1 Purpose

The COmparison of Continuous Optimizers (COCO) software is a benchmarking software to render easier experiments in the field of continuous optimization. A post-processing Python package generates tables and figures to be included in a research paper template presenting all results.

The COCO software was used for the GECCO workshops named Black-Box Optimization Benchmarking (BBOB-2009 and BBOB-2010). The efforts of BBOB-2009 resulted in thirty-eight accepted workshop papers presenting results of state-of-the-art algorithms.

SF is with the Research Center PPE, University of Applied Science Vorarlberg, Hochschulstrasse 1, 6850 Dornbirn, Austria

RR is with the TAO Team of INRIA Saclay–Île-de-France at the LRI, Université-Paris Sud, 91405 Orsay cedex, France

Available at http://coco.gforge.inria.fr
The COCO software provides:

1. a single generic function interface \texttt{fgeneric} to the benchmark functions of BBOB-2010, coded in \texttt{MATLAB}/\texttt{GNU Octave}, \texttt{C/C++} and \texttt{PYTHON},
2. Java Native Interface classes to use \texttt{fgeneric} in \texttt{JAVA},
3. the \texttt{PYTHON} post-processing module \texttt{bbob.pproc},
4. \LaTeX{} templates to generate papers, and
5. the corresponding documentation.

The practitioner in BBO who wants to benchmark one or many algorithms on the BBOB-2010 testbeds has to download \texttt{COCO}, interface the algorithms to call the test functions in the testbed and use the post-processing tools. The most substantial part is to render the interface of the considered algorithms compatible with the interface of our objective/fitness function implementation in \texttt{fgeneric}.

We describe the different steps for obtaining a complete workshop paper for an algorithm, thus allowing us to present the architecture of \texttt{COCO}. We also present additional facilities implemented for the comparison of the results of the many algorithms submitted. Section 2 presents the experimental framework software used to generate benchmarking data. Section 3 describes the post-processing facilities of \texttt{COCO}, namely the \texttt{PYTHON} package \texttt{bbob.pproc}. Section 4 briefly describes the process of compiling a paper regrouping all the post-processed results.

## 2 Experimental Framework Software

The experimental framework software mainly consists in the implementation of the methodology presented in [1]. The software is centered on the interface function, \texttt{fgeneric}.

We describe the format of the output data files and the content of the files as they are written by \texttt{fgeneric}. These files are to be analysed with the provided post-processing tools that are described in Section 3. To display an example of the use of \texttt{fgeneric}, we provide two example scripts. Executing the \texttt{MATLAB} scripts provided in Listings 2 and 3 results in testing an algorithm — \texttt{MY_OPTIMIZER} in the examples, see Listing 1 — on the noiseless testbed of BBOB-2010 and displaying measures of the time complexity of an algorithm respectively. In Listing 2, lines 6 to 10 set variables used by \texttt{fgeneric}. The whole set of experiment on the noiseless testbed is done by looping over the lines 18 to 36.

The function \texttt{fgeneric} outputs the results of the experiments, also it provides a single interface to any of the test functions of the BBOB-2010 testbeds. Once \texttt{fgeneric} is loaded into memory, the initialization process, see line 21 in Listing 2, sets all variables internal to \texttt{fgeneric}: the test function considered, the instance considered, the output directory. Later calls to \texttt{fgeneric} evaluate the chosen test function at the point $\vec{x}$ given as input argument, see line 11 of Listing 1. Necessary finalization operations are effected by using the command \texttt{fgeneric\textquotesingle finalize\textquotesingle} in \texttt{MATLAB}, see line 31 in Listing 2.
Listing 1: `MY_OPTIMIZER.m`: Monte Carlo search in MATLAB. At each iteration, 200 points are sampled and stored in a matrix of size `DIM` × 200 so as to reduce loops and function calls within MATLAB and therefore improve its efficiency.

```matlab
function MY_OPTIMIZER(FUN, DIM, ftarget, maxfunevals)
    % MY_OPTIMIZER(FUN, DIM, ftarget, maxfunevals)
    % samples new points uniformly randomly in [-5,5]^DIM
    % and evaluates them on FUN until ftarget of maxfunevals
    % is reached, or until 1e8 * DIM fevals are conducted.
    % Relies on FUN to keep track of the best point.
    maxfunevals = min(1e8 * DIM, maxfunevals);
    popsize = min(maxfunevals, 200);
    for iter = 1:ceil(maxfunevals/popsie)
        feval(FUN, 10 * rand(DIM, popsize) - 5);
        if feval(FUN, 'fbest') < ftarget % task achieved
            break;
        end
        % if useful, modify more options here for next start
    end
    % if useful, modify more options here for next start
end
```

In Listing 2, the function $f_8$ is tested in 2, 3, 5, 10, 20, and 40-D. The `while` loop from line 15 to 18 make the runs last thirty seconds.

### 2.1 Running Experiments in C or C++

The interface to `fgeneric` differs from the MATLAB example given in [1], we provide in Listing 4 the equivalent example script in C. A specific folder structure is needed for running an experiment. While creating the folder structure was handled by running `fgeneric` in MATLAB, this is not the case using the C code. This folder structure can be obtained by un-tarring the archive `createfolders.tar.gz` and renaming the output folder or alternatively by executing the Python module `createfolders` before executing any experiment program. Make sure `createfolders.py` is in your current working directory and from the command-line simply do:

```
python createfolders.py PUT_MY_BBOB_DATA_PATH
```

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 np, double *result)` for an array of vectors where `XX` is the concatenation of the `np` candidate vectors and `result` is an array of size `np` which contains the resulting function values.

The code provided can be compiled in C or C++.

### 2.2 Running Experiments in Java

The class `JNIfgeneric` implements an interface for using the C-implementation of `fgeneric`. Methods `fgeneric_string` in C are replaced by `JNIfgeneric_string`, except for the initialization `JNIfgeneric_initBBOB(...)` and finalization `JNIfgeneric_exitBBOB()`.
Listing 2: exampleexperiment.m: script for benchmarking MY_OPTIMIZER, see Listing 1, for BBOB-2010 on the noiseless function testbed in MATLAB/GNU OCTAVE

```matlab
% runs an entire experiment for benchmarking MY_OPTIMIZER
% on the noise-free testbed. fgeneric.m and benchmarks.m
% must be in the path of Matlab/Octave
% CAPITALIZATION indicates code adaptations to be made
addpath('PUT_PATH_TO_BBOB/matlab'); % should point to fgeneric.m etc.
datapath = 'PUT_MY_BBOB_DATA_PATH'; % different folder for each experiment
opt.algName = 'PUT ALGORITHM NAME';
opt.comments = 'PUT MORE DETAILED INFORMATION, PARAMETER SETTINGS ETC';
maxfunevals = '20 * dim'; % SHORT EXPERIMENT, takes overall three minutes
more off; % in octave pagination is on by default
t0 = clock;
rand('state', sum(100 * t0)); % initializes the pseudo-random generator
for dim = [2,3,5,10,20,40] % small dimensions first, for CPU reasons
    for ifun = benchmarks('FunctionIndices') % or benchmarksnoisy(...) for instance = [1:15] % Instances 1 to 15
        fgeneric('initialize', ifun, iinstance, datapath, opt);
        MY_OPTIMIZER('fgeneric', dim, fgeneric('ftarget'), eval(maxfunevals));
        disp(sprintf([' f%d in %d-D, instance %d: FEs=%d,' ...
                     ' fbest-ftarget=%.4e, elapsed time [h]: %.2f'], ...
                     ifun, dim, iinstance, ...
                     fgeneric('evaluations'), ...
                     fgeneric('fbest') - fgeneric('ftarget'), ...
                     etime(clock, t0)/60/60));
        fgeneric('finalize');
    end
    disp([' date and time: ' num2str(clock, ' %.0f')]);
end
disp(sprintf('---- dimension %d-D done ----', dim));
```

3 Post-Processing the Experimental Data

The PYTHON post-processing tool, called `bbob_pproc` in BBOB-2010 generates image files and \LaTeX{} tables from the raw experimental data obtained as described previously in Section 2.

The entire post-processing tool requires that PYTHON is installed on your machine. The minimal software requirements for using the post-processing tool are PYTHON (2.6), Matplotlib (0.99) and Numpy (1.2). Please note that the packages do not work with Python 3.X but only with Python 2.X. The installation of the software is described in Appendix A.

Overview of the `bbob_pproc` Package

We present here the content of the latest version of the `bbob_pproc` package (version 10.7).
Listing 3: `exampletiming.m`: script for measuring the time complexity of `MY_OPTIMIZER`, see Listing 1, for BBOB-2010 in Matlab/Gnu Octave

```matlab
% runs the timing experiment for MY_OPTIMIZER. fgeneric.m
% and benchmarks.m must be in the path of MATLAB/Octave
addpath('PUT_PATH_TO_BBOB/matlab'); % should point to fgeneric.m etc.
more off; % in octave pagination is on by default

timings = [];
runs = [];
dims = [];
for dim = [2,3,5,10,20,40]
    nbrun = 0;
    ftarget = fgeneric('initialize', 8, 1, 'tmp');
    tic;
    while toc < 30 % at least 30 seconds
        MY_OPTIMIZER(@fgeneric, dim, ftarget, 1e5); % adjust maxfunevals
        nbrun = nbrun + 1;
    end % while
timings(end+1) = toc / fgeneric('evaluations');
dims(end+1) = dim; % not really needed
runs(end+1) = nbrun; % not really needed
fgeneric('finalize');
disp(['Dimensions: ' sprintf(' %11d ', dims)]);
    disp([' run: ' sprintf(' %11d ', runs)]);
    disp([' times [s]: ' sprintf(' %11.1e ', timings)]);
end
```

`rungeneric.py` is the main interface of the package that performs different routines listed below,

`rungeneric1.py` post-processes data from one single algorithm and outputs figures and tables included in the templates `template1generic.tex`, `noisytemplate1generic.tex`,

`rungeneric2.py` post-processes data from two algorithms using modules from `bbob_pproc.comp2` and outputs comparison figures and tables included in the template `template2generic.tex`, `noisytemplate2generic.tex`,

`rungenericmany.py` post-processes data from \( n > 0 \) algorithms using modules from `bbob_pproc.compall` and outputs comparison figures and tables included in the template, `template3generic.tex`, `noisytemplate3generic.tex`.

`genericsettings.py` defines generic settings for the output figures and tables,

`bwsettings.py` defines generic settings with figures in grayscale,

`pproc.py` defines the classes `DataSetList` and `DataSet` which are the main data structures that we use to gather the experimental raw data,

`dataoutput.py` contain routine to output instances of `DataSet` in PYTHON-formatted data files,

---

2 for \( n \) larger than ten, modifications of the template will be required
Listing 4: exampleexperiment.c: script for benchmarking MY_OPTIMIZER, for BBOB-2010 on the noiseless function testbed in C

```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 idx_dim, ifun, instance;
  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 (instance = 1; instance <= 15; instance++)
      {
        /* Mandatory for each experiment: set DIM, funcId, instanceId*/
        params.DIM = dim[idx_dim];
        params.funcId = ifun;
        params.instanceId = instance;
        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=%lu, ", ifun, dim[idx_dim],
                instance, fgeneric_evaluations());
        printf(" fbest-ftarget=\%1.4e, elapsed time (h): \%3.2f\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 ----
", dim[idx_dim]);
  }

  return 0;
}
```
readalign.py, bootstrap.py contain routines for the post-processing of the raw experimental data,
pptex.py defines some routines for generating $\LaTeX$ tables,
ppfig.py defines some routines for generating figures,
ppfigdim.py, pptable.py, pprdistrib.py, pplegloss.py are used to produce figures and tables presenting the results of one algorithm,
bbob.pproc.compall is a sub-package which contains modules for the comparison of the performances of algorithms, routines in this package can be called using the interface of rungenericmany.py,
bbob.pproc.comp2 is a sub-package which contains modules for the comparison of the performances of two algorithms, routines in this package can be called using the interface of rungeneric2.py.

3.1 Using the bbob.pproc Package

To perform the post-processing on the experimental data, the bbob.pproc package needs to be downloaded\(^3\) and un-archived.

The main interface is in rungeneric.py and will behave differently depending on the number of folders given as input arguments (each corresponding to the data of a different algorithms).

If one folder, DATAPATH, containing all data generated by the experiments for one algorithm is in the current working directory, the post-processing is executed by the following command:

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

from a shell\(^4\), the folder path_to_postproc_code is the one where the provided post-processing software was un-archived.

The above command create the folder with the default name ppdata/DATAPATH in the current working directory, which contain the post-processed data in the form of figures and $\LaTeX$ files for the tables. This process might take a few minutes.

To run the post-processing directly from a Python shell, the following commands need to be executed:

```
>>> import bbob.pproc as bb
>>> bb.rungeneric.main('DATAPATH'.split())
```

This first command loads bbob.pproc into memory and requires that the path to the package is in the Python search path.

The resulting ppdatapath/DATAPATH folder now contains a number of $\LaTeX$, eps, pdf files.

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/rungeneric.py -h
```

\(^3\)The package can be obtained from http://coco.gforge.inria.fr/.

\(^4\)Note that in Windows the path separator `/` must be used instead of `/`
In particular, this command describes the additional options for the execution of the post-processing. The code documentation can be found in the folder `path_to_postproc_code/pydoc` within the provided software package.

### 3.2 Comparison of Algorithms

The sub-package `bbob_pproc.compall` and `bbob_pproc.comp2` (v10.0 and more recent) from `bbob_pproc` provide facilities for the generation of tables and figures comparing the performances of algorithms tested using COCO.

The post-processing works with data folders as input argument, with each folder corresponding to the data of an algorithm. Supposing you have the folders `ALG1`, `ALG2` and `ALG3` containing the data of algorithms `ALG1`, `ALG2` and `ALG3`, you will need to execute from the command line:

```
python path_to_postproc_code/bbob_pproc/rungeneric.py ALG1 ALG2 ALG3
```

This assumes the folders `ALG1`, `ALG2` and `ALG3` are in the current working directory. In this case, the folders contain a number of files with the pickle extension which contain PYTHON-formatted data or the raw experiment data with the info, dat and tdat extensions. Running the aforementioned command will generate the folder `ppdata` containing comparison figures and tables.

Outputs appropriate to the comparison of only two algorithms can be obtained using `bbob_pproc` by executing from the command line:

```
python path_to_postproc_code/bbob_pproc/rungeneric.py ALG0 ALG1
```

This assumes the folders `ALG0` and `ALG1` are in the current working directory. Running the aforementioned command will generate the folder `ppdata` containing the comparison figures.

To run the post-processing from a Python shell, the following commands need to be executed:

```python
>>> import bbob_pproc as bb
>>> bb.rungeneric.main('ALG1 ALG2 ALG3'.split())
```
or:

```python
>>> import bbob_pproc as bb
>>> bb.rungeneric.main('ALG0 ALG1'.split())
```

The from...import... command loads package into memory and requires that the path to the package is in the PYTHON search path. Call to the `main` method runs the whole post-processing script.

### 3.3 rungeneric1.py, rungeneric2.py, rungenericmany.py

Instead of using the `rungeneric.py` interface, the user can directly use the sub-routines `rungeneric1.py`, `rungeneric2.py` or `rungenericmany.py` to generate some post-processing output. These sub-routines are to be used in the same way as `rungeneric.py`:

```
python path_to_postproc_code/bbob_pproc/rungeneric1.py DATAPATH
```
or:
python path_to_postproc_code/bbob_pproc/rungeneric2.py ALG0 and ALG1
or:
python path_to_postproc_code/bbob_pproc/rungenericmany.py ALG1 ALG2 ALG3...

4 Generating a Paper

template1generic.tex, template2generic.tex and template3generic.tex are the template \LaTeX files that include all the figures and tables presenting the result of an algorithm on the noiseless testbed of BBOB-2010.

noisytemplate1generic.tex, noisytemplate2generic.tex and noisytemplate3generic.tex are the same template \LaTeX files for the noisy testbed of BBOB-2010. If compiled correctly using \LaTeX, it generates documents collecting and organizing the output from bbob_pproc.

To compile a document, one needs:
1. to have a working \LaTeX distribution\footnote{http://www.latex-project.org/},
2. to be in the correct working directory (containing the folder pndata that includes all the output from the bbob_pproc),
3. that templateXgeneric.tex, where X is 1, 2 or 3\footnote{template1.tex for showing the results of one algorithm, template2.tex, template3.tex for showing comparison results}, bbob.bib are in the working directory (all files are provided with the software),

Then the following commands needs to be executed in a shell:

latex templateXgeneric
bibtex templateXgeneric
latex templateXgeneric
latex templateXgeneric

The document templateXgeneric.dvi is then generated in a generic article format. The \BibTeX file bbob.bib includes the references to the BBOB-2010 experimental set-up and documentation.

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- Alvaro Fialho for providing the C++ version of the code.
References


A Installing bbbob_pproc

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 (2.6), Matplotlib (0.98) and Numpy (1.2). 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.

While the bbbob_pproc source files are provided, you need to install Python and its libraries Matplotlib and Numpy. We recommend using Python 2.6 or 2.7 and not a higher version (3.0, 3.1) since the necessary libraries are not available and the code is not verified. The code is currently tested using Python (2.6.5), Matplotlib (0.99.1.1) and Numpy (1.3.0).

A.1 Downloading the Packages

For all operating systems the packages can be found at the following locations:

- Python: http://www.python.org/download/releases/,
- Numpy: http://sourceforge.net/projects/numpy/,

Not all packages are distributed for all operating systems with 64-bit architecture. Installing 32-bit versions of the packages whether your architecture is 32 or 64-bit is possible though.

A.2 Installing on Linux

In most common Linux distributions Python (not Numpy or Matplotlib) is already part of the installation. If not, use your favorite package manager to install Python (package name: python), Numpy (python-numpy) and Matplotlib (package name: python-matplotlib) and their dependencies. If your distribution and repositories are up-to-date, you should have at least Python (2.6.5), Matplotlib (0.99.1.1) and Numpy (1.3.0). Though those are not the most recent versions of each package, they meet the minimal software requirements to make the BBOB-2010 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 Section A.1. A dependency for the Linux version of Matplotlib is libpng, which can be obtained at http://www.libpng.org/. You then need to properly install the downloaded packages before you can use them. Please refer to the corresponding package installation pages.
A.3 Installing on Windows

For installing Python under Windows, please go to the Python link in Section A.1 and download python-2.6.X.msi. This file requires the Microsoft Installer, which is a part of Windows XP and later releases. After installing Python, it is recommended to first install Numpy and then Matplotlib. Both can be installed with the standard .exe files which are respectively

- numpy-1.5.0b1-win32-superpack-python2.6.exe and,
- matplotlib-1.0.0.win32-py2.6.exe.

These files can be obtained from the provided SourceForge links in Section A.1.

A.4 Installing on Mac OS

Mac OS X comes with Python pre-installed, the version might be older than 2.6 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.X.macosx10.3.dmg containing universal binaries from the Python download page, see Section A.1. More information on the update of Python on Mac OS can be found at this location: http://www.python.org/download/mac/. Open the disk image and use the installer. Following this step leave the pre-installed Python on the system and install the MacPython 2.X distribution. MacPython contains a Python installation as well as some Mac-specific extras.

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