IBM Appresentor: automatize the generation, compilation, and profiling of an experimental campaign

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1 Introduction

IBM Appresentor is a design-of-experiments toolchain that automates most of the work required to generate a set of experiments to be profiled with the IBM Platform-Independent Software Analysis tool [1]. It includes makefiles implementing the compilation and instrumentation flow, assuming that IBM Platform-Independent Software Analysis has been installed and its environment variables are set correctly. Additionally, it includes scripts to generate a set of experiments for a target application. These experiments are obtained by varying the set of application parameters. Once the experiment set is generated, compilation instrumentation and execution of all experiments can be booted using commands defined in the makefiles.

The toolchain supports C, C++ and, when using LLVM 3.5.0 and dragonegg, FORTRAN. Parallel applications implemented in OpenMP and MPI are supported. Parallel applications implemented using other paradigms are not supported.

2 Requirements

The scripts to generate the design of experiments require Python. The toolchain has been tested with Python 2.7.14 and Python 2.7.5. The pyDOE package is also required. *IBM Platform-Independent Software Analysis* should be installed along with all its requirements, including LLVM. For FORTRAN applications LLVM 3.5.0, dragonegg, and gcc are required. The toolchain was tested with gcc 4.8.5.

All the requirements of the target application should also be met.

3 Usage

To use the toolchain you can proceed with the following steps:

- 1. If the target application is not released with the DOE-toolchain distribution, install the target application in the benchmark folder (Section 3.1).
- 2. Generate the experiment set (Section 3.2).
- 3. Profile the experiments by compiling and running them (Section 3.2).

3.1 Installing a new benchmark

Let <DOE-root> be the folder where the DOE-toolchain can be found. To install a new application, first of all create a folder:

<DOE-root>/benchmarks/<suiteName>,

where <suiteName> names your benchmark suite. When generating experiments (Section 3.2), you will have to choose a benchmark suite, all applications within that suite will be processed. Then, create the application folder using the application name <appName>:

<DOE-root>/benchmarks/<suiteName>/<appName>.

Within this folder, create the following two subfolders:

- autorun_config
- src

In autorun_config, the definition of the application parameters to be handled by the DOE toolchain must be defined as described in Section 3.1.1. The whole source code of your application should be deployed in the folder src. Source files should be named using extensions that are representative of the source language. Use .c for C files, .cpp or .cc for C++ files, and .f for FORTRAN files.

The IBM Platform-Independent Software Analysis instrumentation is carried out after the linking phase. First, all source files will be compiled to generate LLVM IR files named by replacing the file source file extension with .bc. Second, all these files are linked together into an IR file named main.bc. This file main.bc will be instrumented using IBM Platform-Independent Software Analysis. No source file should be named main.c (or any other extension) because its IR file name would conflict with the overall linked program main.bc.

We recommend to read the parameter passing mechanism described in Section 3.1.2 to clarify how the DOE toolchain will pass compile-time and run-time parameters.

3.1.1 Parameter space definition

The DOE toolchain is meant to automatize the generation and execution of several experiments for a target application. Experiments are executed by varying some compile-time and run-time parameters. Parameters and parameter spaces must be defined in the file:

<DOE-root>/benchmarks/<suiteName>/<appName>/autorun_config/autorun.py,

Different parameter spaces can exist for a single application. For example one may want to have an experimental space for short runs and one for longer runs. Each parameter space has its own name paramSpace> and must be defined in autorun.py. When generating the experiments (Section 3.2) you should specify the name paramSpace> of the space you want to use. We explain the format of autorun.py with an example:

```
####### example configuration file ########
   def getInputParameterRange(dsName):
2
        11 11 11
3
        This Python function defines the input parameter spaces for the
        given application.
5
        Parameters:
6
            dsName: the name of the parameter space.
        Return:
            build_input
                A dictionary listing the compile-time parameters:
10
                    - key -> string, the name of the parameter
11
                    - values -> a list of strings representing the parameter values
            run_input
13
                A dictionary listing the run-time parameters:
14
                    - key -> string, the name of the parameter
                    - values -> a list of strings representing the parameter values
            openmp_threads
17
                The list of parallelism options for OpenMP applications
18
            mpi_procs
                The list of parallelism options for MPI applications
20
21
22
        found = False;
23
24
        if(dsName == "test"):
25
            found = True
26
            print('[INFO] DOE \"test\" found!')
            build_input = {
28
            'PROBLEMSIZE'
                            : [ '33' , '36' , '39' , '42' , '45' ],
29
                        : [ '0.25' , '0.75' , '1.25' , '1.75' , '2.25' ],
            'IDT'
30
                        : [ '9' , '12' , '15' , '18' , '21' ],
            'NITER'
            'PROCS'
                        : [ 'mpi_procs' ]
32
            }
33
            run_input = {
36
```

```
37
            omp_threads = [] # an empty dictionary means no OpenMP
39
           mpi_procs = ['25' , '36', '49', '64' , '81']
40
41
        if(dsName == "alternativeTest"):
            found = True
43
            print('[INFO] DOE \"networkTest\" found!')
44
            build_input = {
                            : [ '36' , '64' , '102' , '162' ],
            'PROBLEMSIZE'
                        : [ '0.5'],
            'IDT'
47
                        : [ '30' ],
            'NITER'
48
            'PROCS'
                        : [ 'mpi_procs' ]
            }
50
51
           run_input = {
            omp_threads = [] # an empty dictionary means no OpenMP
55
56
           mpi_procs = ['9' , '16' , '25' , '36' , '49' , '64']
58
        ############### default, "train
59
        if(not found):# default, "train"
            print('[INFO] DOE not found, using DOE \"train\"!')
           build_input = {
62
63
                           : [ '15', '18', '21', '24', '27'],
            'PROBLEMSIZE'
                        : [ '0.25' , '0.5' , '0.75' , '1.0' , '1.25' ],
            'IDT'
                        : [ '3' , '6' , '9' , '12' , '15' ],
            'NITER'
66
            'PROCS'
                        : [ 'mpi_procs' ]
67
            }
            run_input = {
70
71
72
            omp_threads = []
74
           mpi_procs = ['9', '16', '25', '36', '49']
75
       return build_input, run_input, omp_threads, mpi_procs
77
78
79
   # This function was meant for polyhedral optimizations and is experimental.
   # It defines the dimension of the used data structures.
   # Do not modify it.
   def getDimensionNumber():
83
       return 0
```

The file autorun.py defines the Python function getInputParameterRange

that takes as input the name of the parameter space and returns as output the requires space. In the example there are defined three spaces: 'test' (definition begins at line 25), 'alternativeTest' (definition begins at line 42), and 'train' (definition begins at line 60). If the parameter space name is not found, the 'train' space is returned.

This function returns as output four variables defining the parameter space: built_input, run_input, omp_threads, and mpi_proces. The first part describes the compile-time parameters, the second part describes the run-time parameters and the last two define the parallelism levels for OpenMP and MPI. When declaring parameters in built_input, run_input we recommend not to use the _ character in parameter names. This character will generate problems if the profiles are then processed using IBM Exascale Extrapolator because _ is a special character in the Mathematica language.

3.1.2 Parameter passing

The application parameters are passed from the DOE toolchain to the application in different ways depending on the type of parameter.

Compile-time parameters.

Compile time parameters defined in build_input (in the file autorun.py) are passed through macros handled by the C preprocessor. The macros are defined when clang (or gcc) is invoked by passing the flags -D<paramName> <paramValue> for each parameter listed in the dictionary build_input. The values <paramValue> are varied from experiment to experiment but are always taken from within the list associated to the key <paramName> in the dictionary build_input.

NOTE: The macro definitions -D<paramName> <paramValue> at compile time are handled by the C preprocessor. FORTRAN files will be compiled to LLVM IR by using dragonegg (a gcc plugin). The compile time parameters are still passed as C preprocessor macros as defined previously. Take care that FORTRAN code might include several include statements that are not managed by the C preprocessor. Bugs may arise. Note that, since gcc is used for compiling the FORTRAN files, the C preprocessor is executed also for FORTRAN code. You can replace any include statement with the C preprocessor statement #include if needed.

Run-time parameters.

Parallelism parameters

The variables omp_threads, and mpi_proces returned by getInputParameterRange define the parallelization options for OpenMP an MPI respectively. These variables must be a list of natural number, leave an empty list if the application does not use OpenMP or MPI. The actual parallelism to use may vary from experiment to experiment and it is automatically identified by the DOE toolchain among the ones in the returned list.

When an experiment is run, the number of OpenMP thread is passed as an environment variable while the number of MPI processes is passed when automatically invoking mpirin.

The toolchain enable to pass the value of parallelism parameters also as compile-time or run-time parameter. If you want a parameter paramName> to always assume the value of the parallelism in use, define the list of possible values such to include a single value that refers to the parallelism parameter name. For example in the autorun.py file listed above, the compile time parameter PROCS will always assume the same value of the number of MPI processes in use.

3.1.3 Setting up the Makefile

All the source files that need to be compiled must be specified in an application Makefile:

<DOE-root>/benchmarks/<suiteName>/<appName>/Makefile.

This file must also include the list of the compilation, instrumentation and execution commands defined within the DOE toolchain, and the list of IBM Platform-Independent Software Analysis instrumentation flags. Following an example of this Makefile:

```
include ../../../Makefile.common
2
   ######### GENERAL CONFIGURATION
5
6
   APP_NAME=LU
   LANGUAGE=FORTRAN
9
   SOURCES_C=
10
   SOURCES_CC=
11
   SOURCES_CPP=
   SOURCES_F= bcast_inputs.f
                               error.f exchange_3.f \
13
                           exchange_6.f jacu.f neighbors.f proc_grid.f \
14
                           setbv.f ssor.f buts_vec.f exact.f \
15
                           exchange_4.f init_comm.f l2norm.f nodedim.f \
16
                           read_input.f setcoeff.f subdomain.f \
17
                           blts_vec.f erhs.f exchange_1.f exchange_5.f \
18
                           jacld.f lu.f pintgr.f rhs.f setiv.f \
19
                           verify.f timers.f print_results.f
20
21
   ########## COMPILE-TIME APPLICATION CONFIGURATION
```

```
# Default values of compile-time parameters
   PROBLEMSIZE?=32
   DT?=1
   NITER?=60
26
   PROCS?=9
27
   ######### ADDITIONAL COMPILER FLAGS
29
   APPADDFFLAGS=\
30
           -DPROBLEMSIZE=$(PROBLEMSIZE) -DDT=$(DT) \
31
           -DNITER=$(NITER) -DPROCS=$(PROCS) \
32
            -I$(MPI_DIR)/include
33
   APPADDLDFLAGS=-L$(MPI_DIR)/lib -lmpi -lmpi_mpifh
34
35
36
   include ../../../Makefile.common.PISA
37
   include ../../../Makefile.common.commands
```

For a new application you may copy the above content and edit the following definitions:

- APP_NAME: the name to be given to the executable application executable.
- LANGUAGE: the source-code language (either C, C++, or FORTRAN).
- SOURCES_C: the list of C source files to be compiled.
- SOURCES_CC: the list of C++ source files terminating with the extension .cc to be compiled.
- SOURCES_CPP: the list of C++ source files terminating with the extension .cpp to be compiled.
- SOURCES_F: the list of FORTRAN source files to be compiled.
- The application compile-time parameters (in the example: PROBLEMSIZE, DT, NITER, and PROCS). Assign default values to these variables by using the operator ?= such to enable the DOE toolchain to override their definition when the Makefile will be invoked.
- APPADDFFLAGS: the application specific flags to be passed at compile time. Together with the compile-time flags, you also need to specify the compile-time parameter passing mechanism (i.e. the macro definitions).
- APPADDLDFLAGS: the application-specific flags to be passed to clang when linking.

3.2 Setup and run an experimental campaign

All the scripts automating the setup of an experimental campaign are located in the folder: <DOE-root>/scripts. A design of experiment for a preinstalled benchmark suite can be generated from within the scripts directory by invoking the experiment_generator script. You can execute the command: ./experiment_generator --help for a quick help. The script has to be invoked with the following parameters:

```
> ./experiment_generator --benchmark <suiteName> \
     --designType <doeName> --designSpace <paramSpace>
```

```
[INF0] In order to run the experiment campaign:
[INF0] 1) cd <DOE-root>/results/experiment/<timestamp>
[INF0] 2) make generate.cnls -j20
[INF0] 3) make run.cnls -j8
```

The make command <code>generate.cnls</code> compiles and instruments the application for all the compile-time configurations defined by the DOE and the parameter space. The make command <code>run.cnls</code> executes all the experiments for the required parameter configurations using the instrumented binaries. You can find all the <code>IBM Platform-Independent Software Analysis</code> profiles with the command:

```
> find . -name '*.rawprofile'
```

If you are interested to profile a single application within the benchmark suite, use the make commands gen<appName>.cnls and run<appName>.cnls. You can also profile the native execution of each experiment by invoking

```
> make runnative
```

or

> make runnative<appName>

You will find the set of files storing execution time information with the command:

```
> find . -name '*.time'
```

3.2.1 Design of experiments

The toolchain supports the six following different design of experiments [2]:

 Plackett-Burman: it is a supersaturated DOE based on two parameter levels. For any parameter <parName> in the compile-time, run-time or parallelism parameter whose definition in autorun.py includes more than two levels, the first and last one are used.

- Two-Level-Full-Factorial: it consider that each parameter can assume either a high or a low value. For any parameter cparName in the compiletime, run-time or parallelism parameter whose definition in autorun.py includes more than two levels, the first and last one are used.
- Central-Composite: it is a design of experiments for curve fitting and it considers that each parameter can assume one out of five values {minimum, low, central, high, maximum}. It requires that the parameter space definition in autorum.py lists exactly five values for each parameter (including compile-time, run-time and parallelism parameters).
- Latin-Square: it is a space filling technique that assumes each parameter defined in autorun.py to have the same number of possible values.
- Exhaustive: it consider experiments to cover all possible parameter value combinations, i.e. it covers the whole parameter space.

4 Contributors

- Davide Gadioli
- Giovanni Mariani

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References

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