

Real-Parameter Black-Box Optimization Benchmarking 2009 Software: User Documentation

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1 Purpose

This document provides the user documentation for the Black-Box Optimization Benchmarking (BBOB) tools provided for the GECCO 2009 workshop of the same name.

2 Description of the Software

The software can be downloaded at:

<http://coco.gforge.inria.fr/doku.php?id=bbob-2009>

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It provides:

1. benchmark functions, coded in MATLAB/Octave and C,
2. a single generic function interface to the benchmark functions, coded in MATLAB/Octave and C,
3. the Python post-processing tool,
4. L^AT_EX files templates, and
5. the corresponding documentations.

The first two items are used to run black-box optimization experiments and generate data. The second two are used to obtain a workshop paper collecting the results from the experiments.

3 Running Experiments

A general description of the goals of the workshop, how to perform the experiments using MATLAB/Octave and the output data format can be found in [1].

To run an experiment, an interface function, `fgeneric`, is provided which takes care of writing the output data and which can be used to gather further information on the current experiment (e.g., the current number of function evaluations).

3.1 Running Experiments in C

The interface to `fgeneric` differs from the MATLAB example given in [1], we provide in Figure 1 the equivalent example script in C. A specific folder structure, described in Annex in [1], is needed for running an experiment. This folder structure can be obtained by untarring the archive `createfolders.tar.gz` and renaming the output folder or alternatively by executing the Python¹ module `createfolders` before executing any experiment program. Make sure `createfolders.py` is in your experiment working directory and from the command-line simply do:

```
python createfolders.py PUT_MY_BBOB_DATA_PATH
```

To list other differences between the MATLAB and C code, calls to `fgeneric` specified by a string first argument in MATLAB, are replaced by `fgeneric_string` in C, e.g. `fgeneric('ftarget')` is replaced with `fgeneric.ftarget`. Also, the generic call to `fgeneric(X)` to evaluate candidate vectors is replaced by `fgeneric_evaluate(double * X)` for a single vector and `fgeneric_evaluate_vector(double * XX, unsigned int n, double * result)` for an array of vectors where `XX` is the concatenation of the `n` candidate vectors and `result` is an array of size `n` which contains the resulting function values.

¹To install Python, please refer to 4.1.

Figure 1: exampleexperiment.c: example for benchmarking MY_OPTIMIZER on the noise-free function testbed in C.

```

/*runs an entire experiment benchmarking MY_OPTIMIZER on the noise-free testbed*/

#include <stdio.h>
#include <string.h>
#include <time.h>
#include <stdlib.h>
#include "bbobStructures.h" /* Include all declarations for BBOB calls */

/* include all declarations for your own optimizer here */
void MY_OPTIMIZER(double(*fitnessfunction)(double*), unsigned int dim, double ftarget,
                 unsigned int maxfunevals);

int main()
{
    unsigned int dim[6] = {2, 3, 5, 10, 20, 40};
    unsigned int instances[15] = {1, 2, 3, 4, 5, 1, 2, 3, 4, 5, 1, 2, 3, 4, 5};
    unsigned int idx_dim, ifun, idx_instances;
    clock_t t0 = clock(); time_t Tval;
    ParamStruct params = fgeneric_getDefaultPARAMS();

    srand(time(NULL)); /* used by MY_OPTIMIZER */
    strcpy(params.dataPath, "PUT_MY_BBOB_DATA_PATH");
    /* please run 'python createfolders.py PUT_MY_BBOB_DATA_PATH' beforehand */
    strcpy(params.algName, "PUT ALGORITHM NAME");
    strcpy(params.comments, "PUT MORE DETAILED INFORMATION, SETTINGS ETC");

    for (idx_dim = 0; idx_dim < 6; idx_dim++)
    {
        /*Function indices are from 1 to 24 (noiseless) or from 101 to 130 (noisy)*/
        for (ifun = 1; ifun <= 24; ifun++)
        {
            for (idx_instances = 0; idx_instances < 15; idx_instances++)
            {
                /* Mandatory for each experiment: set DIM, funcId, instanceId*/
                params.DIM = dim[idx_dim];
                params.funcId = ifun;
                params.instanceId = instances[idx_instances];
                fgeneric_initialize(params);

                MY_OPTIMIZER(&fgeneric_evaluate, dim[idx_dim], fgeneric_ftarget(),
                            20*dim[idx_dim]); /* SHORT EXPERIMENTS. */

                printf(" f%d in %d-D, instance %d: FEs=%d,", ifun, dim[idx_dim],
                       instances[idx_instances], fgeneric_evaluations());
                printf(" fbest-ftarget=%.4e, elapsed time [h]: %.2f\n",
                       fgeneric_best() - fgeneric_ftarget(),
                       (double)(clock()-t0)/CLOCKS_PER_SEC/60./60.);

                fgeneric_finalize();
            }
            Tval = time(NULL); printf("    date and time: %s", ctime(&Tval));
        }
        printf("---- dimension %d-D done ----\n", dim[idx_dim]);
    }
    return 0;
}

```

3.2 Collected Data

If run correctly, the data for an experiment will be located in a folder that we will call `DATAPATH` in the following. The post-processing tool described in Section 4 takes `DATAPATH` as input argument and will recursively look for '.info' files which contain the necessary information about the experiments, see Appendix E in [1].

4 Generate the Workshop Paper

The post-processing tool `bbob_pproc` generates image files and \LaTeX tables from the raw experimental data obtained as described previously in 3. These files and tables will be a part of your workshop paper. The workshop paper itself is composed of the description of your algorithm, any additional information you feel necessary to describe your experiments, and the aforementioned results from the post-processing procedure. To ease the procedure of writing the paper we include a \LaTeX template for each testbed within the provided software.

The entire post-processing tool is written in Python and requires Python to be installed on your machine. The minimal software requirements for using the post-processing tool are Python (2.5.2), Matplotlib (0.91.2) and Numpy (1.0.4). In the following, we explain how to obtain and install the required software for different systems (Linux, Windows, Mac OS) and which steps you have to perform to run the post-processing on your data.

4.1 Installing the Software

While the `bbob_pproc` source files are provided, you need to install Python and its libraries Matplotlib and Numpy. We recommend using Python 2.5 and not a higher version (2.6 or 3.0) since the necessary libraries are not (yet) available and the code is not verified. For all operating systems the packages can be found at the following locations:

- Python: <http://www.python.org/download/releases/2.5.4/>,
- Numpy: <http://sourceforge.net/projects/numpy/>,
- Matplotlib: <http://sourceforge.net/projects/matplotlib/>.

We recommend the use of the latest versions of Matplotlib (0.98.5.2) and Numpy (1.2.1).

4.1.1 Linux

In most common Linux distributions Python (but not Numpy or Matplotlib) is already part of the installation. If not, use your favorite package manager to install Python (`python`), Numpy (package name: `python-numpy`) and Matplotlib (package name: `python-matplotlib`) and their dependencies. If your distribution and repositories are up-to-date, you should have Python 2.5.2, Matplotlib 0.91.2 and Numpy 1.0.4. Though those are not the most recent versions of each package, they meet the minimal software requirements to make the BBOB software work. If needed, you can alternatively download sources and compile binaries. Python and the latest versions of Matplotlib and Numpy can be downloaded

from the links in Sect 4.1. A dependency for the Linux version of Matplotlib is libpng, which can be obtained at <http://www.libpng.org/>. You will then need to properly install the downloaded packages before you can use them. Please refer to the corresponding package installation pages.

4.1.2 Windows

For installing Python under Windows, please go to the Python link in Sect 4.1 and download python-2.5.4.msi . This file requires the Microsoft Installer, which is a part of Windows XP and later releases. If you don't have the Microsoft Installer, there is a link for the download provided at the same page. After installing Python, it is recommended to first install Numpy and then Matplotlib. Both can be installed with the standard .exe files (respectively numpy-1.2.1-win32-superpack-python2.5.exe and matplotlib-0.98.5.2.win32-py2.5.exe). These files can be obtained from the provided SourceForge links in Sect 4.1.

4.1.3 Mac OS

Mac OS X comes with Python pre-installed, the version might be older than 2.5 though. It is recommended to upgrade Python by downloading and installing a newer version. To do this, if you have Mac OS X 10.3 and later you can download the disk image file python-2.5.4-macosx.dmg containing universal binaries from the Python download page, see Sect 4.1. More information on the update of Python on Mac OS can be found at this location: <http://www.python.org/download/mac/>². Open the disk image and use the installer³ You will then need to download and install Numpy and Matplotlib from the SourceForge links listed in Sect 4.1.

4.2 Using the bbob_pproc Package

When you have obtained experimental data and want to perform the post-processing you will need to download and un-archive the bbob_pproc package on your system. The package can be obtained from <http://coco.gforge.inria.fr/doku.php?id=bbob-2009>. Then, to post-process the data, make sure the data folder DATAPATH generated by your experiments is in the current working directory and execute:

```
python path_to_postproc_code/bbob_pproc/run.py DATAPATH
```

from a shell or the command window (Windows). Note that in Windows the path separator '\' must be used instead of '/'. The folder path_to_postproc_code is the one where the provided post-processing software was un-archived. The above command will create the folder ppdata in the current working directory, which will contain the post-processed data in the form of figures and L^AT_EX files for the tables. This process might take a few minutes. If the verbosity is off (default), you will only get the following message to indicate that the post-processing is finished:

²The discussion over IDLE for Leopard user (<http://wiki.python.org/moin/MacPython/Leopard>) is not relevant for the use of bbob_pproc package.

³Following this step will leave the pre-installed Python on the system and install the MacPython 2.5.4 distribution. MacPython contains a Python installation as well as some Mac-specific extras.

Output data written to folder `ppdata`.

If you are familiar with Python and want to run the post-processing directly from a Python shell you have to execute:

```
>>> import bbob_pproc
>>> bbob_pproc.main('DATAPATH')
```

This first command requires that the path to `bbob_pproc` is in the Python search path. The resulting `ppdata` folder now contains a number of `TEX`, `eps`, `png` files.

4.2.1 Help for the `bbob_pproc` Package

Additional help for the `bbob_pproc` package can be obtained by executing the following command in a shell:

```
python path_to_postproc_code/bbob_pproc/run.py -h
```

This will give a short summary of the package and additional options for executing the post-processing. The code documentation can be found in the folder `path_to_postproc_code/pydoc` within the provided software package. It gives a detailed explanation of the Python code.

4.3 Using the `LATEX` Templates

The files `templateBBOBarticle.tex` and `templateBBOBnoisyarticle.tex` are provided. If compiled correctly using `LATEX`, they will generate documents collecting and organizing the output from the BBOB post-processing tool. Each of the templates has a given page organization optimized for the presentation of the results for a each testbed. In the case that you have results on both testbeds, we recommend that you generate and submit two separate documents.

To compile a document, you need to make sure that:

1. you have a working `LATEX` distribution on your computer⁴,
2. you are in the working directory (containing the folder `ppdata` containing all the output from the BBOB post-processing tool),
3. `templateBBOBarticle.tex`⁵, `bbob.bib` and `sig-alternate.cls` are in the working directory (all files are provided with the software),

Then, simply execute:

```
latex templateBBOBarticle
bibtex templateBBOBarticle
latex templateBBOBarticle
latex templateBBOBarticle
```

in a shell or open the chosen template file in your favorite `LATEX` editor and compile it with the `LATEX` option. The document `templateBBOBarticle.dvi` will be generated in the format required for a GECCO workshop paper. Now you need to complete this document by filling and adding in the template file

⁴<http://www.latex-project.org/>

⁵If you used the noisy benchmark functions, use `templateBBOBnoisyarticle.tex` instead.

all the meta-information that is needed, your own information as author, a complete description of your algorithm and a description of how you chose your algorithm parameters (see also [1]). As for your bibliographical references you can either add your references to `bbob.bib` or include your *bib*-file in the `bibliography`-statement at the end of the template file. Please refer to <http://www.sheridanprinting.com/typedept/gecco3.htm> and <http://www.acm.org/sigs/publications/sigguide-v2.2sp> for more details on the formatting of your L^AT_EX document.

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References

- [1] N. Hansen, A. Auger, S. Finck, and R. Ros. Real-parameter black-box optimization benchmarking 2009: Experimental setup. Technical Report RR-6828, INRIA, 2009.