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Sumatra: a toolkit for reproducible research

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1.1 Introduction

Lack of replicability in computational studies is, at base, a problem of shortcomings in record keeping. In laboratory-based experimental science, the tradition is to write down all experimental details in a paper notebook. This approach is no longer viable for many computational studies, as the number of details that could have an impact on the final result is so large. Automated or semi-automated tools for keeping track of all the experimental details – the scientist's own code, input and output data, supporting software, the computer hardware used, etc. – are therefore needed.

For the busy scientist, the time investment needed to learn to use these tools, or to

adapt their workflow so as to make use of them, may be one they are reluctant to make, especially since the problems of lack of reproducibility often take some time to manifest themselves. To achieve wide uptake among computational scientists, therefore, tools to support reproducible research should aim to minimise the effort required to learn, adopt and use them (see [1] for a more detailed version of this argument).

Sumatra is a software tool to support reproducible computational research, which aims to make reproducible computational science as easy to achieve (or easier) than nonreproducible research, largely by automating the process of capturing all the experimental details. In practice, this means that using Sumatra should require minimal changes to existing workflows and, given the wide diversity in workflows for computational science, Sumatra should be easy to adapt to different computational environments.

This chapter is intended for two groups of people:

- scientists who are interested in using Sumatra to track the details of their own research;
- developers who are interested in using Sumatra as a library in their own software for reproducible research.

The first section is an extended case study, illustrating how Sumatra may be of use in day-to-day research. This is followed by an in-depth explanation of Sumatra's architecture, including examples of how to use Sumatra as a Python library and how to extend and customize Sumatra.

1.2 Using Sumatra

We will illustrate one way to use Sumatra, and why you might want to use Sumatra, with a story about Alice and Bob. Bob is a graduate student in Alice's lab. When Alice was a graduate student herself, she kept track of the evolution of her code by giving each significant version a different file name, and she included the file name as a label in every figure she generated. Alice used to be quite confident she could, if it were ever necessary, go back and recreate the results from her earlier papers, since she has the original data carefully archived on CD-ROMs. However, after her recent experience with Charlie, she is not so sure. Charlie was a postdoc in Alice's lab, who got some great results, which they wrote up and submitted to a high-profile journal. The reviews were quite positive, but the reviewers asked for some new figures and a change to one of the existing figures. The problem was that when they tried to generate the modified figure, they could not get the results to match: the new graph looked significantly different, and no longer showed the effect they had found. Although Charlie had used the Subversion version control system for his code, he had not been so careful about keeping track of which version of the code had been used for each figure in the manuscript: several of the figures had originally been generated for a poster, and in the rush to get the poster finished in time to send to the printers, Charlie had not had time to keep such careful notes as usual, and had not always remembered to check-in changes in his code to the Subversion repository. Now Charlie has left science for a job with a major bank, and the manuscript is languishing in a drawer.

As a consequence of these experiences, Alice asked Bob, her new graduate student, to try out Sumatra. Sumatra automates the necessary, but tedious and error-prone process of keeping track of which code version was used to produce which output. Bob has his code in a Mercurial version control repository (for the purposes of this chapter, we will use a simplified version of Bob's code. If you would like to follow along, the repository is available

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at http://bitbucket.org/apdavison/ircr2013). Bob downloaded and installed Sumatra according to the instructions at http://neuralensemble.org/sumatra.

Bob normally runs his analysis (of scanning electron microscope images of glass samples) as follows:

```
$ python glass_sem_analysis.py MV_HFV_012.jpg
1699.875 65.0
```

This analyses the image specified on the command line, generates some further images, and prints out some statistics (see the SciPy tutorial at http://scipy-lectures.github.com/ for more details). The output images are saved to a specific subdirectory labelled according to the day on which the code is run, and the individual files are labelled with a timestamp, e.g. "Data/20121025/MV_HFV_012_163953_phases.png".

He creates a new Sumatra project in the same directory, using the smt command-line tool:

```
$ smt init ProjectGlass
$ smt configure -e python -m glass_sem_analysis.py -i . -d Data
```

This creates a new project, and sets "python" as the default executable to be used, "glass_sem_analysis.py" as the default script file, the current directory (".") as the place to look for input data, and a subdirectory "Data" as the place to start looking for output files. (If Bob could not remember the various options to the "smt configure" command, "smt help configure" would tell him).

"smt info" shows the current configuration of Bob's project. Note that it is using the already-existing Mercurial repository in his working directory:

```
$ smt info
Project name
                    : ProjectGlass
Default executable : Python (version: 2.6.7) at /usr/bin/python
                   : MercurialRepository at /home/bob/Projects/Glass
Default repository
Default main file
                    :
                      glass_sem_analysis.py
Default launch mode : serial
Data store (output) : ./Data
           (input)
                    :
Record store
                    : Django record store at
                      /home/bob/Projects/Glass/.smt/records
Code change policy
                    : error
Append label to
                    : None
```

Now to run the analysis using Sumatra:

\$ smt run MV_HFV_012.jpg
1699.875 65.0

. . . .

Since Bob has already specified the executable and script file, all he has to provide is the name of the input data file. The program runs as before and gives the same results, but in addition, Sumatra has captured a great deal of information about the *context* of the computation – exactly which version of the code was used, what the input and output data files were, what operating system and processor architecture were used, etc. Some of this information can be viewed in the console:

\$ smt list -1	
Label	: 20121025-170718
Timestamp	: 2012-10-25 17:07:18
Reason	:
Outcome	:
Duration	: 3.73256802559
Repository	: MercurialRepository at /home/bob/Projects/Glass
Main_File	: glass_sem_analysis.py
Version	: 9d24b099b5f3

Script_Arguments Executable		MV_HFV_012.jpg Python (version: 2.6.5) at /usr/bin/python
Parameters	:	
Input_Data	:	MV_HFV_012.jpg(5d789282b10a0da7a91560f33f8baf7272f7543d)
Launch_Mode	:	serial
Output_Data	:	20121025/MV_HFV_012_170722_phases.png(c9955f84ca3c1912
	:	20121025/MV_HFV_012_170722_sand.png(20bd5420d37ee589f3
	:	20121025/MV_HFV_012_170722_histogram.png(e7884dc5f3e9c
Tags	:	

but in general it is better to use the built-in web browser-based interface, launched with the smtweb command – see Figure 1.1.

Two things in particular should be noted from this figure. The first is that the versions of not only the Python interpreter and Bob's own code, but also the libraries on which Bob's code depends (NumPy, etc.), are captured. The second is that the path of each input and output data file is accompanied by a long hexadecimal string. This is the SHA1 digest, or hash, of the file contents (as used in crypographic applications, and also in version control systems such as Git and Mercurial). If the file contents are changed even slightly, the hash will change, which allows us to check for files being corrupted or accidentally over-written.

Now Bob would like to investigate how his image analysis method is affected by changing its parameters. He thinks this will be easier to keep track of if the parameters are separated out into a separate file, so he modifies his script and adds a new file default_parameters. The script now expects two arguments, first the parameter file, second the input data, and would normally be run using

```
$ python glass_sem_analysis.py default_parameters MV_HFV_012.jpg
```

but Bob wants to run it with Sumatra:

```
$ smt run default_parameters MV_HFV_012.jpg
Code has changed, please commit your changes.
```

Bob has forgotten to commit his changes to the version control repository. Sumatra detects this, and will then either refuse to run (the default, seen here) or will store the differences since the last commit. Bob commits and tries again.

```
$ hg commit -m 'Separated out parameters into separate file'
$ smt run -r 'test separate parameter file' default_parameters MV_HFV_012.jpg
1699.875 65.0
```

Note that he has also used the "-r" flag to note the reason for running this analysis, in case he forgets in future. Have Bob's modifications had any effect on his results? The output statistics are the same, and an inspection of the output data hashes in the web interface shows they have not changed either, so no, the results are unchanged.

We have seen already that Bob has less typing to do when running his analyses with Sumatra, as he has already specified the executable and script file as defaults. This is an example of how Sumatra tries to make it easier to use a tool for reproducible research than not to use one. Another example is the ability to specify parameters on the command line, rather than having to edit the parameter file each time:

So far, Bob has been using Charlie's old computer, running Ubuntu Linux 10.04. The next day, he is excited to find that the new computer Alice ordered for him has arrived. He installs Ubuntu 12.04, together with all the latest versions of the Python scientific libraries. He also copies over his glass analysis data, and migrates the Sumatra project. He tries

```
4
```

Label:	20121025-170718					
Reason:						
ccuson.		//				
Outcome:						
Timestamp:	25/10/2012 17:07:	18				
Duration:	3.73s					
Executable:	Python version 2.6.	5 (/usr/bin/pytho	n)			
Launch mode:	serial					
Repository:	/home/bob/Project	c/Class				
Main file:	glass_sem_analysis.					
	glass_sem_analysis. 9d24b099b5f3	РY				
Version:						
Arguments:	MV_HFV_012.jpg					
Tags:						
/ MV_HFV_012.jp	g 5d789282b10a0da7	a91560f33f8baf7	272f7543d 0 bytes			
		a91560f33f8baf7	272f7543d 0 bytes			
/ MV_HFV_012.jp Output		a91560f33f8baf7	272f7543d 0 bytes			
Output		a91560f33f8baf7	272f7543d 0 bytes			
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Output (/Data 20121025/MV_1 20121025/MV_2 20121025/MV_2 Parame Depend Name dateutil glib ggobject matplotlib mpl_toolkits numpy pytz scipy wx Platfor Name Ubuntu01	t files HFV_012_170722_pha HFV_012_170722_san HFV_012_170722_san HFV_012_170722_hist eters dencies Path /usr/lib/pyth	ses.png c0955 d.png 20bd5 ogram.png e7884 odules/python2.6, odules	f84ca3c19123d24c 420d37ee589f3c2 dc5f3e9ce2c0d713 dc5f3e9ce2c0d713 /dateutil /gtk-2.0/gdib gtk-2.0/gdib gtk-2.0/gobject /matplotlib /mpl_toolkits ges/numpy ges/yt2 ges/scipy ges/scipy ges/scipy a2bit ELF	cc4c87d197514d9 542a12438cb7866; cb3e9ddffb07bf20 inicode/wx System type Linux	9974b image/png 34.3 KB 10c image/png 27.7 KB Version 1.4.1 unknown 0.99.1.1 unknown 1.3.0 2010b 0.7.0 2.8.10.1 (gtk2-unicode) Release 2.6.32-24-generic	
Output /Data 20121025/WV_1 20121025/WV_2 20121025/WV_2 Parame Depend Name Jateutil Jib gobject matplotilb	t files HFV_012_170722_pha HFV_012_170722_san HFV_012_170722_hist eters dencies Path /usr/lib/pymc /usr/lib/pymc /usr/lib/pymc /usr/lib/pymc /usr/lib/pymc /usr/lib/pymc /usr/lib/pythc /usr/lib	ses.png c0955 d.png 20bd5 ogram.png e7884 odules/python2.6, odules	f84ca3c19123d24c 420d37ee589f3c2! dc5f3e9ce2c0d713 dc5f3e9ce2c0d713 //dateutil /gtk-2.0/gobject /matjotlib /mpl_toolikis ges/numpy ges/pytz ges/scipy ges/wx-2.8-gtk2-u Architecture 32bit ELF aobase.py:272: DC	cc4c87d197514d9 542a12438cb7866; cb3e9ddffb07bf20 inicode/wx System type Linux	9974b image/png 34.3 KB 10c image/png 27.7 KB Version 1.4.1 unknown 0.99.1.1 unknown 1.3.0 2010b 0.7.0 2.8.10.1 (gtk2-unicode) Release 2.6.32-24-generic	

FIGURE 1.1

Record of a computation captured with Sumatra, displayed in the web browser interface.

to run the analysis script, but gets an error: in the latest version of NumPy, the return format of the histogram() function has changed. This is straightforward to fix (see https: //bitbucket.org/apdavison/ircr2013/changeset/924a39a), so now Bob can commit and try again:

Has the upgrade affected Bob's results?

```
$ smt diff 20121025-172833 20121026-174545
Record 1
                         : 20121025-172833
Record 2
                         : 20121026-174545
Executable differs
                         : no
Code differs
                         : yes
  Repository differs
                         : no
  Main file differs
                         : no
  Version differs
                         : yes
  Non checked-in code
                         : no
  Dependencies differ
                         : yes
Launch mode differs
                         : no
Input data differ
                         : no
Script arguments differ : no
Parameters differ
                         : no
Data differ
                         : yes
```

OK, Bob knew he had changed the code because of the new histogram() function, and he knew the dependencies had changed, because of the operating system upgrade, but it was a bit disappointing to see the output data are different. Using the web browser, we can look at the results from the two simulations (one from Ubuntu 10.04, one from Ubuntu 12.04) side by side (Figure 1.2) – visually there is no difference, just a tiny change in the margins, probably due to the upgraded matplotlib package.

Alice puts her head round the door to ask how Bob is getting on with Sumatra. So far, Bob is happy. His productive workflow has hardly changed – in fact he has a little bit less to type, since Sumatra stores the names of the default executable and default script for him, and he can modify parameters quickly on the command line rather than having to open up the parameter file in his editor. The web browser interface lets him quickly browse and search through his results (Figure 1.3), and compare different runs side-by-side. And he feels much more confident that he will be able to replicate his results in the future.

Alice tries Sumatra out for herself the following week. Alice wants to use one of Bob's figures in a grant application, but Bob is on vacation, and she wants to make a few small changes to the figure. She copies Bob's Sumatra record store (which by default was created as the file .smt/records in a subdirectory of Bob's working directory) to the lab network file server, so that she can access Bob's records and Bob in turn will be able to see her results when he returns, and sets up a new project on her MacBook:

\$ smt init -s /Volumes/shared/glass/smt_records ProjectGlass \$ smt configure -e python -m glass_sem_analysis.py -i . -d Data

Before starting her own modifications, she re-runs Bob's last analysis:

```
$ smt repeat 20121026-174545
The new record does not match the original. It differs as follows.
Record 1
                        : 20121026-174545
Record 2
                         : 20121026-174545_repeat
Executable differs
                         : no
Code differs
                         : yes
  Repository differs
                         : no
  Main file differs
                         : no
  Version differs
                         : no
  Non checked-in code
                         : no
```

 $\mathbf{6}$

	121025-172833					
	20121025-17	72833		20121026-12	74545	
Reason:	Parameters an	e now in a separate file		Fixed to work	with the new numpy.histogram() function	
Outcome:						
imestamp:	25/10/2012 1	17:28:33		26/10/2012	17:45:45	
uration:	3.85s			3.85s		
xecutable:		n 2.6.5 (/usr/bin/python)			n 2.7.3 (/usr/bin/python)	
aunch mode:	serial	12.0.5 (/usi/bii/pytioli)		serial	2.7.5 (/us//bii//pytion/	
epository:	/home/bob/P	have in the following of the second sec		/home/bob/P	trainets (Class	
lain file:						
ersion:	glass_sem_ana 432ff7ef3f45	alysis.py		glass_sem_an 924a39a0d24		
ersion:	432tt/et3t45			924a39a0d24	c	
	Digest: c9955	184cu3c19123d24ccc4<87d197514d9e01e	3.0 2.8 2.6 2.4 2.2 2.0 1.8 1.6 1.4 1.2 1.0	Digest: 7f8ed	DcGef97bB317af8e2d9ab9f856a193c2687	3.0 2.8 2.6 2.4 2.2 2.0 1.8 1.6 1.4 1.2 1.0
ependencies:	Marrie	Park	Manadan		Prot	Manatan
	Name dateutil	Path /usr/lib/pymodules/python2.6/dateutil	Version 1.4.1	Name PIL	Path /usr/lib/python2.7/dist-packages/PIL	Version unknown
		/usr/lib/pymodules/python2.6/gtk-		PIL PyQt4	/usr/lib/python2.7/dist-packages/PiL /usr/lib/python2.7/dist-packages/PyQt4	unknown
	glib	2.0/glib	unknown	apport	/usr/lib/python2.7/dist-packages/apport	unknown
	gobject	/usr/lib/pymodules/python2.6/gtk- 2.0/gobject	unknown	apt	/usr/lib/python2.7/dist-packages/apport	unknown
				dateutil	/usr/lib/python2.7/dist-packages/dateutil	1.5
	matnlotlib		0 99 1 1	dateutii		1.5
	matplotlib mpl_toolkits	/usr/lib/pymodules/python2.6/matplotlib	0.99.1.1	glib	/usr/lib/python2.7/dist-packages/glib	unknown
	mpl_toolkits	/usr/lib/pymodules/python2.6/matplotlib /usr/lib/pymodules/python2.6/mpl_toolkits	unknown			
	mpl_toolkits numpy	/usr/lib/pymodules/python2.6/matplotlib /usr/lib/pymodules/python2.6/mpl_toolkits /usr/lib/python2.6/dist-packages/numpy	unknown 1.3.0	glib	/usr/lib/python2.7/dist-packages/glib	unknown
	mpl_toolkits numpy pytz	/usr/lib/pymodules/python2.6/matplotlib /usr/lib/pymodules/python2.6/mpl_toolkits /usr/lib/python2.6/dist-packages/numpy /usr/lib/python2.6/dist-packages/pytz	unknown 1.3.0 2010b	glib gobject	/usr/lib/python2.7/dist-packages/glib /usr/lib/python2.7/dist-packages/gobject /usr/lib/pymodules/python2.7/matplotlib	unknown unknown 1.1.1rc
	mpl_toolkits numpy	/usr/lib/pymodules/python2.6/matplotlib /usr/lib/pymodules/python2.6/mpl_toolkits /usr/lib/python2.6/dist-packages/numpy /usr/lib/python2.6/dist-packages/pytz /usr/lib/python2.6/dist-packages/scipy	unknown 1.3.0 2010b 0.7.0	glib gobject matplotlib	/usr/lib/python2.7/dist-packages/glib /usr/lib/python2.7/dist-packages/gobject /usr/lib/pymodules/python2.7/matplotlib	unknown unknown 1.1.1rc
	mpl_toolkits numpy pytz	/usr/lib/pymodules/python2.6/matplotlib /usr/lib/python2.6/mpl_toolkits /usr/lib/python2.6/dist-packages/numpy /usr/lib/python2.6/dist-packages/pytz /usr/lib/python2.6/dist-packages/scipy /usr/lib/python2.6/dist-packages/wx-2.8-	unknown 1.3.0 2010b 0.7.0 2.8.10.1 (gtk2-	glib gobject matplotlib mpl_toolkits	/usr/lib/python2.7/dist-packages/glib /usr/lib/python2.7/dist-packages/gobject /usr/lib/pymodules/python2.7/matplotlib /usr/lib/pymodules/python2.7/mpl_toolkits	unknown unknown 1.1.1rc unknown
	mpl_toolkits numpy pytz scipy	/usr/lib/pymodules/python2.6/matplotlib /usr/lib/pymodules/python2.6/mpl_toolkits /usr/lib/python2.6/dist-packages/numpy /usr/lib/python2.6/dist-packages/pytz /usr/lib/python2.6/dist-packages/scipy	unknown 1.3.0 2010b 0.7.0 2.8.10.1	glib gobject matplotlib mpl_toolkits nose numpy	/usr/lib/python2.7/dist-packages/glib /usr/lib/python2.7/dist-packages/gobject /usr/lib/pymodules/python2.7/matplotlib /usr/lib/python2.7/dist-packages/nose /usr/lib/python2.7/dist-packages/numpy	unknown unknown 1.1.1rc unknown 1.1.2
	mpl_toolkits numpy pytz scipy	/usr/lib/pymodules/python2.6/matplotlib /usr/lib/python2.6/mpl_toolkits /usr/lib/python2.6/dist-packages/numpy /usr/lib/python2.6/dist-packages/pytz /usr/lib/python2.6/dist-packages/scipy /usr/lib/python2.6/dist-packages/wx-2.8-	unknown 1.3.0 2010b 0.7.0 2.8.10.1 (gtk2-	glib gobject matplotlib mpl_toolkits nose numpy pytz	/usr/lib/python2.7/dist-packages/glib /usr/lib/python2.7/dist-packages/gobject /usr/lib/python2.7/matplotlib /usr/lib/python2.7/dist-packages/nose /usr/lib/python2.7/dist-packages/numpy /usr/lib/python2.7/dist-packages/pytz	unknown unknown 1.1.1rc unknown 1.1.2 1.6.1 2011k
	mpl_toolkits numpy pytz scipy	/usr/lib/pymodules/python2.6/matplotlib /usr/lib/python2.6/mpl_toolkits /usr/lib/python2.6/dist-packages/numpy /usr/lib/python2.6/dist-packages/pytz /usr/lib/python2.6/dist-packages/scipy /usr/lib/python2.6/dist-packages/wx-2.8-	unknown 1.3.0 2010b 0.7.0 2.8.10.1 (gtk2-	glib gobject matplotlib mpl_toolkits nose numpy pytz scipy	/usr/lib/python2.7/dist-packages/glib /usr/lib/python2.7/dist-packages/gobject /usr/lib/python2.7/dist-packages/nose/ /usr/lib/python2.7/dist-packages/nose /usr/lib/python2.7/dist-packages/nose /usr/lib/python2.7/dist-packages/pytz /usr/lib/python2.7/dist-packages/scipy /usr/lib/python2.7/dist-packages/scipy /usr/lib/python2.7/dist-packages/scipy	unknown unknown 1.1.1rc unknown 1.1.2 1.6.1 2011k 0.9.0
	mpl_toolkits numpy pytz scipy	/usr/lib/pymodules/python2.6/matplotlib /usr/lib/python2.6/mpl_toolkits /usr/lib/python2.6/dist-packages/numpy /usr/lib/python2.6/dist-packages/pytz /usr/lib/python2.6/dist-packages/scipy /usr/lib/python2.6/dist-packages/wx-2.8-	unknown 1.3.0 2010b 0.7.0 2.8.10.1 (gtk2-	glib gobject matplotlib mpl_toolkits nose numpy pytz	/usr/lib/python2.7/dist-packages/glib /usr/lib/python2.7/dist-packages/gobject /usr/lib/python2.7/matplotlib /usr/lib/python2.7/dist-packages/nose /usr/lib/python2.7/dist-packages/numpy /usr/lib/python2.7/dist-packages/pytz	unknown unknown 1.1.1rc unknown 1.1.2 1.6.1 2011k 0.9.0 0.6
	mpl_toolkits numpy pytz scipy	/usr/lib/pymodules/python2.6/matplotlib /usr/lib/python2.6/mpl_toolkits /usr/lib/python2.6/dist-packages/numpy /usr/lib/python2.6/dist-packages/pytz /usr/lib/python2.6/dist-packages/scipy /usr/lib/python2.6/dist-packages/wx-2.8-	unknown 1.3.0 2010b 0.7.0 2.8.10.1 (gtk2-	glib gobject matplotlib mpl_toolkits nose numpy pytz scipy	/usr/lib/python2.7/dist-packages/glib /usr/lib/python2.7/dist-packages/gobject /usr/lib/python2.7/dist-packages/nose/ /usr/lib/python2.7/dist-packages/nose /usr/lib/python2.7/dist-packages/nose /usr/lib/python2.7/dist-packages/pytz /usr/lib/python2.7/dist-packages/scipy /usr/lib/python2.7/dist-packages/scipy /usr/lib/python2.7/dist-packages/scipy	unknown unknown 1.1.1rc unknown 1.1.2 1.6.1 2011k 0.9.0

FIGURE 1.2

Excerpts from a side-by-side comparison of two computation records, one run on Ubuntu 10.04, the other on Ubuntu 12.04.

Dependencies differ	:	yes
Launch mode differs	:	no
Input data differ	:	no
Script arguments differ	:	no
Parameters differ	:	no
Data differ	:	no

She has slightly different versions of the dependencies on her MacBook, but the results are unchanged. Alice can now proceed to reformat the figures, confident that her computing environment is consistent with that of her graduate student. Since the grant application is being written in LATEX, Alice can also use the sumatra LATEX package to automatically pull images from the Sumatra record store into her document, with automatic cross-checking of SHA1 hashes to ensure the image is indeed the correct one and has not been accidentally over-written.

Dummy	titl	e
-------	------	---

🖶 Projec	ctGlass 🐚 – Searc	h				~ Q	Search					
New record ProjectGlass										1 - 8 of 8		
Repository	Label	Tag	Reason	Outcome	Duration	Date	Time	Executable name	Executable version	Main file	Version	Arguments
Glass	20121026-174545	Precise, default,	Fixed to work with the new numpy.histogram(function		3.85s	26/10/2012	17:45:45	Python	2.6.5	glass_sem_analysis.py	924a3	<parameters> MV_HFV_012.jpg</parameters>
Glass	20121026-172642	error,	Running on Ubuntu 12.04	TypeError: histogram() got an unexpected keyword argument 'new'		26/10/2012	17:26:42	Python	2.6.5	glass_sem_analysis.py	432ff	<parameters> MV_HFV_012.jpg</parameters>
Glass	20121025-173606	Lucid,	No filtering, but more cleaning		2.82s	25/10/2012	17:36:06	Python	2.6.5	glass_sem_analysis.py	432ff	<parameters> MV_HFV_012.jpg</parameters>
Glass	20121025-173350	Lucid,	Trying a different colourmap ("hot")	"Copper" is nicer	4.08s	25/10/2012	17:33:50	Python	2.6.5	glass_sem_analysis.py	432ff	<parameters> MV_HFV_012.jpg</parameters>
Glass	20121025-173036	Lucid,	No filtering		2.83s	25/10/2012	17:30:36	Python	2.6.5	glass_sem_analysis.py	432ff	<pre><parameters> MV_HFV_012.jpg</parameters></pre>
Glass	20121025-172833	Lucid, default,	Parameters are now in a separate file		3.85s	25/10/2012	17:28:33	Python	2.6.5	glass_sem_analysis.py	432ff	<pre><parameters> MV_HFV_012.jpg</parameters></pre>
Glass	20121025-170718	Lucid, default,	Image file now specified on command line		3.73s	25/10/2012	17:07:18	Python	2.6.5	glass_sem_analysis.py	9d24b	MV_HFV_012.jpg
Glass	20121025-163949	Lucid, default,	First run with Sumatra		3.76s	25/10/2012	16:39:49	Python	2.6.5	glass_sem_analysis.py	89af3	

FIGURE 1.3

List of computation records in the Sumatra web browser interface.

In conclusion, we hope to have demonstrated that by using Sumatra, Alice and Bob have improved the reproducibility of their computational experiments, enhanced communication within their lab, and increased the manageability of their projects, with minimal effort and minimal change to their existing workflow.

1.3 Design criteria

In introducing the architecture of Sumatra so that others can build upon and extend it, we begin by describing the constraints we wish Sumatra to satisfy, before describing, in the following section, its current architecture.

The design of Sumatra is driven by two principles:

- 1. there is a huge diversity in computational science workflows;
- software to assist reproducibility must be very easy to use, or only the very conscientious will use it.

To elaborate on the first issue, of workflow diversity, different scientists may launch computations from the command-line, in interactive notebooks, in graphical interfaces, in web-based tools. Computations may be launched serially, as batch jobs, as distributed computations, for immediate execution or queued for deferred execution, on local machines, small clusters, supercomputers, grids, or in the cloud. Projects may be solo or collaborative efforts. Different workflows may be used for different components of a project or during different phases of a project (e.g. exploration vs preparation of final published figures).

Given this diversity, it is unlikely there is a single software tool to support reproducible research which will be optimal for all possible workflows. At the same time, there is a considerable amount of functionality that is required whatever the workflow, e.g. unambiguous identification of exactly which code has been run. Sumatra is therefore designed as a core library of loosely-coupled components for common functionality, easily extensible and customizable, so people can adapt Sumatra to their own use cases, and so other people can built other tools on top of Sumatra.

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Such a library is potentially useful to tool developers, but will not on its own promote reproducibility: it must be integrated into scientists' existing workflows, so that the barrier to adoption is as low as possible. Sumatra also, therefore, provides tools, built on top of the core library, that wrap around or work alongside widely-used types of workflow. Three such tools are available at the time of writing: smt, which supports workflows built around running individual computations on the command line; smtweb, which provides a browser-based tool for browsing and querying the results of previous computations; and a LaTeX package which allows the automated inclusion of figures generated by a Sumatra-tracked computation in documents, with hyperlinks to the provenance information. The use of these tools was demonstrated in the previous section. In the future, further tools may be developed to support more interactive workflows.

Given the above constraints, Sumatra must enable a scientist to easily respond to the following questions:

- what code was run?
 - which executable?
 - * name, location, version, compilation options
 - which script?
 - * name, location, version
 - * options, parameters
 - * dependencies (name, location, version)
- what were the input data?
 - name, location, content
- what were the outputs?
 - data, logs, stdout/stderr
- who launched the computation?
- when was it launched/when did it run? (queueing systems)
- where did it run?
 - machine name(s), other identifiers (e.g. IP addresses)
 - processor architecture
 - available memory
 - operating system
- why was it run?
- what was the outcome? (interpreted in terms of the ongoing project)
- which project was it part of?

1.4 Architecture

This section gives an overview of Sumatra's architecture, intended for readers who may be interested in extending or building upon Sumatra, or applying some of its methods in their own approaches to replicability. More fine-grained detail is available in the online documentation at http://neuralensemble.org/sumatra. Sumatra has a modular design, with the coupling between modules made as loose as possible. Within modules, a common motif to provide flexibility and configurability is to use abstract base classes to define a common interface, which are then subclassed to provide different implementations of a given type of functionality (e.g. version control, data storage). The principal classes in the core Sumatra library, and their composition, are shown in Figure 1.4. More detail about the individual modules, classes and their interactions is given in the following sections.

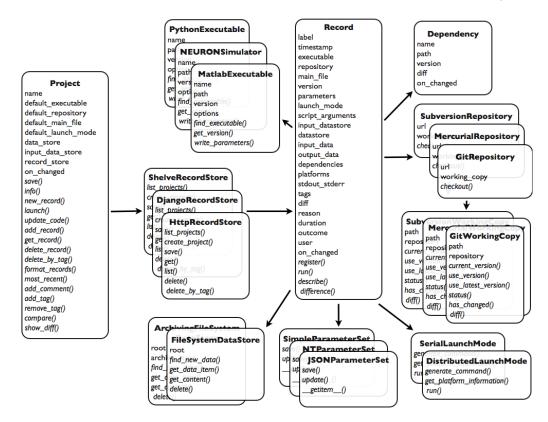


FIGURE 1.4

The principal classes in Sumatra with their attributes and methods. The arrows denote the relationship "contains an instance or instances of". Not shown, for reasons of space, are the classes DataKey, DataItem, RecordDifference, PlatformInformation, Formatter. Not all sub-classes are shown.

1.4.1 Code versioning and dependency tracking

To ensure replication, we need to capture identifying information about all of the code that was run. Where code is modular, this means capturing the local file system path of each library/module/package that is included/imported by the "main" file (its "dependencies"), together with, if possible, the version of the module, so that (i) the environment could be recreated in future, (ii) if failing to replicate with more up-to-date versions of libraries in future, we can investigate what has changed. This must be done recursively, of course, if a dependency itself has dependencies.

Finding the dependencies requires, in general, being able to parse the programming language used (although in future it may be possible to use a tool such as CDE [5, 6] to determine which dependencies are loaded at run-time). Sumatra therefore requires a "dependency finder" module to be provided for each programming language used. At the time of writing, such modules are all distributed within Sumatra, i.e. as modules dependency_finder.python, dependency_finder.matlab, etc., but a plugin architecture is planned, so that users can easily extend Sumatra where the language they are using is not supported.

Version information may be provided in many ways, some of which are dependent on the programming language used, others independent. As an example of the former, Python modules often define a variable called <u>__version__</u>, VERSION or version, or a function called

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get_version(). Two examples of the latter are: obtaining the version from a VCS; obtaining the version from a package management system (such as apt, on Debian). Sumatra's strategy, therefore, is that each dependency_finder module provide a list of functions, each implementing one heuristic for finding versions, e.g. find_versions_by_attribute(), or find_versions_by_version_control(). Each of these is tried in turn, and the first version found is the one used (the order is important: generally a version obtained from a VCS is more reliable/precise than a version obtained from a variable defined within the code).

It may happen that some of the code under version control has been modified since the last commit. In this scenario, it is usually best to abort the computation and to commit the changes before proceeding. However, there may be good reasons for not wanting to commit, and so Sumatra also provides the option of storing the "diff" between the VCS working copy and the last commit.

Given the variety of VCSs in use, Sumatra's strategy is to wrap each VCS so as to provide a uniform interface. For each VCS supported by Sumatra, the versioncontrol module contains a sub-module containing two classes - a subclass of versioncontrol.base.Repository and a subclass of versioncontrol.base.WorkingCopy. Sumatra does not require all the functionality of VCSs, and is not intended to replace the normal methods of interacting with a VCS for code development. The Repository subclass has two roles: storing the repository URL, and obtaining a fresh checkout/clone of the code base from a remote server (even the latter is not strictly necessary). The functionality required of the WorkingCopy subclass is more extensive: determine the current version; determine whether any of the code has been modified; determine the diff between the working code and the last commit; determine whether a given file is under version control; change the working copy to an older or newer version (for replicating previous computations and then returning to the most recent state of the code base).

In general, the difference between distributed and centralized version control systems is not important for Sumatra. The only difference is that, for distributed VCSs, the repository used is always a local one, and it is therefore often useful, for the purposes of future replication and open science, to store the URL of the "upstream" repository, often a public repository on a remote server.

1.4.2 Data handling

Replicability of a computational result requires knowing what the input data (if any) were, and it requires storing the output data so that future replication attempts can be checked against the original results. Inputs to a program can be subdivided into data, and configuration/parameters. These can generally be distinguished in that data could be processed by a different program, while parameters are tightly tied to the code. Sumatra attempts to distinguish parameter/configuration files from input data files by the structure of the data; as a fall-back, parameters will be treated as input data. Parameter file handling is described below.

Data may be stored in many ways: in individual files on a local or remote file system, in a relational database, in a remote resource accessed over the internet by some API. However it is stored, the most important thing to know about data is its content. However, it would be redundant for Sumatra to store a separate copy of each input and output data item, especially given the potentially enormous size of data items in many scientific disciplines. Sumatra therefore stores an identifier for each data item, which enables retrieval of the item from whichever data store – the file system, a relational database, etc. – is used. In the case of the file system, for example, the identifier consists of the file system path relative to a user-defined root directory together with the SHA1 hash of the file contents. The latter is needed to catch overwriting or corruption of files. To handle different ways of storing data, Sumatra defines an abstract DataStore class, which is then subclassed: for example, the FileSystemDataStore which is used to work with data stored on a local file system. The minimal functionality required of a DataStore subclass is: find new content, given a time stamp (used to link output data to a given computation); return a data item object, given the item's identifier ("key"); return the contents of a data item; delete a data item. DataItem objects support obtaining the data contents, and may also contain additional metadata, such as the mimetype.

It is straightforward to add extra functionality to a DataStore subclass. For example, the ArchivingFileSystemDataStore works the same as the plain FileSystemDataStore but in addition copies all the output data files to an archive format. The MirroredFileSystemDataStore allows specifying a URL from which the data file can be retrieved (in addition to the local version). This supports, for example, using Dropbox (https://www.dropbox.com) with a public folder, or FTP, or FigShare (http://figshare.com) to make your data available online.

1.4.3 Storing provenance information

Once Sumatra has captured the context of your computational experiment, it needs to store all this information somewhere. For individual projects, a local database is probably the best way to do this. For collaborative projects, or if you often work while travelling, it may be necessary for this information to be stored in a remote database accessible over the internet. To provide this flexibility, Sumatra defines an abstract RecordStore class, which is then subclassed.

Sumatra currently provides three RecordStore subclasses: ShelveRecordStore, which provides only basic functionality, but has the advantage of requiring no external libraries to be installed; DjangoRecordStore, which uses the Django web framework to store the provenance information in a relational database (SQLite by default, but MySQL, PostgreSQL and others are also supported) and adds the ability to browse the record store using a web browser; and HttpRecordStore, which is a client for storing provenance information in a remote database accessed over HTTP using JSON as the transport format. The server for the HttpRecordStore is not distributed with Sumatra, but such a server is straightforward to implement. Two implementations currently exist – a Django-based implementation at https://bitbucket.org/apdavison/sumatra_server and a MongoDB-based version at https://github.com/btel/Sumatra-MongoDB.

The functionality required of a RecordStore subclass is: support multiple Sumatra projects; list all projects contained in the store; save a Sumatra Record object under a given project; list all the records in a project; retrieve a Record given its identifier (project+label); delete a Record given its identifier; delete all Records which have a given tag; return the most recent record; export a record in JSON format; import a record in the same format; synchronize with another record store so that they both contain the same records for a given project.

1.4.4 Parameter handling

It is a common practice in scientific computing to run a simulation or analysis with different parameters and to compare the results. Given this important use case, Sumatra allows parameters to be handled differently from other input data. If Sumatra is able to recognize a particular parameter file format then (i) the parameters are available for future searching/querying/comparison; (ii) Sumatra can add extra parameters. An important use case of the latter is that Sumatra can add the label/identifier for the current record, for use by the user's code in constructing file names, etc. Sumatra currently supports four parameter

file formats, including simple "key=value" files, JSON, and config/ini-style formats. Implementing support for a new parameter file format is straightforward: define a MyParameterSet class whose constructor accepts either a filename or a text string containing the file contents. The class should also implement method as_dict() which returns parameter names and values in a (possibly nested) Python dict, update(), which functions like dict.update(), and save(), which writes the parameter set to file in the given format.

1.4.5 Launching computations

If your code is written in Python, then you can use Sumatra directly within your scripts, and run your computation with Python as usual. If you are using other tools (or if using Python and you do not want to modify your code) then Sumatra needs to launch your computation in order to be able to capture the context. The challenge here is that there are so many different workflows, so many different ways of launching a computation: from the command line on the local machine, from the command line on a remote machine (using ssh, for example), on a cluster, computing grid or supercomputer using a job manager, as a parallel computation using MPI, or by clicking a button in a graphical interface

To handle this variety, Sumatra follows the usual pattern of defining an abstract base class, LaunchMode, which is then subclassed to support different methods of launching computations. A LaunchMode subclass needs to define a method generate_command() which should return a string which will be executed on the command line. The LaunchMode is also responsible for capturing information about the platform – the operating system, the processor architecture, etc. For computations run on the local machine, the base class takes care of this. For computations run on a remote machine or machines, the LaunchModel subclass must override the get_platform_information() method. Sumatra currently provides SerialLaunchMode and DistributedLaunchMode subclasses.

To generate the launch command, Sumatra may need extra information about the particular executable being used – particular arguments or flags that are needed in different circumstances. Similarly, there may be a build step or other preliminary that is needed before launching the computation. If this is the case, a user may define an Executable subclass which may define any of the attributes pre_run, mpi_options, requires_script, and may optionally redefine the method _get_version(). The user then calls the programs.register_executable() method to register the new subclass with Sumatra.

1.4.6 Putting it all together

Tying all of the foregoing together are the Record class and the Project class. The Record class has two main roles: gathering provenance information when running a computation, and acting as a container for provenance information. When launching a new computation, as diagrammed in Figure 1.5, a new Record object stores the identifiers of any input data, interacts with a WorkingCopy object to check that the code is at the requested version, uses the dependency_finder module to find the list of dependencies (and their versions), and then obtains platform information from the appropriate LaunchMode. It then runs any precursor tasks, such as building the executable, writes a modified parameter file, if necessary, and then passes control to the LaunchMode, which spawns a new process in which it runs the requested computation while capturing the standard output and standard error streams. Once this completes, the Record object calculates the time taken, stores stdout and stderr, asks the DataStore object to find any new data generated by the computation, and stores the identifiers of this output data.

The Project class has one main role: to simplify use of the Sumatra API by storing

Dummy title

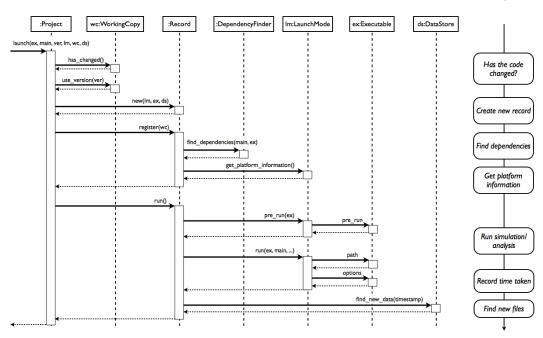


FIGURE 1.5

The flow of control between different Sumatra objects during a computation. Time flows from top to bottom. Each dashed vertical line represents the life time of an object, labelled at the top with the class and, in some cases, an instance name. Solid horizontal arrows represent method calls or attribute access.

default values and providing shortcut functions for frequently-performed tasks. Thus, for example, while creating a new Record object requires passing up to 16 arguments, the Project.new_record() method will often be called with just two - the parameter set and the list of input data items - since most of the others take default values stored by the Project. The smt command accesses Sumatra's functionality almost entirely through an instance of the Project class.

The precise division of responsibilities between the Record and Project class is not critical, and could evolve in future versions of Sumatra to enhance usability of the API.

1.4.7 Search/query/reuse

So far we have talked about the API from the perspective of capturing provenance information. We now consider the use cases of accessing, querying and using the stored provenance information.

As described above, this information is stored in a "record store", represented by a subclass of **RecordStore**, and whose backend may be a flat file, relational database, or web service. The common record store interface allows querying based on record identifiers (project + label) and on tags. Individual record store implementations may allow more sophisticated queries: for example, the DjangoRecordStore allows queries based on Django's object-relational-mapper, or even using plain SQL.

The main use cases for accessing records of previous computations: (i) are comparing the results of similar runs (e.g. examining the effects of parameter changes); (ii) repeating

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a previous computation to check that the results are reproducible; (iii) further processing of results, e.g. further analyses, visualization, inclusion in manuscripts.

The first two of these use cases are supported by the Project.compare() method, which calls Record.difference(), which returns an instance of the RecordDifference class. This class has assorted methods which allow a precise dissection of the differences between two computations.

1.5 Discussion

In this chapter we have presented Sumatra for two (albeit overlapping) audiences: the working computational scientist, and the software developer or scientist-developer who may wish to extend or build-upon Sumatra. In this book as a whole, a number of different tools to support reproducible research have been presented. For a scientist interested in ensuring their research is easily reproducible, when should you use Sumatra and when another tool?

Software for reproducible research can be divided into three general categories: tools for literate programming, workflow management systems, and tools for environment capture.

Literate programming¹ and the closely related "interactive notebook" approach² inextricably bind together code and the results generated by that code, which is clearly hugely beneficial for reproducible research. With some such systems, information about software versions, input data and the computing environment can also be included in the final document. If your literate programming environment or interactive notebook supports Python, you could also use Sumatra via its API to provide this functionality. Scenarios that are generally more difficult to handle with the current generation of literate programming tools and interactive notebooks are: (i) where computations take considerable time (hours or days) to run; (ii) where computations are distributed on parallel hardware or are queued for later execution; (iii) where code is split among many modules, so that the code included in the literate document or notebook is only a small part of the whole.

Visual workflow management or pipeline tools, such as Kepler [8], Taverna [10] and VisTrails [3, 4] are aimed at scientists with limited coding experience, or who prefer visual programming environments. They are particularly straightforward to use in domains where there some standardization of data formats and analysis methods – for example in bioinformatics and in fields that make extensive use of image processing. The main disadvantage is that where there are no pre-existing components for a given need, creating a new component can require considerable effort and a detailed knowledge of the workflow system architecture. Most widely-used systems include provenance tracking either as an integral part or as an optional module.

Environment capture systems, such as Sumatra, are generally the easiest to adopt for an existing workflow. The simplest approach is to capture the entire operating system as a virtual machine (VM) image – see the chapter by Howe [7] in the current volume. A more lightweight alternative to this is CDE[5, 6], which archives only those executables and libraries actually used by the computation. The main disadvantages with such approaches are: (i) your results risk being highly sensitive to the particular configuration of your computer; (ii) it is difficult or impossible to index, search or analyse the provenance information. Sumatra aims to overcome both of these disadvantages by capturing the information needed

¹see for example ref [2], which explains the use of Sweave (http://www.statistik.lmu.de/~leisch/ Sweave/) and Org-mode (http://orgmode.org) for reproducible research, and ref[11] in the current volume.

²for example Mathematica (http://www.wolfram.com/mathematica/), Sage (http://www.sagemath.org) and IPython (http://ipython.org)

to recreate the experimental context, rather than the context itself in binary form. Some combination of Sumatra and CDE would perhaps give the best of both worlds. Integration of CDE is planned in a future version of Sumatra.

In summary, at the time of writing, Sumatra is most suitable for scientists who prefer to write their own code and run it from the command line, especially when factors such as computation time, parallelism, or remote execution make it difficult to work interactively, or where code is highly modular so that literate programming tools capture only the tip of the code iceberg. In any case, Sumatra is fast to set up, easy to use and requires no changes to existing code, so there is little to be lost in trying it out.

We have seen that Sumatra makes it much easier to replicate computational research, in capturing the details of the software and hardware environment that was used. In particular, Sumatra makes it much easier to identify, in the case of failure to reproduce a result, what are the differences between the original and current environments. However, Sumatra cannot guarantee reproducibility, for two reasons. First, there are some details that are not captured. For example, in Figure 1.1 you can see that for some of the dependencies the version is unknown, either because the version information is genuinely not present or because Sumatra does not yet have a heuristic for finding it. Similarly, the compilation procedure and software library versions used to compile third-party programs, such as the Python interpreter, are not currently captured, and it may sometimes be impossible to capture this information. Second, with the passage of time, even if you know the particular versions of the libraries used, these versions may no longer be available, or the particular hardware architecture needed may not even be available. This problem is not restricted to Sumatra, of course. The use of virtual machines and careful archiving of old hardware is one partial solution, while for code that continues to be useful, a program of maintenance and ongoing updates can avoid obsolescence.

In the future, we plan to add support for using Sumatra with interactive notebooks (i.e. supporting a more granular unit of computation than an entire script), automated re-creation of software environments using the captured information, support for pipelines (where the output in one Sumatra record is the input in another), better support for compiled languages and software build systems, and interoperability with other provenance tracking tools, probably using the Open Provenance Model [9].

Sumatra is open source software, and is developed as an open community – if you have ideas or wish to contribute in any way, please join us at http://neuralensemble.org/sumatra.

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