PyExperimentSuite Documentation

Contents

1	Intro	oduction	2		
2	Dov	vnload and Installation	2		
	2.1	Requirements	2		
	2.2	Download from github.com	3		
	2.3	Download with git	4		
	2.4	Successful Installation	4		
3	Using PyExperimentSuite				
	3.1	Typical Workflow	5		
	3.2	Getting Started	5		
	3.3	Implementing reset() and iterate()	7		
	3.4	Optional Restore Functionality	10		
	3.5	PyExperimentSuite for non-Python experiments	12		
4	The Configuration File 13				
	4.1	Basic Parameter Definitions	14		
	4.2	Evaluating Parameter Ranges	15		
	4.3	Parameter Range Combinations	16		
5	Run-Time Options 17				
	5.1	Help	17		
	5.2	Configuration File	18		
	5.3	Number of Cores	18		
	5.4	Deletion of Old Experiments	18		
	5.5	Selected Experiments	18		
	5.6	Browse Experiments	19		
	5.7	Progress of Experiments	19		
6	Res	ult Retrieval	20		
•		Log Files	20		

	6.2	Python Interface
	6.3	Retrieving Information about Experiments
	6.4	Single Values
	6.5	Single Values from Parameter Range Experiments
	6.6	Histories
	6.7	Histories from Parameter Range Experiments
	6.8	Aggregated Histories over Repetitions
7	Exa	mples and Best Practices 26
	7.1	Strings in Iterable Objects
	7.2	Passing Classes as Parameters
	7.3	Passing Lists as Parameters
	7.4	Debugging PyExperimentSuite Scripts

1 Introduction

PyExperimentSuite is an open source software tool written in Python, that supports scientists, engineers and others to conduct automated software experiments on a large scale with numerous different parameters. It reads parameters (or ranges of parameters) from a configuration file, runs the experiments using multiple cores if desired and logs the results in files. Parameter combinations can be evaluated as a grid (each combination of parameters) or in a list (try several defined parameter combinations in row). PyExperimentSuite also supports continuing any experiments where left off when the execution was interrupted (e.g. power failure, process was killed, etc.). The experiment results can be obtained in different ways by a built-in Python interface.

2 Download and Installation

To install PyExperimentSuite, you can either download the package from the Github website, or install it directly with *git*. If you're not familiar with the version management command line tool *git*, follow the steps in section 2.2.

2.1 Requirements

PyExperimentSuite is developed in Python and does not have many dependencies. It does require Python 2.6 or higher, though, because it makes use of the multiprocess-

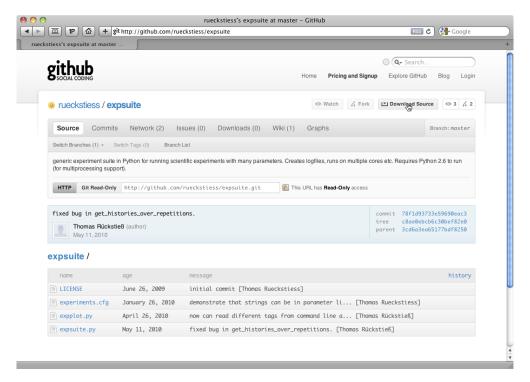


Figure 1: Downloading PyExperimentSuite from github.com. Click on the button "Download Source" on the top left and then select the type of archive you would like to download.

ing module, which was only added in version 2.6. Another dependency is numpy. Other than that, PyExperimentSuite is a stand-alone package.

2.2 Download from github.com

To download and setup PyExperimentSuite manually, go to the the website on which the project is hosted: http://github.com/rueckstiess/expsuite/. Click on the "Download Source" button on the top right and select the type of archive (zip or tar) that you would like to download (see Fig. 1).

After the download is complete, unpack the archive and move the extracted folder to your desired location, preferably in a directory of your \$PYTHONPATH environment variable (otherwise, you might have to update your \$PYTHONPATH environment variable to include the destination path).

2.3 Download with git

If you are familiar with git and have it installed on your system, you can simply clone the project directly from github. Switch to the directory under which you want to install the expsuite directory. This directory should be in your \$PYTHONPATH environment variable (otherwise, you might have to update your \$PYTHONPATH environment variable to include the destination path). Now run this command from the command line:

```
$ git clone http://github.com/rueckstiess/expsuite.git
```

A directory will be created under the current directory and the sources will be downloaded into the directory. Note that you will not be able to push any committed changes back to the github repository, it is read-only.

2.4 Successful Installation

To test, whether the installation was successful, open your Python interpreter by typing python at the command line, and import the package like this:

```
>>> import expsuite
```

If you receive no error messages, installation was successful and you can now use PyExperimentSuite.

3 Using PyExperimentSuite

PyExperimentSuite is most useful for experiments written in Python. The next sections will demonstrate, how you derive from the PyExperimentSuite base class to create your own suite, how you implement the necessary functions reset() and iterate(), and what the restore option on iteration level is and how to make use of it.

If your experiments are not written in Python but you still want to try out PyExperimentSuite, you find some help in Section 3.5.

3.1 Typical Workflow

We will give the common workflow with PyExperimentSuite here at the beginning as an overview of what lies ahead of you. These steps are then explained in detail in the following sections.

In order to successfully use PyExperimentSuite, these steps need to be performed:

- 1. create a class which derives from the PyExperimentSuite class
- 2. add the if __name__ == '__main__': part at the bottom of the file
- 3. fill out the reset () and iterate () methods
- 4. (optional) fill out the <code>save_state()</code> and <code>restore_state()</code> methods and set the <code>restore_supported</code> flag to <code>True</code>
- 5. create or edit the experiments.cfg file and add all the parameters and experiments
- 6. run the suite from the command line
- 7. after completion, open your Python console or create a Python script and use the built-in interface to access your results and visualize or post-process them

3.2 Getting Started

Each new experiment setup requires that you write a few lines of Python code. You need to create a new class that inherits from the PyExperimentSuite base class, fill in the missing reset () and iterate() methods, and optionally take care of saving and loading relevant data for your experiments so the suite can be interrupted and continued at any time.

It is recommended (but not essential), that you create a new folder for your experiment setup. PyExperimentSuite will create subfolders below its execution folder and depending on how many experiments you create, it is easy to loose oversight of all your experiments. Therefore, it is best to keep everything belonging to one set of experiments in one folder.

For a typical experiment setup, create a new file and give it a plausible name, for example suite.py. Import the PyExperiment Suite class and create a new class that inherits from PyExperiment Suite. Next, add the empty method declarations for

the two functions reset () and iterate () as shown in Listing 1. While reset () does not need a return value, iterate () needs to return a dictionary. For now, this can be the empty dictionary. Finally, add the three lines at the bottom of the script for creating the MySuite object and starting it. If you're done with this task, you have completed workflow point 1 and 2 from Section 3.1.

Listing 1: MySuite class definition with empty methods for later implementation.

```
from expsuite import PyExperimentSuite

class MySuite(PyExperimentSuite):

   def reset(self, params, rep):
       pass

   def iterate(self, params, rep, n):
       ret = {}
       return ret

if __name__ == '__main__':
    mysuite = MySuite()
   mysuite.start()
```

If you go to your command line and start the script with

```
$ python suite.py
```

an error message will appear to notify you that the config file experiments.cfg could not be found. To remedy this, create a second file called experiments.cfg in the same folder and add a default section with some parameters, as shown in Listing 2. These three parameters are necessary in all your configuration files and PyExperimentSuite will complain, if any of them are missing.

Listing 2: Configuration file with default section.

```
[DEFAULT]
repetitions = 1
iterations = 100
path = results
```

After saving the configuration file, go back to the command line and try running the script again:

```
$ python suite.py
```

This time, there should be no error message after hitting enter. The script executes all the defined experiments (which is currently *none*) and returns to the command line.

If you want to see a little more feedback from your script, run the help option on the command line with

```
$ python suite.py --help
```

Now you receive a short list of possible commands that your script can be executed with.

3.3 Implementing reset() and iterate()

The main implementation work you have to do for PyExperimentSuite goes in the two functions reset () and iterate(). In order to understand better what they do, have a look at Listing 3. It demonstrates in a simplified code example the execution order of these functions.

Listing 3: Simplified execution loop of an experiment in the suite.

```
# execute this loop for each experiment
for r in range(repetitions):
    reset()
    for i in range(iterations):
        iterate()
```

For each new experiment, the reset () method is called first. Here, all the necessary objects should be created that are required by the experiment. Everything needed for the experiment should be assigned to class variables (with the prefix self.) because it is needed in the iterate() method later on, which doesn't have access to local variables.

Implementation of reset()

What code exactly goes into your reset () method depends entirely on the experiment you want to set up.

Let's say you want to train a Neural Network¹ with a given dataset and test several different learning rates as part of your experiment. Your reset () method will need the code to initialize the Neural Network (perhaps with random weights), tell the network where it can find the dataset and also set the learning rate. Listing 4 shows how this would be implemented in the reset () function. Note that you have to import all the necessary modules and classes before, the listing only shows an example for the reset () definition.

Listing 4: Example of reset() method for training a Neural Network

```
def reset(self, params, rep):
    # create network and randomize weights
    self.network = NeuralNetwork()
    self.network.randomizeWeights()

# give dataset to network
    path = os.path.join('./data/', params['filename'])
    self.network.setDataSet(path)

# set learning rate
    self.network.learningrate = params['learningrate']
```

The network is assigned to a class variable self.network because we will need access to it later on in the iterate() method. Values that are not required any further can be stored in local variables, like the path of the dataset in our example. In this example we make both the filename and the learning rate a parameter of the experiment, which needs to be defined in the configuration file experiments.cfg.

We can then access these parameters through the dictionary params which uses the parameter name defined in experiments.cfg as key. This means, that we need two more parameters in our configuration file. Listing 5 shows the changes to the default section of our original config file from Section 3.2.

In addition to the params dictionary, that is passed to reset () and contains the parameters of the current experiment, there is an second parameter rep. This integer contains the index of the current repetition, starting from 0 and increasing for each repetition, up to params ['repetitions']-1. Sometimes this number might be of use to you, for example if you have to load a particular datafile for each seperate repetition, or if the initialization of each experiment needs to be unique. You could for

¹You don't need to know what a Neural Network is or how it works in order to understand this example. Just think of it as a black box that has some internal parameters (called *weights*) that need to be adjusted according to a given dataset, which is presented to the network repeatedly. This provess is called *training*. It further contains a free parameter called *learning rate* that determines how fast the network will learn. Finding the optional learning rate is not trivial and is often done by trial-and-error.

example seed a random number generator with a combination of the system time and this repetition number, to guarantee a unique random sequence for each repetition.

Listing 5: Configuration file for the neural network example.

```
[DEFAULT]
repetitions = 1
iterations = 100
path = results
filename = 'stockmarket.data'
learningrate = 0.01
```

Implementation of iterate()

The iterate() method will be called repeatedly during the execution of one single repetition of one experiment. It is a single step or cycle within the experiment. Most experiments have a natural separation into repeated iterations already. If your experiment does not seem to be iterative, you might have to introduce some separation into steps manually. If you feel like this is not possible at all, you can also use a single iteration for the whole experiment, setting the iterations parameter in the config file to 1.

Just as in the reset () method, the params dictionary is also passed to the iterate () method. The second parameter passed to this method is rep, which indicates the current index of repetition, starting from 0. iterate () has a third parameter, n, which represents the current iteration index. The variable n starts from 0 and counts up to params ['iterations']-1.

While the reset () method did not return any values, the iterate() method is expected to return a dictionary. The dictionary should contain the relevant information, that you want to store in the log files. This can be some measure of progress of your experiment, an error, some identifying strings or the running index n.

Put the relevant values, together with a descriptive key, into a dictionary and return it at the end of the function:

```
ret = {'first_value':value1, 'another_value':value2}
return ret
```

Make sure that none of the keys contains spaces, because the key-value pairs them-

selves are stored with spaces as delimiters in the log files. If you choose a key that contains spaces by accident, PyExperimentSuite will rename it and replace the spaces with underscores, and issue a warning.

More about how to retrieve this information after your experiments have finished can be found in Section 6.

To continue the Neural Network example, we will now execute a learning step in the iterate() function. We assume that our NeuralNetwork class offers two methods, train() and test(). The first will execute a training step in the network and return the training error. The latter then runs the network on an independent test dataset and evaluates the test error. We'd like to store both values, together with the current iteration step, in the log files. Listing 6 demonstrates how to achieve this. The keys in the dictionary are strings, with which we will later be able to retrieve the results. Use unique, descriptive labels for these keys.

Listing 6: Example of iterate() method for training a Neural Network

```
def iterate(self, params, rep, n):
    trainerr = self.network.train()
    testerr = self.network.test()
    ret = {'n':n, 'trainerror':trainerr, 'testerror':testerr}
    return ret
```

The suite is now ready to execute the experiments. By calling the script from the command line, it parses the configuration file, runs the required number of experiments, repetitions and iterations, and stores the results in log files.

At this point you have completed step 3 of the workflow checklist from Section 3.1.

3.4 Optional Restore Functionality

So far, we implemented everything necessary to successfully run the PyExperimentSuite. Repetitions will be spread over the available cores on your machine, speeding up the execution by a factor of how many cores you have (and be willing to share with Py-ExperimentSuite).

If you interrupt the suite while it is executing the experiments, by killing the process or even by a power failure, the suite is left in a somewhat undefined state.

Without any further implementation work from you, all it can do is to abandon the already executed iterations in the current repetition and start the repetition again. Py-ExperimentSuite does that automatically: it will delete the already logged iterations of the current repetition (or repetitions, if you use multiple cores) and restart them from iteration step 0, by calling the reset() method again. Already completed repetitions are of course not affected and will remain in the log files.

Still, some experiments can be very time-consuming and even losing a few iteration steps might be inacceptable to you. In this case, you need to set the class variable restore_supported to True and implement a few more lines of code, namely the two methods save_state() and restore_state(). The save_state() method is called after each iteration step has completed, and its task is to save all relevant information needed to continue from this iteration step to the disk. This can be done by pickleing² crucial objects to a file, or otherwise saving the current state of the experiment. The current repetition and iteration indices are again passed to the function as variables rep and n. This might be useful if you want to store not only the current iteration state but perhaps want to create a history of all iteration steps. The suite can easily be extended to start from any particular point in time, not just the last one.

If the variable restore_supported is set to True, the suite checks upon start if there are unfinished experiments on iteration level. It then calls the restore_state() method with the appropriate parameters, repetition and iteration index. The task of the restore_state function is to load the correct data from files and restore the class objects to that state. Before the restore_state function is called, a call to reset() has already been made. All you have to do is change the initialized class members to the values you saved in the save_state() method.

Let's continue the NeuralNetwork example and fill out the two methods, shown in Listing 7. First of all, we need to enable the restore_supported flag, so the PyExperimentSuite knows that it should call the appropriate functions to save and restore the data. Next, we implement the save_state() method. We create a file handle for file write access. The location of the file can be anywhere, but it is probably best to place it in the current experiment's folder, where the log files are stored. This location can be retrieved by concatenating the parameters' path and name fields. os.path.join() is perfectly suited for this. As we might execute several repetitions simultaneously on different cores, it is absolutely necessary to have one file per repetition, thus using the repetition index rep in the file name. In this example we only save the last executed iteration per repetition, so we don't need to create seperate files for each iteration. We use the module cPickle to serialize the whole object, dump it in the file, and close the file afterwards.

²pickle, and its faster C implementation cPickle are Python packages that can serialize objects and save them to disk, and restore the files and create objects again. See the Python documentation for more information about these modules.

Restoring the data is straight-forward as well. Instead of write access, we create a file handle with read access to the file we want to restore. Again, the repetition index is necessary as we have one file for each repetition. Using cPickle's load() function, we import the stored data and assign it to the class variable self.network.

That's all the necessary steps and you have completed workflow item 4 from the list in Section 3.1. If you run the script now, and abort or kill it during runtime, the most data you could lose is one single iteration (the one that was currently being calculated, before the results could have been saved to disk).

When starting PyExperimentSuite again, it will load the last available iteration and continue from where it left off.

Listing 7: Save and restore methods for the Neural Network example.

3.5 PyExperimentSuite for non-Python experiments

Although PyExperimentSuite is most useful for your Python experiments, it is possible to use it for all kinds of automated scripts, even in other languages. The only condition would be, that the experiment can be run from the command line. This includes any C programs, shell scripts or other tools that you might want to evaluate.

In order to make them work with PyExperimentSuite, you need to make use of Python's command line modules, like <code>sys</code>, <code>os</code> and others. Basically, you need to call the necessary scripts from within Python, and find a way to grab the return value. A good possibility might be the function <code>os.open()</code> or <code>os.popen3()</code>, which lets you access stdin. stdout and stderr streams as file handles.

This feature has not been tested much and might not work in all environments. I would therefore classify this feature as experimental for now. Feedback on successful usage of non-Python experiments is more than welcome.

4 The Configuration File

So far, we have talked a lot about how to set up your python suite script and how to implement the necessary methods. This Section will give more details about the configuration file which is workflow step 5 from Section 3.1.

The default configuration file is called experiments.cfg and is expected to sit in the same folder as your suite.py script (or whatever name you choose for it instead). The configuration script is parsed by the python module ConfigParser³ and follows its syntax.

Basically, the file is separated in different sections, that start with a section header in square brackets, like [sectionname]. Following the section header are entries of the form name:value or name=value, each on a single line. Leading whitespace is removed from values. Lines beginning with # or; are ignored and can be used as comments.

A special section with header [DEFAULT] can be provided. Any other section will inherit all key-value pairs from this section, unless it defines another value for an existing key, in which case the new value will be chosen instead.

For the PyExperimentSuite, it is recommended to create one <code>[DEFAULT]</code> section for all parameters that are common to all experiments (like the path to store the results for example), and to create a separate section for each different experiment. The three keys <code>repetitions</code>, <code>iterations</code> and <code>path</code> need to exist in all experiments, so it makes sense to define default values for them in the <code>[DEFAULT]</code> section.

There is one more parameter that takes a special role for PyExperimentSuite, and that is experiment. We will discuss its relevance further down in this section. All other parameters are entirely dependent on the kind of experiments you want to conduct and need to be choosen accordingly.

³http://docs.python.org/library/configparser.html

4.1 Basic Parameter Definitions

Acceptable values include any type that can be evaluated to a Python int, float, string, or object. In fact, PyExperimentSuite will try to evaluate the given value using the Python function <code>eval()</code>, and if no errors occur, interpret it as its Python pendant. This means, that everything that can be evaluated to an integer will be treated as one, the same is true for floats. More complex values, like sin(2.) or 0.5*pi are also valid expressions.

Everything, that cannot be evaluated without error, will be interpreted as a string. This means that both strings with and without quotation marks will be interpreted as a string in Python. It is your choice to add quotation marks (single or double) or not. If you do, they will be stripped away.

Lists and iterable objects play a special role as explained in 4.2 below and will, by default, not be passed to the parameters as their Python equivalent.

Listing 8 demonstrates some valid key-value pairs.

Listing 8: Examples for valid key-value pairs for the configuration file.

```
[DEFAULT]
path = ./results/
repetitions = 1
iterations = 100
filename = 'stockmarket.data'
learningrate = 0.01
[experiment1]
path = "./testruns/"
alpha = 0.1
beta = 2.
gamma = sin(2.)
delta = 0.5*pi
[experiment2]
comment = "This is experiment 2"
alpha = 0.1
beta = [0.5, 1.0, 1.5]
[experiment3]
comment = "Experiment 3 uses a grid for the parameter combinations"
experiment = 'grid'
```

```
alpha = arange(0.0, 1.0, 0.2)
beta = [0.5, 1.0, 2.0]

[experiment4]
comment = "Experiment 4 uses a list for the parameter combinations."
experiment = list
alpha = arange(0.0, 1.0, 0.2)
beta = [0.5, 1.0, 2.0]

[experiment5]
comment = "Experiment 5 will assign the whole lists to parameters."
experiment = 'single'
alpha = arange(0.0, 1.0, 0.2)
beta = [0.5, 1.0, 2.0]
```

4.2 Evaluating Parameter Ranges

If your value evaluates to a python object that is iterable (e.g. lists, numpy arrays, generator objects, etc.) PyExperimentSuite will not assign the object to the parameter by default. Instead, it assumes that you want to try a range of values in separate experiments. This is very handy if you need to find the optimal value for certain parameters, because you can just define the range and let PyExperimentSuite do all the testing. If you need to assign a list or iterable object to a parameter and do not want PyExperimentSuite to create several parameter range experiments in this way, you need to set your experiment type to 'single'. Section 7.3 explains how to do that. Per default, the elements of iterable objects are passed to the parameter individually in seperate sub-experiments.

Log files are usually stored in a directory named ./path/name/ where path is the path parameter from the configuration file and name is the section name of the experiment. If you define a parameter as a range, PyExperimentSuite will create another level of sub-directories under ./path/name/. Each sub-directory has a unique name that consists of the parameter names and values of the ranges.

Experiment 2 from Listing 8 would create the following three paths:

```
./results/experiment2/beta0.5/
./results/experiment2/beta1.0/
./results/experiment2/beta1.5/
```

This also helps identify the folders of certain experiments, if you should ever look for the log files. In the normal use case, you will most likely not have to go and find these files by yourself but use the Python interface to retrieve your results instead. More about the interface can be found in Section 6.

All other non-iterable parameters are kept constant for the experiments. This is also true for the repetitions parameter, which means that each of these "sub-experiments" are repeated an equal number of times, defined by repetitions in the config file.

4.3 Parameter Range Combinations

Should you define more than one parameter in your config file that evaluates to an iterable object, you have two choices: try every single combination of all parameter choices or run successive experiments with the parameters of the same list indices. Figure 2 explains the difference graphically.

The first choice is called a *grid* experiment, and is also set as the default. PyExperimentSuite will create sub-experiments for every single combination of all parameters.

Experiment 3 in Listing 8 defines two iterable parameters, alpha and beta. Numpy's arange () evaluates to a list with the values 0.0, 0.2, 0.4, 0.6 and 0.8 for alpha, and beta takes on the values of 0.5, 1.0 and 2.0. In total, PyExperimentSuite will run $5 \cdot 3 = 15$ experiments, and the folders for the log files will contain the key-value pairs in their name, e.g. ./results/experiment3/alpha0.2beta1.0/. Note the additional parameter experiment, which is set to the value 'grid'.

The second option for running an experiment with parameter range combinations is a *list* experiment. Here, each experiment will contain values from both lists at the same indices. The number of sub-experiments created by this process is the length of the shortest list. In case of experiment 4 in Listing 8, PyExperimentSuite will create 3 experiments with the following parameters:

```
alpha = 0.0, beta = 0.5
alpha = 0.2, beta = 1.0
alpha = 0.3, beta = 2.0
```

The list for alpha would contain more values, but since there are not corresponding values in the beta list, the other experiments are discarded.

To control which type of experiment you want to conduct, you can define the special parameter experiment and assign grid or list to it (with or without quotation marks), as shown in Listing 8. The default value is grid. If you don't want to create multiple experiments but use the assigned iterable object as value, use single as the experiment value.

Of course both mechanisms are not limited to two lists but can have arbitrary numbers of iterable objects. Just keep in mind that in the grid case, the number of experiments is the product of all list lengths, and in the list case, it is the length of the shortest list.

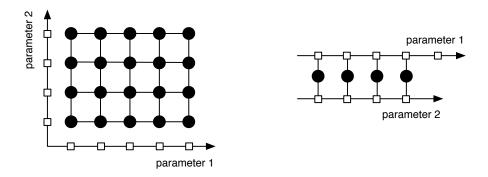


Figure 2: Two different types of experiments with parameter range combinations. The white squares represent discrete values of a parameter range, which is shown as an arrow. The black circles are experiments with a certain parameter combination. On the left side, a grid experiment is evaluated. On the right side, a list experiment is shown.

5 Run-Time Options

When calling your suite.py script from the command line, several options are available. The most important of them is the -h or --help flag. Calling the script with that option will present a list of all the available command line options. In this section, we will describe each of them and explain their usage. After reading this chapter, you are able to complete workflow step 6 of Section 3.1.

5.1 Help

```
$ python suite.py -h
$ python suite.py --help
```

This option will not execute the script but instead show a list of available options for the suite.

5.2 Configuration File

```
$ python suite.py -c <config-file>
$ python suite.py --config=<config-file>
```

Specify your configuration file with the experiment definitions with this option. The default file is experiments.cfg.

5.3 Number of Cores

```
$ python suite.py -n <number>
$ python suite.py --numcores=<number>
```

This option lets you restrict PyExperimentSuite to a certain number of cores. If you specify this option, PyExperimentSuite will at most use the number of cores given. If this option is not present, PyExperimentSuite will use all the available cores it can find on the local machine.

When debugging your script, it is recommended to use only one core (with option -n 1). Details are explained in Section 7.4.

5.4 Deletion of Old Experiments

```
$ python suite.py -d
$ python suite.py --del
```

With this option you can tell PyExperimentSuite to delete all existing experiments before starting to parse the configuration file. This is useful if you made a mistake in the definition of your config file and need to restart the whole experiment set from scratch. This cannot be undone, so be careful when to use this option.

5.5 Selected Experiments

```
$ python suite.py -e <experiment>
$ python suite.py --experiment=<experiment>
```

If you don't want to execute all experiments defined in the config file but only one or a few selected ones, you can use this option. Specify each experiment you want to run explicitly with the -e option. If this option is not present, all defined experiments will be executed.

5.6 Browse Experiments

```
$ python suite.py -b
$ python suite.py --browse
$ python suite.py -B
$ python suite.py --Browse
```

The browse option does not execute the script (no experiments are started or continued) but returns a list of current experiments and additional information. The lower-case version of this option tells you each experiment's name, start and end time, the number of repetitions and iterations, and the progress of the experiment in percent. The upper-case version of this option is more verbose and additionally tells you all the parameters for this experiment.

```
experiment ./results/experiment2/beta0.20alphano
    started 2010-07-28 00:10:22
    ended 2010-07-28 00:10:22
repetitions 1
iterations 10
    progress 100%
```

5.7 Progress of Experiments

```
$ python suite.py -p
$ python suite.py --progress
```

Similar to the browse option, this progress option does not start or continue any experiments. It only shows a list of experiments and a progress bar of their completion. This is helpful if you have many experiments in the queue and quickly want to check how far the whole suite is.

6 Result Retrieval

This section explains what you can do after your experiments have finished. It will explain where the log files are stored, what you will find in the log files, and how you can easily retrieve any relevant information within Python, without writing your own log file parser.

6.1 Log Files

The log files that PyExperimentSuite creates contain the key-value pairs that the iterate() function (ref. Section 3.3) returned as dictionary. Each line of a log file corresponds to one single iteration. They key-value pairs are stored in this format:

```
key1:value1 key2:value2 key3:value3 ...
```

For each repetition, a seperate log file is created, and each log file is named #.log where # is the repetition number.

As the order of Python dictionary entries is unspecified, so is the order of the keyvalue pairs in the log file. The log file of our ongoing Neural Network example could look like this:

```
trainerror:0.32441 testerror:0.45531 n:0 trainerror:0.29245 testerror:0.42648 n:1 trainerror:0.27552 testerror:0.42796 n:2 trainerror:0.23443 testerror:0.39441 n:3 ...
```

The log file would be stored at ./results/nn-learningrate/0.log and there would be only one log file for the one repetition we requested.

If your experiment contained any parameter range evaluations with lists or other iterable objects (see Section 4.2 and 4.3), PyExperimentSuite will have created subdirectories under ./path/name/ with unique names based on the evaluated parameter combination. Within the sub-directories, the log files are stored as described before.

PyExperimentSuite not only stores the log files in the experiment folder, but one additional file: the configuration file of that single experiment. It is named experiment.cfg without the s, to distinguish it from the original experiments.cfg file. If you open

it, you will see that all the parameters that this particular experiment ran with, are there. There is no default section.

The existance of this additional configuration file has three reasons: it is a reminder for you, if you have forgotten which parameters exactly you evaluated in this experiment. PyExperimentSuite will also consult this file if you call the option -B or --Browse to tell you the value of all the parameters. Lastly, by specifying the -c or --config option, you can run only this particular experiment again.

Experiments with parameter ranges create an intermediate experiment.cfg file in the ./path/name folder and one for each ./path/name/sub-dir folder as well.

You can use these log files to post-process your results, or plot or otherwise display them. In order to do this, you would usually have to write a parser, that can read the key:value syntax. PyExperimentSuite ships with a build-in parser already, that is accessible via a Python interface. The next section explains, how to retrieve your stored results.

6.2 Python Interface

PyExperimentSuite comes with several functions, that are intended to be used to retrieve data after the experiments have run. This section describes the final workflow step 7 from the list in Section 3.1.

All the functions intended to use for data retrieval start with the prefix <code>get_</code>. They are described in the following sections. Throughout the examples, we assume that you entered a Python console, imported your the class MySuite from <code>suite.py</code> and created a mysuite object, like shown in Listing 9. For post-processing examples, we also import numpy and matplotlib functions. While matplotlib is incredibly useful to visualize your results, it is not a prerequisite for the main functionality of PyExperimentSuite but simply a part of the following examples.

Listing 9: Example to retrieve the results stored in log files.

```
>>> from suite import MySuite
>>> from numpy import *
>>> from matplotlib import pyplot as plt
>>> mysuite = MySuite()
```

6.3 Retrieving Information about Experiments

The first data retrieval function described here is <code>get_exps()</code>. This function goes through all existing subdirectories and checks if they contain an <code>experiment.cfg</code> file, and are thus identified as experiments. It returns the full paths of all existing experiments (finished or not) under the current directory as a list.

The function has an optional parameter path, that specifies the directory of where to start with the search.

```
>>> mysuite.get_exps()
>>> mysuite.get_exps(path='.')
```

Another useful function is get_exp() (note the missing s).

```
>>> mysuite.get_exp(name, path='.')
```

This function takes the name of an experiment, as you put it in the config file section in square brackets, goes through all subdirectories and tries to locate the experiment. It then returns the full path of the found experiment. This function is useful if you know your experiment name but not the path anymore. It also has an optional parameter path to specify the directory of where to start the search.

The third function that gives you information about your experiments is get_params ().

```
>>> mysuite.get_params(exp)
```

This function takes the location of an experiment and returns all the parameters of the experiment, as listed in the experiment.cfg file. The return type is a dictionary.

6.4 Single Values

The most atomic data retrieval task is to access a single value of a single iteration. You could, for example, want to find the value of testerror after 25 iterations in the first repetition of experiment nn-learningrate. The function get_value() does exactly that.

```
>>> mysuite.get_value(exp, rep, tags, which)
```

The function takes four parameters: exp is the directory of the experiment you would like to access, rep specifies the repetition number. tags can be a string or a list of

strings. If it is a string, it needs to contain the key which you would like to access. This is the key you added to the dictionary in method iterate(). If you want to access several keys at once, pass a list of strings to the function, and if you want to retrieve all keys, simply pass the string 'all'. Finally, you need to specify, which value(s) you would like to access. The choices are the strings 'last', 'min', 'max' or an integer. 'last' will return the value from the very last iteration. 'min' and and 'max' return the minimal or maximal value over all iterations. If you pass an integer to the which argument, you will get the value at this specific iteration. For the afore-mentioned example, the call would be

If you requested only one key, the return value will be a scalar. If you requested several or all keys, you will get a dictionary with key-value pairs as a result.

6.5 Single Values from Parameter Range Experiments

Retrieving a single value from a sub-experiment created by parameter ranges works the same, except the experiment location is a bit longer. Use the full experiment path ./path/name/sub-exp to retrieve single values.

There is, however, a second function, that is made to work with parameter ranges, and it is particular useful with grid experiments. If you have tested a grid of several parameters, and you would like to retrieve values along one particular parameter axis and fixing the other ranges, the function <code>get_values_fix_params()</code> might be of use to you.

```
>>> mysuite.get_values_fix_params(exp, rep, tags, which, **kwargs)
```

It uses the same syntax as get_value() but has an additional **kwargs argument, which means you can pass an arbitrary number of additional keyword-value pairs in the function. The keywords expected here are the keys you would like to keep fixed to a certain value. Basically, for each keyword you add, you slice the (possibly multi-dimensional) grid along one axis and only consider experiments, that lie on that axis. Refer to the left side of Figure 2 again for visual support.

Let's have a look at experiment 3 from Listing 8, which defined two parameter ranges in a grid fashion:

```
alpha = arange(0.0, 1.0, 0.2)
```

```
beta = [0.5, 1.0, 2.0]
```

This definition will create 15 experiments (because alpha evaluates to [0.0, 0.2, 0.4, 0.6, 0.8]). If this was part of our Neural Network example, and we were looking for the smallest test error under the condition that alpha was 0.2, this is how we would call the function:

As you can see in the function call, this function actually returns a tuple of two different things: the actual values, and the experiment parameters. Both return values are lists of equal lengths. The first one contains results in the form that you would expect from get_value(): either scalars or dictionaries of key-value pairs. the second return value contains equally many dictionaries: each dictionary consists of all the parameters of the experiments that matched the criteria given by **kwarg.

You can add as few or many conditions to the function as you like, not just one. Each condition limits the returned experiments and values further. If you call the function without any conditions, it will go over all sub-experiments. As an example, if you evaluated several different parameter ranges in a grid and just want to find the overall lowest test error, call this function without any **kwargs arguments.

I recommend that you create a small toy grid experiment with PyExperimentSuite and call this function with different **kwargs parameters, to get a grasp of how this function exactly works.

6.6 Histories

Another common task with PyExperimentSuite is to retrieve a whole history over all iterations of a certain key, for example if you want to plot how the error of an optimization problem slowly decreased with each iteration. The function plot_history() can help you with this.

```
>>> mysuite.get_history(exp, rep, tags)
```

This function expects the location of your experiment, the index of the repetition you would like to access, and the keys you want to retrieve, either a single string, a list of strings, or the string 'all'.

The function returns a list of all values of that key, or a dictionary of lists with the corresponding keys.

In case you want to plot the history with the seperate module matplotlib, you could call:

6.7 Histories from Parameter Range Experiments

Just as with the <code>get_value()</code> method described above, histories can also be returned from parameter range experiments. If you know the exact sub-experiment already, you can pass the full path to the <code>get_history()</code> function and will receive the history.

If you want to retrieve several histories over a set of parameter range experiments, the function get_histories_fix_params() is what you need.

```
>>> mysuite.get_histories_fix_params(exp, rep, tags, **kwargs)
```

It works exactly like the single value version <code>get_values_fix_params()</code> but returns history lists instead of single values.

6.8 Aggregated Histories over Repetitions

There is another useful history-retrieving function, which is most useful if you executed many repetitions of the same experiment. One reason to do this, is because your experiment might be of stochastic nature and you want to average over all these repetitions.

```
>>> mysuite.get_histories_over_repetitions(exp, tags, aggregate)
```

This function requires the location of your experiment, the key(s) you are interested in (again as single string, list of string or the string 'all'), and an aggregation function. Typical aggregation functions could be sum, mean, max, etc. The function will not return the history of one single repetition, but use all repetitions and maps the values of each repetition to the aggregation function. Let's say the aggregation function is called aggr() and the value at repetition r and iteration i is defined as v_i^r . Repetitions range from i to i and iterations range from i to i. The result is then a new list, that

contains values with same iteration indices, to which aggr () has been applied:

$$[\mathsf{aggr}(v_0^0, v_0^1, \dots, v_0^R), \mathsf{aggr}(v_1^0, v_1^1, \dots, v_1^R), \dots, \mathsf{aggr}(v_I^0, v_I^1, \dots, v_I^R)]$$

One common use of this function is to average histories over all repetitions with the aggregation function mean (imported from numpy):

7 Examples and Best Practices

This section contains some more examples on how to use PyExperimentSuite and demonstrates some less common cases that you might be confronted with.

7.1 Strings in Iterable Objects

In Section 4 you learned about the configuration file and the syntax of defining parameters:

```
parameter = value
```

You also read that strings don't need to be surrounded by quotation marks, because everything that cannot be evaluated to a pythonic integer, float or object will be interpreted as strings. If you want to use parameter ranges over strings, however, the quotation marks become necessary again:

```
parameter = ['string1', 'string2', 'string3']
```

Why is this the case? PyExperimentSuite will again try to evaluate the value of the parameter—as a whole. It would stumble over the unquoted literals, assuming they were undefined variables, and conclude that it can't evaluate the list. The result would be a parameter that interprets the whole definition of the list as a single string, including the square brackets around it.

7.2 Passing Classes as Parameters

A common usecase of PyExperimentSuite is to test different algorithms or classes against each other. One possibility is of course to define a flag, that your reset () method uses to distinguish between the different program alternatives.

Let's say you have this line in your config file:

```
newalgorithm = [0, 1]
```

PyExperimentSuite will execute two experiments, once with newalgorithm set to 0 and once where it is set to 1. Your reset () method could the look like Listing 10.

Listing 10: Comparing different classes against one another with if-statement.

```
def reset(self, params, reps):
    if params['newalgorithm']:
        # initialize new algorithm
        self.algorithm = NewAlgorithmClass()
    else:
        # initialize old algorithm
        self.algorithm = OldAlgorithmClass()
```

Another, perhaps more elegant way is to use Python's eval() function⁴ to test different classes. Define the parameter in the config file as a list of strings (but do use quotation marks, as explained in Section 7.1):

```
algorithm = ['NewAlgorithmClass', 'OldAlgorithmClass']
```

Now you can write a much shorter reset () method (Listing 11). Note the additional + '()' in the eval() statement, which is necessary to create a new object from the class instead of assigning the class itself to self.algorithm. If you have additional parameters to pass to the algorithm class upon initialization, you can do that as well within the eval() statement and string concatenation.

Listing 11: Comparing different classes against one another with eval-statement.

```
def reset(self, params, reps):
    # convert string to class
    self.algorithm = eval(params['algorithm'] + '()')
```

⁴Although you often read that eval() is evil and that you shouldn't use it, I find it perfectly suited for this task and don't see the disadvantage of using it here. But the decision is up to you.

7.3 Passing Lists as Parameters

While the parameter range mechanism with iterable objects is convenient, sometimes you might want to pass an actual list as a single parameter without evaluating it as a range. You can tell PyExperimentSuite that your lists should not be transformed into parameter range experiments, by defining the parameter experiment in your config file and assigning it the string single.

```
experiment = 'single'
alpha = [0.5, 1.0, 1.5]
```

Now the list is assigned to the parameter as a whole and no sub-experiments are created. In the above case, you could access params['alpha'] within your reset () or iterate() methods and would get a list containing the three values 0.5, 1.0, 1.5.

It is currently not possible to mix grid, list and single experiments, i.e. one list will be returned as is where the other lists form a grid.

7.4 Debugging PyExperimentSuite Scripts

If you request more than one core to execute the experiments, PyExperimentSuite will use the multiprocessing package and create a worker pool for the required number of processes. If you specifically request to use only 1 core (with the runtime option -n 1), PyExperimentSuite will not go through the multiprocessing package but run the experiment directly from the main process. This is especially useful if you are debugging your script, because you will get the exact location in the code where an exception occured. Using multiple processes, this information is not available and you will have trouble finding any bugs.