Computing skills

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LIV.DAT advanced researcher skills school 2021/03/24

Overview

- Introduction: issues and goals
- Discussion of survey results
- Interacting with the Machine
- Data storage and management
- Software development and programming
- Break*
- Version control
- Multi-processing and GPU acceleration
- Presentation: making good plots
- Jobs and skills outside of academia

!!!This is a group discussion and not 1-hour lecture.Feel free to contribute at any time!

The dangers of poor computing practices

- The Climate Research Unit (UEA) emails were hacked and released to the public in 2009
- "Climategate" ensued: scientists were manipulating the data say conspiracy theorists
- No evidence of data manipulation was found
- In reality, poor data management and programming practices were highlighted
- · Some emails contained discussion of bugs in the code
- Although the code was fine and the science checks out, the reputation was damaged
- First hand admissions from scientists: "Yup, my awful programming strikes again"
- With programming becoming an ever larger fraction of researchers' activities, it is important to apply the same rigour as you do with publications
- · Paper retractions due to mistakes happen all the time

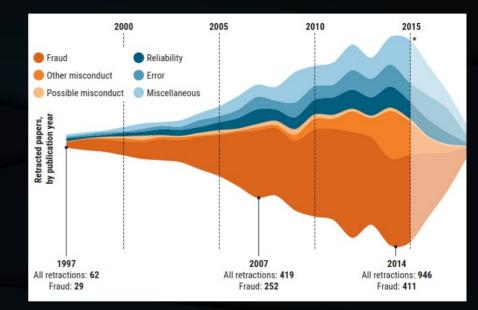


Image source: sciencemag.org

https://www.sciencemag.org/news/2018/10/what-massi ve-database-retracted-papers-reveals-about-science-p ublishing-s-death-penalty

Modern scientist vs. software developer

Issues

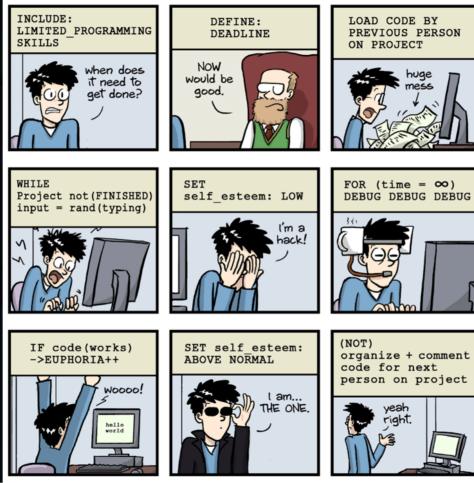
• Difference?

- In science, software is viewed as a tool and not the product. Papers are.
- There is no code review. If the result matches our expectations then it's working correctly. Is it?
- Scientists are often self-taught (>90%)
- There is no reward for good code
- Projects grow organically with no clear requirements at the beginning
- Much of the code is never used again

Aims

- Software engineering/development has figured out many of the problems
- Practices exist that make code development faster, easier, better
- Can we learn something from them?
- Improve productivity, collaborate easier, have more confidence in results
- Mistakes in science can be extremely costly, how can we minimise them?
- Shake off the label of "bad programmer" if leaving academia

PROGRAMMING FOR NON-PROGRAMMERS



JORGE CHAM © 2014

WWW.PHDCOMICS.COM

Goals of this workshop

- Identify the skills you already have but didn't appreciate enough
- Identify the areas of computing which you are lacking in
- Gain awareness of best practice
 - The carpentry projects are good places to start*
- Have a collection of sources to refer to
- Find out what employers are looking for in industry

Theory **Computation** develop models, simulate models. hypothesis, test hypothesis, observations, predictions, conclusions. visualizations, Experiment measurements, observations, collect data, test hypothesis, ...

*https://datacarpentry.org/ https://software-carpentry.org/

Survey results



Interacting with the Machine

- Shell or "Terminal" is a program used to interact with the operating system
- Holds your commands in memory to be used again
- About 90% of the internet runs Linux servers
- AWS, Google cloud, and Azure all support Linux environments so the Shell is not going away anytime soon
- Various flavours: bash, csh, tcsh, ksh, etc.
- Simple use involves: Is, pwd, cd, cp, mv, etc.

J=1	egidi	jus@egidijus	-MS-7B89: ~/work	Q			×
Hello World egidijus@egidij /bin/bash egidijus@egidij bash_script.sh Desktop Documents egidijus@egidij egidijus@egidij /home/egidijus/ egidijus/local :/usr/games:/us egidijus@egidij /home/egidijus/	us-MS-7889:-\$ ec us-MS-7889:-\$ ec Downloads hobby_projects mintconda3 us-MS-7889:-\$ cd us-MS-7889:-\$ cd us-MS-7889:-{wor work/cuda-11.1.0 /bin:/usr/local/ games:/s us-MS-7889:-/wor work/cuda-11.1.0 us-MS-7889:-/wor	ho \$SHELL Music Pictures Public work/ k\$ echo \$P _2/bin/:/h sbin:/usr/ nap/bin k\$ echo \$L _2/lib64/:	snap Templates TensorFlow-Tut ATH ome/egidijus/mi local/bin:/usr/ D_LIBRARY_PATH	orial nicond	a3/con	dabir	

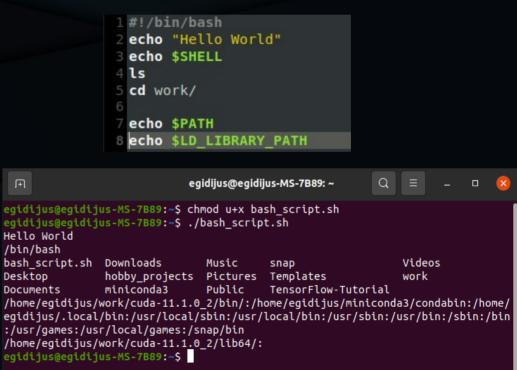
Shell scripts

You can type your commands one-by-one

F	egidijus@egidijus	-MS-7B89: ~/work	Q			×
egidijus@egidijus-MS-7B89: Hello World egidijus@egidijus-MS-7B89: /bin/bash egidijus@egidijus-MS-7B89: bash_script.sh Downloads Desktop hobby_proj Documents miniconda3 egidijus@egidijus-MS-7B89: /home/egidijus/work/cuda-1 egidijus@egidijus/work/cuda-1 egidijus@egidijus-MS-7B89: /home/egidijus/work/cuda-1 egidijus@egidijus-MS-7B89:	-\$ echo \$SHELL -\$ ls Music ects Pictures Public -\$ cd work/ ~/work\$ echo \$P 1.1.0_2/bin/:/h bcal/sbin:/usr/ es:/snap/bin ~/work\$ echo \$L 1.1.0_2/lib64/:	snap Templates TensorFlow-Tut PATH nome/egidijus/mi 'local/bin:/usr/ .D_LIBRARY_PATH	orial	a3/con	dabir	

https://maker.pro/linux/tutorial/basic-linux-commands-for-beginners

Or you can save them in a file and execute them all



Environment variables

- Within Linux, global locations and values are stored in the form of environment variables to be looked up by shell applications
- Most commonly used are: PATH, LD_LIBRARY_PATH, PYTHONPATH, etc.
- How you control these depends on your shell
- They can be set once from the terminal or added to a file to be loaded at startup
- In my case, they should be stored in "~/.profile" or "~/.bashrc" if it doesn't exist

egidijus@egidijus-MS-7B89:~\$ export PATH="/home/egidijus/Music/:\$PATH" egidijus@egidijus-MS-7B89:~\$ echo \$PATH

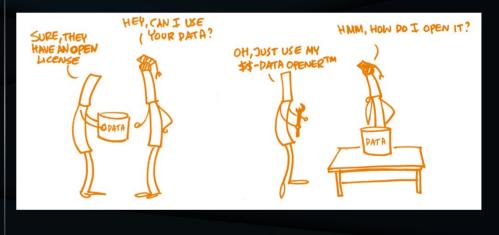
/home/egidijus/Music/:/home/egidijus/work/cuda-11.1.0_2/bin/:/home/egidijus/mini conda3/condabin:/home/egidijus/.local/bin:/usr/local/sbin:/usr/local/bin:/usr/sb in:/usr/bin:/sbin:/bin:/usr/games:/usr/local/games:/snap/bin

	.dropbox	Public
	.dropbox-dist	.python_history
.apport-ignore.xml	.gnupg	snap
.bash_aliases	hobby_projects	.ssh
.bash_history	.ipython	.sudo_as_admin_successful
.bash_logout	.jupyter	Templates
.bashrc	.keras	TensorFlow-Tutorial
bash_script.sh	.kite	.thunderbird
bogofilter	.local	Videos
cache	miniconda3	.virtual_documents
. conda	.mozilla	.virtualenvs
.condarc	Music	.vscode
config		.wget-hsts
Desktop	Pictures	work
docker	.pki	
Documents	.profile	
Downloads	.psensor	

https://linuxize.com/post/how-to-set-and-list-environment-variables-in-linux/

Data storage and management: forms of good practice

- Save the raw data without any changes preserve it for others to start where you did
- Save a version of the data that *you* want to use - change format, improve readability, save time
- Save data-point values for easy and fast plotting – you *will* have to remake that plot
- Favour open, non-proprietary format CSV, JSON, HDF
- Record the steps taken to transform data and keep it with the data
- If data storage and manipulation is a significant part of your project, it may be worth reading up on data engineering



What format is best?

- There is no one "best" format for storing data
- This depends on your needs and resources
- Compressing the data can yield great size results at the cost of speed
- If you never plan to visually inspect the data, do you need it in human-readable format?
- Consider the overheads: CSV is quite efficient at storing small amounts of data*

	read_s	write_s	size_ratio_to_CSV
storage			
CSV	17.900	69.00	1.000
CSV.gzip	18.900	186.00	0.047
Pickle	0.173	1.77	0.374
HDF_fixed	0.196	2.03	0.435
HDF_tab	0.230	2.60	0.437
HDF_tab_zlib_c5	0.845	5.44	0.035
HDF_tab_zlib_c9	0.860	5.95	0.035
HDF_tab_bzip2_c5	2.500	36.50	0.011
HDF_tab_bzip2_c9	2.500	36.50	0.011

https://www.architecture-performance.fr/ap_blog/loading-data-into-a-pandas-dataframe-a-performance-study/

https://stackoverflow.com/a/37012035

Best practices in programming

Write programs for people, not computers

- Clear, short chunks of facts
 - A program should not require its readers to hold more than a handful of facts in memory at once
- Make names consistent, distinctive, and meaningful
- Make code style and formatting consistent
 - Will depend on your code language but most have a style guide

```
Wilson et al. 2014 "Best practices in scientific computing"
```

```
create(path.archiveList.xFilesFactor=None,aggregationMethod=None,sparse=False,useFallocate=False):
#.Set.default.param
if vFilesEactor is None.
 xFilesFactor = 0.5
if aggregationMethod is None:
  aggregationMethod = 'average
validateArchiveList(archiveList)
if os.path.exists(path)
raise InvalidConfiguration("File %s already exists!" % path)
 fh = open(path, 'wb')
 ·if LOCK:
    fcntl.flock( fh.fileno(), fcntl.LOCK EX )
  aggregationType = struct.pack( longFormat, aggregationMethodToType.get(aggregationMethod, 1) )
  oldest = max([secondsPerPoint * points for secondsPerPoint, points in archiveList])
  fh_write(nackedMetadata)
  headerSize = metadataSize + (archiveInfoSize * len(archiveList))
  archiveOffsetPointer = headerSize
  for secondsPerPoint.points in archiveList:
    archiveInfo = struct.pack(archiveInfoFormat, archiveOffsetPointer, secondsPerPoint, points)
    fh.write(archiveInfo)
    archiveOffsetPointer += (points * pointSize)
  if CAN FALLOCATE and useFallocate:
    remaining = archiveOffsetPointer - headerSize
    fallocate(fh, headerSize, remaining)
  elif sparse:
   fh.seek(archiveOffsetPointer - 1)
    fh.write('\x00')
    remaining = archiveOffsetPointer - headerSize
    chunksize = 16384
    zeroes = '\x00' * chunksize
    while remaining > chunksize
      fh.write(zeroes)
     remaining -= chunksize
    fh.write(zeroes[:remaining])
  if AUTOFLUSH
   fh.flush()
   os.fsync(fh.fileno())
```

https://julien.danjou.info/python-bad-practice-concrete-case/

fh.close()

Let the computer do the work

- Make the computer repeat tasks: loops, functions
- Keep functions to doing one task at a time easier to test
- Use existing modules/code rather than writing things from scratch
- Save recent commands in a file for reuse (write a script)
- If your work depends on repeated use of the same workflow: build it automatically
- Personal advice: DO NOT comment lines out as a way of controlling behaviour

This code would benefit so much from functions and loops

1 f = h5py.File('snapshot_011_z001p017_z1p0_paper_plot_quantities_10p0lMstar_30kpc_14p0lM200c_incBCG_field.hdf5', 'r')
2 SFR = np.array(f('SFR'])
3 fq_field = np.size(SFR[np.where(SFR < -12)])/np.size(SFR)
4 M_star_field = np.array(f('M_star'])
5
6 f = h5py.File('snapshot_011_z001p017_z1p0_paper_plot_quantities_10p0lMstar_30kpc_14p0lM200c_incBCG_cluster.hdf5', 'r')
7 SFR = np.array(ff('SFR'])
8 fq_cluster = np.size(SFR[np.where(SFR < -12)])/np.size(SFR)
9 M_star_cluster = np.array(ff('M_star'])
9 M_star_cluster = np.array(ff('M_star'])</pre>

Forgetting to rename on of these instances is only a matter of time

Make incremental changes

- Work in small steps with frequent feedback and course correction
 - Really helps if you structure your code well (small functions)
- Use a version control system
 - You rarely know if the changes you make will be positive
 - More on version control later

- One example of incremental development may be:
 - Pseudocode draft
 - Define variables you know you are going to need, in the format you think you need them
 - Empty functions for functionality
 - Finalise function logic, returning intermediate values to test functionality
 - Test final function with various inputs
 - Add comments and optimise if possible

Don't repeat yourself or others

- Data input, constants, variable values must have a single, authoritative source
- Store parameter values in a separate file if you have many
- Modularise, rather than copy/paste
 - OOP may come into play here
- Re-use and don't re-write
 - On a wider scale: modules
 - On a small scale: worth the time searching for that bit of code you wrote in the past
- Plan for mistakes
 - functions/classes are much easier to unit-test
 - Implement assertions (check on inputs, format, etc.)

11	<pre>def compute_fq_from_snapshot(filename):</pre>
12	
13	····Inputs: (string) file name
14	<pre>・・・・Computes quenched fraction given a snapshot file name</pre>
15	••••Outputs: (tuple) quenched fraction, stellar mass
16	
17	····f·=·h5py.File(filename,·'r')
18	<pre>SFR = np.array(f['SFR'])</pre>
19	····M_star·=·np.array(f['M_star'])
20	<pre>fq = np.size(SFR[np.where(SFR < -12)])/np.size(SFR)</pre>
21	store fa M stor

Optimise software only after it works correctly

- Clean up comments, general tidy up
- Use a profiler to find bottlenecks
- Consider parallelising if it takes a while to run
- Write in the highest level language possible, then translate to something more efficient if needed
 - e.g. start in python and translate to C/FORTRAN
 - The overall structure is unlikely to change
 - Evidence shows that you work faster in high level language or one you are familiar with

Oram, A. 2010 "Two comparisons of programming languages"

import time import concurrent.futures from PIL import Image, ImageFilter

img_names = [

```
photo-1516117172878-fd2c41f4a759.jpg',
    'photo-1532009324734-20a7a5813719.jpg',
     'photo-1524429656589-6633a470097c.ipg'.
    'photo-1530224264768-7ff8c1789d79.jpg',
    'photo-1564135624576-c5c88640f235.ipg'.
    'photo-1541698444083-023c97d3f4b6.jpg',
    'photo-1522364723953-452d3431c267.jpg',
    'photo-1513938709626-033611b8cc03.jpg',
    'photo-1507143550189-fed454f93097.jpg',
    'photo-1493976040374-85c8e12f0c0e.jpg'.
    'photo-1504198453319-5ce911bafcde.jpg',
     'photo-1530122037265-a5f1f91d3b99.jpg',
     'photo-1516972810927-80185027ca84.ipg'.
    'photo-1550439062-609e1531270e.jpg',
     photo-1549692520-acc6669e2f0c.jpg
t1 = time.perf counter()
size = (1200, 1200)
```

- def process_image(img_name):
- img = Image.open(img_name)
- img = img.filter(ImageFilter.GaussianBlur(15))
- img.thumbnail(size)
- img.save(f'processed/{img_name}')
-print(f'{img_name} was processed...')

```
t2 = time.perf_counter()
```

print(f'Finished in {t2-t1} seconds')

Document design and purpose, not mechanics

- Some comments is usually better than no comments at all
- Proper, meaningful comments is what really makes a difference
- Document the purpose of the code rather than what it does – this should be conveyed by the code itself
- Refactor code if it helps explain how it works
- Embed documentation within the code itself
- If excessive documentation is required, consider a format specifically designed for it, e.g. Jupyter notebooks

No comments at all

```
r = n / 2;
while ( abs( r - (n/r) ) > t ) {
  r = 0.5 * ( r + (n/r) );
}
System.out.println( "r = " + r );
```

Good and unnecessary comments

```
// square root of n with Newton-Raphson approximation
r = n / 2;
while ( abs( r - (n/r) ) > t ) [ // while result is greater than tolerance
r = 0.5 * ( r + (n/r) );
]
System.out.println( "r = " + r );
```

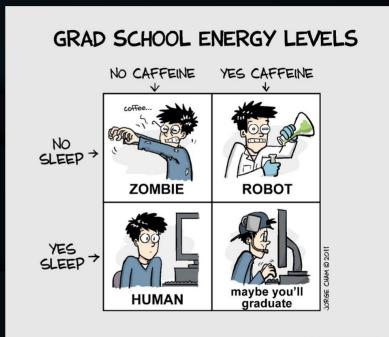
Descriptive names instead of comments

https://blog.codinghorror.com/coding-without-comments/

A working break

Task: find a job you would consider outside of academia, list:

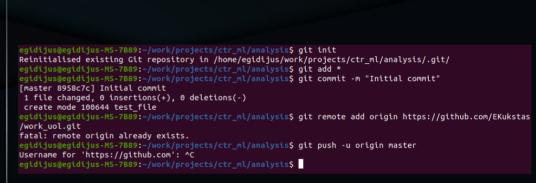
- job title,
- website you found it on,
- one key technical skill it requires
- https://docs.google.com/document/ d/1mRyHjTynmQFqVC2W-IxIUXY ZVKOI8-FgRliIXGyhqbg/edit?usp= sharing



WWW. PHDCOMICS. COM

Version control: Git

- Git stores metadata on your code and tracks changes
- Dated "images" of your code are available to access should something go wrong
- Github, Gitlab, Gitbucket, etc. are online repositories where your code can be stored and shared with others
 - This allows for collaborative development where other people can propose changes/additions to your code
 - You then get to decide whether to "merge" the changes or not
- · Git is very powerful and not always intuitive
- Practical advice: at the very least, commit your code regularly as a form of backup for yourself, whatever the state



https://rogerdudler.github.io/git-guide/

Parallel computing

- Most CPUs these days have multiple cores
- Scientific computing clusters have 100s or even 1000s of cores
- Most processes are designed to only use a single core
 - Explicit instructions have to be passed in or with your code to use multiple cores
- It is quite easy to implement and can speed up your code multiple times over
- Specific tasks, such as machine learning can take advantage of the GPU, as well
 - You will need to install specific drivers and set your code up to use it, which can be more involved
 - The rewards can be big if your problem is well suited (matrix operations, etc.)

img names = | img names = [photo-1516117172878-fd2c41f4a759.jpg', photo-1532009324734-20a7a5813719.jpg', photo-1532009324734-20a7a5813719.jpg', photo-1524429656589-6633a470097c.ipg'. 'photo-1530224264768-7ff8c1789d79.jpg'. 'photo-1530224264768-7ff8c1789d79.jpg', photo-1564135624576-c5c88640f235.jpg', 'photo-1564135624576-c5c88640f235.jpg', 'photo-1541698444083-023c97d3f4b6.ipg'. photo-1541698444083-023c97d3f4b6.ipg'. 'photo-1522364723953-452d3431c267.ipg'. 'photo-1522364723953-452d3431c267.jpg', photo-1513938709626-033611b8cc03.jpg', 'photo-1513938709626-033611b8cc03.jpg', 'photo-1507143550189-fed454f93097.ipg'. 'photo-1507143550189-fed454f93097.ipg'. photo-1493976040374-85c8e12f0c0e.jpg', photo-1493976040374-85c8e12f0c0e.jpg', photo-1530122037265-a5f1f91d3b99.jpg'. 'photo-1516972810927-80185027ca84.jpg', 'photo-1550439062-609e1531270e.jpg', 'photo-1549692520-acc6669e2f0c.jpg 'photo-1550439062-609e1531270e.jpg', photo-1549692520-acc6669e2f0c.jpg t1 = time.perf counter() t1 = time.perf counter() size = (1200, 1200) size = (1200, 1200) def process image(img name): img = Image.open(img name) def process image(img name): img = Image.open(img name) img = img.filter(ImageFilter.GaussianBlur(15)) img = img.filter(ImageFilter.GaussianBlur(15)) img.thumbnail(size) img.save(f'processed/{img name}') print(f'{img name} was processed...') img.thumbnail(size) img.save(f'processed/{img name}') for img in img names: print(f'{img name} was processed...') process image(img) t2 = time.perf counter() with concurrent.futures.ProcessPoolExecutor() as executor: executor.map(process image, img names) print(f'Finished in {t2-t1} seconds') t2 = time.perf counter()print(f'Finished in {t2-t1} seconds')

Finished in 12.15791793000244 seconds

Finished in 2.2747356240033696 seconds

https://youtu.be/fKl2JW_qrso http://bit.ly/multiprocess-code

Presentations

- Use whatever tool you like, but make sure it works the same way on any system
 - re. PowerPoint and LibreOffice Impress
 - PDF is a good format
- Plots are the ultimate product make sure they are of highest quality
 - Frequently presented at talks / on posters by other people if they're good
- Make sure that:
 - The message is as clear as possible without a lengthy caption (out of context colleague test)
 - Labels are clear and large enough
 - Consider colour blindness: choose your colours carefully and favour line-types over different colours
 - Favour vector images over raster
 - Plot contours instead of many scatter points, rasterize if absolutely necessary
- Avoid plots from simulation packages you have little control and they're usually poor. Save the data instead and plot it yourself
- Python, R, Matlab all have good plotting modules where you have full control

Example parameter changes for Python's matplotlib

```
plt.style.use('default')
plt.rcParams['legend.handlelength'] = lhandle
```

https://towardsdatascience.com/matplotlib-styles-for-scientific-plot ting-d023f74515b4

Concluding remarks

- Appreciate the importance of programming in your professional lives
 - Reproducibility, accountability depend on it
- Continue to improve: look at your code from start of the PhD
- Don't expect to implement all of the advice now
- Be aware of what good practices are
- It will help you in the long run