Introduction and overview

What is "software engineering"?

 Software DESIGN: creating the high-level architecture of your code, choosing the optimal data structures and algorithms based on the specific applications and limitations

• **Software DEVELOPMENT:** writing, testing, and debugging your code, consistent with the software design that has been laid out

• **Software MAINTENANCE:** organizing, modifying, updating, and further documenting your code once it has been released

For this week...

• Software Design creating the high-level architecture of your code, choosing the optimal data structures and algorithms based on the specific applications and limitations

- **Software DEVELOPMENT:** writing, testing, and debugging your code, consistent with the software design that has been laid out
- **Software MAINTENANCE:** organizing, modifying, updating, and further documenting your code once it has been released

There is a big need for rigorous software reproducibility

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TECHNOLOGY FEATURE 24 August 2020

Challenge to scientists: does your ten-year-old code still run?

Missing documentation and obsolete environments force participants in the Ten Years Reproducibility Challenge to get creative.

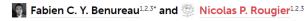
Jeffrey M. Perkel

PROTOCOLS article

Front. Neuroinform., 04 January 2018 | https://doi.org/10.3389/fninf.2017.00069



Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions



¹INRIA Bordeaux Sud-Ouest, Talence, France

EDITORIAL

Ten Simple Rules for Taking Advantage of Git and GitHub

Yasset Perez-Riverol^{1*}, Laurent Gatto², Rui Wang¹, Timo Sachsenberg³, Julian Uszkoreit⁴, Felipe da Veiga Leprevost⁵, Christian Fufezan⁶, Tobias Ternent¹, Stephen J. Eglen⁷, Daniel S. Katz⁸, Tom J. Pollard⁹, Alexander Konovalov¹⁰, Robert M. Flight¹¹, Kai Blin¹², Juan Antonio Vizcaíno^{1*}

PLOS BIOLOGY

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PERSPECTIVE

Challenges and recommendations to improve the installability and archival stability of omics computational tools

Serghei Mangul o o, Thiago Mosqueiro o, Richard J. Abdill, Dat Duong, Keith Mitchell, Varuni Sarwal, Brian Hill, Jaqueline Brito, Russell Jared Littman, Benjamin Statz , Angela Ka-Mei Lam, Gargi Dayama, Laura Grieneisen, [...], Ran Blekhman [view all]

https://doi.org/10.1038/d41586-020-02462-7 https://doi.org/10.1371/journal.pcbi.1004947 https://doi.org/10.3389/fninf.2017.00069 https://doi.org/10.1371/journal.pbio.3000333

²Institut des Maladies Neurodégénératives, Université de Bordeaux, Centre National de la Recherche Scientifique UMR 5293, Bordeaux, France

³LaBRI, Université de Bordeaux, Bordeaux INP, Centre National de la Recherche Scientifique UMR 5800, Talence, France

REPRODUCIBILITY CHECKLIST

Code Workflows based on point-and-click interfaces, such as Excel, are not reproducible. Enshrine your computations and data manipulation in code.

Document Use comments, computational notebooks and README files to explain how your code works, and to define the expected parameters and the computational environment required.

Record Make a note of key parameters, such as the 'seed' values used to start a random-number generator. Such records allow you to reproduce runs, track down bugs and follow up on unexpected results.

Test Create a suite of test functions. Use positive and negative control data sets to ensure you get the expected results, and run those tests throughout development to squash bugs as they arise.

Guide Create a master script (e.g. a 'run.sh' file) that downloads required data sets and variables, executes your workflow and provides an obvious entry point to the code.

Archive GitHub is a one option. Archiving services such as Zenodo and Software Heritage promise long-term stability.

Track Use version-control tools such as Git to record your project's history.

Package Create ready-to-use computational environments using containerization tools (for example, Docker, Singularity), web services (Code Ocean, Gigantum, Binder) or virtual-environment managers (Conda).

Automate Use continuous-integration services (ex. Travis CI) to automatically test your code over time, and in various computational environments.

Simplify Avoid niche or hard-to-install third-party code libraries that can complicate reuse.

Verify Check your code's portability by running it in a range of computing environments.

nature

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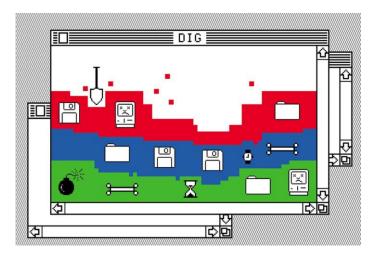
nature > technology features > article

TECHNOLOGY FEATURE | 24 August 2020

Challenge to scientists: does your ten-year-old code still run?

Missing documentation and obsolete environments force participants in the Ten Years Reproducibility Challenge to get creative.

Jeffrey M. Perkel



"Of the 43 articles they proposed reproducing, 28 resulted in reproducibility reports..."

The three pillars of the workshop...

I. Readable, robust, and reproducible code







III. Test-driven development and optimization



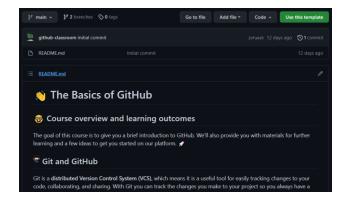


(Jacob's) coding best practices for Python



- 1. Follow the PEP8 style-guide
- 2. Write modularly
- 3. Comment extensively
- 4. Handle exceptions
- Use argparse/configparse for input handling
- 6. Include thorough and consistent documentation
- 7. Log your errors
- 8. Write informative README's





The Zen of Python

```
>>> import this
The Zen of Python, by Tim Peters
Beautiful is better than ugly.
Explicit is better than implicit.
Simple is better than complex.
Complex is better than complicated.
Flat is better than nested.
Sparse is better than dense.
Readability counts.
Special cases aren't special enough to break the rules.
Although practicality beats purity.
Errors should never pass silently.
Unless explicitly silenced.
In the face of ambiguity, refuse the temptation to guess.
There should be one -- and preferably only one -- obvious way to do it.
Although that way may not be obvious at first unless you're Dutch.
Now is better than never.
Although never is often better than *right* now.
If the implementation is hard to explain, it's a bad idea.
If the implementation is easy to explain, it may be a good idea.
Namespaces are one honking great idea -- let's do more of those!
```

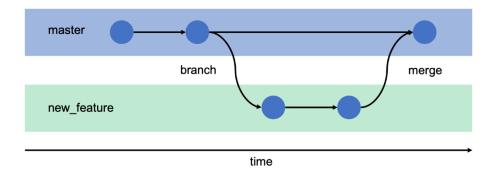
Version Control

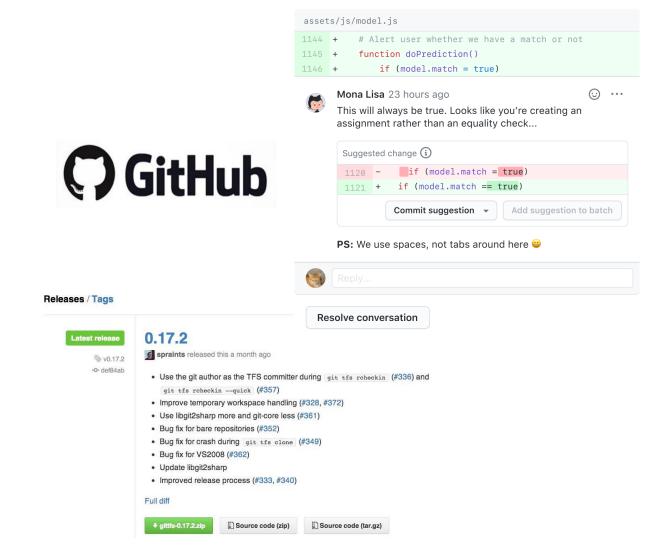
A process by which you track and document the incremental changes of your software over the course of development

```
TZ=PST8PDT git log-compact --decorate --graph -n 17 v2.6.1
   2015-09-28 ===
                  (tag: v2.6.1) Git 2.6.1
    358560 15:34 jch (tag: v2.5.4) Git 2.5.4
 * a2558fb8 15:30 jch (tag: v2.4.10) Git 2.4.10

* 6343e2f6 15:28 jch Sync with 2.3.10
                         (tag: v2.3.10, maint-2.3) Git 2.3.10
                           Merge branch 'jk/xdiff-memory-limits' into maint-2.3
                          merge-file: enforce MAX_XDIFF_SIZE on incoming files
                             Merge branch 'jk/transfer-limit-redirection' into maint-2.3
          === 2015-09-25 ===
                            http: limit redirection depth
                             http: limit redirection to protocol-whitelist
                             transport: refactor protocol whitelist code
            === 2015-09-28 ===
            df37727a 14:33 jch Merge branch 'jk/transfer-limit-protocol' into maint-2.3
        === 2015-09-23 ===
           fccbb 11:35 jk submodule: allow only certain protocols for submodule fetches
```







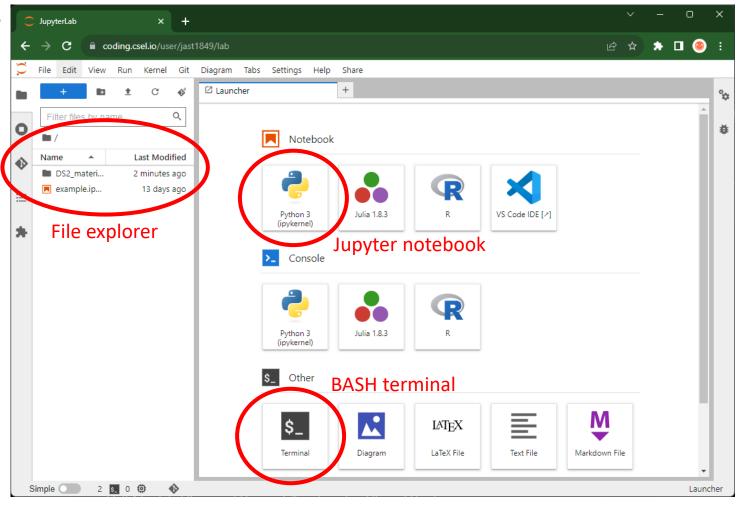
Our "dev environment"

We will use a remote environment (CSEL) provided by the CS department.

Think of this like any other linux/unix computer system (like AWS, fiji, or your laptop)

Can be accessed at:

https://coding.csel.io/





Coding @ Computer Science Education Lab

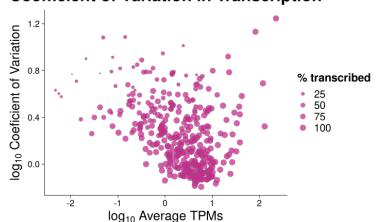
Department of Computer Science University of Colorado, Boulder

Jupyter notebooks

Enable you to seamlessly combine live code with visualizations, equations, and narrative.

```
In [10]: options(repr.plot.width=12, repr.plot.height=8)
         txpt2 <- ggplot(hg38_tpms_hermit,
                aes(x=log(mean, base=10),
                    y=log(coefvar, base=10),
                   size=percent_transcribed)) +
         ggtitle("Coefficient of Variation in Transcription") +
         xlab(expression(paste(log[10], 'Average TPMs'))) +
         ylab(expression(paste(log[10], 'Coeficient of Variation', sep=""))) +
         labs(size="% transcribed") +
         geom_point(color = 'gray40',
                    fill = 'maroon3',
                    shape = 21,
                    alpha = 0.75) +
         theme_cowplot(24) +
         theme(plot.title = element_text(hjust = 0.5),
               title = element_text(size = 34, face = "bold"),
               axis.title = element text(size = 30, face = "bold"),
               axis.text.x = element_text(size = 20),
               axis.text.y = element_text(size = 20),
              legend.title = element text(size = 26),
              legend.text = element text(size = 24))
```

Coefficient of Variation in Transcription



Normalized counts for genes in dbNascent

These are human counts for genes in human genome based on nascent trancription of 880 high QC samples.

- . Less than %50 of bidirectionals that are in TSS regions
- Sample QC 1,2 and 3

The normaliztion was performed using counts from both genes and bidirectionals as follows:

$$ext{TPM}_{ ext{i}} = rac{r_i/l_i}{\sum_1^j r_j/l_j} * 10$$

where r_i are the mapped reads for transcript i (for all genes and bidirectional transcripts), l_i is the transcript length and $\sum_j r_j / l_i$ sums all j length normalized transcripts. The ratio is multiplied by a scaling factor of 10^6 . The counts for genes was normalized over the 5' truncated transcripts (750 bp at the 5' end of gene removed).

Lastly, genes with no transcription in ALL samples were also filtered out from this table.

```
In [3]: hg38_tpms <- data.table::fread('/Users/rusi2317/projects/meta_analysis_qc/hg38/processed_data/final_co
    unts/genes_5ptrunc_bidirectionals_tpm.tsv')
    hg38_tpms$num_zeros <- rowSums(hg38_tpms[,6:885] == 0)
    hg38_tpms$percent_zero_samples <- (hg38_tpms$num_zeros/880)*100
    hg38_tpms$percent_transcribed <- 100-hg38_tpms$percent_zero_samples
    dim(hg38_tpms)
    head(hg38_tpms)</pre>
```

282165 · 896

A data.table: 6 x 896

chromosome	start	end	GeneID	Length	SRZ1950491	SRZ1950493	SRZ1950495	SRZ1950497	SRZ19!
<chr></chr>	<int></int>	<int></int>	<chr></chr>	<int></int>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
chr1	77562416	77681908	ZZZ3	119493	5.89606855	8.4220380	6.065659968	6.340954238	6.41280
chr17	4004445	4142280	ZZEF1	137836	2.99959752	3.7204037	3.492450878	3.340299643	3.75287
chr7	143382095	143391111	ZYX	9017	1.34000571	0.9263288	0.368001010	0.500336067	0.85069
chr1	52727203	52827336	ZYG11B	100134	2.28827793	2.6288455	1.661646325	1.882821794	2.26278
chr1	52843510	52894995	ZYG11A	51486	0.02560166	0.0000000	0.009207122	0.009223826	0.01146
chr3	126458901	126475169	ZXDC	16269	4.48314983	5.1963559	4.705705105	4.101245248	5.37988

You can interact with these components modularly.

Our learning objectives

1. Get comfortable using Git + GitHub

2. Get comfortable using Jupyter notebooks

3. Practice Python basics and best practices

4. Learn the basics of testing and optimization

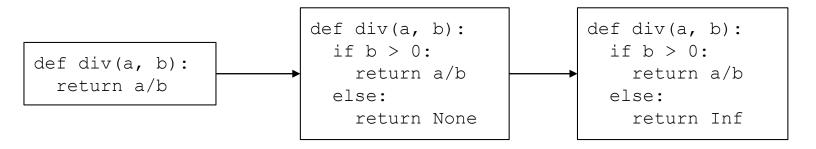
What will be covering this week?

Tentative schedule (topics subject to change)

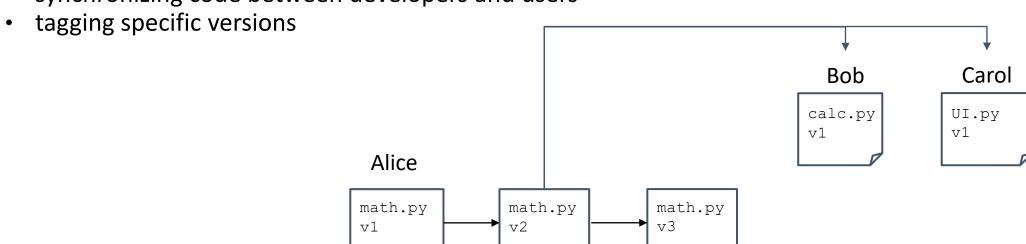
Dates	Hour	Topic	
Friday 7/28	1	Software engineering motivation/overview, CSEL, JupyterLab	
	2	Version control, Git, GitHub intro	
	3	Git, GitHub	
Monday 7/31	1	Primer (variables, data types, whitespace, operations)	
	2	Primer (functions, import, iterables, loops)	
	3	Primer (if/then, str-parse, read files/stdin, cmd line variables)	
Tuesday 8/1	1	Python best practices (PEP8/pycodestyle, modular coding, commenting/documentation)	
	2	Python best practices (arg/configparser, exception handling, error logging)	
	3	Python best practices (exception handling)	
Wednesday 8/2	1	Common packages: numpy, scipy, pandas	
	2	Unit testing	
	3	Test driven development	
Thursday 8/3	1	Optimization (timeit, getsizeof)	
	2	Optimization (cProfile)	
	3	GitHub revisited	

Version Control

Version control manages changes to files for a project



- reverting back to an old version
- allowing developers to test changes without losing the original
- synchronizing code between developers and users



Version control software







Software repository hosting company







A couple definitions

• Repository ("repo"): An archive containing all the saved states of all the files you are tracking.

 "Commit": The action of recording the state of your file(s) to the archive

Basic Git workflow

An itemized change to a file(s)

Hidden archival files that contain the files' full history

LOCAL repository

A collection of files you directly edit



COMMANDS

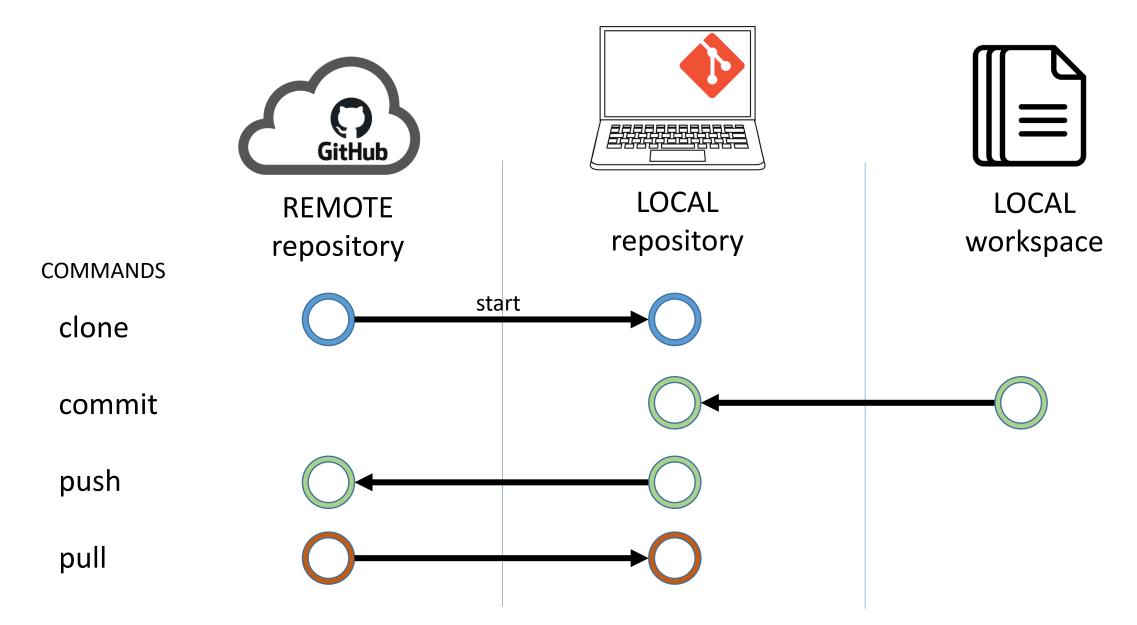
add

commit





Git + GitHub workflow

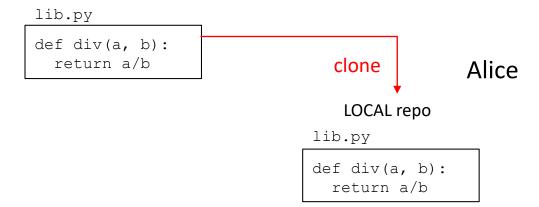


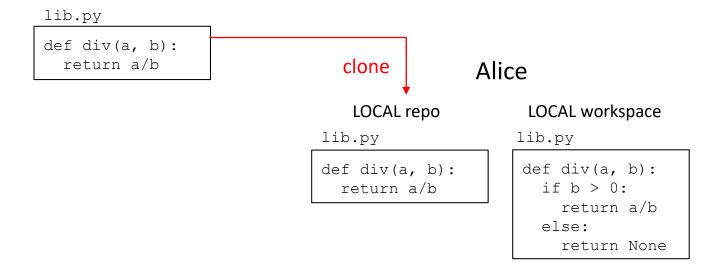


lib.py

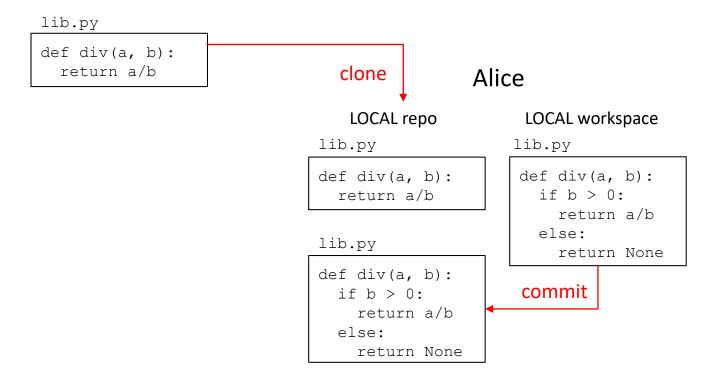
def div(a, b):
 return a/b

^{*} NOTE: we are ignoring branches for now

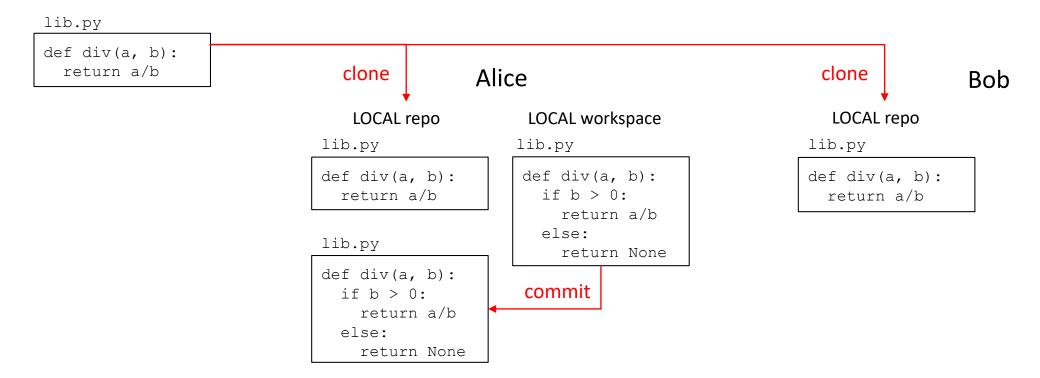


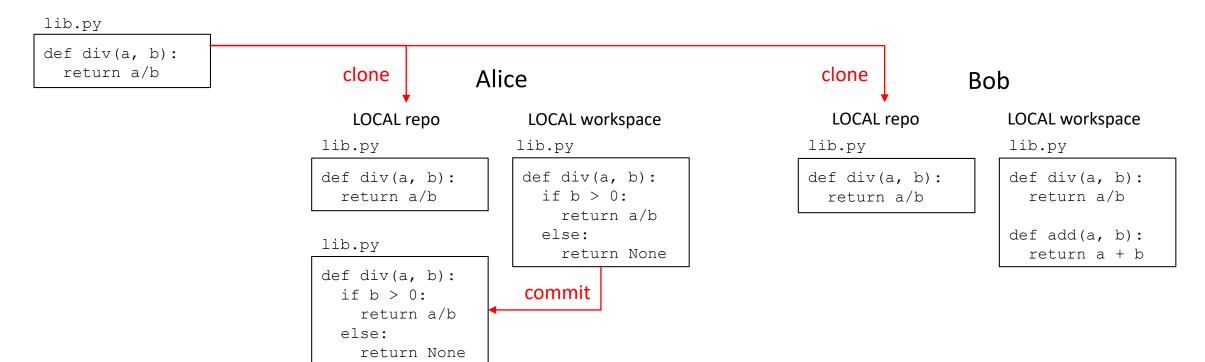


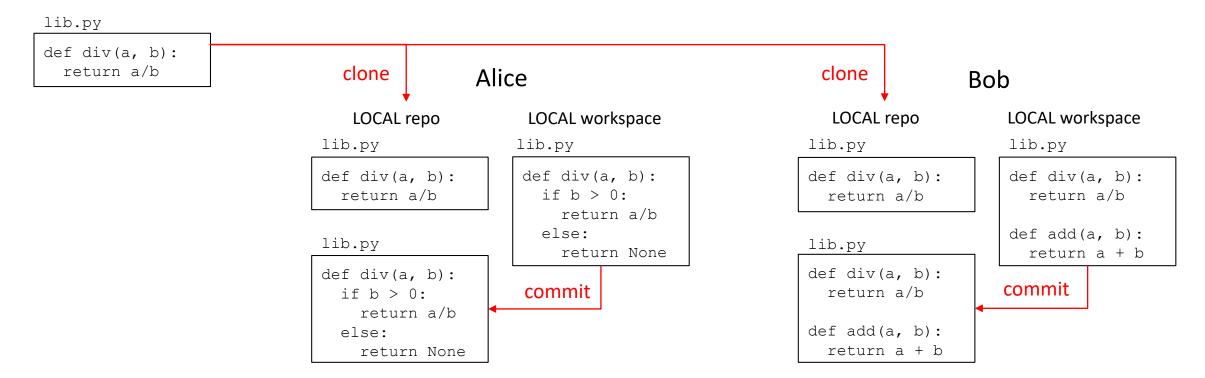
^{*} NOTE: we are ignoring branches for now

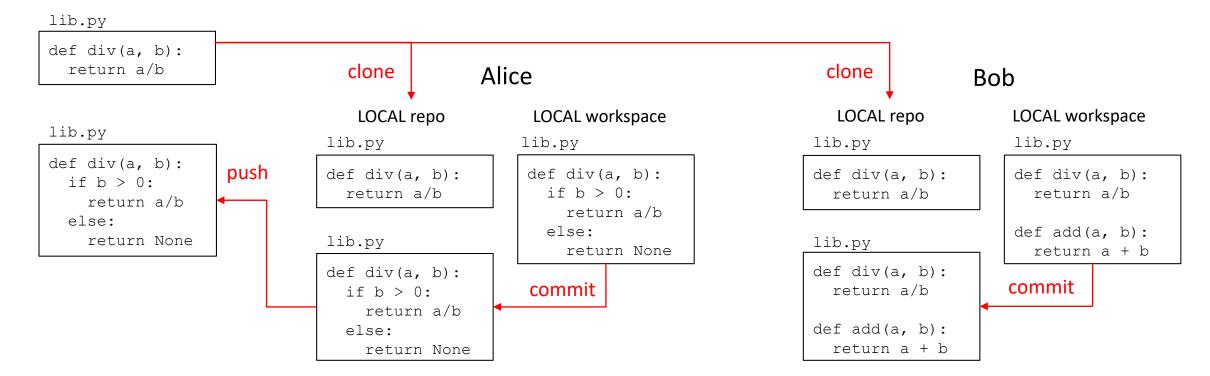


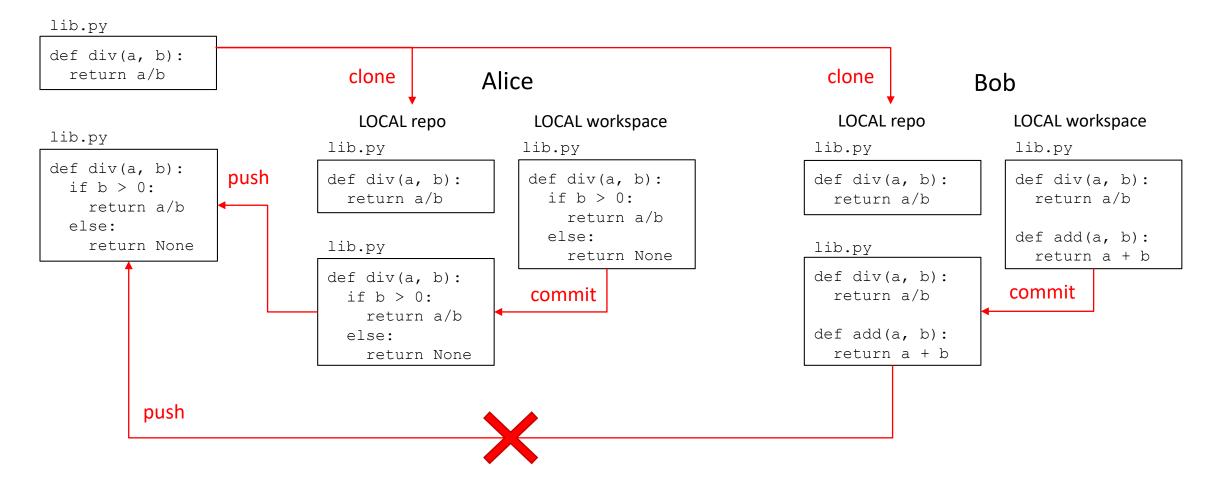
^{*} NOTE: we are ignoring branches for now



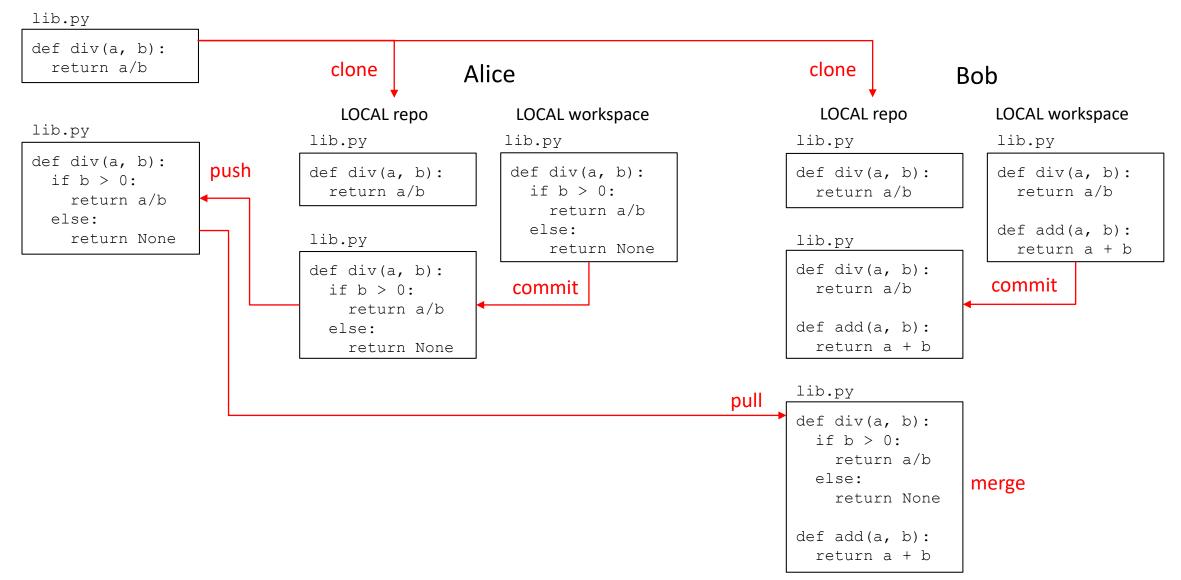




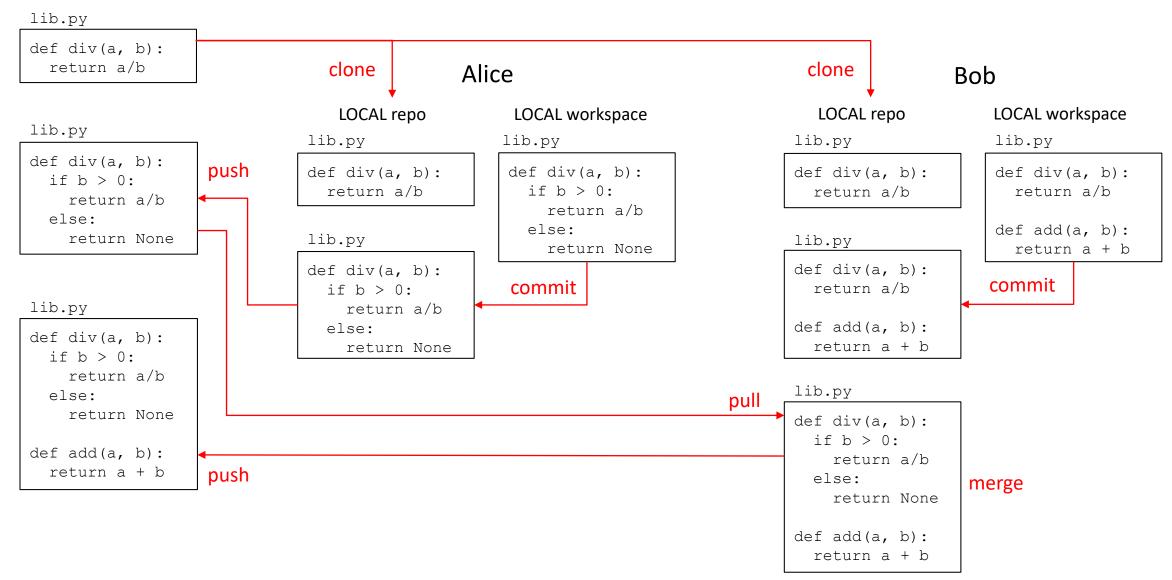




^{*} NOTE: we are ignoring branches for now

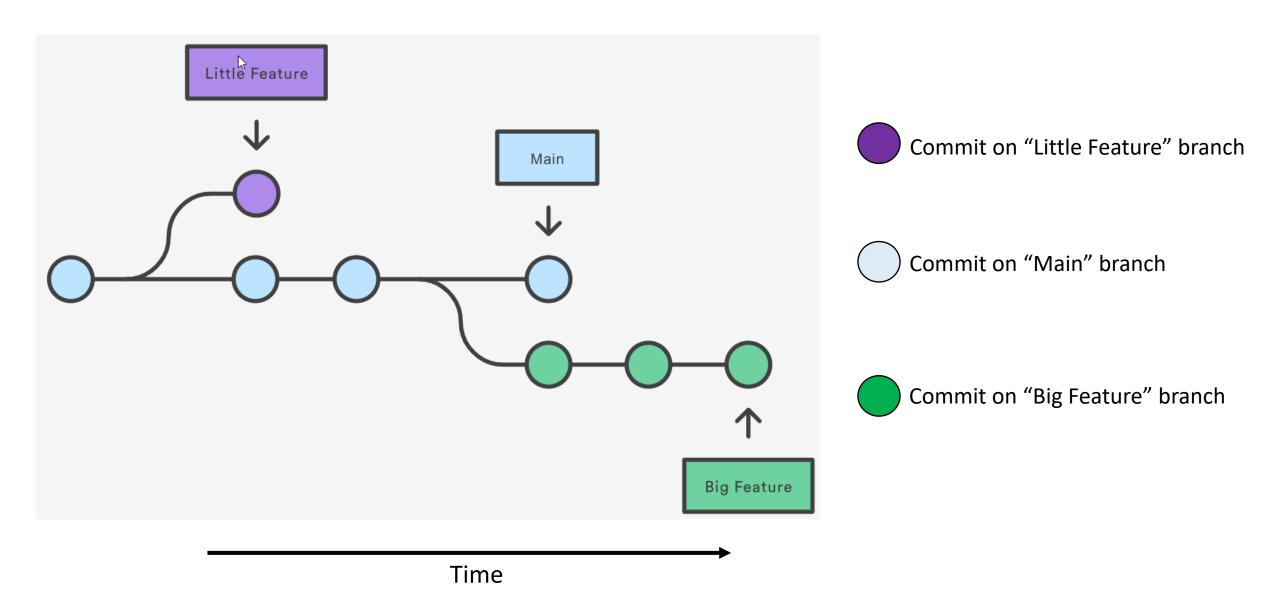


^{*} NOTE: we are ignoring branches for now



^{*} NOTE: we are ignoring branches so far

Git branching



Git branching



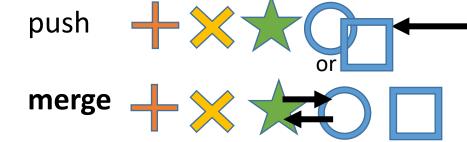
commands r

REMOTE repository

branch

checkout

commit

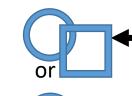




LOCAL repository

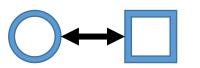








LOCAL workspace



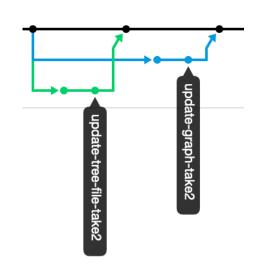


Git branch workflow

- 1. Check out the main branch (do NOT create new branches from non-main branches)
- 2. Create and checkout a new branch.
- 3. Run git push -set-upstream origin <newbranchname> to track the branch in your remote repo
- 4. Edit your files and make all the commits as usual.
- 5. Run git push -set-upstream origin <newbranchname> to upload the commits to the branch on remote
- 6. Go to the repo on GitHub and create a "pull request" to start to merge your dev branch to main.
- 7. Fix any merge conflicts if they exist.
- 8. Back on your computer, check out the main branch and run git pull, to copy the changes in the remote repo to your local repo.

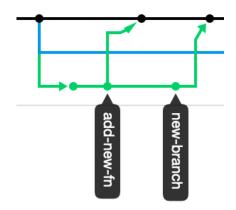
Git branching

From your GitHub repo page going to Insights \rightarrow Networks will show a graphical representation of your repo. The example at right shows the main branch in black and two dev branches (green and blue), both of which have been merged back into the main.

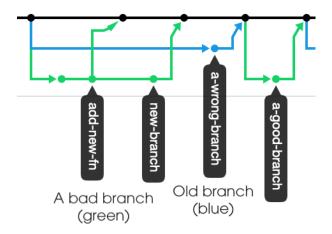


Examples of bad branching:

Creating a branch from another dev branch (instead of from main)



Creating a new branch from an older commit on main.



Markdown syntax

Header 1 ## Header 2 ### Header 3 ...

Header 1

Header 2

Header 3

```
Monospace code block:

def my_func(x):
    y = x + 3
    return y
...
```

```
**Bold text**
__Bold text__
```

```
*Italic text*
_Italic text_
```

Horizontal lines:

Enumerate and itemize

- 1. Pie
- 2. Cake
- 3. Flan

- > Block quote
- >> More indenting
- >>> And even more indenting

- Broccoli
- Carrot
- Lettuce