        WR1(JMAX,K) = WR1(JMAX,K)
    CONTINUE
    WR1(J,K) = WR1(J,K)
    CONTINUE
```

In the initialization routine, the assignment JM = JMAX - 1 occurs, and this relation holds for the rest of the program. Given this symbolic relationship and array kill analysis, the DO 15 loop may be safely parallelized by privatizing WR1 and two other arrays. The
ability to detect and propagate this type of relationship can greatly improve the precision of dependence analysis.

In other cases, symbolic values are read, making it virtually impossible for static analysis to determine actual dependences. Consider the following fragment from the program dpmin.

\[
\begin{align*}
\text{DO } 300 & \quad \text{N} = 1, \text{ NBA} \\
13 & = \text{IT}(N) \\
15 & = \text{JT}(N) \\
16 & = \text{KT}(N) \\
\vdots \\
F(13 + 1) & = F(13 + 1) - D_{T1} \\
F(13 + 2) & = F(13 + 2) - D_{T2} \\
F(13 + 3) & = F(13 + 3) - D_{T3} \\
F(13 + 4) & = F(13 + 4) - D_{T4} \\
F(13 + 5) & = F(13 + 5) - D_{T5} \\
F(13 + 6) & = F(13 + 6) - D_{T6} \\
F(13 + 7) & = F(13 + 7) - D_{T7} \\
F(13 + 8) & = F(13 + 8) - D_{T8} \\
F(13 + 9) & = F(13 + 9) - D_{T9} \\
\end{align*}
\]

The arrays IT(N), JT(N) and KT(N) are read from a file, so the system assumes dependences connect all references to F.

To assist the user in deriving assertions that eliminate spurious dependences, the system may be able to derive breaking conditions that eliminate a particular dependence or class of dependences. In the above, a breaking condition for loop-carried dependences between instances of F(13+1) is that IT(N) is a permutation array (i.e., all values for I3+1 are unique). While possible, it would take significantly more analysis for the system to derive breaking conditions to eliminate all dependences on F and parallelize the loop. The system must recognize that if IT(N) is a function satisfying the constraint IT(I) + 3 ≤ IT(I+1) (similarly for JT and KT), IT(NBA) + 3 ≤ JT(I) and JT(NBA) + 3 ≤ KT(I), all dependences may be eliminated.

A similar approach is being pursued by Pugh and Wonnacott [32]. They derive relational constraints on variables during dependence testing using a variant of integer programming, and these are presented to the user in their implementation.

5 Program Transformation

5.1 Background

Ped supports a large set of transformations proven useful for introducing, discovering, and exploiting parallelism and for enhancing memory hierarchy use [27]. Figure 2 shows a taxonomy of the transformations available in Ped. Transformations are applied according to a power steering paradigm: the user specifies the transformations to be made, and the system provides advice and carries out the mechanical details. The system advises whether the transformation is applicable (is syntactically correct), safe (preserves the semantics of the program) and profitable (contributes to parallelization). The complexity of many transformations makes correct application difficult and tedious by hand. Thus, power steering provides safe, profitable and correct application of transformations and incremental updates of dependence information to reflect the modified program.

Reordering transformations change the order in which statements are executed, either within or across loop iterations. They expose or enhance loop-level parallelism and improve data locality. Dependence breaking transformations eliminate storage-related dependences that inhibit parallelism. They often introduce new storage and convert loop-carried dependences to loop-independent dependences. Memory optimizing transformations expose reuse of memory locations in registers or cache.

5.2 Evaluation

Table 4 lists the transformations used for parallelizing each program. The rows describe the existing transformations used and the additional transformations needed.

It is notable that only a few of Ped’s transformations were used. The most commonly used transformation was scalar expansion, which transforms a scalar into an array to eliminate loop-carried dependences. Loop unrolling was the only other transformation used more than once. For the most part, the transformations in the table are commonly used in vectorization; however, in one example loop interchange of an imperfect loop nest was required.

As compared with the previous section, users were much better at reproducing analysis not provided by the system than at determining which transformations could improve parallelization. The users and their assistants mentioned that selecting among the large number of transformations is too overwhelming. It was not clear to them which transformations to attempt for a given loop nest.

5.3 Obstacles to Parallelization

To assist users in program parallelization, they requested more automated assistance for applying transformations and two additional transformations.

Transformation Guidance. While few transformations were performed on these programs, it was not be-
cause opportunities for transformation did not exist. Indeed, a substantial amount of research has demonstrated the value of loop transformations in exposing parallelism. Cursory re-examination of the programs reveals opportunities for many transformations that enable parallelization of outer loops. These opportunities include fusion and interchange in pueblo3d and distribution in dpmin and neoss. However, when the users were confronted with the selection of transformations, they did not know which ones to explore.

Several users want the transformation selection to include only those which are safe and profitable for the currently selected loop. This structure would save them from sifting through the entire list of transformations for each loop. As a simple extension to the current system, it could evaluate the safety of all the transformations for a particular loop on demand and present only the safe ones. However, determining what transformations are profitable is much more difficult. Profitability not only depends on machine specifics, but on subsequent transformations.

According to the users, transformation advice should incorporate the compiler’s parallel code generation algorithms for a particular architecture. Ideally, a user would select the architecture and request parallelization at the loop, subroutine or program level. The system would then automatically perform parallelization or describe the impediments to a desired parallelization. Impediments would be presented in a systematic fashion based on the relative importance of a loop or subroutine. The user could evaluate any impediments and correct overly conservative assumptions, thus enabling semi-automatic parallelization. Several users stressed the importance of providing consistent analysis and parallelization algorithms between the compiler and interactive tool.

**Complex Control Flow.** Three programs, neoss, nxsns and dpmin, were written in dialects of Fortran that do not support structured if statements and that require do loops to execute at least one iteration regardless of loop bounds. Possibly to compensate for constructs lacking in the language, programmers introduced complex control flow involving goto statements. Consider the following loop with gotos from neoss and the structured equivalent produced by hand during the workshop.

```plaintext
DO 50 K= . . . 
  IF (DERV(K) - RES(RR+1)) 100, 10, 10 
  10 CONTINUE
  GOTO 101
  100 <b3>
  101 <b4>
  50 CONTINUE

DO 50 K= . . . ü structured version
  IF (DERV(K) .GE. RES(RR+1)) THEN
    ELSE
    ENDIF
    <b4>
  50 CONTINUE
```

The gotos make it difficult for the users to understand the original loop. Other variations of if-then-else constructs formed with gotos also appear in these programs. However, users were able to further transform and parallelize a loop of this sort after control flow was simplified by hand.

To assist users in this process, the simplification of complex control flow can be automated by recognizing and substituting structured idioms for unstructured control-flow when appropriate. The need for this transformation is unique to an interactive setting. It is not necessary in completely automatic systems or internally in interactive tools because control dependence suffices to understand control flow regardless of the language constructs.

**Interprocedural Transformations.** The program spec77 has a number of loops containing procedure calls in the key procedure gloop. Interprocedural analysis indicates that the loops may be safely parallelized, but the loops have at most twelve iterations, limiting the number of possible parallel threads. The procedures invoked in these loops however contain outer loops with
many more iterations that may also safely execute in parallel. A solution that combines the granularity of the outer loop with the parallelism of the loop in the procedure is to perform loop interchange across the procedure boundary [23]. In some cases, loops in gloop contained multiple calls so the loops of the called procedures were first fused before applying interchange.

In order to enable transformations such as loop interchange and loop fusion across procedure boundaries, we must be able to move a loop into or out of a procedure invocation. We call these interprocedural transformations loop embedding and loop extraction, respectively [23]. Steve Poole brought spec77 to the workshop because he was familiar with this work and wanted to perform these transformations in PED. Embedding and extraction are not currently implemented in PED.

6 Related Work
A few other papers report on the effectiveness of existing automatic parallelizing compilers on large applications [9, 6, 12, 13, 33] and of interactive tools [9]. Two of these suggest compiler-programmer interaction to achieve parallelization [9, 33].

Blume and Eigenmann explore the effectiveness of KAP [29] applied to Perfect benchmark programs on an 8-processor Alliant [6]. Half of the programs demonstrate little or no improvement following parallelization. Using the transformation algorithms in KAP, scalar expansion is the only transformation that consistently improved performance on several codes. The compiler often fails to parallelize important loops, such as loops containing procedure calls, and sometimes parallelizes loops with insufficient granularity. These failures are in part because the compiler does not perform interprocedural analysis.

Eigenmann et al. present novel approaches for improving parallelization in four Perfect programs [12, 13]. These techniques include run-time dependence testing and aggressive use of synchronization to guard critical sections.

Singh and Hennessy examine parallelization of three programs on an 8-processor Alliant [33]. They observe that certain programming styles interfere with compiler analysis. These include linearized arrays and specialized use of the boundary elements in an array. To aid the compiler in selecting appropriate loops to parallelize, they suggest user access to profiling information and assertion facilities that allow specifying ranges of symbolic variables.

Cheng and Pase consider 25 programs running on an 8 processor Cray Y-MP, using Cray fpp, KAP/Cray and Forge to introduce parallelism [9]. Most of the parallel versions demonstrate a speedup of less than 2 over their vector counterparts. When using Forge, the only interactive tool, they offer two suggestions. First, the user should be given insight about what loops to parallelize, either through profiling or performance estimation. Second, they want the system to query for unknown scalar variable values and use these assertions in analysis to eliminate dependences.

Because PED contains some of the features recommended by these studies, our evaluation reveals how they work in practice. For example, most of these studies find interprocedural analysis to be essential, but missing from the compilers under investigation. In PED, interprocedural analysis is found effective, but more advanced analysis such as interprocedural symbolic propagation is also needed.

Our evaluation is also distinguished because it examines the interactive parallelization process with outside users. The existence of an advanced interactive tool allowed us to go beyond the comments provided by the studies of automatic parallelizers to investigate the appropriate level for compiler and tool interaction. For instance, users deleted dependences in PED, as suggested previously [25, 33], but requested higher-level assertions and guidance. We also uncovered the need for new features such as access to compiler transformation algorithms and control flow simplification.

7 Conclusion
In an interactive system, the combination of user expertise with the sophisticated analysis and transformations used by parallelizing compilers requires powerful mechanisms and careful engineering. Through user evaluation of the ParaScope Editor, an interactive parallel programming tool, we have established several essential user requirements. Users prefer that the user interface tie the compiler analysis to the source code and provide facilities such as view filtering of source code and analysis results, navigational assistance, and transformation guidance. To assist users in refining program analysis for use in compilation, a facility that communicates high-level information to the tool in the form of assertions is needed. Advanced analysis, such as interprocedural array section analysis, symbolic analysis and array privatization, can substantially reduce the amount of work the user must perform. By continuing extensive evaluations, improvements and additions, the ParaScope Editor aspires to meet user requirements.

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References