Data processing & automatization

ENBIT Workshop Brussels, May 2018









Message from "Captain Obvious"

ENBIT Workshop Brussels, May 2018









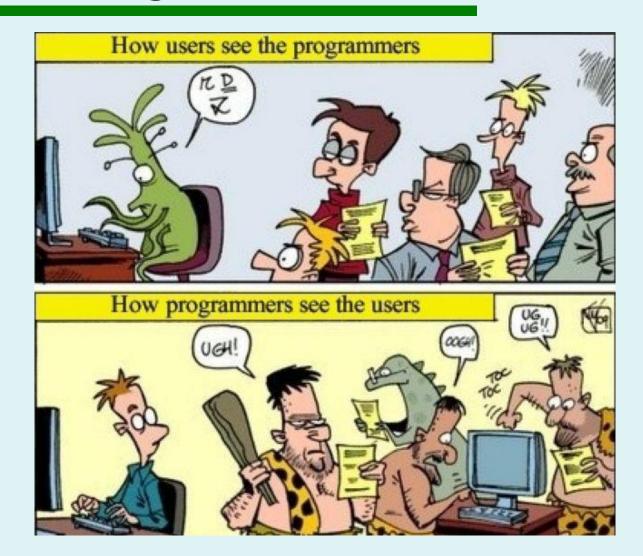
Job done?

Methods people vs. Clinical people:

- M-people → happy when...
 - there is a theoretical solution,
 - evaluated and tested on some data,
 - relies on "a few" parameters to be adjusted
- C-people → happy when...
 - there is a 1-button software solution
 - works on any data
 - ...all the time

Find some common ground!

Programmers vs. Users



Find some common ground!

Why?

Science relies on data and its analysis.

→ use & write scientific software!

Do we know

- what we want?
 - → Mostly yes.
- how to calculate it?
 - → We are working on it.
- how to build the "tool"?
 - → Usually done "as it flows"!
- → Software/code development best practices!

Better code writing goal

Improve

- productivity of scientific programming,
- reliability of the resulting code.

- → speed up result production
- → boost confidence in results
- → ensure results reproducibility
- → increase your scientific impact

Best practices

- 1. Write programs for people, not computers
- 2. Let the computer do the work
- 3. Make incremental changes
- 4. Don't repeat yourself or others
- 5. Plan for mistakes
- 6. Optimize software only after it works correctly
- 7. Document design and purpose, not mechanics
- 8. Collaborate

Code & Document

- 1. Write programs for people
- 7. Document design and purpose, not mechanics

- Make names consistent, distinctive, and meaningful.
- Make code style, input/output and formatting consistent
- Break programs up into "simple modules"
- Document interfaces and reasons, not implementations (40% of file content!).

Code & Automatize

- 2. Let the computer do the work
- 4. Don't repeat yourself or others
- never change data manually!
- do not type commands more than once
- script code for a "re-do this" call
- turn scripts into functions
- modularize code rather than copy-pasting bits.
- re-use code instead of rewriting it.

Code & develop

- 3. Make incremental changes
- 5. Plan for mistakes
- 8. Collaborate

- use a version control system.
- put everything that has been created manually in version control.
- automated testing of the code, in part or whole (unit, integration, regression tests)
- like manuscript writing, have colleagues review the code and/or write the code together

Can we trust our tools & results?

		Data	
		Same	Different
Code	Same	Reproducible	Replicable
	Different	Robust	Generalisable

How to evaluate/optimize?

Several issues:

- Gold standard reference
 - → rely on expert manual marking?
 - → inter- & within-rater variability?
- Measure the match between prediction & GS
 - → which metric(s)?
 - → which one(s) really matter(s)?
- Parameter setting & optimization
 - → train-test split & cross-validation or independent test set.
 - → danger of overfitting & double dipping !

Data format & organization

Stick to **open data format**:

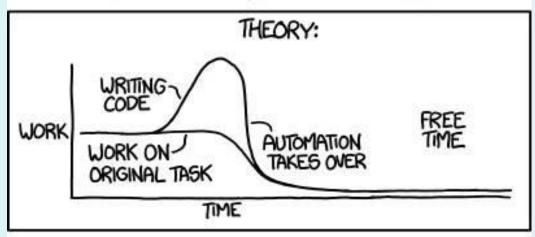
- Text & meta-data:
 - text → markdown file (.md)
 - array → tab-separated value file (.tsv)
 - key/value & structure → JSON file (.json)

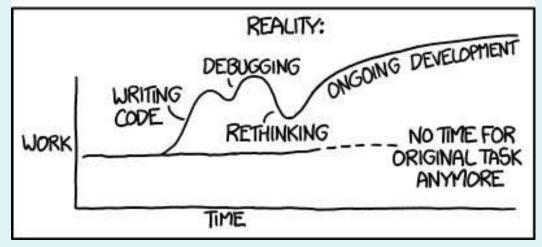
- Images: NIfTI + JSON file (.nii + .json)
 Using BIDS formatting?
- .md, .tsv & .json files should be versioned!

Thank you for your attention! Any question?

...and don't forget

"I SPEND A LOT OF TIME ON THIS TASK.
I SHOULD WRITE A PROGRAM AUTOMATING IT!"





References

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