From flop to success in academic software development

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Ínría



"Most lines of code written by programmers in academia never reach an audience" *G. Varoquaux, Nov 18th 2014*





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Technical problems: making softwareMarketing problems: unknown users

This talk [TL;DR]*

Choose your battles

projects that solve a problem

Win them

software production

* Too Long, Didn't Read



Please allow me to introduce myself

I'm a man of wealth and taste I've been around for a long, long year

Physicist gone bad Neuroscience, Machine learning

Worked in a software startup Enthought: scientific computing consulting in Python

Coder (done my share of mistake)
 Mayavi, scikit-learn, joblib...

Scipy community Chair of scipy and EuroScipy conferences

Researcher (PI) at INRIA

Software for scientific research



Reproducible science: enabling falsification

Replicating

A 3rd party redoing the work

Code and data made available

Reproducing

New analysis on different data / code coming to the same conclusion

Reusing

Applying the approach to a new problem

Let us enable reusable research

Reproducible science: enabling falsification

Replicating

A 3rd party redoing the work

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Reproducing

New analysis on different data / code coming to the same conclusion

Arguments for BSD license No strings attached

Applying the Can tinker with it

Let us enable reusable research

Reusing

Imagine a circle that contains human knowledge



Courtesy of Matt Might, via Stefan van der Waalt

By the time you finish elementary school, you know a little



Courtesy of Matt Might, via Stefan van der Waalt

High school takes you a little bit further



Courtesy of Matt Might, via Stefan van der Waalt

With a bachelors degree, you gain a speciality



Courtesy of Matt Might, via Stefan van der Waalt

A master's degree deepens this speciality



Courtesy of Matt Might, via Stefan van der Waalt

Research papers take you to the edge of human knowledge



Courtesy of Matt Might, via Stefan van der Waalt

Once you are at the boundary, you focus



Courtesy of Matt Might, via Stefan van der Waalt

You push at the boundary for a few years



Courtesy of Matt Might, via Stefan van der Waalt

And one day it yields



Courtesy of Matt Might, via Stefan van der Waalt

That dent you've made, is called a PhD

Courtesy of Matt Might, via Stefan van der Waalt

Of course, the world looks different to you now



Courtesy of Matt Might, via Stefan van der Waalt

But don't forget the big picture



Courtesy of Matt Might, via Stefan van der Waalt

This is an optimistic view



This is an optimistic view



Translationnal computational science

Computational science

The use of computers and mathematical models to address scientific research

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In medecine: bring bench science to medical practice

Translationnal computational science

Computational science

The use of computers and mathematical models to address scientific research

Translationnal science

In medecine: bring bench science to medical practice



Pick a problem to work on

Take the "easy" route

There needs to be a market screeming for the software (in academia and in industry)

Refine your vision

Pull, not push Design driven be need



Having an impact



Having an impact



Project idea

A software implementing: *i)* machine learning and *ii)* neuroimaging and *iii)* a graphical user interface and *iv)* 3D plotting

Project idea

A software implementing:

 i) machine learning
 and ii) neuroimaging
 and iii) a graphical user interface
 and iv) 3D plotting



Define project scope and vision



- Break down projects by expertise
- Don't solve hard problems
- Know the software landscape
- Don't target markets that will not yield contributors

Need a vision = elevator pitch



Define project scope and vision



■ Break down projects by expertise

- Don't solve hard problems
- Know the software landscape
- Don't target markets that will not yield contributors

Need a vision = elevator pitch

Your research (PhD) probably does not qualify ⇒ need to cherry-pick contributions Open source and community development

Code maintenance too expensive to be alonescikit-learn $\sim 300 \text{ email/month}$ nipy $\sim 45 \text{ email/month}$ joblib $\sim 45 \text{ email/month}$ mayavi $\sim 30 \text{ email/month}$

Gmail ▼ COMPOSE Inbox (53,064) Starred Starred "Hey Gael, I take it you're too busy. That's okay, I spent a day trying to install XXX and I think I'll succeed myself. Next time though please don't ignore my emails, I really don't like it. You can say, 'sorry, I have no time to help you.' Just don't ignore."

Open source and community development

Code maintenance too expensive to be alonescikit-learn $\sim 300 \, \text{email/month}$ nipy $\sim 45 \, \text{email/month}$ joblib $\sim 45 \, \text{email/month}$ mayavi $\sim 30 \, \text{email/month}$

Your "benefits" come from a fraction of the code
■ Data loading? Maybe?
■ Standard algorithms? Nah

Share the common code...

...to avoid dying under code

Code becomes less precious with time And somebody might contribute features

Community development in scikit-learn



 More than 200 contributors
 ~ 12 core contributors
 1 full-time INRIA programmer from the start



Estimated cost of development: \$ 6 millions COCOMO model, http://www.ohloh.net/p/scikit-learn

Communities: many eyes makes code fast


Having an impact

You need a community



6 steps to a community-driven project

- 1 Focus on **quality**
- 2 Build great docs and examples
- 3 Use github
- **4** Limit the technicality of your codebase
- **5** Releasing and packaging matter
- 6 Focus on your contributors, give them credit, decision power

http://www.slideshare.net/GaelVaroquaux/ scikit-learn-dveloppement-communautaire G Varoquaux



What's in a scientific-computing environment



The scientific workflow

agile

Interaction...
 → script...
 → module...
 ☆ interaction again...

Consolidation, progressively

Low tech and short turn-around times



Python, what else?

Interactive language
 Easy to read / write
 General purpose



Python, what else?

Interactive language Easy to read / write General purpose Old virtual machine / compiler Younger languages promissing (Julia) but will they get adoption beyond science?

Choose your weapons

Python, what else?

Use numpy arrays scikit-learn scikit-image

It's about plugin things together

Software architecture for science

"Scriptability" is paramount

In an application: MVC (model, view, controller)

Model	View	View		Controller		
Numerical or	Ouput:	graphs,	Input:	dialogs,		
data-processing	or files		or an A	or an API		
core	Must	enable	Avoid in	Avoid input as files:		
	headles	headless use		essive		

Dialogs should never be far from the code Dialog generation: traits, IPython widgets

Reactive programming:

dialogs modify object, and the model updates

Don't own the main

Software architecture for science

"Scriptability" is paramount

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Dialogs should never be far from the code

Dialog In Mayavi: script generation for free

Reactive programming:

dialogs modify object, and the model updates

Don't own the main

Quality is free*



* This is a book, by Philip Crosby



You need quality

Quality will give you users

Bugs give you bad rap

Quality will give you developers Contribute to learn and improve

Quality will make your developers happy People need to be proud of their work

Do less, do better

Goes against the grant-system incentive



Quality: what & how

Great documentation

Simplify, but don't dumb down

Focus on what the user is trying to solve

Great APIs

- Example-based development
- If something is hard to explain, rethink the concepts
- Limit the number of different concepts and objects
- Consistency, consistency, consistency

Good numerics

Write tests based on mathematical propertiesWhen a user finds an instability, write a new test

Quality: what & how

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- Simplify, but don't dumb down
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Great APIs

Example-based development
 If somet Quality enables reuse oncepts
 Limit the Beyond mere reproducibility
 Consistency, consistency, consistency

Good numerics

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Be productive



Be productive





Limited resources are good

Need success in the short term, not the long term

- The startup culture: fail fast Quickly identify non-viable projects
- The simpest solution that works is the best



Short cycles, limited ambitions



Keep coming back to your usersRelease early, release often

Simplicity

Complexity increase superlinearly

[An Experiment on Unit Increase in Problem Complexity, Woodfield 1979]

25% increase in problem complexity

 \Rightarrow 100% increase in code complexity



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The 80/20 rule

80% of the usecases can be solved with 20% of the lines of code

Avoid feature creep



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 \Rightarrow 100% increase in code complexity

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Avoid feature creep

Use objects sparingly

Don't use classes for the sake of it



Software engineering



Software development is an industrial process

It's time to adopt engineering practices

Amateur practices that work for small projects do not scale



Software engineering good practices

Coding convention, good naming

Version control

Use git + github

Unit testing If it's not tested, it's broken or soon will be.

Make a package, with controlled dependencies and compilation



Things we did right (maybe)



Mayavi: 3D visualization in Python

Success factors

Building upon VTK

Component model (UI)

Internals open to the world

 \Rightarrow from interaction to scripting

Great power

Limiting factors

 Building upon VTK A lot of complexity
 Codebase too complex and object-oriented (bound to VTK)
 Users of GUIs do not turn into developers

Composition is an API killer

Mayavi: 3D visualization in Python

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joblib: computational workflow patterns

Parallel for loop

>>> from joblib import Parallel, delayed
>>> Parallel(n_jobs=2)(delayed(sqrt)(i**2)
...
for i in range(8))

[0.0, 1.0, 2.0, 3.0, 4.0, 5.0, 6.0, 7.0]On-demand dispatch to ease memory consumptionThreading and processes backends

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joblib: computational workflow patterns

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[0.0, 1.0, 2.0, 3.0, 4.0, 5.0, 6.0, 7.0]

Memoize pattern

mem = joblib.Memory(cachedir='.')

- g = mem.cache(f)
- b = g(a) # computes a using f c = g(a) # retrieves results from store

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joblib: computational workflow patterns

Success factors

Simplicity of usePatterns we really, really need (pull not push)



Simplicity of use

Patterns we really, really need (pull not push)

Limiting factor

 Vision of the project unclear
 Positioning with regards to landscape unclear (parallel computing world fuzzy)
 Tricky code incide

Tricky code inside



Right project vision Machine learning without learning the machinery Black box that can be opened Right trade-off between "just works" and versatility (think Apple vs Linux) We're not going to solve all the problems for you I don't solve hard problems Feature-engineering, domain-specific cases... Python is a programming language. Use it. Cover all the 80% usecases in one package

- Right project vision
- High-level programming
 - Optimize algorithmes, not for loops
 - Know perfectly Numpy and scipy
 - Use Cython, quad not ${\rm C}/{\rm C}++$



scikit-learn: machine learning in Python

Success factors

- Right project vision
- High-level programming
- Good API design
 - separate data from operations

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- High-level programming
- Good API design
 - separate data from operations
 - Object API exposes a data-processing language



fit, predict, transform, score, partial_fit

- Instantiated without data but with all parameters



- Right project vision
- High-level programming
- Good API design
- Great community
 - Github + code review



scikit-learn: machine learning in Python

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Success factors

- Right project vision
- High-level programming
- Good API design
- Great community
- Great documentation
- **Limiting factors**
- Tricky numerical code
- **Our own success** ⇒ huge volume



G Varoquaux

Choose the project well Not all battles can be fought Make sure that there is a market Don't solve (too many) hard problems





- **1** Choose the project well
- 2 Reach a community

Users: market your project Developers: community-driven development





- **1** Choose the project well
- 2 Reach a community
- **3** Make good software

With quality, software engineering Usability matters





- **1** Choose the project well
- 2 Reach a community
- **3** Make good software



