A ROSE-Based End-to-End Empirical Tuning System for Whole Applications
Draft User Tutorial

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ROSE Tutorial (pdf)
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## Contents

1 Introduction 3

2 Preparation 5
   2.1 Using HPCToolkit ............................................ 5
   2.2 Using gprof .................................................... 5

3 Code Triage and Transformations 7
   3.1 Invoking Code Triage ........................................... 7
   3.2 Tool Interface ................................................ 8
   3.3 Kernel Extraction: Outlining .................................. 9
   3.4 Transformations for Autotuning ............................... 11

4 Empirical Tuning 15
   4.1 Parameterized Transformation Tools .......................... 15
   4.2 Search Engines ................................................. 17
   4.3 An Example Search ............................................. 19

5 Working with Active Harmony 20

6 User-Directed Autotuning 23
   6.1 Manual Code Triage ............................................ 23
   6.2 Parameterized ROSE Loop Translators ........................ 25
   6.3 Connecting to the Search Engine .............................. 26
   6.4 Results ....................................................... 28

7 Summary 28

8 Appendix 29
   8.1 Patching Linux Kernels with perfctr ........................ 29
   8.2 Installing BLCR ............................................... 30
1 Introduction

This document describes an on-going project aimed for a ROSE [10]-based end-to-end empirical tuning (also called autotuning) system. It is part of an SciDAC PERI [11] project to enable performance portability of DOE applications through an empirical tuning system. Our goal is to incorporate a set of external tools interacting with ROSE-based components to support the entire life-cycle of automated empirical optimization. We strive to keep this document up-to-date for better communication among project participants. This document is not meant to reflect the final design or implementation choices.

Currently, the ROSE-based autotuning system (shown in Fig. 1) is designed to work in three major phases.

1. Preparation
   The preparation phase uses external performance tools to collect basic performance metrics of a target application.

2. Code triage, kernel extraction and transformations
   This phase is carried out by a set of ROSE-based modules. A ROSE tool interface module reads in both source files of the application and the performance data to construct an abstract syntax tree (AST) representation of the input code annotated with performance information.

   Then a code triage module is followed to locate problematic targets (e.g. loops) within the application. A set of potentially beneficial optimizations and/or their configurations for each target is chosen (manually for now) based on program analysis.

   After that, a ROSE AST Outliner extracts a selected target into a stand-alone kernel, which will in turn be compiled into a dynamically loadable library routine. The application will be transformed accordingly and compiled to be a binary executable. This binary executable calls the...
outlined routine, collects performance metrics for the call. Optionally, a checkpointing/restarting library can be used to shorten the execution by stopping (checkpointing) and restarting immediately before calling the outlined function.

3. **Empirical tuning**

The final phase does the actual empirical tuning. First of all, the potentially beneficial optimizations and their corresponding configurations are represented as points within an integral search space which can be handled by a search engine. The search engine adopts some search policy to evaluate points in the search space and search for a point corresponding a transformation strategy leading to the best performance.

During this phase, multiple versions of the target kernel are generated by a parameterized translator from the searched points and compiled into dynamically loadable library routines. The performance of the kernel variants are measured one by one as the checkpointed binary is restarted multiple times and calls multiple versions of the dynamically loadable library routine.

We give some details about the system design and the current implementation status in the following sections.

A list of our current hardware/software configurations is given below:

- A Dell Precision T5400 workstation with two sockets, each a 3.16 GHz quad-core Intel Xeon X5460 processor, and total 8 GB memory;
- Red Hat Enterprise Linux release 5.3 (Tikanga) Linux x86 kernel 2.6.18-92.el5.perfctr SMP PREEMPT;
- PAPI 3.6.2;
- the Rice HPCToolkit version TRUNK-4.9.0=1280 (the latest release does not support the 32-bit machine we use, so we use an older version);
- ROSE compiler version 0.9.4a, revision 6701 (providing a tool interface, outliner, loop translator, etc.);
- Berkeley Lab Checkpointing and Restarting library V. 0.8.2;
- POET from Univ. Texas San Antonio (We use its latest CVS version actually, not sure the exact release version number);
- the GCO search engine from Univ. of Tennessee (UTK) (We got a package from UTK directly, not sure the release number);
- a C version jacobi iteration program, used as a simple example input code for autotuning.
- the SMG2000 [5] (Semicoarsening Multigrid Solver) benchmark from the ASCI Purple, used as an example real application.
2 Preparation

The preparation phase provides basic information about a target application’s performance characteristics. Such information can be obtained by many performance tools. Currently, we accept performance data generated by both HPC-Toolkit and GNU gprof.

2.1 Using HPCToolkit

The HPCToolkit [3], developed at the Rice University, is an open source profile-based performance analysis tool which samples the executions of optimized applications. No code instrumentation is needed to use HPCToolkit. But debugging information (by compiling with the -g option if GCC is used) in the binary executables is needed for the tool to associate performance metrics with source language constructs.

After installation, a typical session of using HPCToolkit is given below:

```bash
% Prepare the executable with debugging information
gcc -g smg2000.c -o smg2000
% Sample one or more events for the execution, use wall clock here
hpcrun -e WALLCLK -- ./smg2000 -n 120 120 120 -d 3
% Convert the profiling result into a XML format
hpcproftt -p -D /home/liao6/svnrepos/benchmarks/smg2000 ./smg2000 \
smg2000.WALLCLK.tux268.llnl.gov.10676.0x0 > result.xml
```

Fig. 2 shows the profiling results of SMG2000 using HPCToolkit. A statement in a loop takes more than 46% execution time, which makes the loop dominant, most expensive loop of the entire program.

2.2 Using gprof

GNU gprof can generate line-by-line performance information for an application. A typical session to generate such information is given below:

```bash
[liao@codes]$ gcc -g seq-pi.c -pg
[liao@codes]$ ./a.out
[liao@codes]$ gprof -l -L a.out gmon.out &>profile.result
```

The option -l tells gprof to output line-by-line profiling information. -L causes gprof to output full file path information, which is needed for ROSE to accurately match performance data to source code.

An excerpt of an output file for smg2000 looks like the following:

```text
Flat profile:
Each sample counts as 0.01 seconds.

<table>
<thead>
<tr>
<th>% cumulative</th>
<th>name</th>
</tr>
</thead>
<tbody>
<tr>
<td>time</td>
<td>seconds</td>
</tr>
<tr>
<td>35.01</td>
<td>13.08</td>
</tr>
<tr>
<td>9.06</td>
<td>16.46</td>
</tr>
<tr>
<td>8.40</td>
<td>19.60</td>
</tr>
<tr>
<td>7.67</td>
<td>22.46</td>
</tr>
<tr>
<td>5.97</td>
<td>24.70</td>
</tr>
<tr>
<td>5.27</td>
<td>26.66</td>
</tr>
<tr>
<td>3.86</td>
<td>27.73</td>
</tr>
<tr>
<td>2.28</td>
<td>28.59</td>
</tr>
</tbody>
</table>
```

TODO: update the text when the latest release of HPCToolkit works on 32-bit platforms.
Figure 2: Profiling results of SMG2000 using HPCToolkit

2.07 29.36 0.78 hypre_CyclicReduction (/home/liao/smg2000/struct_ls/cyclic_reduction.c:1061 @ 8054450)
1.79 30.03 0.67 hypre_SemiRestrict (/home/liao/smg2000/struct_ls/semi_restrict.c:262 @ 8056a8c)
1.67 30.66 0.62 hypre_SemiInterp (/home/liao/smg2000/struct_ls/semi_interp.c:294 @ 8055d6c)
1.12 31.07 0.42 hypre_CyclicReduction (/home/liao/smg2000/struct_ls/cyclic_reduction.c:1133 @ 8054b2f)
0.96 31.43 0.36 hypre_CyclicReduction (/home/liao/smg2000/struct_ls/cyclic_reduction.c:912 @ 80531a6)
0.87 31.76 0.33 hypre_StructAxpy (/home/liao/smg2000/struct_mv/struct_axpy.c:69 @ 806642c)
0.80 32.06 0.30 hypre_CyclicReduction (/home/liao/smg2000/struct_ls/cyclic_reduction.c:1002 @ 8053a60)
0.78 32.35 0.29 hypre_SMGResidual (/home/liao/smg2000/struct_ls/smg_residual.c:236 @ 804d14b)
0.72 32.62 0.27 hypre_CyclicReduction (/home/liao/smg2000/struct_ls/cyclic_reduction.c:1063 @ 8054462)
0.60 32.84 0.23 hypre_SMGResidual (/home/liao/smg2000/struct_ls/smg_residual.c:240 @ 804d146)
0.59 33.06 0.22 hypre_SemiRestrict (/home/liao/smg2000/struct_ls/semi_restrict.c:269 @ 8056a8c)
0.59 33.28 0.22 hypre_SMGResidual (/home/liao/smg2000/struct_ls/smg_residual.c:240 @ 804d146)
0.51 33.48 0.19 hypre_StructVectorSetConstantValues (/home/liao/smg2000/struct_vector.c:578 @ 806f6dc)
0.48 33.66 0.18 hypre_SMGSetupInterp0p (/home/liao/smg2000/struct_ls/smg_setup_interp.c:292 @ 804ea04)
0.46 33.83 0.17 hypre_SemiInterp (/home/liao/smg2000/struct_ls/semi_interp.c:227 @ 8055d6c)
0.40 33.98 0.15 hypre_CyclicReduction (/home/liao/smg2000/struct_ls/cyclic_reduction.c:855 @ 8052bd4)
0.40 34.12 0.15 hypre_StructMatrixInitializeData (/home/liao/smg2000/struct_mv/struct_matrix.c:369 @ 80678b0)
3 Code Triage and Transformations

The second phase (shown in Fig. 3) includes code triage and a set of code transformations. Code triage relies on a ROSE-based tool interface to read in both source files and performance information of the input application. It then conducts various automated or user-directed analyses to identify problematic code segments, such as loops. Finally, the identified code segments are extracted (also called outlined) into separated routines so they can be individually optimized by empirical methods.

![Figure 3: Phase 1 and 2 of the autotuning system](image)

### 3.1 Invoking Code Triage

The source code for code triage is located in `rose/projects/autoTuning/autoTuning.C`. It already has initial implementation to call ROSE’s tool interface, conduct simple code triage, and finally extract kernels using ROSE’s AST outliner.

With the input application and its performance result available, users can invoke the ROSE-based code triage by using the following command:

```
autoTuning -c jacobi.c -rose:hpct:prof jacobi-raw.xml \
-rose:autotuning:triage_threshold 0.8 -rose:outline:output_path "tests"
```

The command above provides an input source file and its corresponding XML-format performance data generated by HPCToolkit. It asks the code triage program to find the most time-consuming loops which account for just...
above 80% of the total execution time. The identified loops will be automatically extracted to separated, source files and saved into an output path named tests.

Users can also enable code triage only without calling outlining. The performance data can come from GNU gprof. An example is given below:

```bash
# example command line to perform code triage only.
autoTuning -c jacobi.c -rose:autotuning:triage_only -rose:gprof:linebyline jacobi.gprof.txt
```

The output is a list of abstract handles and their corresponding execution time percentage:

```
The abstract handles for hot statements exceeding the threshold are:
Project::SourceFile<name, /home/liao6/jacobi.c>::ExprStatement<position, 193.9-194.76>
  0.382
Project::SourceFile<name, /home/liao6/jacobi.c>::ExprStatement<position, 196.9-196.45>
  0.3643
Project::SourceFile<name, /home/liao6/jacobi.c>::ExprStatement<position, 188.9-188.29>
  0.11
```

The abstract handles for enclosing loops for hot statements exceeding the threshold are:

```
Project::SourceFile<name, /home/liao6/jacobi.c>::ForStatement<position, 190.5-198.7>
  0.8189
```

The above example command identifies a list of the most time-consuming statements and loops and reports them using abstract handles. The report will end once the sum of execution time of the statements or loops reach or exceed a preset threshold (default is 75% of the total execution time).

We explain some details for the implementation of code triage and autotuning-related transformations in the following subsections.

### 3.2 Tool Interface

ROSE has a performance tool interface, called ROSE-HPCT, in its distribution to accept performance results generated by external performance tools. Basically, it reads in the XML files generated from HPCToolkit and attaches performance metrics to the ROSE AST representing the corresponding source code. It can handle macro expansions during the metric match process. When necessary, all performance metrics are also propagated from statement levels to loop, function, and file levels. Similarly, it also accepts the line-by-line performance data generated by GNU gprof. Detailed information about ROSE-HPCToolKit can be found in Chapter 44 of the ROSE Tutorial.

The code triage program uses the following code to invoke ROSE-HPCT.

```c
int main(int argc, char * argv[]) {
  vector<string> argvList (argv, argv+argc);
  // Read into the XML files
  RoseHPCT::ProgramTreeList_t profiles = RoseHPCT::loadHPCTProfiles (argvList);
  // Create the AST tree
  SgProject * project = frontend(argvList);
  //Attach metrics to AST, last parameter is for verbose
  RoseHPCT::attachMetrics(profiles, project, project->get_verbose()>0);
  //...
}
```
3.3 Kernel Extraction: Outlining

Each of the identified tuning targets, often loops, will be extracted from the original code to form separated functions (or routines). The ROSE AST outliner is invoked to generate such functions. This kernel extraction step can be automatically invoked by the code triage program or manually initiated by users via the outliner’s command line interface.

The ROSE AST outliner handles multiple input languages, including C, C++ and recently Fortran. It also provides both command line and programming interfaces for users to specify the targets to be outlined. Detailed information of using the ROSE outliner can be found in Chapter 27 of the ROSE Tutorial. You can also refer to a paper [8] for the algorithm we use to outline kernels. For the code triage program, the programming interface of the Outliner is used as below:

```cpp
// recognize options for outliner
Outliner::commandLineProcessing(argvList);
...
SgForStatement* target_loop = findTargetLoop(hot_node);
if (isOutlineable (target_loop))
    outline(target_loop);
```

The ROSE AST outliner accepts user options to further guide its internal behavior.

- **rose:outline:parameter** will wrap all parameters of the outlined function into an array of pointers.
- **rose:outline:new_file** will separate the outlined function into a new source file to facilitate external tools for further handling.
- **rose:outline:use_dlopen** will transform the code to use `dlopen()` and `dlsym()` to dynamically load the outlined function from a library.

For the SMG2000 example, the most time-consuming loop is outlined and a call to it is used to replace the loop in the original code. The loop is actually expanded from a macro `hypre_BoxLoop3For()`. ROSE is able to identify it after a bottom up metrics propagation phase in ROSE-HPCT. The kernel extraction’s result is shown in the following code listing:

```cpp
// A prototype of the outlined function is inserted at the beginning of the code
void **_out_argv1_1527[21];
void ***_out_argv = *(void **_out_argv1_1527_+0); // void **_out_argv1_1527_+0 = ((void *)(hypre::nx));
*(void **_out_argv1_1527_+1) = ((void *)(hypre::ny));
*(void **_out_argv1_1527_+2) = ((void *)(hypre::nx));
*(void **_out_argv1_1527_+3) = ((void *)(hypre::sz3));
*(void **_out_argv1_1527_+4) = ((void *)(hypre::sy3));
*(void **_out_argv1_1527_+5) = ((void *)(hypre::sx3));
*(void **_out_argv1_1527_+6) = ((void *)(hypre::sz2));
*(void **_out_argv1_1527_+7) = ((void *)(hypre::sy2));
*(void **_out_argv1_1527_+8) = ((void *)(hypre::sx2));
*(void **_out_argv1_1527_+9) = ((void *)(hypre::sz1));
*(void **_out_argv1_1527_+10) = ((void *)(hypre::sy1));
*(void **_out_argv1_1527_+11) = ((void *)(hypre::sz1));
*(void **_out_argv1_1527_+12) = ((void *)(loopk));
*(void **_out_argv1_1527_+13) = ((void *)(loopj));
*(void **_out_argv1_1527_+14) = ((void *)(loopi));
*(void **_out_argv1_1527_+15) = ((void *)(rp));
*(void **_out_argv1_1527_+16) = ((void *)(xp));
*(void **_out_argv1_1527_+17) = ((void *)(Ap));
*(void **_out_argv1_1527_+18) = ((void *)(kr));
*(void **_out_argv1_1527_+19) = ((void *)(xi));
```

```cpp
// The target loop is replaced by a call to the outlined function
```

```cpp
void _out_4755(void **out_argv) {
```

```cpp
// A prototype of the outlined function is inserted at the beginning of the code
void OUT_1_6755(void **out_argv);
```

```cpp
// The target loop is replaced by a call to the outlined function
```
The ROSE outliner uses a variable cloning method to avoid using pointer dereferencing within the outlined computation kernels. The method is based
on usage of a variable that is passed by reference and accessed via pointer-
dereferencing by classic outlining algorithms. Such a variable is used either by
value or by address within an outlining target. For the C language, using the
variable by address occurs when the address operator is used with the variable
(e.g., &X). C++ introduces one more way of using the variable by address:
associating the variable with a reference type (TYPE & Y = X; or using the
variable as an argument for a function’s parameter of a reference type). If
the variable is not used by its address, a temporary clone variable of the same
type (using TYPE clone;) can be introduced to substitute its uses within the
outlined function. The value of a clone has to be initialized properly (using
clone = *parameter;) before the clone participates in computation. After the
computation, the original variable must be set to the clone’s final value (using
*parameter = clone). By doing this, many pointer dereferences introduced by
the classic algorithm can be avoided.

For easier interaction with other tools, the outlined function is separated into
a new source file, usually named after the function name. The ROSE outliner
will recursively find and copy all dependent type declarations into the new file to
make it compilable. For the SMG2000 example, a file named out_1_6755___orig.c
is generated and it only contains the function body of OUT_1_6755(()=>). This file
will be transformed by a parameterized optimizer into a kernel variant named
OUT_1_6755__.c and further compiled into a dynamically loadable routine.

A sample makefile to generate the .so file is given below:

```
[liao@localhost smg2000] cat makefile-lib
lib:OUT_1_6755__.so
  gcc -c -fpic OUT_1_6755__.c
  gcc -shared -lc -o OUT_1_6755__.so OUT_1_6755__.o
clean:
  rm -rf OUT_1_6755__.so OUT_1_6755__.o
```

3.4 Transformations for Autotuning

The call site of the outlined function in the source code has to be further trans-
formed to support empirical optimization. These transformations include:

- using dlopen() to open a specified .so file containing the outlined target
  and calling the function found in the file,
- adding performance measuring code to time the call of the outlined target
  function,
- transforming code to support checkpointing the execution right before
dlopen() opening the library source file so multiple variants of the file
  can be used to test optimization choices empirically when the program is
  restarted multiple times.

3.4.1 Calling a Function Found in a .so File

As mentioned earlier, the outlined function containing the target kernel is stored
in a separated source file, which will be transformed into a kernel variant and
then compiled to a dynamically loadable library. The original source code has

FIXME: We are working on using liveness analysis to further eliminate unnecessary value assignments.
to be transformed to open the library file, find the outlined function, and finally call it using the right parameters. An example of the resulting transformation on the function containing the outlined loop is given below:

```c
#include "dlfcn.h"

/* handle to the shared lib file */
void *FunctionLib;

/* a prototype of the pointer to a loaded routine */
void (*OUT__1__6755__)(void **out_argv);

int hypre_SMGResidual( void *residual_vdata,
    hypre_StructMatrix *A,
    hypre_StructVector *x,
    hypre_StructVector *b,
    hypre_StructVector *r)
{
    // .......... 
    const char *dlError = dlerror();
    FunctionLib = dlopen("OUT__1__6755__.so",RTLD_LAZY);
    if (dlError)
        printf("cannot open .so file!\n");
    exit(1);

    /* Find the first loaded function */
    OUT__1__6755__ = dlsym(FunctionLib,"OUT__1__6755__.");
    if (dlError)
        printf("cannot find OUT__1__6755__.()\n");
    exit(1);

    /* Call the outlined function by the found function pointer */
    (*OUT__1__6755__)(out_argv);

    // save timing information into a temporary file
    remove("/tmp/peri.result");
    time1=time_stamp();
    time2=time_stamp();
    FILE* pfile = fopen("/tmp/peri.result","a");
    if (pfile != NULL)
    {
        fprintf(pfile,"%f\n",time2-time1);
        fclose(pfile);
    }

3.4.2 Timing the Call

The call to the outlining target kernel should be timed to get the evaluation results during the empirical optimization. We instrument the call as follows to get the performance evaluation of a kernel variant. More accurate and less intrusive methods based on hardware counters can also be used in the future.

// save timing information into a temporary file
remove("/tmp/peri.result");
time1=time_stamp();
// parameter wrapping statements are omitted here
(*OUT__1__6755__) (...out_argv);

time2=time_stamp();
FILE* pfile;
pfile=fopen("/tmp/peri.result","a");
if (pfile != NULL)
{
    fprintf(pfile,"%f\n",time2-time1);
    fclose(pfile);
}
```
3.4.3 Checkpointing and Restarting

In order to efficiently evaluate hundreds or even thousands of kernel variants, we use a checkpointing and restarting method to measure the time spent on calling the kernel without unnecessarily repeating the execution before and after calling the kernel. This allows the full context (state) of the application to be used in the evaluation of the kernel performance.

The Berkeley Lab Checkpoint/Restart (BLCR) library [2] was selected for its programmability, portability and robustness. BLCR is designed to support asynchronous checkpointing, which means a running process is notified by another process to be checkpointed first, but the exact checkpointing action happens on an indeterministic point later on. This default behavior is not desired for our purpose since we want an accurate source position to do the checkpointing. Fortunately, BLCR’s library does provide some internal functions to help us initiate synchronous (immediate) checkpointing, though not well documented.

After some trial and error rounds, we use the following code transformation to have a synchronous checkpointing using the blcr-0.8.2 release. The idea is to have a running program to notify itself to initiate a checkpointing. The program is then blocked until the request is served.

To support BLCR, we have transformed the original source code in two locations. The first location is the file where the main() function exists. We add the necessary header files and initialization codes for using BLCR. For example:

```c
/* Code addition in the file containing main() */

#define BLCR_CHECKPOINTING
#define O_LARGEFILE 0

int main( int argc, char *argv[] )
{
    #if BLCR_CHECKPOINTING
        // initialize the blcr environment
        cr = cr_init();
        cr_info_init();
    #endif
    //......
}
```

FIXME: need to discuss the drawbacks, e.g. missing cache warmup; and how that might be addressed in the future by pushing back the checkpoint start and transformations to the checkpointed program (later).
The second place is the actual source line to initiate the checkpointing. A checkpoint argument structure is filled out first to customize the behavior we want, including the scope, target, memory dump file, consequence, and so on. A blocking phase is put right after `cr_request_checkpoint()` to have an immediate checkpointing. Our goal is to stop the executable right before opening the .so file so a different kernel variant can be compiled into the .so file each time. The execution will be restarted multiple times so multiple variants can be evaluated this way.

```c
#ifdef BLCR_CHECKPOINTING
// #include <sys/types.h>
#include <unistd.h>
#include <sys/stat.h>
#include <fcntl.h>
#include "libcrr.h"
#ifndef OLARGEFILE
#define OLARGEFILE 0
#endif
static int g_checkpoint_flag = 0;
extern cr_client_id_t cr;
#endif

int hypre_SMGResidual (void *residual_vdata,
    hypre_StructMatrix *A,
    hypre_StructVector *x,
    hypre_StructVector *b, hypre_StructVector *r)
{
    // ............
#if BLCR_CHECKPOINTING
    int err;
    cr_checkpoint_args_t cr_args;
    cr_checkpoint_handle_t cr_handle;
    cr_initialize_checkpoint_args_t (&cr_args);
    cr_args.cr_scope = CRSCOPEPROC;  // checkpoint an entire process
    cr_args.cr_target = 0;  // checkpoint self
    cr_args.cr_signal = SIGKILL;  // kill self after checkpointing
    cr_args.cr_fd = open("dump.y", O_WRONLY | O_CREAT | OLARGEFILE, 0400);
    if (cr_args.cr_fd < 0)
    {
        printf("Error: cannot open file for checkpointing context\n"");
        abort();
    }
    printf("Checkpointing: stopping where\n");
    err = cr_request_checkpoint (&cr_args, &cr_handle);
    if (err < 0)
    {
        printf("cannot request checkpointing! err=%d\n", err);
        abort();
    }
    // block self until the request is served
    cr_enter_cs (cr);
    cr_leave_cs (cr);
    printf("Checkpointing: restarting where\n");
    FunctionLib =dlopen("OUT_1.6755.so", RTLD_LAZY);
    // ignore the code to find the outlined function and to time the execution here
    // ...
    (*OUT_1.6755_)("out_argv1",1527);
    // timing code ignored here
    // ...
    // stop after the target finishes its execution
    exit(0);
#endif
```
Only these minimum transformations are needed to build the target application to support BLCR. We choose the static linking method to support BLCR as follows. The BLCR library is linked with the final executable.

```bash
smg2000: smg2000.o
  @echo "Linking" "$@ "...
  "$CC" -o smg2000 smg2000.o "$LFLAGS" -lcr
```

The checkpointed binary has to be executed once to generate a memory dump file (process image), which can then be reused multiple times to restart the execution immediately before the line where `dlopen()` is invoked to evaluate multiple variants of the optimization target kernel.

```bash
{liao@localhost test} ./smg2000 -n 120 120 120 -d 3
current process ID is 6028
Running with these driver parameters:
(nx, ny, nz) = (120, 120, 120)
(Px, Py, Pz) = (1, 1, 1)
(bx, by, bz) = (1, 1, 1)
(cx, cy, cz) = (1.000000, 1.000000, 1.000000)
(n_pre, n_post) = (1, 1)
dim = 3
 solver ID = 0
========================================================================================================
Struct Interface:
========================================================================================================
Struct Interface:
  wall clock time = 0.550000 seconds
  cpu clock time = 0.540000 seconds
Checkpointing: stopping here ..
Killed
```

A context dump file named dump.yy will be generated by BLCR as the checkpointing result. This dump file will be used to restart the execution using the command: `cr_restart dump.yy`.

## 4 Empirical Tuning

The actual empirical tuning (shown in Fig. 4) is conducted via the cooperation of several components: a search engine, a parameterized transformation tool, and the previously introduced checkpointing/restarting library. The basic idea is that: 1) a set of pre-selected transformations and their corresponding configuration ranges are given (by the code triage module) and converted into an integral search space; 2) the search engine evaluates points from the search space by driving the parameterized transformation tool to generate kernel variants and restarting the checkpointed binary to run the variants one by one.

### 4.1 Parameterized Transformation Tools

Several choices exist to generate kernel variants, they include POET [12], CHiLL [6], and the ROSE loop translators. We take POET as an example here.
POET (Parameterized Optimizations for Empirical Tuning) developed by Dr. Qing Yi under contract with University of Texas at San Antonio (UTSA), is an efficient language and tool to express hundreds or thousands of complex code transformations and their configurations using a small set of parameters. It is especially relevant to the evaluation of large-scale search spaces as part of empirical tuning and is orthogonal to any specific search strategy.

Using command line options and a configuration file, users can direct POET to apply a set of specified transformations with desired configurations on selected code portions. Also, the target kernel has to be instrumented to aid POET in the process. Detailed POET user instructions can be found at its official website [4]. For example, the SMG2000 kernel has the following format to support POET:

```plaintext
A default transformation parameter, unrolling factor, is also given in the file. But this parameter is usually superseded by a command line parameter, the following command line specifies unrolling 5 times.
/home/liao/download/qing/POET/src/pcg -punrollI=5 -L/home/liao/download/qing/POET/lib my.pt
```

The POET configuration file (my.pt) we use to optimize SMG2000's kernel is shown below. In this file, loop unrolling is specified to be performed on the target within a source file named out_1_6755_orig.c. The result will be saved inside a file named out_1_6755_.c.

```plaintext
#include Cfront.code
#include opt.pi

<parameter unrollI type=1 default=2 message="unroll factor for innermost loop 1"/>
<trace target, nestI/>
<input from="OUT_1_6755_orig.c" to=target parse=FunctionDefn/>
<eval
   INSERT(nestI, target);
   UnrollLoops[factor=unrollI](nestI[Nest.body], nestI);
/>
<output from=target to="OUT_1_6755_.c"/>
```

A default transformation parameter, unrolling factor, is also given in the file. But this parameter is usually superseded by a command line parameter, the following command line specifies unrolling 5 times.

```
/home/liao/download/qing/POET/src/pcg -punrollI=5 -L/home/liao/download/qing/POET/lib my.pt
```

**Figure 4:** Phase 3 of the autotuning system
void OUT_1_6755_ ( void **out_argv )
{
    int Ai = *((int *) ( out_argv[20] ));
    int xi = *((int *) ( out_argv[19] ));
    int ri = *((int *) ( out_argv[18] ));
    double *Ap = *((double **) ( out_argv[17] ));
    double *xp = *((double **) ( out_argv[16] ));
    double *rp = *((double **) ( out_argv[15] ));
    int loopi = *((int *) ( out_argv[14] ));
    int loopj = *((int *) ( out_argv[13] ));
    int loopk = *((int *) ( out_argv[12] ));
    int hypre_sx1 = *((int *) ( out_argv[11] ));
    int hypre Symfony = *((int *) ( out_argv[10] ));
    int hypre_sx2 = *((int *) ( out_argv[9] ));
    int hypre_sx3 = *((int *) ( out_argv[8] ));
    int hypre_sx4 = *((int *) ( out_argv[7] ));
    int hypre_sx5 = *((int *) ( out_argv[6] ));
    int hypre_sx6 = *((int *) ( out_argv[5] ));
    int hypre_sx7 = *((int *) ( out_argv[4] ));
    int hypre_sx8 = *((int *) ( out_argv[3] ));
    int hypre_nx = *((int *) ( out_argv[2] ));
    int hypre_ny = *((int *) ( out_argv[1] ));
    int hypre_nz = *((int *) ( out_argv[0] ));
    for ( loopk = 0; loopk < hypre_nz; loopk++)
    {
        for ( loopj = 0; loopj < hypre_ny; loopj++ )
        {
            for ( loopi = 0; loopi < hypre_nx; loopi++ ) // @BEGIN(nestI)
            {
                rp[ri] = ((Ap[Ai]) * (xp[xi]));
                Ai += hypre_sx1;
                xi += hypre_sx2;
                ri += hypre_sx3;
            }
        }
    }
    Ai += (hypre_sx1 - (hypre_nx * hypre_sx1));
    xi += (hypre_sx2 - (hypre_nx * hypre_sx2));
    ri += (hypre_sx3 - (hypre_nx * hypre_sx3));
    // @END(nestI : Nest)
    Ai += (hypre_sx1 - (hypre_sx4 * hypre_sx1));
    xi += (hypre_sx2 - (hypre_sx4 * hypre_sx2));
    ri += (hypre_sx3 - (hypre_sx4 * hypre_sx3));
}
// some code omitted here...

Figure 5: the SMG 2000 kernel

4.2 Search Engines

Currently, we adopt the GCO (Generic Code Optimization) search engine [13] from University of Tennessee at Knoxville as the external search engine used in our system. It has been connected with a specific version of POET (not yet fully updated to the latest POET release unfortunately) to explore code transformations using several popular search policies, such as random search, exhaustive search, simulated anneal search, genetic algorithm, and so on.

The search engine interacts with the parameterized optimization tool (POET) via a bash script, usually named as eval.xxx where xxx indicates the target application. This script is manually generated currently and does the

FIXME: TODO
The generation of .pt files is not yet automated currently.
following tasks:

1. specifies the search space’s dimension number and lower, upper bound for each dimension,

2. specifies the number of executions for each evaluation. This will help exclude some executions disturbed by system noises,

3. validation of the number of command line options for this script, the number should match the number of dimensions of the search space so each value corresponding one dimension. All the options together mean a valid search point within the search space.

4. converts the search point into transformation parameters understandable by POET. Some transformation choices are obtained by interpreting integer values in a custom way, such as the order of loop interchanging.

5. generates a kernel variant by invoking POET with right parameters to conduct the corresponding transformations on the target kernel,

6. compiles the generated kernel variant into a dynamically loadable shared library file (a .so file),

7. restarts the checkpointed binary to evaluate the kernel variant. This step is repeated multiple times as configured and the shortest execution time is reported as the evaluation result for this particular transformation setting (a point).

A full example script for SMG2000 is given below.

```bash
#!/bin/bash

DIM 1
LB 1
UB 32

ITER=4

# command line validation
# should have x parameters when calling this script
# x= number of dimensions for each point
if [ "$1" = "" ]; then
    echo "0.0"
    exit
fi

# convert points to transformation parameters
# Not necessary in this example

# remove previous variant of the kernel and result
/bin/rm -f out.6755* /tmp/peri.result

# generate a variant of the kernel using the transformation parameters
/home/liao/download/qing/POET/src/pcg -punroll1=$1 -L/home/liao/download/qing/POET/lib my.pt > /dev/null

# build a .so for the kernel variant
# To redirect stdout to NULL is required
# since the search engine looks for stdout for the evaluation result
make -f makefile-lib > /dev/null

cp OUT.6755* .so /home/liao/svnrepos/benchmarks/smg2000/struct.la /.

# generate a program to execute and timing the kernel
```

18
As we can see, the evaluation of a kernel variant needs the cooperation of three parties.

1. the transformed target application providing a performance measurement (timing) for the call to the variant,
2. the eval_smg script choosing the best execution after several times of execution using the same kernel variant,
3. the search engine retrieving the information returned from eval_smg as the evaluation result of a variant and proceeding the search accordingly.

### 4.3 An Example Search

We use the random search policy of the UTK search engine to demonstrate a sample search process. The search engine chooses the maximum evaluation value as the best result by default. So a reciprocal of a timing result is indicated by an environment variable GCO\_SEARCH\_MODE to be the evaluation result.

The UTK search engine also accepts a upper time limit for a search session. We use 1 minute in this example by adding 1 as the last parameter.

```bash
[liao@localhost smg2000]$ export GCO_SEARCH_MODE=reciprocal
[liao@localhost smg2000]$ ../search/random_search ./eval_smg 1
Checkpointing: restarting here...
Case: ./eval_smg 12 --
Got the evaluation result: 50.7846
Checkpointing: restarting here...
Case: ./eval_smg 13 --
Got the evaluation result: 49.65
Checkpointing: restarting here...
Case: ./eval_smg 11 --
Got the evaluation result: 50.9502
Checkpointing: restarting here...
Case: ./eval_smg 31 --
Got the evaluation result: 49.8107
Checkpointing: restarting here...
Case: ./eval_smg 15 --
Got the evaluation result: 49.8703
```
Checkpointing: restarting here ..
Case: ./eval_smg 14 --
Got the evaluation result: 49.645
Checkpointing: restarting here..
Case: ./eval_smg 25 --
Got the evaluation result: 50.0551
Checkpointing: restarting here ..
Case: ./eval_smg 18 --
Got the evaluation result: 49.7018
skipping already visited point 31 , value = 49.810719
Checkpointing: restarting here..
Case: ./eval_smg 32 --
Got the evaluation result: 49.5221
Checkpointing: restarting here ..
Case: ./eval_smg 22 --
Got the evaluation result: 49.6475
Checkpointing: restarting here ..
Case: ./eval_smg 6 --
Got the evaluation result: 51.261
Checkpointing: restarting here ..
Case: ./eval_smg 30 --
Got the evaluation result: 50.0475
skipping already visited point 18 , value = 49.701789
skipping already visited point 14 , value = 49.645038
Checkpointing: restarting here ..
Case: ./eval_smg 4 --
Got the evaluation result: 51.435
Time limit reached...

Random Search Best Result: Value=49.522112, Point=32
Total Number of evaluations: 13

In the sample search above, a one-dimension search space (loop unrolling factor) was examined. Within the one-minute time limit, points were randomly chosen by the search engine and three of them were redundant. Apparently, the UTK search engine was able to skip redundant evaluations. In the end, a point (32) had the best value (reciprocal of timing) which means for the target smg2000 kernel, unrolling 32 times generated the best performance.

Similarly, other search policies can be used by replacing random_search with exhaustive_search, anneal_search, ga_search, simplex_search, etc.

5 Working with Active Harmony

We describe how our end-to-end empirical tuning framework can be adapted to work with another search engine, namely the Active Harmony system [11, 7]. The work is done with generous help from Ananta Tiwari at University of Maryland.

Active Harmony allows online, runtime tuning of application parameters which are critical to the application performance. Domain-specific knowledge is usually needed to identify those application parameters. An active harmony server will be running to manage the values of different application parameters and to conduct the search. Applications have to be modified to communicate with the server in order to send performance feedback for one specific set of parameter values (current point) and get the next set of parameter values (next point). Currently, it supports a search strategy based on the Nelder-Mead simplex method [9].

For SMG2000, we generate a set of unrolled versions (OUT_{1...6755}_X()) for the target kernel and treat the function suffix X as the tunable parameter.
As a result, the .so file contains all unrolled versions.

A configuration file is needed for Active Harmony to know the parameters to be tuned and their corresponding ranges. The following file is used to specify a tunable parameter named unroll with a range from 1 to 32. obsGoodness (observed goodness - or performance) and predGoodness (predicted goodness) are related to a GUI window showing during the execution. They do not impact the performance tuning and the workings of the search algorithm.

```plaintext
harmonyApp smg2000Unroll {
   { harmonyBundle unroll {
        int {1 32 1}
   }
   { obsGoodness 1 5000}
   { predGoodness 1 5000}
}
```

We don’t use BLCR since an online tuning method is used for Active Harmony. The code transformation to work with Active Harmony is shown below. The basic idea is to call a set of Harmony interface functions to startup communication with the server (harmony_startup()), send a configuration file (harmony_application_setup_file()), define a parameter to be tuned (harmony_add_variable()), report performance feedback (harmony_performance_update()), and finally request the next set of values of the tunable parameters (harmony_request_all()).

```c
#include "hclient.h"
#include "hsockutil.h"
static int har_registration = 0;
static int *unroll = NULL;

int hypre_SMGResidual(...) {
    // initialize the search engine for the first call
    if (har_registration == 0) {
        printf("Starting Harmony...\n");
        harmony_startup(0);
        printf("Sending setup file!\n");
        harmony_application_setup_file("harmony.tcl");
        printf("Adding a harmony variable...\n");
        unroll = (int *)harmony_add_variable("smg2000Unroll","unroll",VAR_INT);
        har_registration++;
    }

    // Load the .so files
    char nextUnroll[255];
    if (g_execution_flag == 0) {
        printf("Opening the .so file...\n");
        FunctionLib = dlopen("./unrolled_versions.so",RTLD_LAZY);
        dlError = dlerror();
        if (dlError) {
            printf("cannot open .so file!\n");
            exit(1);
        }
    } // end if (flag ==0)

    // Use the value provided Harmony to find a kernel variant
    snprintf(nextUnroll, "OUT_-1_-6755_604", *unroll);
    printf("Trying to find: %s...\n", nextUnroll);
    OUT_-1_-6755_604 = (void (*)(void**)) dlsym( FunctionLib, nextUnroll);

    // timing the execution of a variant
    timel=time_stamp();
    //... parameter wrapping is omitted here
```
Figure 6 shows a screen shot of using Active Harmony to tuning the SMG2000 benchmark. The top panel gives a graphical representation of the search space (one dimension named unroll) and the tuning timeline with evaluation values (x-axis is time, y-axis is evaluation values). The left-bottom shell window is the client application being tuned. The right-bottom shell windows shows the activities of the server.

We have found that online/runtime search engines like the Active Harmony can be extremely expensive if the tuned kernel are invoked thousands of times. For SMG2000, it took hours to finish the tuning using a 120x120x120 input data set. The major reason is that for each call of the kernel, a bidirectional communication between the client application and the server has to be finished. Another reason is that the current tuning process is embedded into the thousands of calls of the kernel so that points are always evaluated even when some of them have already been evaluated before.
6 User-Directed Autotuning

While fully automated autotuning can be feasible in some cases, many applications need users' expertise to obtain significant performance improvements. We revisit the SMG2000 benchmark in this section to see how one can use our autotuning system to have a user-directed empirical tuning process.

6.1 Manual Code Triage

The SMG2000 kernel (shown in Fig. 5) that is automatically identified by our simple code triage may not be the best tuning target. It is actually only a portion of a bigger computation kernel that is very hard to be automatically identified. Also, the bigger kernel has a very complex form which impedes most compilers or tools for further analyses and transformations. With the help from Rich Vuduc, who is a former postdoc with us, we manually transform the code (via code specification) to obtain a representative kernel which captures the core computation algorithm of the benchmark. We then put the outlining pragma (`#pragma rose outline`) before the new kernel and invoked the ROSE outliner to separate it out into a new source file, as shown below.

```c
#include "autotuning_lib.h"
static double time1, time2;
void OUT_1_6119_2(void **out_argv);
typedef int hypre_MPI_Comm;
typedef int hypre_MPI_Datatype;
typedef int hypre_Index [3];
typedef struct hypre_Box_struct
{
  hypre_Index imin;
  hypre_Index imax;
} hypre_Box;

typedef struct hypre_BoxArray_struct
{
  hypre_Box *boxes;
  int size;
  int alloc_size;
} hypre_BoxArray;

typedef struct hypre_RankLink_struct
{
  int rank;
  struct hypre_RankLink_struct *next;
} hypre_RankLink;

typedef hypre_RankLink *hypre_RankLinkArray[3][3][3];

typedef struct hypre_BoxNeighbors_struct
{
  hypre_BoxArray *boxes;
  int *procs;
  int *ids;
  int first_local;
  int num_local;
  int num_periodic;
  hypre_RankLinkArray *rank_links;
} hypre_BoxNeighbors;

typedef struct hypre_StructGrid_struct
{
  hypre_MPI_Comm comm;
  int dim;
  hypre_BoxArray *boxes;
```
typedef struct hypre_StructStencil {  
    hypre_Index *shape;  
    int size;  
    int max_offset;  
    int dim;  
    int ref_count; 
} hypre_StructStencil;

typedef struct hypre_ComTypeEntry {  
    hypre_Index imin;  
    hypre_Index imax;  
    int offset;  
    int dim;  
    int length_array[4];  
    int stride_array[4]; 
} hypre_ComTypeEntry;

typedef struct hypre_ComType {  
    hypre_ComTypeEntry **comm_entries;  
    int num_entries; 
} hypre_ComType;

typedef struct hypre_ComPkg {  
    int num_values;  
    hypre_MPI_Comm comm;  
    int num_sends;  
    int num_recvs;  
    int *send_procs;  
    int *recv_procs;  
    hypre_ComType **send_types;  
    hypre_ComType **recv_types;  
    hypre_MPI_Datatype *send_mpi_types;  
    hypre_MPI_Datatype *recv_mpi_types;  
    hypre_ComType *copy_from_type;  
    hypre_ComType *copy_to_type; 
} hypre_ComPkg;

typedef struct hypre_StructMatrix {  
    hypre_MPI_Comm comm;  
    hypre_StructGrid *grid;  
    hypre_StructStencil *user_stencil;  
    hypre_StructStencil *stencil;  
    int num_values;  
    hypre_BoxArray *data_space;  
    double *data;  
    int data_allocated;  
    int data_size;  
    int **data_indices;  
    int symmetric;  
    int **symm_elements;  
    int num_ghost[6];  
    int global_size;  
    hypre_ComPkg *comm_pkg;  
    int ref_count; 
} hypre_StructMatrix;

void OUT_1_6119(void **out_argv) {

\textbf{static int} counter = 0;

hypre_StructMatrix *A = *((hypre_StructMatrix **)(&out_argv[20]));
int ri = *((int *)(&out_argv[19]));
double *rp = *((double **)(&out_argv[18]));
int stencil_size = *((int *)(&out_argv[17]));
int i = *((int *)(&out_argv[16]));
int (*dxp_s)[15UL] = *((int *)(stencil_size[15UL]));
int hypre_syl = *((int *)(out_argv[14]));
int hypre_sz1 = *((int *)(out_argv[13]));
int hypre_sz2 = *((int *)(out_argv[12]));
int hypre_sz3 = *((int *)(out_argv[11]));
int hypre_mz = *((int *)(out_argv[10]));
int hypre_mx = *((int *)(out_argv[9]));
int hypre_my = *((int *)(out_argv[8]));
int hypre_mz = *((int *)(out_argv[7]));
int si = *((int *)(out_argv[6]));
int ii = *((int *)(out_argv[5]));
int jj = *((int *)(out_argv[4]));
int kk = *((int *)(out_argv[3]));
const double *Ap0 = *((const double **)(out_argv[2]));
const double *xp0 = *((const double **)(out_argv[1]));

at_begin_timing(); // begin timing

for (si = 0; si < stencil_size; si++)
for (kk = 0; kk < hypre_mz; kk++)
for (jj = 0; jj < hypre_my; jj++)
for (ii = 0; ii < hypre_mx; ii++)
	rp[(si + ii) + (jj + hypre_sz3) + (kk + hypre_sz1) + (kk + hypre_syl) +
A[si][ii][jj]]
* xp0[(si + ii) + (jj + hypre_sz2) + (kk + hypre_sz2) + (dxp_s)[si]][ii]];

at_end_timing(); // end timing

*((int *)(out_argv[2])) = kk;
*((int *)(out_argv[3])) = jj;
*((int *)(out_argv[4])) = ii;
*((int *)(out_argv[5])) = si;
*((double **)(out_argv[18])) = rp;
*((hypre_StructMatrix **)(out_argv[20])) = A;

As we can see, the new kernel directly and indirectly depends on some user
defined types. The ROSE outliner was able to recursively find and copy them
to the new file in a right order.

\subsection{Parameterized ROSE Loop Translators}

ROSE provides several standalone executable programs (loopUnrolling, loop-
Interchange, and loopTiling under ROSE_INSTALL/bin) for loop transformations. So autotuning users can use them via command lines with abstract handles to create desired kernel variants. Detailed information about the parameterized loop translators can be found in Chapter 50 of the \texttt{ROSE Tutorial}.

These translators use ROSE's internal loop translation interfaces (declared
within the SageInterface namespace). They are:

- \texttt{bool loopUnrolling (SgForStatement *loop, size_t unrolling_factor)}: This function needs two parameters: one for the loop to be unrolled and the other for the unrolling factor.

- \texttt{bool loopInterchange (SgForStatement *loop, size_t depth, size_t lexico-Order)}: The loop interchange function has three parameters, the first one to specify a loop which starts a perfectly-nested loop and is to be
interchanged, the 2nd for the depth of the loop nest to be interchanged, and finally the lexicographical order for the permutation.

- **bool loopTiling (SyForStatement *loopNest, size_t targetLevel, size_t tileSize):** The loop tiling interface needs to know the loop nest to be tiled, which loop level to tile, and the tile size for the level.

For efficiency concerns, these functions only perform the specified translations without doing any legitimacy check. It is up to the users to make sure the transformations won’t generate wrong code.

Example command lines using the programs are given below:

```bash
# unroll a for statement 5 times. The loop is a statement at line 6 within an input file.
loopUnrolling -c inputloopUnrolling.C \ 
-rose:loopunroll:abstract_handle "Statement<position,6>" -rose:loopunroll:factor 5

# interchange a loop nest starting from the first loop within the input file,
# interchange depth is 4 and
# the lexicographical order is 1 (swap the innermost two levels)
loopInterchange -c inputloopInterchange.C -rose:loopInterchange:abstract_handle \ 
"ForStatement<numbering,1>" -rose:loopInterchange:depth 4 \ 
-rose:loopInterchange:order 1

# tile the loop with a depth of 3 within the first loop of the input file
# tile size is 5
loopTiling -c inputloopTiling.C -rose:loopTiling:abstract_handle \ 
"ForStatement<numbering,1>" -rose:loopTiling:depth 3 -rose:loopTiling:tilesize 5
```

### 6.3 Connecting to the Search Engine

We applied several standard loop optimizations to the new kernel. They are, in the actual order applied, loop tiling on i, j and k levels (each level has a same tiling size from 0 to 55 and a stride of 5), loop interchange of i, j, k and si levels (with a lexicographical permutation order ranging from 0 to 4! -1), and finally loop unrolling on the innermost loop only. For all optimizations, a parameter value of 0 means no such optimization is applied. So the total search space has 14400 (12 * 4! * 50) points.

The bash script used by the GCO search engine to conduct point evaluation is given below. Note that a transformation command needs to consider previous transformations’ side effects on the kernel. We also extended GCO to accept strides (using #ST in the script) for dimensions of search space.

```bash
$ cat eval_smg_combined
#!/bin/bash

# number of executions to find the best result for this variant
ITER=3

# command line validation
# should have x parameters when calling this script
# x= number of dimensions for each point
if [ "$3" = "" ]; then
    echo "Fatal error: not enough parameters are provided for all search dimensions";
    exit
fi
```

26
# convert points to transformation parameters
# Not necessary in this example

# target application path
APP_PATH=/home/liao6/svnrepos/benchmarks/smg2000
KERNEL_VARIANT_FILE=OUT__1__6119__.c

# remove previous variant of the kernel and result
/bin/rm -f $APP_PATH/struct_ls/$KERNEL_VARIANT_FILE /tmp/peri.result $APP_PATH/*.so *.so

# ------------ tiling i, j, k ----------------------
# first tiling is always needed.
loopTiling -c $APP_PATH/struct_ls/OUT__1__6119__perfectNest.c -rose:loopTiling:abstract_handle "ForStatement<numbering,1>" -rose:loopTiling:depth 4 -rose:loopTiling:tilesize $1 -rose:output $KERNEL_VARIANT_FILE
if [ $1 -ne 0 ]; then
  fi
# -------------- interchange si, k, j, i--------------
if [ $1 -ne 0 ]; then
else
  # No tiling happens, start from 1
fi
# ------------ unrolling innermost only -------------------
# generate a variant of the kernel using the transformation parameters
# unrolling the innermost level, must redirect to avoid mess up search engine
if [ $1 -ne 0 ]; then
  loopUnrolling -c $KERNEL_VARIANT_FILE -rose:loopunroll:abstract_handle "ForStatement<numbering,7>" -rose:loopunroll:factor $3 -rose:output $KERNEL_VARIANT_FILE > /dev/null 2>&1
else
  loopUnrolling -c $KERNEL_VARIANT_FILE -rose:loopunroll:abstract_handle "ForStatement<numbering,4>" -rose:loopunroll:factor $3 -rose:output $KERNEL_VARIANT_FILE > /dev/null 2>&1
fi

# build a .so for the kernel variant
# To redirect stdout to NULL is required
# since the search engine looks for stdout for the evaluation result
make -f makefile-lib filename=OUT__1__6119__.so

cp OUT__1__6119__.so $APP_PATH/struct_ls/

# generate a program to execute and timing the kernel
# Handled by ROSE

best_time="999999999.0"

# run the program multiple times
for (( i="1" ; i <= "$ITER" ; i = i + "1" ))
do
  # The tuned kernel will write timing information into /tmp/peri.result
  # To make the search engine looks for stdout for the evaluation result
  # make -f makefile-lib filename=OUT__1__6119__; cp OUT__1__6119__.so $APP_PATH/struct_ls/
  # The tuned kernel will write timing information into /tmp/peri.result
  # $APP_PATH/test/smg2000 -n 120 120 120 -d 3 > /dev/null
  if [ $? -ne 0 ]; then
    echo "Error: program finishes abnormally!"
    exit 1
  else
    test -f /tmp/peri.result
    if [ $? -ne 0 ]; then
      echo "Error: The temp file storing timing information does not exist!"
      exit 1
    fi
    time='tail -1 /tmp/peri.result | cut -f 1'
  fi
done

# select the smaller one

27
6.4 Results

The new SMG2000 kernel is invoked thousands of times during a typical execution. So instead of using checkpointing/restarting, we used a counter to reduce the point evaluation time. The counter was set to be 1600, which means the execution is terminated once the kernel is be called 1600 times. By doing this, an exhaustive search using GCO became feasible within 40 hours for an input data size of $120 \times 120 \times 120$.

The best performance was achieved at point (0,8,0), which means loop interchange using the lexicographical number 8 (corresponding to an order of $[k, j, si, i]$) improved the performance while tiling and unrolling did not help at all. The best searched point achieved a 1.43x speedup for the kernel (1.18 for the whole benchmark) when compared to the execution time using Intel C/C++ compiler v. 9.1.045 with option `-O2` on the Dell T5400 workstation.

7 Summary

The work presented is ongoing work, and focused on the whole program empirical tuning and the automation of all the required steps to make that work for realistic HPC applications in C, C++, and Fortran. The work to has a few steps that are likely not so easy to fully automate, namely the manipulation of the Makefiles and bash scripts that are required to support the empirical tuning, but it is likely this can be simplified in the future.

The work presented is also immediately focused on providing an infrastructure for the empirical tuning and less on the empirical tuning of any specific applications. SMG2000 was selected somewhat at random, since it is moderately large and we have demonstrated for many years that we can compile it easily with ROSE.
8 Appendix

This section gives quick instructions for installing some needed software packages to establish the end-to-end autotuning system. Please consult individual software releases for detailed installation guidance.

8.1 Patching Linux Kernels with perfctr

This is required before you can install PAPI on Linux/x86 and Linux/x86_64 platforms. Take PAPI 3.6.2 as an example, the steps to patch your kernel are:

- Find the latest perfctr patch which matches your Linux distribution from papi-3.6.2/src/perfctr-2.6.x/patches. For Red Hat Enterprise Linux 5 (or CentOS 5), the latest kernel patch is patch-kernel-2.6.18-92.el5-redhat.

- Get the Linux kernel source rpm package which matches the perfctr kernel patch found in the previous step. You can find kernel source rpm packages from one of the many mirror sites. For example, wget http://altruistic.lbl.gov/mirrors/centos/5.2/updates/SRPMS/kernel-2.6.18-92.1.22.el5.src.rpm

- Install the kernel source rpm package. With a root privilege, simply type:
  
rpm -ivh kernel*.src.rpm

  The command will generate a set of patch files under /usr/src/redhat/-SOURCES.

- Generate the kernel source tree from the patch files. This step may require the rpm-build and redhat-rpm-config packages to be installed first.

  yum install rpm-build redhat-rpm-config # with the root privilege

  cd /usr/src/redhat/SPECS

  rpmbuild -bp --target=i686 kernel-2.6.spec # for x86 platforms

  rpmbuild -bp --target=x86_64 kernel-2.6.spec #for x86_64 platforms

- Copy the kernel source files and create a soft link. Type

  cp -a /usr/src/redhat/BUILD/kernel-2.6.18/linux-2.6.18.1686 /usr/src

  ln -s /usr/src/linux-2.6.18.1686 /usr/src/linux

- Now you can patch the kernel source files to support perfctr. Type

  cd /usr/src/linux

  /path/to/papi-3.6.2/src/perfctr-2.6.x/update-kernel \ 
     --patch=2.6.18-92.el5-redhat

- Configure your kernel to support hardware counters.
cd /usr/src/linux
make clean
make mrproper
yum install ncurses-devel
make menuconfig

Enable all items for performance-monitoring counters support under the menu item of processor type and features.

- Build and install your patched kernel.

make -j4 & make modules -j4 & make modules_install & make install

- Configure perfctr as a device that is automatically loaded each time you boot up your machine.

cd /home/liao6/download/papi-3.6.2/src/perfctr-2.6.x
cp etc/perfctr.rules /etc/udev/rules.d/99-perfctr.rules
cp etc/perfctr.rc /etc/rc.d/init.d/perfctr
chmod 755 /etc/rc.d/init.d/perfctr
/sbin/chkconfig --add perfctr

With the kernel patched, it is straightforward to install PAPI.

cd /home/liao6/download/papi-3.6.2/src
./configure
make
make test
make install # with a root privilege

8.2 Installing BLCR

BLCR (the Berkeley Lab Checkpoint/ Restart library)’s installation guide can be found at http://upc-bugs.lbl.gov/blcr/doc/html/BLCR_Admin_Guide.html. We complement the guide with some Linux-specific information here. It is recommended to use a separate build tree to compile the library.

mkdir buildblcr
cd buildblcr
# explicitly provide the Linux kernel source path
../blcr-0.8.2/configure --with-linux=/usr/src/linux-2.6.18.i686
make
# using a root account
make install
make insmod check

# Doublecheck kernel module installation
# You should find two module files: blcr_imports.ko blcr.ko
ls /usr/local/lib/blcr/2.6.18-prep/

To configure your system to load BLCR kernel modules at bootup:
# copy the sample service script to the right place
cp blcr-0.8.2/etc/blcr.rc /etc/init.d/.

# change the module path inside of it
vi /etc/init.d/blcr.rc

module_dir=@MODULE_DIR@  
module_dir=/usr/local/lib/blcr/2.6.18-prep/

# run the blcr service each time you boot up your machine
chkconfig --level 2345 blcr.rc on

# manually start the service
# error messages like "FATAL: Module blcr not found." can be ignored.
/etc/init.d/blcr.rc restart

Unloading BLCR: FATAL: Module blcr not found.
FATAL: Module blcr_imports not found.

[ OK ]

Loading BLCR: FATAL: Module blcr_imports not found.
FATAL: Module blcr not found.

[ OK ]

References


**Things to Fix in Documentation**

| TODO: update the text when the latest release of HPCToolkit works on 32-bit platforms | 5 |
| We are working on using liveness analysis to further eliminate unnecessary value assignments. | 11 |
| need to discuss the drawbacks, e.g. missing cache warmup; and how that might be addressed in the future by pushing back the checkpoint start and transformations to the checkpointed program (later). | 13 |
| TODO: the input is manually changed from the kernel generated by the autoTuning translator. POET expects normalized loops with special tags, integer loop control variables and ++ operator is not allowed. We will discuss with Qing to either drop these restrictions or use ROSE to normalize the loops automatically. | 16 |
| TODO: the generation of .pt files it not yet automated currently. | 17 |
| This is not an ideal way to tune the application, we will explore better alternatives. | 20 |