

Manual for Queensland University of Technology (QUT) High Performance Computer (HPC)

Prepared for the Department of Health Western Australia (DOHWA)

James Hogg

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This document stands as an introduction for staff at the Department of Health Western Australia to successfully operate the Queensland University of Technologies (QUT) High Performance Computing (HPC) system. This manual is relatively specific to the Bayesian modelling project. For a more general manual we strongly recommend the HPC guide written by Ethan Goan (a current QUT PhD student). His guide is available in the `resources\Other_resources` folder in this repository.

The QUT HPC facilities provide access to a large amount of high quality computational resources (listed below). These resources are allocated to users in a job-by-job basis. You'll learn how this works in this manual.

- 212 compute nodes
- 3780 Intel Xeon Cores
- Approx. 200G B RAM per Compute Nodes
- 34 TB of main storage
- 1800 TB additional storage in file store
- 24 Tesla GPUs
- Visualisations services and more

Note this manual is introductory, and useful troubleshooting resources for the HPC include Stack Overflow and ChatGPT.

Accessing the HPC

Open Command Prompt (from your computers search bar) and type

```
ssh <your_qut_id>@lyra.qut.edu.au    # generic
ssh n1234567@lyra.qut.edu.au        # specific example
```

where you will replace `<your_qut_id>` with your personal QUT ID. The specific example above uses `n1234567`. By hitting ENTER the console will then prompt you to enter your password. **IMPORTANT:** *While typing your password, the console will not show you any dots. It will appear as if the console is not recognizing that you're inputting your password. Be assured that it is!*

Once you've typed your password and hit ENTER you will now be inside the Lyra head node. Your screen might look something like the one below.

```
C:\Users\n1234567>ssh n1234567@Lyra.qut.edu.au
* ===== +
* Access to this computer system is only for authorised QUT staff and students *
* and external persons authorised by QUT. *
* ----- *
* WARNING:          It is a criminal offence to: *
*                   i. Obtain access to data without authority. *
*                   ii. Damage, delete, alter or insert data without authority. *
* ----- *
* Communications on or through QUT's computer systems are monitored and *
* recorded to secure effective system operation and for other lawful purposes. *
+ ===== +

+ ----- Please login using your QUT Access Username and Password ----- +

n1234567@Lyra.qut.edu.au's password:
Last login: Tue Aug 15 04:44:18 2023 from 172.26.218.129
```

*

```
  /---\  _  _  _  _  _
 | | | | | | | | | |
 | | | | | | | | | |
```


that are available from the QUT HPC - there are absolutely loads of these! If all were loaded every time we opened Lyra, we would be waiting for a *long* time. Instead, we have to tell Lyra what modules we'd like to use. In the example above I told Lyra that I would like to use R version 4.0.3. At the time of writing, this is the most recent version of R available to HPC users. Note that **foss** stands for "Free and Open Source Software" and 2020b denotes the year and specific release of R.

Your first HPC job

Installing R packages

As you might expect loading the R module only loads base R packages. To install R packages you'll need to exit the Lyra head node and enter a hotel room (aka, a HPC job).

The way *most* HPCs work is that users submit a request asking for specific computational resources for a specific time period. Then, users wait for the HPC (or more specifically the Portable Batch System (PBS)) to find and allocate the required computational resources. Depending on your specifications you could be waiting seconds or days for the computational resources to become available. **Remember:** *You and other researchers are sharing the HPC so try to never request more resources than you need. This common courtesy ensures that you don't have to wait hours for a small HPC job.*

To request a job, we use the `qsub` command, which I *think* means to submit a computational query. The example code below shows how to ask for 1 hour (`walltime`), 1 CPU (`ncpus`) and 2 GB of RAM (`mem`).

```
qsub -I -S /bin/bash -l walltime=1:00:00,ncpus=1,mem=2gb
```

You can of course edit this to match your needs. For example,

```
qsub -I -S /bin/bash -l walltime=0:20:00,ncpus=1,mem=2gb
```

only asks for 20 minutes of computational resources.

Once you type the code above, you should see a printout similar to below.

```
n1234567@Lyra01:~> qsub -I -S /bin/bash -l walltime=1:00:00,ncpus=1,mem=2gb
qsub: waiting for job 5030193.pbs to start
qsub: job 5030193.pbs ready

n1234567@c14n007:~>
```

This indicates that you are now inside a "node" of the HPC that will automatically exit (aka shutdown) after 1 hour. Notice how the prefix to the `~>` symbol has changed from `n1234567@Lyra01:` (the head node) to `n1234567@c14n007:` (the job).

Returning to the task at hand, we'd like to run some R code. Using what we just learnt we load the R module and open R.

As you might expect the permissions on the HPC are strict and confusing. You **CANNOT** install R packages into the root directory. If you try, R will send a series of errors your way. Instead, you must create a new folder in your personal directory (on the HPC) to hold your R packages. I've called my folder `r_lib`. You can create your folder in the usual way (via the mounted drive on your local computer), by running

```
dir.create("r_lib")
```

in R or

```
mkdir r_lib
```

in Unix. You choose!

Next, we need to tell R to install and search for packages in your newly created folder. This is achieved by running

```
.libPaths("r_lib")
```

at the start of *all* R sessions on the HPC. Repeating for emphasis - the `.libPaths("r_lib")` command must be used at the beginning of every R session. Without this command, R won't find any of your packages and your HPC job will fail! Now that we have told R where to install your packages, the process is as simple as running `install.packages("<package_name>")`.

Once you are finished installing your packages type `q()` to exit R and then `logout` to end the HPC job. At this stage you'll be returned to the Lyra head node.

Congratulations, you've just successfully submitted an interactive HPC job to use R.

HPC jobs

There are two main types of jobs you can submit to the HPC; interactive and batch jobs. Interactive jobs provide you with an interactive R session like what we just saw. **NOTE:** *You will need to have your interactive session open for the job to continue.*

Batch jobs run in the background. They are intended for jobs that will need to run for longer, or when multiple jobs need to be submitted simultaneously. Unlike interactive jobs, after you submit a batch job, you can close your connection to HPC altogether (and even shut your local computer down!) without affecting the jobs processing on the HPC.

Submitting an R script

We've already learned how to submit interactive jobs, but these can be a little cumbersome, as we need to manually type all our code into command prompt. What if we had an R script that we'd like to run on the HPC. Of course, first we need to load all the R scripts and data into our personal HPC directory; I do this via the mounted drive on my local computer. **NOTE:** *The HPC cannot access data that is not in your HPC personal directory.*

To submit an R script to run on the HPC, we must write a corresponding `.sub` script which specifies the computational resources we need and what we'd like to run. These files, written in Unix code, have many names, but in this manual you can think of them as configuration files. Below is an example.

```
1  #!/bin/bash -l           // The Shebang - don't ask me why!
2  #PBS -N myfile           // NAME of job on Lyra
3  #PBS -l ncpus=1          // number of cores
4  #PBS -l mem=2GB          // memory allocation for job
5  #PBS -l walltime=1:00:00 // time allocation for job
6  #PBS -e myjob_errors     // where to save the error file
7  #PBS -o myjob_output     // where to save the output file
```

NOTE: *Do not include the backslashes and comments, these cause headaches for Unix. I have included them for instructional purposes only.*

When the HPC reads the above `.sub` file, it will know that we are requesting a HPC job called `myfile` which requires 1 CPU, 2 GB of RAM and 1 hours of run time. I discuss choosing memory and walltime values and lines 6-7 [here](#) and [here](#). Although the above `.sub` file is technically complete, it doesn't have any code yet. To run an R script using a `.sub` script we can add the following code to the configuration file.

```
1  module load r/4.0.3-foss-2020b
2  module load gdal/3.2.1-foss-2020b
3  R -e "source('QUTHPC_training/single/ms.R')"
```

The example above is extremely simple. On lines 1 and 2 we load the R module and the `gdal` module, which is required when handling spatial data in R. On the 3rd line we tell Unix to open R and then run what ever

code is inside the quotation marks. Technically you could write *all* your R code here, however I strongly recommend against this.

Instead, we tell R to run the entire `ms.R` script. Now that we have successfully written a `.sub` file, we can use it to create a HPC job.

```
n1234567@Lyra01:~> qsub myfile.sub
5059221.pbs
```

By running `qsub <sub_file>`, we tell the PBS system to read the file for (1) the computational resources we required and (2) the code we wish to run. When running a single `.sub` file, the PBS will acknowledge your request for a job by printing a unique Job ID. In the example above this is `5059221.pbs`. If you do not see a Job ID, then there is something wrong with your `.sub` file.

Monitoring the status of jobs

To monitor the status of your submitted jobs run the following command. Remember spaces are important in Unix.

```
qstat -u <your_qut_id>
```

Of course in practice you will replace `<your_qut_id>` with your QUT ID number. This command will provide an output structured like this.

```
n1234567@Lyra01:~> qstat -u n1234567
```

pbs:

| Job ID | Username | Queue | Jobname | SessID | NDS | TSK | Req'd Memory | Req'd Time | Elap S | Time |
|-------------|----------|-------|------------|--------|-----|-----|--------------|------------|--------|------|
| 5005627.pbs | n1234567 | quick | asra1_Ast* | -- | 1 | 1 | 2gb | 01:00 | Q | -- |
| 5005628.pbs | n1234567 | quick | asra1_CHD* | -- | 1 | 1 | 2gb | 01:00 | Q | -- |

I describe each column below. Some of these you may be able to guess anyway!

- **Job ID:** The unique job ID provided by the PBS.
- **Username:** Your QUT username that you used to log into the HPC system.
- **Queue:** Based on the computational resources you requested, the PBS categorizes the job using simple descriptive language. For the example above, where we only requested 1 hour of Walltime, the PBS has assigned this job **quick**. Other options include **long**.
- **Jobname:** The assigned job name that we fed to the HPC via the second line of the `.sub` file.
- **SessID:** NA
- **NDS:** Number of cores requested
- **TSK:** NA
- **Req'd Memory:** memory allocation for job
- **Req'd Time:** time allocation for job
- **S:** Status of the job. Options include Q (queued), R (running) and E (elapsed).
- **Elap Time:** Elapsed time for the job. Only shows a value once **S = R**. Once **Elap Time** is equal to **Req'd Time** the job stops — regardless of if it has finished or not! For this reason it is always best to slightly overestimate the run time of your jobs.

If some of these values don't look right, stop the job and edit your configuration file accordingly before resubmitting. To stop a queued or running job, type

```
qdel <jobid> // include the .pbs at the end
```

When your models are running the `qstat` call will look like this.

```
n1234567@Lyra01:~> qstat -u n1234567
```

pbs:

| Job ID | Username | Queue | Jobname | SessID | NDS | TSK | Req'd Memory | Req'd Time | Elap S | Time |
|-------------|----------|-------|------------|--------|-----|-----|-----------------|---------------|-----------|-------|
| 5005629.pbs | n1234567 | quick | asra1_Ast* | 47118 | 1 | 1 | 2gb | 01:00 | R | 00:01 |
| 5005630.pbs | n1234567 | quick | asra1_CHD* | 47168 | 1 | 1 | 2gb | 01:00 | R | 00:01 |

Most of the time, your use of the HPC head node will be to monitor your jobs, so ensure you're accustomed to the `qstat` command.

Choosing `walltime` and `mem`

There is a real art in setting memory and walltime values when running jobs on the HPC.

Walltime is the total amount of time it will take to complete your job. This includes loading packages, data, compiling, fitting and summarising the model, and saving any output. You must leave sufficient time for all these tasks to be complete as the HPC will stop the job once the `walltime` has been met. Like most of our scripts, the last commands are always to save our results. To avoid running models for days and then having the results discarded it is best practice to save model results earlier in your scripts if possible. If you wish to save space, you can always tell R at the very end of your script to delete a file.

To help guess how long a model will take, I recommend making use of the `jf$MCMCrecommendations` function which allows one to approximate the run time of a model based on a shorter run of the same model. Of course, this function is an approximation to the run time of the model *only*, it cannot indicate how long model summarising or saving will take.

Unfortunately, setting the memory value is arguably more difficult. Unlike your local R session which will just run slower if it exceeds your memory, the HPC will just end your job — discarding any modelling done up to that point. To avoid this, it is best practice to run your models with a high value of memory (say 10GB) to start with and then reduce this as necessary.

You might think that just setting memory very high is the pragmatic solution. **Please don't do this!** Firstly, by asking for computational resources you don't need, you slow the HPC down for everyone. Secondly, requesting large amounts of RAM (memory) will make it take substantially longer for the PBS to allocate you a job. For example, if you request 100GB, you might have to wait an hour for the PBS to allocate you this job.

In R you can assess the size of your objects (e.g. your `draws` or `fit` objects) using

```
format(object.size(<object_name>), units = "Mb")
```

Remember that R holds all objects in memory. Thus, by summing the size of all the (“large”) objects that you'll need to have simultaneously available in R, you can glean an initial approximation to the memory value you should specify in your configuration `.sub` file.

I find that an iterative process for deciding walltime and memory is most efficient. First, set a reasonably large value for both. After a successful HPC job is complete, the output and error files (see [below](#)) provide detailed information about the computational resources the HPC *actually* used, irrespective of what you requested. By viewing these values, you can find an optimal upper bound for both.

The output file

After a job, Lyra creates two reports: an output file and an error file. Here is an example of a output file from Lyra. In (almost!) all cases, Lyra will create an output and error file regardless of whether the job was successful. Note that when I refer to success I mean the R script finished running and the required files were successfully saved.

```

1 R version 4.0.3 (2020-10-10) -- "Bunny-Wunnies Freak Out"
2 Copyright (C) 2020 The R Foundation for Statistical Computing
3 Platform: x86_64-pc-linux-gnu (64-bit)
4
5 R is free software and comes with ABSOLUTELY NO WARRANTY.
6 You are welcome to redistribute it under certain conditions.
7 Type 'license()' or 'licence()' for distribution details.
8
9 Natural language support but running in an English locale
10
11 R is a collaborative project with many contributors.
12 Type 'contributors()' for more information and
13 'citation()' on how to cite R or R packages in publications.
14
15 Type 'demo()' for some demos, 'help()' for on-line help, or
16 'help.start()' for an HTML browser interface to help.
17 Type 'q()' to quit R.
18
19 > .libPaths('r_lib');
20 > base_folder='QUTHPC_training';
21 > model_spec='asra1';
22 > sex='Female';
23 > condition='Asthma';
24 > Rfile='asra1_Asthma_Female';
25 > cur_date='20230815';
26 > niter=800;
27 > nburnin=400;
28 > thin=1;
29 > nchains=4;
30 > source('QUTHPC_training/ms.R');
31 |-----|-----|-----|
32 |-----|-----|-----|
33 |-----|-----|-----|
34 |-----|-----|-----|
35 |-----|-----|-----|
36 |-----|-----|-----|
37 |-----|-----|-----|
38 |-----|-----|-----|
39 >
40 >
41 PBS Job 5005634.pbs
42 CPU time : 00:04:10
43 Wall time : 00:05:50
44 Mem usage : 2097152kb

```

The output file is very useful in determining the computational resources that were *actually* used by the HPC during the running of the job. These details are available from line 41 onward. Note that CPU time and walltime are unique metrics. We're only interested in the walltime as this is what we set in the configuration file.

In the example above, the HPC job ended after 5 mins and 50 seconds and used 2.097152 Gb of memory. Using the iterative method proposed above, we might consider setting the memory at 3GB and the walltime at 10 minutes.

The error file

Here is an example of the error file produced after a successful HPC job. We ran a simple `nimble` model.

```
Attaching packages                                tidyverse 1.3.1
ggplot2 3.3.5      purrr   0.3.4
tibble  3.1.5      dplyr   1.0.7
tidyr   1.1.4      stringr 1.4.0
readr   2.1.1      forcats 0.5.1
Conflicts                                tidyverse_conflicts()
dplyr::filter() masks stats::filter()
dplyr::lag()     masks stats::lag()
Linking to GEOS 3.9.1, GDAL 3.2.1, PROJ 7.2.1
Loading required package: sp
Loading required package: spData
To access larger datasets in this package, install the spDataLarge
package with: `install.packages('spDataLarge',
repos='https://nowosad.github.io/drat/', type='source')`
nimble version 0.12.2 is loaded.
For more information on NIMBLE and a User Manual,
please visit https://R-nimble.org.

Attaching package: 'nimble'

The following object is masked from 'package:stats':

    simulate

Attaching package: 'Matrix'

The following objects are masked from 'package:tidyr':

    expand, pack, unpack

This is bayesplot version 1.8.1
- Online documentation and vignettes at mc-stan.org/bayesplot
- bayesplot theme set to bayesplot::theme_default()
  * Does _not_ affect other ggplot2 plots
  * See ?bayesplot_theme_set for details on theme setting
Defining model
Building model
Setting data and initial values
Running calculate on model
  [Note] Any error reports that follow may simply reflect missing values in model variables.
Checking model sizes and dimensions
  [Note] All model variables are initialized.
Compiling
  [Note] This may take a minute.
  [Note] Use 'showCompilerOutput = TRUE' to see C++ compilation details.
Compiling
  [Note] This may take a minute.
  [Note] Use 'showCompilerOutput = TRUE' to see C++ compilation details.
Starting sampling for 400 iterations for each of 4 chains.
Running chain 1 ...
```

```

Running chain 2 ...
Running chain 3 ...
Running chain 4 ...
Sampling took 0.92 mins
Median Rhat: 1
0.01% of Rhats larger than 1.01
Max Rhat = 1.01 (sigma)
0.01% of ess_bulk are too small
Min ess_bulk = 311.19 (sigma)
0.01% of ess_tail are too small
Min ess_tail = 285.74 (sigma)
Average posterior draws per minute:
Progress ... -> Point estimates...
Progress ... -> Standard errors...
Progress ... -> Highest density intervals...
Progress ... -> Point estimates...
Progress ... -> Standard errors...
Progress ... -> Highest density intervals...

```

Unlike the output file, the error file is useful for diagnosing errors produced by R during the HPC job. This is because the output file provides some of the output from R like you would see in RStudio on your local machine.

Below is an example error file for when a job is stopped early due to an R error.

```

Attaching packages                                tidyverse 1.3.1
ggplot2 3.3.5      purrr   0.3.4
tibble  3.1.5      dplyr   1.0.7
tidyr   1.1.4      stringr 1.4.0
readr   2.1.1      forcats 0.5.1
Conflicts                                tidyverse_conflicts()
dplyr::filter() masks stats::filter()
dplyr::lag()    masks stats::lag()
Linking to GEOS 3.9.1, GDAL 3.2.1, PROJ 7.2.1
Loading required package: sp
Loading required package: spData
To access larger datasets in this package, install the spDataLarge
package with: `install.packages('spDataLarge',
repos='https://nowosad.github.io/drat/', type='source')`
nimble version 0.12.2 is loaded.
For more information on NIMBLE and a User Manual,
please visit https://R-nimble.org.

```

Attaching package: 'nimble'

The following object is masked from 'package:stats':

```
simulate
```

Attaching package: 'Matrix'

The following objects are masked from 'package:tidyr':

```
expand, pack, unpack
```

```

This is bayesplot version 1.8.1
- Online documentation and vignettes at mc-stan.org/bayesplot
- bayesplot theme set to bayesplot::theme_default()
  * Does _not_ affect other ggplot2 plots
  * See ?bayesplot_theme_set for details on theme setting
Error in file(filename, "r", encoding = encoding) :
  cannot open the connection
Calls: source -> withVisible -> eval -> eval -> source -> file
In addition: Warning message:
In file(filename, "r", encoding = encoding) :
  cannot open file 'QUTHPC_training/r_src/asra2.R': No such file or directory
Execution halted

```

As you can see, the job was stopped early due to a directory issue.

HPC errors

Hopefully you now agree that the error and output files are super helpful. But, wait there is more! Lyra also produces output when jobs are stopped due to exceeded computational resources. If your memory allocation is exceeded, you might observe the following line in the error file.

```

/var/spool/PBS/mom_priv/jobs/5059208.pbs.SC: line 23: 47174 Killed

```

While in the output file you might see

```

Mem usage : 2097152kb

```

which has clearly exceeded the memory of 2GB we requested in the .sub file. Remember there are 1 million kB in a GB.

If your job has continued to run past its allocated walltime, the PBS will automatically stop the job. In this case, the error file might say

```

=>> PBS: job killed: walltime 77 exceeded limit 60

```

While in the output file you'll see

```

CPU time   : 00:00:59
Wall time  : 00:01:28
Mem usage  : 1029864kb

```

which clearly shows that we've exceeded the 1 minute of resources we requested.

Sometimes an error will occur that is unique and potentially one that cannot be described in the standard error and output files. In these instances you will receive an automated email from the HPC system that may look like this.

PBS JOB 5005625.pbs



root <erereasearch@qut.edu.au>

To James Hogg

PBS Job Id: 5005625.pbs

Job Name: asra1_CHD_Persons

Post job file processing error; job 5005625.pbs on host cl4n007

If this happens please contact me. In most cases, this is a result of an error in the HPC system. However in some cases, if your specification of output and error files (in terms of directory) is incorrect, the PBS emails you instead.

Batch running of jobs

So far we've described running a single R script (or a single job) on the QUT HPC. In this section, we'll describe how Unix code can be used to create *and* run multiple jobs simultaneously. For example, we might wish to run all combinations of the models, sexes and conditions. Using Unix code, we can achieve this with a single command.

First though, we must define a new file type, called a shell script (.sh file), that's purpose is to create and submit .sub files to the PBS. Below is a complex example of a shell script, which I will guide you through.

```
1  #!/bin/bash
2
3  base_folder='QUTHPC_training/batch'
4
5  # Model
6  for model_spec in asra1 asra2 ## Name of file that runs the model
7  do
8
9      # Sex
10     for sex in Persons Female Male
11     do
12
13         # Condition
14         for condition in Asthma CHD
15         do
16
17             # create the unique file name
18             Rfile=$model_spec'_'$condition'_'$sex
19
20             # get the current date
21             cur_date=$(date +%Y%m%d%H%M)
22
23             # create directories
24             mkdir -p $base_folder/sub_src/$cur_date
25             mkdir -p $base_folder/outputs/lyra/$cur_date/errors
26             mkdir -p $base_folder/outputs/lyra/$cur_date/out
27             mkdir -p $base_folder/outputs/$condition/$sex
```

```

28
29     # Set loop output file with full directory
30     loop_output_file=$base_folder'/outputs/'$condition'/'$sex'/'$Rfile
31
32     # create the unique .sub script files
33     file=$Rfile'.sub'
34
35     # paste the commands in the .sub scripts
36     cat > $base_folder/sub_src/$cur_date/$file <<EOF
37 #!/bin/bash -l
38 #PBS -N $Rfile
39 #PBS -l ncpus=1
40 #PBS -l mem=2GB
41 #PBS -l walltime=0:20:00
42 #PBS -e $base_folder/outputs/lyra/$cur_date/errors/$Rfile
43 #PBS -o $base_folder/outputs/lyra/$cur_date/out/$Rfile
44
45 module load r/4.0.3-foss-2020b
46 module load gdal/3.2.1-foss-2020b
47
48 R -e ".libPaths('r_lib');
49 base_folder='$base_folder';
50 loop_output_file='$loop_output_file';
51 model_spec='$model_spec';
52 sex='$sex';
53 condition='$condition';
54 cur_date='$cur_date';
55 niter=800;
56 nburnin=400;
57 thin=1;
58 nchains=4;
59 source('$base_folder/ms.R');"
60 EOF
61
62     # run each script
63     qsub $base_folder/sub_src/$cur_date/$file
64
65 done
66 done
67 done

```

To define every unique combination of models, sexes and conditions, we define three nested ‘for-loops’. These are visible in lines 6-7, 10-11 and 14-15. The loops are closed on lines 67-68. Just like in R, the ‘for-loop’ Unix syntax creates a dynamic variable which takes on a new value for each round. Consider the ‘for-loop’ defined on line 6. Here, the variable `model_spec` takes on the value `asra1` for the first round and then `asra2` for the second. To reference a dynamic variable in Unix, we use `$model_spec`.

After defining the ‘for-loops’, we create a unique file name, called `Rfile`, on line 18. This is created by joining the dynamic variables defined in the ‘for-loops’. For the first loop `Rfile` will equal

```
asra1_Asthma_Persons
```

Line 21 gives the current date and time. This is useful to document our error and output files from Lyra. Following this, from lines 24 to 29 we create a variety of directories (using the `mkdir -p` command) into which our `.sub` files and results will be saved.

Then in lines 32 and 35, we create additional dynamic variables that will be used later. For the first loop `loop_output_file` will be equal to

```
QUTHPC_training/batch/outputs/Asthma/Persons/asra1_Asthma_Persons
```

This is extremely helpful to define in the `.sh` script as we can use `loop_output_file` in R rather than referring to each of the dynamic variables in the ‘for-loops’.

Lines 39-61 should (hopefully!) look familiar to you. This is where we dynamically create the `.sub` scripts that are specific to the ‘for-loops’. The codes on either side of this (line 38 and 62) are the Unix code required to create the `.sub` file. Please observe how frequently we call the dynamic variables we’ve created in the code prior. The constant use of `$<variable>` should make this obvious.

Line 65 should also be familiar. It takes the `.sub` file created by the shell script and then creates a corresponding HPC job.

Now that we have created a `.sh` script, we can run it on the Lyra head node using

```
bash <file_name> // include the .sh at the end
```

Please observe the new command we’ve used; the `bash` command. This command is coupled with a `.sh` script, just like the `qsub` command is coupled with a `.sub` script. If the `.sh` script has been correctly setup, upon running the `bash` command the PBS will provide a list of new Job IDs for all the submitted jobs. In the example above, this would be a list of 12.

I hope you’re impressed! By using a `.sh` script, we can run many jobs simultaneously using a single line of code. Furthermore, we can now close our connection to Lyra, feeling confident that our models will continue to run without us.

If the jobs were successful you’ll be able to find your results in your personal HPC directory, as specified in the `.sh` script.