SBpipe documentation

Release 4.6.0

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USER MANUAL

1.1 Metadata

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Citation:

Dalle Pezze, P and Le Novère, N. (2017) *BMC Systems Biology* **11**:46. SBpipe: a collection of pipelines for automating repetitive simulation and analysis tasks. DOI:10.1186/s12918-017-0423-3 (https://doi.org/10.1186/s12918-017-0423-3)

1.2 Introduction

This package contains a collection of pipelines for dynamic modelling of biological systems. It aims to automate common processes and speed up productivity for tasks such as model simulation, single/double parameter scan, and parameter estimation.

1.2.1 Requirements

In order to use SBpipe, the following software must be installed:

- Python 2.7+ or 3.4+ https://www.python.org/
- R 3.3.0+ https://cran.r-project.org/

SBpipe can work with the following simulators:

- COPASI 4.19+ http://copasi.org/ (for model simulation, parameter scan, and parameter estimation)
- Python (directly or as a wrapper to call models coded in any programming language)

If LaTeX/PDF reports are also desired, the following software must also be installed:

• LaTeX 2013

1.2.2 How to install SBpipe on GNU/Linux

Install COPASI

As of 2016, COPASI is not available as a package in GNU/Linux distributions. Users must add the path to COPASI binary files manually editing the GNU/Linux \$HOME/.bashrc file as follows:

```
# Path to CopasiSE (update this accordingly)
export PATH=$PATH:/path/to/CopasiSE/
```

The correct installation of CopasiSE can be tested with:

```
# Reload the .bashrc file
$ source $HOME/.bashrc

$ CopasiSE -h
COPASI 4.19 (Build 140)
```

Install LaTeX

Users are recommended to install LaTeX/texlive using the package manager of their GNU/Linux distribution. On GNU/Linux Ubuntu machines the following package is required:

```
texlive-latex-base
```

The correct installation of LaTeX can be tested with:

```
$ pdflatex -v
pdfTeX 3.14159265-2.6-1.40.16 (TeX Live 2015/Debian)
kpathsea version 6.2.1
Copyright 2015 Peter Breitenlohner (eTeX)/Han The Thanh (pdfTeX).
```

Install SBpipe via Miniconda3/Anaconda3

Users need to download and install Anaconda3 (https://www.continuum.io/downloads) or Miniconda3 (https://conda.io/miniconda.html).

1st Method

The following procedure will install SBpipe in a conda environment:

```
# create a new environment `sbpipe`
$ conda create -n sbpipe

# activate the environment. The following line can be
# added to the .bashrc file to skip the activation
# of this environment every time SBpipe is used.
$ source activate sbpipe

# install sbpiper (necessary R dependencies) and sbpipe
$ conda install sbpiper -c conda-forge -c defaults
$ conda install sbpipe -c pdp10 -c conda-forge -c fbergmann -c defaults
# install snakemake (optional)
$ conda install -c bioconda snakemake
```

No further step is needed.

2nd Method

Alternatively, for those users who would like to have an easy access to the package documentation and test suite, it could be useful to store SBpipe in a custom path. To do so, SBpipe should be downloaded from the website or cloned using git. From a GNU/Linux shell, type:

```
$ cd path/to/sbpipe

# install dependencies into isolated environment using Anaconda3/Miniconda3
$ conda env create --name sbpipe --file environment.yaml

# activate environment. The following line can be
# added to the .bashrc file to skip the activation
# of this environment every time SBpipe is used.
$ source activate sbpipe

# install snakemake (optional)
$ conda install -c bioconda snakemake
```

To run sbpipe from any shell, users need to add 'sbpipe/scripts' to their PATH environment variable by adding the following lines to their \$HOME/.bashrc file:

```
# SBPIPE (update this accordingly)
export PATH=$PATH:/path/to/sbpipe/scripts
```

The .bashrc file should be reloaded to apply the previous edits:

```
# Reload the .bashrc file
$ source $HOME/.bashrc
```

Install SBpipe's dependencies via provided scripts

For this type of installation, SBpipe should be downloaded from the website or cloned using git. Users need to make sure that the package python-pip and r-base are installed. The correct installation of Python and R can be tested by running the commands:

```
$ python -V
Python 3.6.4
$ pip -V
pip 9.0.1 from /home/ariel/.local/lib/python3.6/site-packages (python 3.6)

$ R --version
R version 3.4.1 (2017-06-30) -- "Single Candle"
Copyright (C) 2017 The R Foundation for Statistical Computing
Platform: x86_64-pc-linux-gnu (64-bit)
```

The next step is the installation of SBpipe dependencies. To install Python dependencies on GNU/Linux, run:

```
$ cd path/to/sbpipe
$ ./install_pydeps.py
```

To install SBpipe R dependencies on GNU/Linux, run:

```
$ cd path/to/sbpipe
$ R
# Inside R environment, answer 'y' to install packages locally
> source('install_rdeps.r')
```

Finally, to run sbpipe from any shell, users need to add 'sbpipe/scripts' to their PATH environment variable by adding the following lines to their \$HOME/.bashrc file:

```
# SBPIPE (update this accordingly)
export PATH=$PATH:/path/to/sbpipe/scripts
```

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The .bashrc file should be reloaded to apply the previous edits:

```
# Reload the .bashrc file
$ source $HOME/.bashrc
```

NOTE: If R package dependencies must be compiled, it is worth checking that the following additional packages are installed in your machine: build-essential, liblapack-dev, libblas-dev, libcairo-dev, libssl-dev, libcurl4-openssl-dev, and gfortran. These can be installed using the package manager coming with your distribution. Other packages might be needed, depending on R dependencies. After installing these packages, install_rdeps.r must be executed again.

Python bindings code for COPASI models (optional)

If Python bindings for COPASI are installed, SBpipe automatically checks whether the COPASI model can be loaded and executed, before generating the data. As of January 2018, this code is released for Python 2.7 and Python 3.6 on the COPASI website and Anaconda Cloud.

1.2.3 How to install SBpipe on Windows

Installation of COPASI and LaTeX

Windows users need to install the Windows versions of COPASI and LaTeX MikTeX https://miktex.org/.

Install MINGW

We advise users to install Git for Windows https://git-for-windows.github.io/ as a simple Shell (MINGW) running on Windows. Leave the default setting during installation.

Preparation of SBpipe and COPASI with MINGW

Once Git for Windows is started, a Shell-like window appears and enables users to run commands. The first step is to clone SBpipe from GitHub using the command:

```
$ git clone https://github.com/pdp10/sbpipe.git
```

We now need to set up the SBpipe environment variable:

```
$ touch .bashrc
$ wordpad .bashrc
```

A Wordpad window should be visible, loading the file .bashrc . The following lines must be copied into this file:

```
#!/bin/bash/
# SBPIPE
export PATH=$PATH:~/sbpipe/scripts

# COPASI (update this accordingly. Use \ to escape spaces)
export PATH=/path/to/copasi/bin/:$PATH

# Optional: activate Anaconda3 environment for SBpipe automatically source activate sbpipe
```

Save the file and close wordpad. Now you should reload the .bashrc file to apply the previous changes:

```
# Reload the .bashrc file
$ source $HOME/.bashrc
```

Install SBpipe via Miniconda3/Anaconda3

See corresponding section for GNU/Linux.

1.2.4 Verify SBpipe installation

The correct installation of SBpipe and its dependencies can be verified by running the following commands. For the correct execution of all tests, LaTeX must be installed.

```
# SBpipe version:
$ sbpipe -V
sbpipe 4.6.0
```

```
# run model simulation using COPASI (see results in tests/copasi_models):
$ cd path/to/sbpipe/tests
$ nosetests test_copasi_sim.py --nocapture --verbose
```

```
# run all tests:
$ nosetests test_suite.py --nocapture --verbose
```

```
# regenerate the manuscript figures (see results in tests/insulin_receptor):
$ nosetests test_suite_manuscript.py --nocapture --verbose
```

1.3 How to use SBpipe

1.3.1 Preliminary configuration steps

Pipelines using COPASI

Before using these pipelines, a COPASI model must be configured as follow using COPASI UI:

pipeline: simulation

- Tick the flag executable in the Time Course Task.
- Select a report template for the Time Course Task.
- Save the report in the same folder with the same name as the model but replacing the extension .cps with .csv (extensions .txt, .tsv, or .dat are also accepted by SBpipe).

pipelines: single or double parameter scan

- Tick the flag executable in the Parameter Scan Task.
- Select a report template for the Parameter Scan Task.
- Save the report in the same folder with the same name as the model but replacing the extension .cps with .csv (extensions .txt, .tsv, or .dat are also accepted by SBpipe)

pipeline: parameter estimation

- Tick the flag executable in the Parameter Estimation Task.
- Select the report template for the Parameter Estimation Task.

• Save the report in the same folder with the same name as the model but replacing the extension .cps with .csv (extensions .txt, .tsv, or .dat are also accepted by SBpipe)

For tasks such as parameter estimation using COPASI, it is recommended to move the data set into the folder Models/ so that the COPASI model file and its associated experimental data files are stored in the same folder.

Pipelines running Python models

pipelines: model simulation

- The model coded in Python must be functional and invokable via python command.
- The program must receive the report file name as input argument (see examples in sbpipe/tests/).
- The program must save the report to file including the *Time* column. Report fields must be separated by TAB, and row names must be discarded.

pipeline: parameter estimation

- The model coded in Python must be functional and invokable via python command.
- The program must receive the report file name as input argument (see examples in sbpipe/tests/).
- The program must save the report to file. This includes the objective value as first column column, and the estimated parameters as following columns. Rows are the evaluated functions. Report fields must be separated by TAB, and row names must be discarded.

Python as a wrapper Users can use Python as a wrapper to execute models coded in ANY programming language. The following Python model is essentially a wrapper invoking an R model called <code>sde_periodic_drift.r</code>. This Python wrapper and <code>sde_periodic_drift.r</code> are stored in the <code>Models/</code> folder. The configuration file calls the Python wrapper. This wrapper code must receive the report file name as input argument and forward it to the R script. This R script will run a model and store the results in the received report file name. These data must be stored as described above.

Python wrapper sde_periodic_drift.py. This runs sde_periodic_drift.r

```
import os
import sys
import subprocess
import shlex
# This is a Python wrapper used to run an R model.
# The R model receives the report_filename as input
# and must add the results to it.
# Retrieve the report file name
report_filename = "sde_periodic_drift.csv"
if len(sys.argv) > 1:
   report_filename = sys.argv[1]
command = 'Rscript --vanilla ' + os.path.join(os.path.dirname(__file__), 'sde_
→periodic_drift.r') + \
          ' ' + report_filename
# Block until command is finished
subprocess.call(shlex.split(command))
```

Configuration file invoking the Python wrapper sde_periodic_drift.py

```
generate_data: True
analyse_data: True
generate_report: True
project_dir: "."
simulator: "Python"
model: "sde_periodic_drift.py"
```

```
cluster: "local"
local_cpus: 7
runs: 14
exp_dataset: ""
plot_exp_dataset: False
exp_dataset_alpha: 1.0
xaxis_label: "Time"
yaxis_label: "#"
```

1.3.2 How to run SBpipe

SBpipe is executed via the command *sbpipe*. The syntax for this command and its complete list of options can be retrieved by running *sbpipe -h*. The first step is to create a new project. This can be done with the command:

```
$ sbpipe --create-project project_name
```

This generates the following structure:

Models must be stored in the Models/ folder. COPASI data sets used by a model should also be stored in Models. To run SBpipe, users need to create a configuration file for each pipeline they intend to run (see next section). These configuration files should be placed in the root project folder. In Results/ users will eventually find all the results generated by SBpipe.

Each pipeline is invoked using a specific option (type sbpipe -h for the complete command set):

```
# runs model simulation.
sbpipe -s config_file.yaml

# runs parameter estimation.
sbpipe -e config_file.yaml

# runs single parameter scan.
sbpipe -p config_file.yaml

# runs double parameter scan
sbpipe -d config_file.yaml
```

1.3.3 Pipeline configuration files

Pipelines are configured using files (here called configuration files). These files are YAML files. In SBpipe each pipeline executes three tasks: data generation, data analysis, and report generation. These tasks can be activated in each configuration files using the options:

generate_data: Trueanalyse_data: Truegenerate_report: True

The <code>generate_data</code> task runs a simulator accordingly to the options in the configuration file. Hence, this task collects and organises the reports generated from the simulator. The <code>analyse_data</code> task processes the reports to generate plots and compute statistics. Finally, the <code>generate_report</code> task generates a LaTeX report containing the computed plots and invokes the utility <code>pdflatex</code> to produce a PDF file. This modularisation allows users to analyse the same data without having to re-generate it, or to skip the report generation if not wanted.

Pipelines for parameter estimation or stochastic model simulation can be computationally intensive. SBpipe allows users to generate simulated data in parallel using the following options in the pipeline configuration file:

cluster: "local"local_cpus: 7runs: 250

The cluster option defines whether the simulator should be executed locally (local: Python multiprocessing), or in a computer cluster (sge: Sun Grid Engine (SGE), lsf: Load Sharing Facility (LSF)). If local is selected, the local_cpus option determines the maximum number of CPUs to be allocated for local simulations. The runs option specifies the number of simulations (or parameter estimations for the pipeline param_estim) to be run.

Assuming that the configuration files are placed in the root directory of a certain project (e.g. project_name/), examples are given as follow:

Example 1: configuration file for the pipeline *simulation*

```
# True if data should be generated, False otherwise
generate_data: True
# True if data should be analysed, False otherwise
analyse_data: True
# True if a report should be generated, False otherwise
generate_report: True
# The relative path to the project directory
project_dir: "."
# The name of the configurator (e.g. Copasi, Python)
simulator: "Copasi"
# The model name
model: "insulin_receptor_stoch.cps"
# The cluster type. local if the model is run locally,
# sge/lsf if run on cluster.
cluster: "local"
# The number of CPU if local is used, ignored otherwise
local_cpus: 7
# The number of simulations to perform.
# n>: 1 for stochastic simulations.
runs: 40
# An experimental data set (or blank) to add to the
# simulated plots as additional layer
exp_dataset: "insulin_receptor_dataset.csv"
# True if the experimental data set should be plotted.
plot_exp_dataset: True
# The alpha level used for plotting the experimental dataset
exp_dataset_alpha: 1.0
# The label for the x axis.
xaxis_label: "Time [min]"
# The label for the y axis.
yaxis_label: "Level [a.u.]"
```

Example 2: configuration file for the pipeline *single parameter scan*

```
# True if data should be generated, False otherwise
generate_data: True
# True if data should be analysed, False otherwise
analyse_data: True
# True if a report should be generated, False otherwise
generate_report: True
# The relative path to the project directory
project_dir: "."
# The name of the configurator (e.g. Copasi, Python)
simulator: "Copasi"
# The model name
```

```
model: "insulin_receptor_inhib_scan_IR_beta.cps"
# The variable to scan (as set in Copasi Parameter Scan Task)
scanned_par: "IR_beta"
# The cluster type. local if the model is run locally,
# sge/lsf if run on cluster.
cluster: "local"
# The number of CPU if local is used, ignored otherwise
local_cpus: 7
# The number of simulations to perform per run.
# n>: 1 for stochastic simulations.
runs: 1
# The number of intervals in the simulation
simulate__intervals: 100
# True if the variable is only reduced (knock down), False otherwise.
ps1_knock_down_only: True
# True if the scanning represents percent levels.
ps1_percent_levels: True
# The minimum level (as set in Copasi Parameter Scan Task)
min_level: 0
# The maximum level (as set in Copasi Parameter Scan Task)
max level: 100
# The number of scans (as set in Copasi Parameter Scan Task)
levels_number: 10
# True if plot lines are the same between scans
# (e.g. full lines, same colour)
homogeneous_lines: False
# The label for the x axis.
xaxis_label: "Time [min]"
# The label for the y axis.
yaxis_label: "Level [a.u.]"
```

Example 3: configuration file for the pipeline *double parameter scan*

```
# True if data should be generated, False otherwise
generate_data: True
# True if data should be analysed, False otherwise
analyse_data: True
# True if a report should be generated, False otherwise
generate_report: True
# The relative path to the project directory
project_dir: "."
# The name of the configurator (e.g. Copasi, Python)
simulator: "Copasi"
# The model name
model: "insulin_receptor_inhib_dbl_scan_InsulinPercent__IRbetaPercent.cps"
# The 1st variable to scan (as set in Copasi Parameter Scan Task)
scanned_par1: "InsulinPercent"
# The 2nd variable to scan (as set in Copasi Parameter Scan Task)
scanned_par2: "IRbetaPercent"
# The cluster type. local if the model is run locally,
# sge/lsf if run on cluster.
cluster: "local"
# The number of CPU if local is used, ignored otherwise
local_cpus: 7
# The number of simulations to perform.
# n>: 1 for stochastic simulations.
# The simulation length (as set in Copasi Time Course Task)
sim_length: 10
```

Example 4: configuration file for the pipeline parameter estimation

```
# True if data should be generated, False otherwise
generate_data: True
# True if data should be analysed, False otherwise
analyse_data: True
# True if a report should be generated, False otherwise
generate_report: True
# True if a zipped tarball should be generated, False otherwise
generate_tarball: True
# The relative path to the project directory
project dir: "."
# The name of the configurator (e.g. Copasi, Python)
simulator: "Copasi"
# The model name
model: "insulin_receptor_param_estim.cps"
# The cluster type. local if the model is run locally,
# sge/lsf if run on cluster.
cluster: "local"
# The number of CPU if local is used, ignored otherwise
local_cpus: 7
# The parameter estimation round which is used to distinguish
# phases of parameter estimations when parameters cannot be
# estimated at the same time
round: 1
# The number of parameter estimations
# (the length of the fit sequence)
# The threshold percentage of the best fits to consider
best_fits_percent: 75
# The number of available data points
data_point_num: 33
# True if 2D all fits plots for 66% confidence levels
# should be plotted. This can be computationally expensive.
plot_2d_66cl_corr: True
# True if 2D all fits plots for 95% confidence levels
# should be plotted. This can be computationally expensive.
plot_2d_95cl_corr: True
# True if 2D all fits plots for 99% confidence levels
# should be plotted. This can be computationally expensive.
plot_2d_99cl_corr: True
# True if parameter values should be plotted in log space.
logspace: True
# True if plot axis labels should be plotted in scientific notation.
scientific_notation: True
```

Additional examples of configuration files can be found in:

```
sbpipe/tests/insulin_receptor/
```

1.3.4 How to run SBpipe with Snakemake

SBpipe can also be executed using Snakemake (https://snakemake.readthedocs.io). Snakemake offers an infrastructure for running software pipelines using declarative rules.

Snakemake can be installed manually via package manager or using the conda command:

```
# Install snakemake (note: it requires python 3+ to run)
$ conda install -c bioconda snakemake
```

SBpipe pipelines for parameter estimation, single/double parameter scan, and model simulation are also implemented as snakemake files (which contain the set of rules for each pipeline). These are:

· sbpipe_pe.snake

- sbpipe_ps1.snake
- sbpipe_ps2.snake
- sbpipe_sim.snake

and are stored on the root folder of SBpipe. The advantage of using snakemake as pipeline infrastructure is that it offers an extended command sets compared to the one provided with the standard sbpipe. For details, run

```
snakemake -h
```

Snakemake also offers a strong support for dependency management at coding level and reentrancy at execution level. The former is defined as a way to precisely define the dependency order of functions. The latter is the capacity of a program to continue from the last interrupted task. Benefitting of dependency declaration and execution reentrancy can be beneficial for running SBpipe on clusters or on the cloud.

Under the current implementation of SBpipe snakefile, the configuration files described above require the additional field:

```
# The name of the report variables
report_variables: ['IR_beta_pY1146']
```

which contain the names of the variables exported by the simulator. For the parameter estimation pipeline, report_variables will contain the names of the estimated parameters.

For the parameter estimation pipeline, the following option must also be added:

```
# An experimental data set (or blank) to add to the
# simulated plots as additional layer
exp_dataset: "insulin_receptor_dataset.csv"
```

A complete example of configuration file for the parameter estimation pipeline is the following:

```
simulator: "Copasi"
model: "insulin_receptor_param_estim.cps"
round: 1
runs: 4
best_fits_percent: 75
data_point_num: 33
plot_2d_66cl_corr: True
plot_2d_95cl_corr: True
plot_2d_99cl_corr: True
logspace: True
scientific_notation: True
report_variables: ['k1','k2','k3']
exp_dataset: "insulin_receptor_dataset.csv"
```

NOTE: As it can be noticed, a configuration files for SBpipe using snakemake requires less options than the corresponding configuration file using SBpipe directly. This because Snakemake files is more automated than SBpipe. Nevertheless, the removal of those additional options is not necessary for running the configuration file using Snakemake.

Examples of configuration files for running SBpipe using Snakemake are in tests/snakemake.

Examples of commands running SBpipe pipelines using Snakemake are:

See snakemake -h for a complete list of commands.

1.4 How to report bugs or request new features

SBpipe is a relatively young project and there is a chance that some error occurs. The following mailing list should be used for general questions:

```
sbpipe AT googlegroups.com
```

All the topics discussed in this mailing list are also available at the website:

https://groups.google.com/forum/#!forum/sbpipe

To help us better identify and reproduce your problem, some technical information is needed. This detail data can be found in SBpipe log files which are stored in \${HOME}/.sbpipe/logs/. When using the mailing list above, it would be worth providing this extra information.

Issues and feature requests can also be notified using the github issue tracking system for SBpipe at the web page:

https://github.com/pdp10/sbpipe/issues.

CHAPTER

TWO

DEVELOPER MANUAL

Mailing list: sbpipe AT googlegroups.com

Forum: https://groups.google.com/forum/#!forum/sbpipe

2.1 Introduction

This guide is meant for developers and contains guidelines for developing this project.

2.2 Development model

This project follows the Feature-Branching model. Briefly, there are two main branches: master and develop. The former contains the history of stable releases, the latter contains the history of development. The master branch contains checkout points for production hotfixes or merge points for release-x.x.x branches. The develop branch is used for feature-bugfix integration and checkout point in development. Nobody should directly develop in here.

2.2.1 Conventions

To manage the project in a more consistent way, here is a list of conventions to follow:

- Each new feature is developed in a separate branch forked from *develop*. This new branch is called *featureNUMBER*, where *NUMBER* is the number of the GitHub Issue discussing that feature. The first line of each commit message for this branch should contain the string *Issue #NUMBER* at the beginning. Doing so, the commit is automatically recorded by the Issue Tracking System for that specific Issue. Note that the sharp (#) symbol is required.
- The same for each new bugfix, but in this case the branch name is called bugfixNUMBER.
- The same for each new hotfix, but in this case the branch name is called hotfixNUMBER and is forked from *master*.

2.2.2 Work flow

The procedure for checking out a new feature from the develop branch is:

```
$ git checkout -b feature10 develop
```

This creates the feature 10 branch off develop. This feature 10 is discussed in *Issue #10* in GitHub. When you are ready to commit your work, run:

```
$ git commit -am "Issue #10, summary of the changes. Detailed
description of the changes, if any."
$ git push origin feature10  # sometimes and at the end.
```

As of June 2016, the branches master and develop are protected and a status check using Travis-CI must be performed before merging or pushing into these branches. This automatically forces a merge without fast-forward. In order to merge any new feature, bugfix or simple edits into master or develop, a developer must checkout a new branch and, once committed and pushed, merge it to master or develop using a pull request. To merge feature10 to develop, the pull request output will look like this in GitHub Pull Requests:

```
base:develop compare:feature10 Able to merge. These branches can be automatically merged.
```

A small discussion about feature 10 should also be included to allow other users to understand the feature.

Finally delete the branch:

```
$ git branch -d feature10  # delete the branch feature10 (locally)
```

2.2.3 New releases

The script release.sh at the root of the package allows to release a new version of SBpipe or update the last github tag. This script also creates and uploads a new SBpipe package for anaconda cloud.

The following two sections describe how to release a new version for SBpipe, manually.

How to release a new tag

When the develop branch includes all the desired feature for a release, it is time to checkout this branch in a new one called release-x.x.x. It is at this stage that a version is established. Only bugfixes or hotfixes are applied to this branch. When this testing/correction phase is completed, the master branch will merge with the release-x.x.x branch, using the commands above. To record the release add a tag:

```
git tag -a v1.3 -m "PROGRAM_NAME v1.3"
```

To transfer the tag to the remote server:

```
git push origin v1.3 # Note: it goes in a separate 'branch'
```

To see all the releases:

```
git show
```

How to release a new SBpipe conda package (Anaconda Cloud)

This is a short guide for building SBpipe as a conda package. Anaconda (or Miniconda) must be installed. In order to proceed, the package conda-build must be installed:

```
$ conda install conda-build

# DON'T FORGET TO SET THIS so that your built package is not uploaded automatically
$ conda config --set anaconda_upload no
```

The recipe for SBpipe is already prepared (file: meta.yaml). To create the conda package for SBpipe:

```
$ cd path/to/sbpipe
$ conda-build conda_recipe/meta.yaml -c pdp10 -c conda-forge -c fbergmann -c_

defaults
```

To test this package locally:

```
# install
$ conda install sbpipe --use-local
# uninstall
$ conda remove sbpipe
```

To upload the package to the Anaconda cloud repository:

```
anaconda upload ~/miniconda/conda-bld/noarch/sbpipe-x.x.x-py_y.tar.bz2
```

2.3 Package structure

This section presents the structure of the SBpipe package. The root of the project contains general management scripts for installing Python and R dependencies (install_pydeps.py and install_rdeps.r), and installing SBpipe (setup.py). Additionally, the logging configuration file (logging_config.ini) is also at this level.

In order to automatically compile and run the test suite, Travis-CI is used and configured accordingly (.travis.yml).

The project is structured as follows:

These folders will be discussed in the next sections. In SBpipe, Python is the project main language. Instead, R is essentially used for computing statistics (see section configuration file in the user manual) and for generating plots. This choice allows users to run these scripts independently of SBpipe if needed using an R environment like Rstudio. This can be convenient if further data analysis are needed or plots need to be annotated or edited. The R code for SBpipe is distributed as a separate R package and installed as a dependency using the provided script (install_rdeps.r) or conda. The source code for this package can be found here: https://github.com/pdp10/sbpiper

2.3.1 docs

The folder docs/contains the documentation for this project. The user and developer manuals in markdown format are contained in docs/source. In order to generate the complete documentation for SBpipe, the following packages must be installed:

- python-sphinx
- pandoc
- · texlive-fonts-recommended
- texlive-latex-extra

By default the documentation is generated in html and LaTeX/PDF. Instruction for generating or cleaning SBpipe documentation are provided below.

To generate the source code documentation:

```
$ cd path/to/sbpipe/docs
$ ./gen_doc.sh
```

To clean the documentation:

```
$ cd path/to/sbpipe/docs
$ ./cleanup_doc.sh
```

The complete source code documentation for this project is stored in docs/build/html (html format) and docs/build/latex (LaTeX/PDF format). A shortcut to the documentation in html format is available at the page docs/index.html.

2.3.2 sbpipe

This folder contains the source code of the project SBpipe. At this level a file called __main__.py enables users to run SBpipe programmatically as a Python module via the command:

```
$ python sbpipe
```

Alternatively sbpipe can programmatically be imported within a Python environment as shown below:

```
$ cd path/to/sbpipe
$ python
# Python environment
>>> from sbpipe.main import sbpipe
>>> sbpipe(simulate="my_model.yaml")
```

The following subsections describe sbpipe subpackages.

рl

The subpackage sbpipe.pl contains the class Pipeline in the file pipeline.py. This class represents a generic pipeline which is extended by SBpipe pipelines. These are organised in the following subpackages:

- create: creates a new project
- ps1: scan a model parameter, generate plots and report;
- ps2: scan two model parameters, generate plots and report;
- pe: generate a parameter fit sequence, tables of statistics, plots and report;
- sim: generate deterministic or stochastic model simulations, plots and report.

All these pipelines can be invoked directly via the script <code>sbpipe/scripts/sbpipe</code>. Each SBpipe pipeline extends the class <code>Pipeline</code> and therefore must implement the following methods:

```
# executes a pipeline
def run(self, config_file)

# process the dictionary of the configuration file loaded by Pipeline.load()
def parse(self, config_dict)
```

- The method run() can invoke Pipeline.load() to load the YAML config_file as a dictionary. Once the configuration is loaded and the parameters are imported, run() executes the pipeline.
- The method parse() parses the dictionary and collects the values.

report

The subpackage sbpipe.report contains Python modules for generating LaTeX/PDF reports.

simul

The subpackage <code>sbpipe.simul</code> contains the class <code>Simul</code> in the file <code>simul.py</code>. This is a generic simulator interface used by the pipelines in SBpipe. This mechanism uncouples pipelines from specific simulators which can therefore be configured in each pipeline configuration file. As of 2016, the following simulators are available in SBpipe:

- Copasi, package sbpipe.simul.copasi, which implements all the methods of the class Simul;
- Python, package sbpipe.simul.python.

Pipelines can dynamically load a simulator via the class method Pipeline.get_simul_obj(simulator). This method instantiates an object of subtype Simul by refractoring the simulator name as parameter. A simulator class (e.g. Copasi) must have the same name of their package (e.g. copasi) but start with an upper case letter. A simulator class must be contained in a file with the same name of their package (e.g. copasi). Therefore, for each simulator package, exactly one simulator class can be instantiated. Simulators can be configured in the configuration file using the field simulator.

tasks

The subpackage sbpipe.tasks contains the Python scripts to invoke the single SBpipe tasks. These are invoked by the rules in the SBpipe snakemake files. These snakemake files are:

- · sbpipe_pe.snake
- sbpipe_ps1.snake
- sbpipe_ps2.snake
- · sbpipe sim.snake

and are stored on the root folder of SBpipe.

utils

The subpackage sbpipe.utils contains a collection of Python utility modules which are used by sbpipe. Here are also contained the functions for running commands in parallel.

2.3.3 scripts

The folder scripts contains the scripts: cleanup_sbpipe and sbpipe. sbpipe is the main script and is used to run the pipelines. cleanup_sbpipe.py is used for cleaning the package including the test results.

2.3.4 tests

The package tests contains the script test_suite.py which executes all sbpipe tests. It should be used for testing the correct installation of SBpipe dependencies as well as reference for configuring a project before running any pipeline. Projects inside the folder sbpipe/tests/ have the SBpipe project structure:

- Models: (e.g. models, COPASI models, Python models, data sets directly used by Copasi models);
- Results: (e.g. pipelines results, etc).

Examples of configuration files (*.yaml) using COPASI can be found in sbpipe/tests/insulin_receptor/.

To run tests for Python models, the Python packages numpy, scipy, and pandas must be installed. In principle, users may define their Python models using arbitrary packages.

As of 2016, the repository for SBpipe source code is github.com. This is configured to run Travis-CI every time a git push into the repository is performed. The exact details of execution of Travis-CI can be found in Travis-CI configuration file <code>sbpipe/.travis.yml</code>. Importantly, Travis-CI runs all SBpipe tests using nosetests.

2.4 Miscellaneous of useful commands

2.4.1 Git

Startup

```
# clone master
$ git clone https://github.com/pdp10/sbpipe.git
# get develop branch
$ git checkout -b develop origin/develop
# to get all the other branches
$ for b in `git branch -r | grep -v -- '->'`; do git branch
--track ${b##origin/} $b; done
# to update all the branches with remote
$ git fetch --all
```

Update

```
# ONLY use --rebase for private branches. Never use it for shared
# branches otherwise it breaks the history. --rebase moves your
# commits ahead. For shared branches, you should use
# `git fetch && git merge --no-ff`
$ git pull [--rebase] origin BRANCH
```

Managing tags

```
# Update an existing tag to include the last commits
# Assuming that you are in the branch associated to the tag to update:
git tag -f -a tagName
# push your new commit:
git push
# force push your moved tag:
git push -f --tags
# rename a tag
git tag new old
git tag -d old
git push origin :refs/tags/old
git push --tags
# make sure that the other users remove the deleted tag. Tell them(co-workers) to...
→run the following command:
git pull --prune --tags
# removing a tag remotely and locally
git push --delete origin tagName
git tag -d tagName
```

File system

```
$ git rm [--cache] filename
$ git add filename
```

Information

```
$ git status
$ git log [--stat]
$ git branch # list the branches
```

Maintenance

```
$ git fsck  # check errors
$ git gc  # clean up
```

Rename a branch locally and remotely

```
git branch -m old_branch new_branch # Rename branch locally
git push origin :old_branch # Delete the old branch
git push --set-upstream origin new_branch # Push the new branch, set
local branch to track the new remote
```

Reset

```
git reset --hard HEAD  # to undo all the local uncommitted changes
```

Syncing a fork (assuming upstreams are set)

```
git fetch upstream
git checkout develop
git merge upstream/develop
```

CHAPTER

THREE

SOURCE CODE

3.1 Python modules

3.1.1 sbpipe package

Subpackages

sbpipe.pl package

Subpackages

sbpipe.pl.create package

Submodules

sbpipe.pl.create.newproj module

This module initialises the folder tree for a new project.

Parameters

- models_folder the folder containing the models
- working_folder the folder to store the results

run (project_name)

Create a project directory tree.

Parameters project_name - the name of the project

Returns 0

Module contents

sbpipe.pl.pe package

Submodules

sbpipe.pl.pe.parest module

This module provides the user with a complete pipeline of scripts for running model parameter estimations

```
classmethod analyse_data (simulator, model, inputdir, outputdir, fileout_final_estims, fileout_param_estim_details, fileout_param_estim_summary, sim_plots_dir, best_fits_percent, data_point_num, cluster='local', plot_2d_66cl_corr=False, plot_2d_95cl_corr=False, plot_2d_99cl_corr=False, logspace=True, scientific_notation=True)
```

The second pipeline step: data analysis.

Parameters

- **simulator** the name of the simulator (e.g. Copasi)
- model the model name
- inputdir the directory containing the simulation data
- **outputdir** the directory to store the results
- **fileout_final_estims** the name of the file containing final parameter sets with the objective value
- **fileout_all_estims** the name of the file containing all the parameter sets with the objective value
- **fileout_param_estim_details** the name of the file containing the detailed statistics for the estimated parameters
- **fileout_param_estim_summary** the name of the file containing the summary for the parameter estimation
- **sim_plots_dir** the directory of the simulation plots
- best_fits_percent the percent to consider for the best fits
- data_point_num the number of data points
- cluster local, lsf for Load Sharing Facility, sge for Sun Grid Engine.
- plot_2d_66cl_corr True if 2 dim plots for the parameter sets within 66% should be plotted
- plot_2d_95cl_corr True if 2 dim plots for the parameter sets within 95% should be plotted
- plot_2d_99cl_corr True if 2 dim plots for the parameter sets within 99% should be plotted
- logspace True if parameters should be plotted in log space
- scientific_notation True if axis labels should be plotted in scientific notation

Returns True if the task was completed successfully, False otherwise.

The first pipeline step: data generation.

Parameters

- **simulator** the name of the simulator (e.g. Copasi)
- model the model to process
- inputdir the directory containing the model
- cluster local, lsf for load sharing facility, sge for sun grid engine
- local cpus the number of cpu
- **runs** the number of fits to perform
- **outputdir** the directory to store the results
- sim_data_dir the directory containing the simulation data sets

Returns True if the task was completed successfully, False otherwise.

classmethod generate_report (model, outputdir, sim_plots_folder)

The third pipeline step: report generation.

Parameters

- model the model name
- outputdir the directory to store the report
- **sim_plots_folder** the folder containing the plots

Returns True if the task was completed successfully, False otherwise.

```
parse (my_dict)
run (config_file)
```

Module contents

sbpipe.pl.ps1 package

Submodules

sbpipe.pl.ps1.parscan1 module

This module provides the user with a complete pipeline of scripts for computing single parameter scans.

The second pipeline step: data analysis.

Parameters

• model – the model name

- knock_down_only True for knock down simulation, false if also scanning over expression.
- outputdir the directory containing the results
- sim_data_folder the folder containing the simulated data sets
- **sim_plots_folder** the folder containing the generated plots
- runs the number of simulations
- local_cpus the number of cpus
- **percent_levels** True if the levels are percents.
- min level the minimum level
- max_level the maximum level
- levels_number the number of levels
- homogeneous_lines True if generated line style should be homogeneous
- cluster local, 1sf for Load Sharing Facility, sge for Sun Grid Engine.
- **xaxis_label** the name of the x axis (e.g. Time [min])
- yaxis_label the name of the y axis (e.g. Level [a.u.])

Returns True if the task was completed successfully, False otherwise.

classmethod generate_data (simulator, model, scanned_par, cluster, local_cpus, runs, simulate_intervals, single_param_scan_intervals, inputdir, outputdir)

The first pipeline step: data generation.

Parameters

- **simulator** the name of the simulator (e.g. Copasi)
- model the model to process
- scanned_par the scanned parameter
- cluster local, 1sf for Load Sharing Facility, sge for Sun Grid Engine.
- local_cpus the number of CPU.
- runs the number of model simulation
- **simulate_intervals** the time step of each simulation
- single_param_scan_intervals the number of scans to perform
- inputdir the directory containing the model
- **outputdir** the directory to store the results

Returns True if the task was completed successfully, False otherwise.

classmethod generate_report (*model*, *scanned_par*, *outputdir*, *sim_plots_folder*) The third pipeline step: report generation.

Parameters

- model the model name
- scanned_par the scanned parameter
- outputdir the directory containing the report
- sim_plots_folder the folder containing the plots

Returns True if the task was completed successfully, False otherwise.

parse (my_dict)

run (config_file)

Module contents

sbpipe.pl.ps2 package

Submodules

sbpipe.pl.ps2.parscan2 module

This module provides the user with a complete pipeline of scripts for computing double parameter scans.

 $\begin{tabular}{ll} \textbf{classmethod analyse_data} (model, scanned_par1, scanned_par2, input dir, output dir, cluster='local', local_cpus=1, runs=1) \\ \end{tabular}$

The second pipeline step: data analysis.

Parameters

- model the model name
- scanned_par1 the first scanned parameter
- scanned_par2 the second scanned parameter
- inputdir the directory containing the simulated data sets to process
- outputdir the directory to store the performed analysis
- cluster local, 1sf for Load Sharing Facility, sge for Sun Grid Engine.
- local_cpus the number of CPU.
- runs the number of model simulation

Returns True if the task was completed successfully, False otherwise.

 ${\tt classmethod\ generate_data}\ (simulator,\ model,\ sim_length,\ input dir,\ output dir,\ cluster,\ lo-cal_cpus,\ runs)$

The first pipeline step: data generation.

Parameters

- **simulator** the name of the simulator (e.g. Copasi)
- model the model to process
- sim_length the length of the simulation
- inputdir the directory containing the model
- **outputdir** the directory to store the results
- cluster local, 1sf for Load Sharing Facility, sge for Sun Grid Engine.
- local cpus the number of CPU.
- runs the number of model simulation

Returns True if the task was completed successfully, False otherwise.

```
classmethod generate_report (model, scanned_par1, scanned_par2, outputdir, sim_plots_folder)
```

The third pipeline step: report generation.

Parameters

- model the model name
- scanned_par1 the first scanned parameter
- scanned_par2 the second scanned parameter
- outputdir the directory containing the report
- **sim_plots_folder** the folder containing the plots.

Returns True if the task was completed successfully, False otherwise.

```
parse (my_dict)
run (config_file)
```

Module contents

sbpipe.pl.sim package

Submodules

sbpipe.pl.sim.sim module

This module provides the user with a complete pipeline of scripts for running model simulations

```
classmethod analyse_data (simulator, model, inputdir, outputdir, sim_plots_dir, exp_dataset, plot_exp_dataset, exp_dataset_alpha=1.0, cluster='local', local_cpus=2, xaxis_label=", yaxis_label=")

The second pipeline step: data analysis.
```

Parameters

- **simulator** the name of the simulator (e.g. Copasi)
- model the model name
- inputdir the directory containing the data to analyse
- outputdir the output directory containing the results
- \bullet $\mbox{\tt sim_plots_dir}$ – the directory to save the plots
- **exp_dataset** the full path of the experimental data set
- plot_exp_dataset True if the experimental data set should also be plotted
- exp_dataset_alpha the alpha level for the data set
- cluster local, lsf for Load Sharing Facility, sge for Sun Grid Engine.
- local_cpus the number of CPUs.
- **xaxis** label the label for the x axis (e.g. Time [min])
- yaxis_label the label for the y axis (e.g. Level [a.u.])

Returns True if the task was completed successfully, False otherwise.

```
classmethod generate_data(simulator, model, inputdir, outputdir, cluster='local', local\_cpus=2, runs=1)

The first pipeline step: data generation.
```

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Parameters

- **simulator** the name of the simulator (e.g. Copasi)
- model the model to process
- inputdir the directory containing the model
- outputdir the directory containing the output files
- cluster local, 1sf for Load Sharing Facility, sge for Sun Grid Engine.
- local_cpus the number of CPUs.
- runs the number of model simulation

Returns True if the task was completed successfully, False otherwise.

classmethod generate_report (model, outputdir, sim_plots_folder)

The third pipeline step: report generation.

Parameters

- model the model name
- outputdir the output directory to store the report
- **sim_plots_folder** the folder containing the plots

Returns True if the task was completed successfully, False otherwise.

```
parse (my_dict)
run (config_file)
```

Module contents

Submodules

sbpipe.pl.pipeline module

Bases: object

Generic pipeline.

Parameters

- models_folder the folder containing the models
- working_folder the folder to store the results
- **sim_data_folder** the folder to store the simulation data
- **sim_plots_folder** the folder to store the graphic results

```
get_models_folder()
```

Return the folder containing the models.

Returns the models folder.

```
get_sim_data_folder()
```

Return the folder containing the in-silico generated data sets.

Returns the folder of the simulated data sets.

```
{\tt get\_sim\_plots\_folder}\ (\ )
```

Return the folder containing the in-silico generated plots.

Returns the folder of the simulated plots.

classmethod get_simul_obj(simulator)

Return the simulator object if this exists. Otherwise throws an exception. The simulator name starts with an upper case letter. Each simulator is in a package within *sbpipe.simulator*.

Parameters simulator – the simulator name

Returns the simulator object.

get_working_folder()

Return the folder containing the results.

Returns the working folder.

classmethod load(config)

Safely load a YAML configuration file and return its structure as a dictionary object.

Parameters config – a YAML configuration file

Returns the dictionary structure of the configuration file

Raise yaml.YAMLError if the config cannot be loaded.

parse (config_dict)

Read a dictionary structure containing the pipeline configuration. This method is abstract.

Returns a tuple containing the configuration

run (config_file)

Run the pipeline.

Parameters config_file – a configuration file for this pipeline.

Returns True if the pipeline was executed correctly, False otherwise.

Module contents

sbpipe.report package

Submodules

sbpipe.report.latex_reports module

Initialize a Latex header with a title and an abstract.

Parameters

- pdftitle the pdftitle for the LaTeX header
- title the title for the LaTeX header
- abstract the abstract for the LaTeX header

Returns the LaTeX header

```
sbpipe.report.latex_reports.latex_report (outputdir, plots_folder, model_noext, file-name_prefix, caption=False)
```

Generate a generic report.

Parameters

- **outputdir** the output directory
- plots_folder the folder containing the simulated plots

- model_noext the model name
- **filename_prefix** the prefix for the LaTeX file
- caption True if figure captions (=figure file name) should be added

Generate a report for a parameter estimation task.

Parameters

- **outputdir** the output directory
- plots_folder the folder containing the simulated plots
- model_noext the model name
- filename_prefix the prefix for the LaTeX file

Generate a report for a single parameter scan task.

Parameters

- **outputdir** the output directory
- plots_folder the folder containing the simulated plots
- **filename_prefix** the prefix for the LaTeX file
- model noext the model name
- scanned_par the scanned parameter

Generate a report for a double parameter scan task.

Parameters

- **outputdir** the output directory
- plots_folder the folder containing the simulated plots
- **filename_prefix** the prefix for the LaTeX file
- model noext the model name
- scanned_par1 the 1st scanned parameter
- scanned_par2 the 2nd scanned parameter

Generate a report for a time course task.

Parameters

- outputdir the output directory
- plots_folder the folder containing the simulated plots
- model_noext the model name
- **filename_prefix** the prefix for the LaTeX file

sbpipe.report.latex_reports.pdf_report (outputdir, filename)
Generate a PDF report from LaTeX report using pdflatex.

Parameters

- **outputdir** the output directory
- filename the LaTeX file name

Module contents

sbpipe.simul package

Subpackages

sbpipe.simul.copasi package

Submodules

sbpipe.simul.copasi.copasi module

```
class sbpipe.simul.copasi.copasi.Bases: sbpipe.simul.simul.Simul(page 32)
Copasi simulator.
```

model_checking (model_filename, task_name=")

Check whether the Copasi model can be loaded and executed correctly.

Parameters

- model filename the COPASI filename
- task_name the task to check

Returns boolean indicating whether the model could be loaded and executed successfully

```
pe (model, inputdir, cluster, local_cpus, runs, outputdir, sim_data_dir, output_msg=False)
```

- ps2 (model, sim_length, inputdir, outputdir, cluster='local', local_cpus=1, runs=1, output_msg=False)
- sim (model, inputdir, outputdir, cluster='local', local_cpus=1, runs=1, output_msg=False)

sbpipe.simul.copasi.model checking module

Parameters

- model_filename the filename to a COPASI file
- data model the COPASI data model structure

Returns a boolean indicating whether the model could be loaded successfully

Check whether the COPASI model task can be executed

Parameters

• model_filename – the filename to a COPASI file

- task_name the task to check
- data_model the COPASI data model structure
- task the COPASI task data structure

Returns a boolean indicating whether the model task can be executed correctly

Check whether the COPASI model task can be executed

Parameters

- model filename the filename to a COPASI file
- task name the task to check
- data_model the COPASI data model structure.

Returns a boolean indicating whether the model task can be executed correctly

Perform a basic model checking for a COPASI model file.

Parameters

- model filename the filename to a COPASI file
- task_name the task to check

Returns a boolean indicating whether the model could be loaded successfully

```
sbpipe.simul.copasi.model_checking.severity2string(severity)
```

Return a string representing the severity of the error message :param severity: an integer representing severity :return: a string of the error message

Module contents

sbpipe.simul.python package

Submodules

sbpipe.simul.python.python module

```
class sbpipe.simul.python.python.Python
    Bases: sbpipe.simul.pl_simul.PLSimul(page 31)
    Python Simulator.
```

Module contents

Submodules

sbpipe.simul.pl_simul module

```
class sbpipe.simul.pl_simul.PLSimul (lang, lang_err_msg, options)
    Bases: sbpipe.simul.simul.Simul (page 32)
```

A generic simulator for models coded in a programming language.

```
get_lang()
```

Return the programming language name :return: the name

get_lang_err_msg()

Return the error if the programming language is not found :return: the error message

get_lang_options()

Return the options for the programming language command :return: the options. Return None, if no options are used.

pe (model, inputdir, cluster, local_cpus, runs, outputdir, sim_data_dir, output_msg=False)

ps1 (model, scanned_par, simulate_intervals, single_param_scan_intervals, inputdir, outputdir, cluster='local', local_cpus=1, runs=1, output_msg=False)

ps2 (model, sim_length, inputdir, outputdir, cluster='local', local_cpus=1, runs=1, output_msg=False)

replace_str_in_report (report)

sim(model, inputdir, outputdir, cluster='local', local_cpus=1, runs=1, output_msg=False)

sbpipe.simul.simul module

```
class sbpipe.simul.simul.Simul
```

Bases: object

Generic simulator.

get_all_fits (path_in='.', path_out='.', filename_out='all_estimates.csv')

Collect all the parameter estimates. Results are stored in filename_out.

Parameters

- path_in the path to the input files
- path_out the path to the output files
- **filename_out** a global file containing all fits from independent parameter estimations.

Returns the number of retrieved files

get_best_fits (path_in='.', path_out='.', filename_out='final_estimates.csv')

Collect the final parameter estimates. Results are stored in filename_out.

Parameters

- path_in the path to the input files
- path_out the path to the output files
- **filename_out** a global file containing the best fits from independent parameter estimations.

Returns the number of retrieved files

get sim columns(path in='.')

Return the columns to analyse (sim task)

Parameters path_in - the path to the input files

pe (*model*, *inputdir*, *cluster*, *local_cpus*, *runs*, *outputdir*, *sim_data_dir*, *output_msg=False*) parameter estimation.

Parameters

- model the model to process
- inputdir the directory containing the model

- cluster local, 1sf for load sharing facility, sge for sun grid engine
- local_cpus the number of cpu
- runs the number of fits to perform
- outputdir the directory to store the results
- sim_data_dir the directory containing the simulation data sets
- output_msg print the output messages on screen (available for cluster='local' only)
- ps1 (model, scanned_par, simulate_intervals, single_param_scan_intervals, inputdir, outputdir,
 cluster='local', local_cpus=1, runs=1, output_msg=False)
 Single parameter scan.

- model the model to process
- scanned_par the scanned parameter
- **simulate_intervals** the time step of each simulation
- single_param_scan_intervals the number of scans to perform
- inputdir the directory containing the model
- outputdir the directory to store the results
- cluster local, lsf for Load Sharing Facility, sge for Sun Grid Engine.
- local_cpus the number of CPU used.
- runs the number of model simulation
- output_msg print the output messages on screen (available for cluster='local' only)
- ps1_postproc (model, scanned_par, simulate_intervals, single_param_scan_intervals, outputdir)

Perform post processing organisation to single parameter scan report files.

Parameters

- model the model to process
- scanned_par the scanned parameter
- **simulate_intervals** the time step of each simulation
- single_param_scan_intervals the number of scans to perform
- outputdir the directory to store the results
- ps2 (model, sim_length, inputdir, outputdir, cluster='local', local_cpus=1, runs=1, output_msg=False)
 Double paramter scan.

- model the model to process
- sim_length the length of the simulation
- inputdir the directory containing the model
- **outputdir** the directory to store the results
- cluster local, 1sf for Load Sharing Facility, sge for Sun Grid Engine.
- local_cpus the number of CPU.
- runs the number of model simulation

• output_msg - print the output messages on screen (available for cluster='local' only)

ps2_postproc (model, sim_length, outputdir)

Perform post processing organisation to double parameter scan report files.

Parameters

- model the model to process
- **sim_length** the length of the simulation
- **outputdir** the directory to store the results

replace_str_in_report (report)

Replaces strings in a report file.

Parameters report – a report file with its absolute path

sim (model, inputdir, outputdir, cluster='local', local_cpus=1, runs=1, output_msg=False)
Time course simulator.

Parameters

- model the model to process
- inputdir the directory containing the model
- **outputdir** the directory containing the output files
- cluster local, lsf for Load Sharing Facility, sge for Sun Grid Engine.
- local_cpus the number of CPU.
- runs the number of model simulation
- **output_msg** print the output messages on screen (available for cluster='local' only)

Module contents

sbpipe.tasks package

Submodules

sbpipe.tasks.generate_data module

```
sbpipe.tasks.generate_data.generate_data(infile, copasi=False)
Replicate a copasi model and adds an id.
```

Parameters

- infile the input file
- copasi True if the model is a Copasi model

```
sbpipe.tasks.generate_data.main(argv=None)
```

```
sbpipe.tasks.generate_data.run_copasi_model(infile)
Run a Copasi model
```

Parameters infile - the input file

```
sbpipe.tasks.generate_data.run_generic_model(infile)
    Run a generic model
```

Parameters infile – the input file

sbpipe.tasks.pe collect module

Collect the results so that they can be processed. :param inputdir: the input folder containing the data :param outputdir: the output folder to stored the collected results :param fileout_final_estims: the name of the file containing the best estimations :param fileout_all_estims: the name of the file containing all the estimations :param copasi: True if COPASI was used to generate the data.

sbpipe.tasks.pe_combine_param_ple_stats module

Combine the statistics for the parameter estimation details

Parameters

- plots_dir the directory to save the generated plots
- **fileout_param_estim_details** the name of the file containing the detailed statistics for the estimated parameters

sbpipe.tasks.pe_ds_preproc module

Parameter estimation pre-processing. It renames the data set columns, and applies a log10 transformation if logspace is TRUE. If all.fits is true, it also computes the confidence levels.

Parameters

- filename the dataset filename containing the fits sequence
- param_names the list of estimated parameter names
- **logspace** true if the data set shoud be log10-transformed.
- all_fits true if filename contains all fits, false otherwise
- data_point_num the number of data points used for parameterise the model. Ignored if all.fits is false
- **fileout_param_estim_summary** the name of the file containing the summary for the parameter estimation. Ignored if all.fits is false

sbpipe.tasks.pe_objval_vs_iters_analysis module

Analysis of the Objective values vs Iterations.

- model_name the model name without extension
- **filename** the filename containing the fits sequence
- plots_dir the directory for storing the plots

sbpipe.tasks.pe parameter density analysis module

```
sbpipe.tasks.pe_parameter_density_analysis.main(argv=None)
sbpipe.tasks.pe_parameter_density_analysis.pe_parameter_density_analysis(model_name,
                                                                                         file-
                                                                                         name,
                                                                                         pa-
                                                                                         ram-
                                                                                         e-
                                                                                         ter,
                                                                                         plots_dir,
                                                                                         thres='BestFits',
                                                                                         best_fits_percent=100
                                                                                         out_param_estim_sur
                                                                                         logspace=True,
                                                                                         sci-
                                                                                         en-
                                                                                         tific notation=True)
```

Parameter density analysis.

Parameters

- model name the model name without extension
- **filename** the filename containing the fits sequence
- parameter the name of the parameter to plot the density
- plots_dir the directory for storing the plots
- **thres** the threshold used to filter the dataset. Values: "BestFits", "CL66", "CL95", "CL99", "All".
- best_fits_percent the percent of best fits to analyse. Only used if thres="BestFits".
- **fileout_param_estim_summary** the name of the file containing the summary for the parameter estimation. Only used if thres!="BestFits".
- logspace true if the parameters should be plotted in logspace
- **scientific_notation** true if the axis labels should be plotted in scientific notation

sbpipe.tasks.pe_parameter_pca_analysis module

PCA for the best fits of the estimated parameters.

- model_name the model name without extension
- **filename** the filename containing the fits sequence
- plots_dir the directory for storing the plots
- best_fits_percent the percent of best fits to analyse. Only used if thres="BestFits".

sbpipe.tasks.pe postproc module

```
sbpipe.tasks.pe_postproc.generic_postproc(infile, outfile, copasi=True)
Perform post processing file editing for the pe pipeline
```

Parameters

- infile the model to process
- outfile the directory to store the results
- copasi True if the model is a Copasi model

```
sbpipe.tasks.pe_postproc.main(argv=None)
```

```
\verb|sbpipe.tasks.pe_postproc.pe_postproc| \textit{infile}, \textit{outfile}, \textit{copasi=True})|
```

Perform post processing file editing for the pe pipeline

Parameters

- infile the model to process
- outfile the directory to store the results
- copasi True if the model is a Copasi model

sbpipe.tasks.pe sampled 2d ple analysis module

```
sbpipe.tasks.pe_sampled_2d_ple_analysis.main(argv=None)
sbpipe.tasks.pe_sampled_2d_ple_analysis.pe_sampled_2d_ple_analysis(model_name,
                                                                                  file-
                                                                                  name,
                                                                                  pa-
                                                                                  rame-
                                                                                  ter1,
                                                                                  pa-
                                                                                  rame-
                                                                                  ter2,
                                                                                  plots_dir,
                                                                                  thres='BestFits',
                                                                                  best_fits_percent=100,
                                                                                  out_param_estim_summary="
                                                                                  logspace=True,
                                                                                  scien-
                                                                                  tific_notation=True)
```

2D profile likelihood estimation analysis.

- model_name the model name without extension
- **filename** the filename containing the fits sequence

- parameter1 the name of the first parameter
- parameter2 the name of the second parameter
- plots_dir the directory for storing the plots
- **thres** the threshold used to filter the dataset. Values: "BestFits", "CL66", "CL95", "CL99", "All".
- best_fits_percent the percent of best fits to analyse. Only used if thres="BestFits".
- **fileout_param_estim_summary** the name of the file containing the summary for the parameter estimation. Only used if thres!="BestFits".
- logspace true if the parameters should be plotted in logspace
- scientific_notation true if the axis labels should be plotted in scientific notation

sbpipe.tasks.pe_sampled_ple_analysis module

Run the profile likelihood estimation analysis.

Parameters

- model_name the model name without extension
- **filename** the filename containing the fits sequence
- parameter the parameter to compute the PLE analysis
- plots_dir the directory to save the generated plots
- **fileout_param_estim_summary** the name of the file containing the summary for the parameter estimation
- **logspace** true if parameters should be plotted in logspace
- **scientific_notation** true if the axis labels should be plotted in scientific notation

sbpipe.tasks.preproc module

```
sbpipe.tasks.preproc.copasi_preproc(infile, outfile, task_name)
Replicate a copasi model and adds an id.
```

- infile the input file
- outfile the output file
- task_name the name of the task (Copasi models)

```
\begin{tabular}{ll} {\bf sbpipe.tasks.preproc.generic\_preproc\,(\it infile,\,outfile)}\\ {\bf Copy\,\,the\,\,model\,\,file} \end{tabular}
```

- infile the input file
- outfile the output file

```
sbpipe.tasks.preproc.main(argv=None)
```

sbpipe.tasks.preproc.preproc(infile, outfile, task_name, copasi=False)

Replicate a copasi model and adds an id.

Parameters

- infile the input file
- outfile the output file
- task_name the name of the task (Copasi models)
- copasi True if the model is a Copasi model

sbpipe.tasks.ps1_analyse_plot module

Plot model single parameter scan time courses (Python wrapper).

Parameters

- model_name the model name without extension
- **inhibition_only** true if the scanning only decreases the variable amount (inhibition only)
- inputdir the input directory containing the simulated data
- outputdir the output directory that will contain the simulated plots
- repeat the simulation number
- percent_levels true if scanning levels are in percent
- min_level the minimum level
- max_level the maximum level
- levels_number the number of levels
- homogeneous_lines true if lines should be plotted homogeneously
- **xaxis_label** the label for the x axis (e.g. Time [min])
- yaxis_label the label for the y axis (e.g. Level [a.u.])

sbpipe.tasks.ps1 analyse plot homogen module

```
sbpipe.tasks.ps1_analyse_plot_homogen.main(argv=None)
```

Plot model single parameter scan time courses using homogeneous lines (Python wrapper).

Parameters

- model name the model name without extension
- inputdir the input directory containing the simulated data
- outputdir the output directory that will contain the simulated plots
- repeat the simulation number
- **xaxis_label** the label for the x axis (e.g. Time [min])
- yaxis_label the label for the y axis (e.g. Level [a.u.])

sbpipe.tasks.ps1 postproc module

```
sbpipe.tasks.ps1\_postproc. \textbf{generic\_postproc} (infile, outfile, scanned\_par, simulate\_intervals, single\_param\_scan\_intervals, copasi=True)
```

Perform post processing organisation to single parameter scan report files.

Parameters

- infile the model to process
- outfile the directory to store the results
- **scanned_par** the scanned parameter
- **simulate_intervals** the time step of each simulation
- single_param_scan_intervals the number of scans to perform
- copasi True if the model is a Copasi model

```
sbpipe.tasks.ps1_postproc.main(argv=None)
```

sbpipe.tasks.ps1_postproc.**ps1_header_init** (*report*, *scanned_par*) Header report initialisation for single parameter scan pipeline.

Parameters

- report a report
- **scanned_par** the scanned parameter

:return a list containing the header or an empty list if no header was created.

```
sbpipe.tasks.ps1\_postproc.ps1\_postproc (infile, outfile, scanned\_par, simulate\_intervals, single\_param\_scan\_intervals, copasi=True) \\ Perform post processing organisation to single parameter scan report files.
```

- infile the model to process
- outfile the directory to store the results
- **scanned_par** the scanned parameter
- **simulate_intervals** the time step of each simulation
- single_param_scan_intervals the number of scans to perform

• copasi – True if the model is a Copasi model

sbpipe.tasks.ps2_analyse_plot module

Plot model double parameter scan time courses (Python wrapper).

Parameters

- model the model name without extension
- scanned_par1 the 1st scanned parameter
- scanned_par2 the 2nd scanned parameter
- inputdir the input directory
- **outputdir** the output directory
- run the simulation number

sbpipe.tasks.ps2_postproc module

sbpipe.tasks.ps2_postproc.generic_postproc(infile, outfile, sim_length, copasi=True)
Perform post processing organisation to double parameter scan report files.

Parameters

- infile the model to process
- outfile the directory to store the results
- \bullet $\mbox{sim_length}$ the length of the simulation
- copasi True if the model is a Copasi model

```
sbpipe.tasks.ps2_postproc.main (argv=None)
sbpipe.tasks.ps2_postproc.ps2_postproc (infile, outfile, sim_length, copasi=True)
Perform post processing organisation to double parameter scan report files.
```

Parameters

- infile the model to process
- outfile the directory to store the results
- sim_length the length of the simulation
- copasi True if the model is a Copasi model

sbpipe.tasks.sim analyse gen stats table module

```
sbpipe.tasks.sim_analyse_gen_stats_table.main(argv=None)
```

```
sbpipe.tasks.sim_analyse_gen_stats_table.sim_analyse_gen_stats_table (inputdir, out-put-dir, model, out-put-file, xaxis_label, yaxis_label, variable)
```

Plot model simulation time courses (Python wrapper).

Parameters

- **inputdir** the directory containing the data to analyse
- outputdir the output directory containing the results
- model the model name
- outputfile the output file containing the statistics
- **xaxis_label** the label for the x axis (e.g. Time [min])
- yaxis_label the label for the y axis (e.g. Level [a.u.])
- variable the model variable to analyse

sbpipe.tasks.sim_analyse_plot_comb_sims module

```
sbpipe.tasks.sim_analyse_plot_comb_sims.main(argv=None)
sbpipe.tasks.sim_analyse_plot_comb_sims(inputdir,
out-
put-
dir,
model,
exp_dataset,
plot_exp_dataset,
exp_dataset,
axxis_label,
yaxis_label,
vari-
able)
```

Plot model simulation time courses (Python wrapper).

- **inputdir** the directory containing the data to analyse
- outputdir the output directory containing the results
- model the model name
- **exp_dataset** the full path of the experimental data set
- plot_exp_dataset True if the experimental data set should also be plotted
- **exp_dataset_alpha** the alpha level for the data set
- **xaxis_label** the label for the x axis (e.g. Time [min])
- yaxis_label the label for the y axis (e.g. Level [a.u.])
- variable the model variable to analyse

sbpipe.tasks.sim_analyse_plot_sep_sims module

Plot model simulation time courses (Python wrapper).

Parameters

- **inputdir** the directory containing the data to analyse
- outputdir the output directory containing the results
- model the model name
- **exp_dataset** the full path of the experimental data set
- plot_exp_dataset True if the experimental data set should also be plotted
- exp_dataset_alpha the alpha level for the data set
- **xaxis_label** the label for the x axis (e.g. Time [min])
- yaxis_label the label for the y axis (e.g. Level [a.u.])
- **variable** the model variable to analyse

sbpipe.tasks.sim_analyse_summarise_data module

Plot model simulation time courses (Python wrapper).

Parameters

- inputdir the directory containing the data to analyse
- model the model name
- outputfile_repeats the output file containing the model simulation repeats
- variable the model variable to analyse

sbpipe.tasks.sim_postproc module

```
sbpipe.tasks.sim_postproc.generic_postproc (infile, outfile, copasi=True)
Perform post processing file editing for the simulate pipeline
```

- infile the model to process
- outfile the directory to store the results
- copasi True if the model is a Copasi model

```
sbpipe.tasks.sim_postproc.main(argv=None)
```

sbpipe.tasks.sim_postproc.sim_postproc(infile, outfile, copasi=True)

Perform post processing file editing for the simulate pipeline

Parameters

- infile the model to process
- outfile the directory to store the results
- copasi True if the model is a Copasi model

sbpipe.tasks.utils module

Module contents

sbpipe.utils package

Submodules

sbpipe.utils.io module

```
sbpipe.utils.io.files with pattern recur(folder, pattern)
```

Return all files with a certain pattern in folder+subdirectories

Parameters

- folder the folder to search for
- pattern the string to search for

Returns the files containing the pattern.

```
sbpipe.utils.io.get_pattern_pos (pattern, filename)
```

Return the line number (as string) of the first occurrence of a pattern in filename

Parameters

- pattern the pattern of the string to find
- **filename** the file name containing the pattern to search

Returns the line number containing the pattern or "-1" if the pattern was not found

```
sbpipe.utils.io.refresh(path, file_pattern)
```

Clean and create the folder if this does not exist.

Parameters

- path the path containing the files to remove
- **file_pattern** the string pattern of the files to remove

```
sbpipe.utils.io.remove_file_silently(filename)
```

Remove a filename silently, without reporting warnings or error messages. This is not really needed by Linux, but Windows sometimes fails to remove the file even if this exists.

Parameters filename – the file to remove

sbpipe.utils.io.replace_str_in_file (filename_out, old_string, new_string)

Replace a string with another in filename_out

Parameters

- filename_out the output file
- old_string the old string that should be replaced
- new_string the new string replacing old_string

sbpipe.utils.io.replace_str_in_report (report)
Replace nasty strings in COPASI report file.

Parameters report - the report

 $\verb|sbpipe.utils.io.write_mat_on_file| (path, filename_out, data)|$

Write the matrix results stored in data to filename_out

Parameters

- path the path to filename_out
- filename_out the output file
- data the data to store in a file

sbpipe.utils.parcomp module

sbpipe.utils.parcomp.call_proc(params)
Run a command using Python subprocess.

Parameters params - A tuple containing (the string of the command to run, the command id)

sbpipe.utils.parcomp.is_output_file_clean (filename, stream_type='standard output')
Check whether a file contains the string 'error' or 'warning'. If so a message is printed.

Parameters

- filename a file
- **stream_type** 'stderr' for standard error, 'stdout' for standard output.

Returns True

 $sbpipe.utils.parcomp. \textbf{parcomp} (\textit{cmd}, \textit{cmd_iter_substr}, \textit{output_dir}, \textit{cluster='local'}, \textit{runs=1}, \textit{local_cpus=1}, \textit{output_msg=False}, \textit{colnames=[]}) \\ Generic function to run a command in parallel}$

Parameters

- cmd the command string to run in parallel
- **cmd_iter_substr** the substring of the iteration number. This will be replaced in a number automatically
- output_dir the output directory
- cluster the cluster type among local (Python multiprocessing), sge, or lsf
- runs the number of runs. Ignored if colnames is not empty
- local_cpus the number of cpus to use at most
- output_msg print the output messages on screen (available for cluster='local' only)
- colnames the name of the columns to process

Returns True if the computation succeeded.

```
sbpipe.utils.parcomp.quick_debug(cmd, out_dir, err_dir)
```

Look up for error and warning in the standard output and error files. A simple debugging function checking the generated log files. We don't stop the computation because it happens that these messages are more warnings than real errors.

Parameters

- cmd the executed command
- out_dir the directory containing the standard output files
- err_dir the directory contining the standard error files

Returns True

```
sbpipe.utils.parcomp.run_cmd(cmd)
```

Run a command using Python subprocess.

Parameters cmd – The string of the command to run

```
sbpipe.utils.parcomp.run_cmd_block(cmd)
```

Run a command using Python subprocess. Block the call until the command has finished.

Parameters cmd – A tuple containing the string of the command to run

```
sbpipe.utils.parcomp.run_jobs_local(cmd, cmd_iter_substr, runs=1, local_cpus=1, out-
                                           put_msg=False, colnames=[])
```

Run jobs using python multiprocessing locally.

Parameters

- cmd the full command to run as a job
- cmd_iter_substr the substring in command to be replaced with a number
- runs the number of runs. Ignored if colnames is not empty
- local_cpus The number of available cpus. If local_cpus <=0, only one core will be used.
- output_msq print the output messages on screen (available for cluster_type='local'
- colnames the name of the columns to process

Returns True

```
sbpipe.utils.parcomp.run_jobs_lsf(cmd, cmd_iter_substr, out_dir, err_dir, runs=1, col-
                                         names=[])
```

Run jobs using a Load Sharing Facility (LSF) cluster.

Parameters

- cmd the full command to run as a job
- cmd_iter_substr the substring in command to be replaced with a number
- out_dir the directory containing the standard output from bsub
- err_dir the directory containing the standard error from bsub
- runs the number of runs. Ignored if colnames is not empty
- colnames the name of the columns to process

Returns True if the computation succeeded.

```
sbpipe.utils.parcomp.run_jobs_sge(cmd, cmd_iter_substr, out_dir, err_dir, runs=1, col-
                                         names=[])
```

Run jobs using a Sun Grid Engine (SGE) cluster.

Parameters

• cmd – the full command to run as a job

- cmd_iter_substr the substring in command to be replaced with a number
- out_dir the directory containing the standard output from qsub
- err_dir the directory containing the standard error from qsub
- runs the number of runs. Ignored if colnames is not empty
- colnames the name of the columns to process

Returns True if the computation succeeded.

sbpipe.utils.rand module

```
sbpipe.utils.rand.get_rand_alphanum_str(length)
Return a random alphanumeric string

Parameters length - the length of the string

Returns the generated string
```

sbpipe.utils.rand.get_rand_num_str(length)
Return a random numeric string

Parameters length - the length of the string

Returns the generated string

sbpipe.utils.re utils module

```
\verb|sbpipe.utils.re_utils.escape_special_chars| (\textit{text})
```

Escape ^,%, ,[,],(,),{,} from text :param text: the command to escape special characters inside :return: the command with escaped special characters

```
sbpipe.utils.re_utils.nat_sort_key(str)
```

The key to sort a list of strings alphanumerically (e.g. "file10" is correctly placed after "file2")

Parameters str – the string to sort alphanumerically in a list of strings

Returns the key to sort strings alphanumerically

Module contents

Submodules

```
sbpipe.__main__ module
```

```
sbpipe.___main__.main(argv=None)
```

sbpipe.main module

```
sbpipe.main.main(argv=None)
SBpipe main function.
```

Returns 0 if OK, 1 if trouble

```
sbpipe.main.read_file_header(filename)
```

Read the first line of a file

Parameters filename – the file name to read

Returns the first line

SBpipe function.

Parameters

- **create_project** create a project with the name as argument
- **simulate** model simulation using a configuration file as argument
- parameter_scan1 model one parameter scan using a configuration file as argument
- parameter_scan2 model two parameters scan using a configuration file as argument
- parameter_estimation model parameter estimation using a configuration file as argument
- logo True to print the logo
- license True to print the license
- nocolor True to print logging messages without colors
- log_level Set the logging level
- quiet True if quiet (CRITICAL+)
- **verbose** True if verbose (DEBUG+)

Returns 0 if OK, 1 if trouble (e.g. a pipeline did not execute correctly).

```
sbpipe.main.sbpipe_logo()
```

Return sbpipe logo.

Returns sbpipe logo

```
sbpipe.main.set_basic_logger(level='INFO')
```

Set a basic StreamHandler logger. :param level: the level for this console logger

```
sbpipe.main.set_color_logger(level='INFO')
```

Replace the current logging.StreamHandler with colorlog.StreamHandler. :param level: the level for this console logger

```
sbpipe.main.set_console_logger(new_level='NOTSET', current_level='INFO', no-
color=False)
```

Set the console logger to a new level if this is different from NOTSET

Parameters

- new_level the new level to set for the console logger
- current_level the current level to set for the console logger
- nocolor True if no colors shouls be used

```
sbpipe.main.set_logger(level='NOTSET', nocolor=False)
```

Set the logger :param level: the level for the console logger :param nocolor: True if no colors shouls be used

sbpipe.sbpipe config module

```
sbpipe.sbpipe_config.isPyPackageInstalled(package)
```

Utility checking whether a Python package is installed.

Parameters package - a Python package name

Returns True if it is installed, false otherwise.

sbpipe.sbpipe_config.which(cmd_name)
Utility equivalent to which in GNU/Linux OS.

Parameters cmd_name - a command name

Returns return the command name with absolute path if this exists, or None

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