Data Requested...

You request data from the authors of a study you're excited about, hoping to replicate it and build your new study on top of that one's findings.

A month later, they send you a zip file with code and data.

After a week of trying to get it to run based on their instructions, you haven't gotten the code to run without errors.

Should you still begin your study? Why / Why not?
Should you request more time to search for a new direction?
How do you know, that the files you received don't actually work as-is?

Death of the Magic Machine...

After a long holiday, you go back to collecting data, only to find that your data acquisition software won’t start, and you don’t know why! It’s the only computer your department has with that program, and the lab member that originally set it up and modified it left long ago.

After a couple days of searching online, you finally find the key program online.

Should you use the program to collect your data?

How will this new change affect what you do with the existing data?

A little help...

Data collection is taking you all day, leaving little time for analysis, literature review, and your other projects. To help lighten the load, the department has offered to hire a part-time student assistant to help you out.

How much training will be needed before they are up to speed?

How similar will their new data look to your old data? Will this affect your study?

If the new data has collection-related problems (incompleteness, new artifacts, data loss, etc), will you catch them in time to retrain your assistant and get enough new data to recoup the loss?

Is it worth getting help, or should you work extra hours on the weekend to keep the status quo and hopefully still make your deadline?

Collaborating on the analysis...

You request help on the data analysis from a labmate, giving them your existing analysis scripts so they can build on it.

A month later, they come back with two figures--one you requested that shows the pattern you had hoped for, and another that is supposed to be the same analysis as the one done you did before, but there are now differences from your previous findings.

Whose code do you work on from that point on: your original files, or theirs?

Do you accept their figures for your manuscript or keep the one you made?

If you don’t accept their code, do you make them an author on your paper?

Have you tried...

After six months of data collection and analysis, you give a presentation on your results to your lab/department, and someone suggests a change in methods or a new parameter.

What do you do with that feedback?

Do you remember what parameter you used in the first place?

How much effort will it be to change your code and re-run the analysis in the first place?

Accepted, with Revisions...

The manuscript for the computational project you've been working on for 2 years is under review, but the editors want a specific new figure before accepting your paper.

How much work will it be to make the requested changes?

Is it worth submitting to a more-lenient journal?

Do you write a rebuttal letter, explaining that it is too much work or claim that the suggestion is out of scope for your study?

Data Analysis Requested...

Your paper is finally accepted! You move on to a new laboratory and continue doing amazing work, but one day you receive an email requesting the code and data for your prior work.

Do you know where it all is?

How much effort will it be to find it?

How much of it needs to be modified before sending it out?

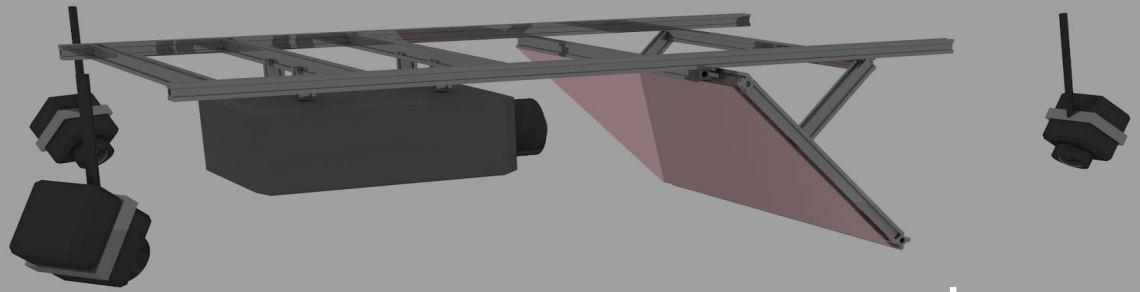
How long does it take for you to reply to the email?

Will it matter if you don’t reply at all?

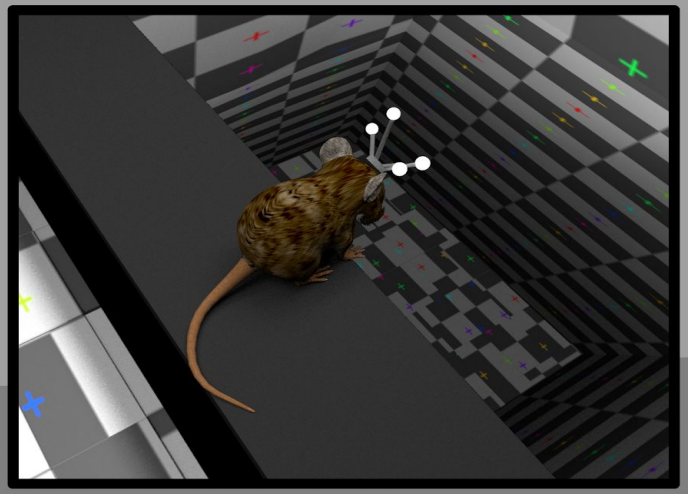
Scientific DevOps:
Designing Reproducible Data Analysis Pipelines with Containerized Workflow Managers

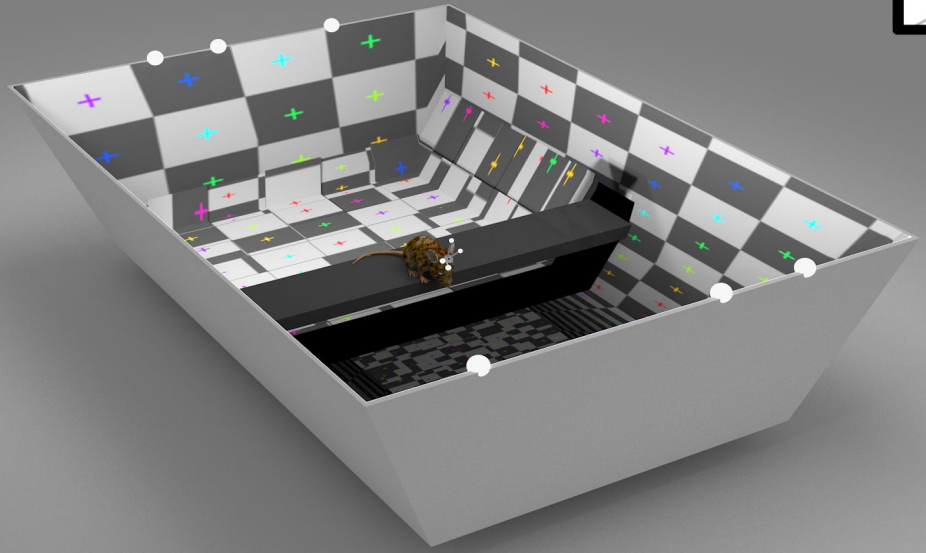
Nicholas Del Grosso
EuroScipy 2019
Bilbao, Spain

These Slides Available at:
github.com/nickdelgrosso/devops_talk_euroscipy2019



Closed-Loop Research and Software Development

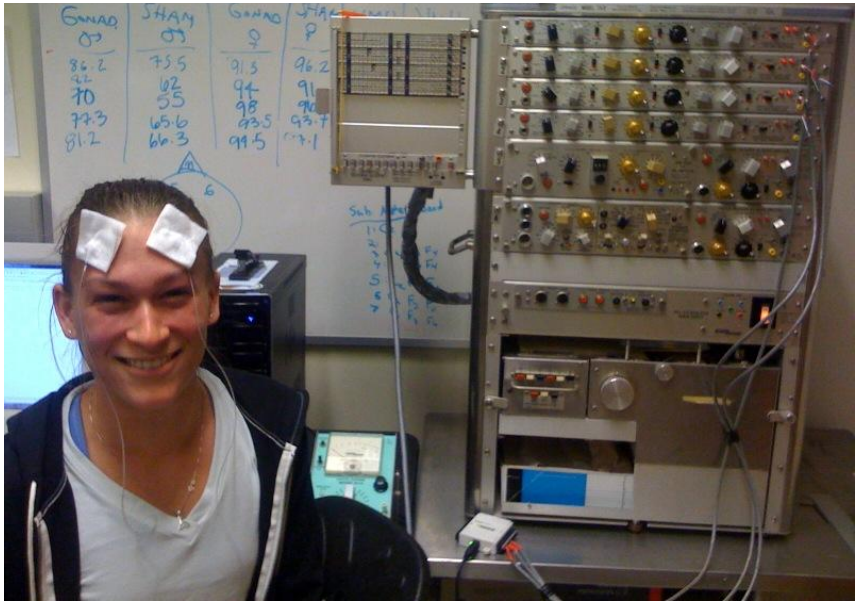




Early Inspiration

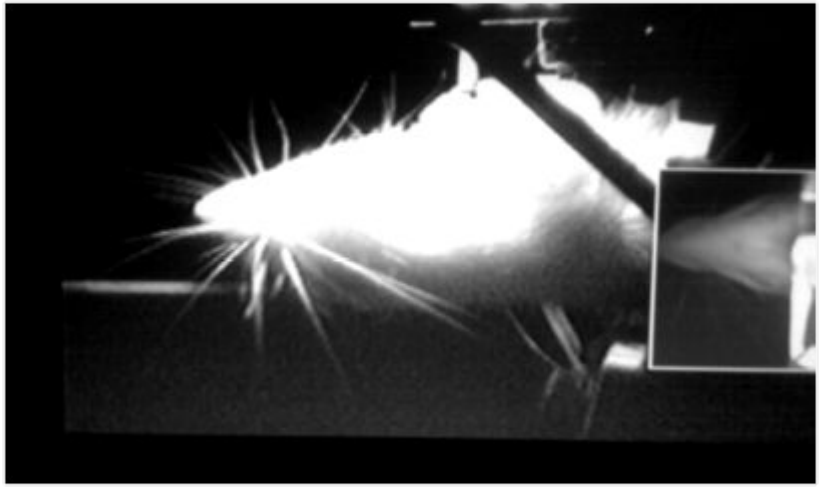
“Teenager moves video icons
just by imagination.”

-Leuthard and Blakely, 2006



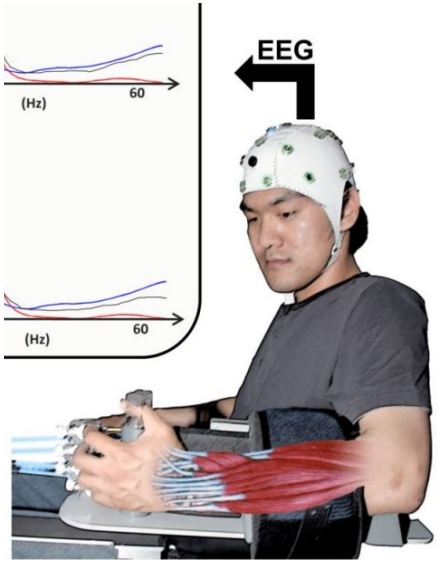
“College Student Learns MATLAB and LabVIEW to build DAQ System for EEG-based BCI.
Fails to ground 20-year-old recording equipment.”

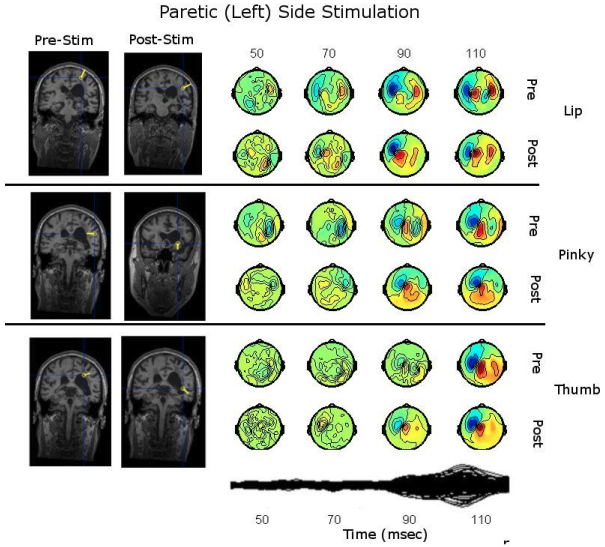
-Nick Del Grosso, 2009

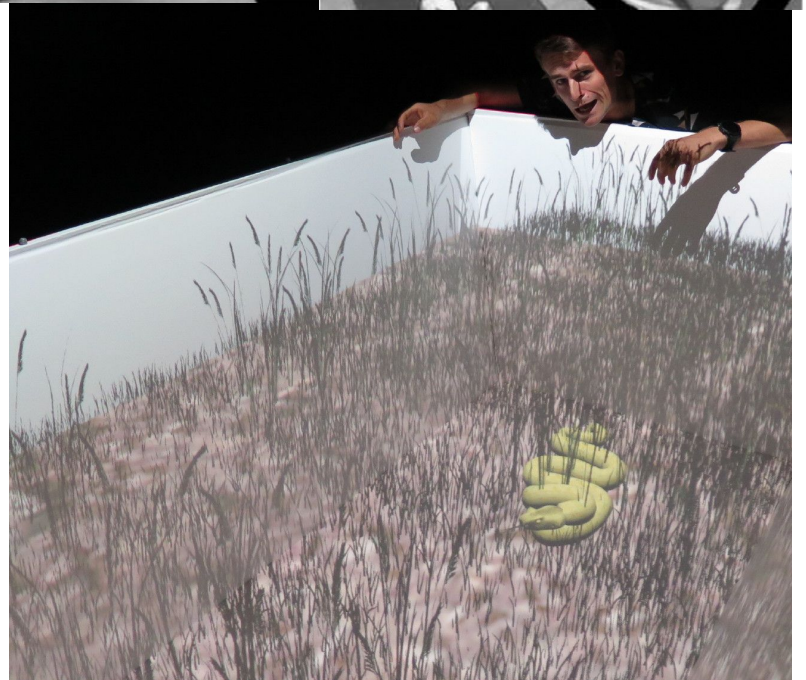



7a


7c


EEG
(Hz)
60
(Hz)
60


Paretic (Left) Side Stimulation
Pre-Stim
Post-Stim
50
70
90
110
Pre
Post
Lip
Pre
Post
Pinky
Pre
Post
Thumb
50
70
90
110
Time (msec)



Nicholas A. Del Grosso

nickdelgrosso

A Passionate Neuroscientist and Pythonista

Munich, Germany

Sign in to view email

PRO

Block or report user

Organizations

Overview | Repositories 125 | Projects 0 | Stars 147 | Followers 70 | Following 26

Pinned

ratcave/ratcave

A Simple Python 3D Graphics Engine extension for Pyglet, Psychopy, and PyGame.

Python ★ 58 ⑂ 19

ratcave/natnetclient

Python NatNet Client for retrieving NaturalPoint's Optitrack da
Motive software using depacketization.

Python ★ 8 ⑂ 10

MS_Booker

A simple web app for booking mass spec machines

Python ★ 1 ⑂ 2

helptranslator

Translates the Python help text to any language, for helping w
programming learning.

Python ★ 6

ratcave/wavefront_reader

A basic Wavefront .obj and .mtl reader module for Python

Python ★ 7 ⑂ 7

dzne_python_workshop

Jupyter Notebook ⑂ 1

424 contributions in the last year

My Latest Concern: The Replication Crisis

Essay

Why Most Published R
Are False

John P. A. Ioannidis

Pow
size
neu

Katherine S. Button[1,2], John P. A. Ioannidis[3], Claire Mokrysz[1], Brian A.
Jonathan Flint[5], Emma S. J. Robinson[6] and Marcus R. Munafò[1]

Abstract | A study with low statistical power has a reduced chance of detecti

False-Positive Psychology: Undisclosed Flexibility in Data Collection and Analysis Allows Presenting Anything as Significant

Joseph P. Simmons[1], Leif D. Nelson[2], and Uri Simonsohn[1]

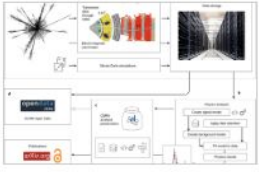
And More Recently: The Reproducibility Crisis

PERSPECTIVE
OPEN ACCESS
15 NOV 2018
Nature Physics

Open is not enough

The solutions adopted by the high-energy physics community to foster reproducible research are examples of best practices that could be embraced more widely. This first experience suggests that reproducibility requires going beyond openness.

Xiaoli Chen, Sünje Dallmeier-Tiessen ··· Sebastian Neubert



WORLD VIEW
24 MAY 2018
Nature

Before reproducibility must come preproducibility

Instead of arguing about whether results hold up, let's push to provide enough information for others 1

Philip B. Stark

EDITORIAL
18 APR 2018
Nature

Checklists work to improve science

Nature authors say a reproducibility checklist is a step in the right direction, but more needs to be done.

WORLD VIEW
16 MAY 2018
Nature

Give every paper a read for reproducibility

I was hired to ferret out errors and establish routines that promote rigorous research, says Catherine Winchester.

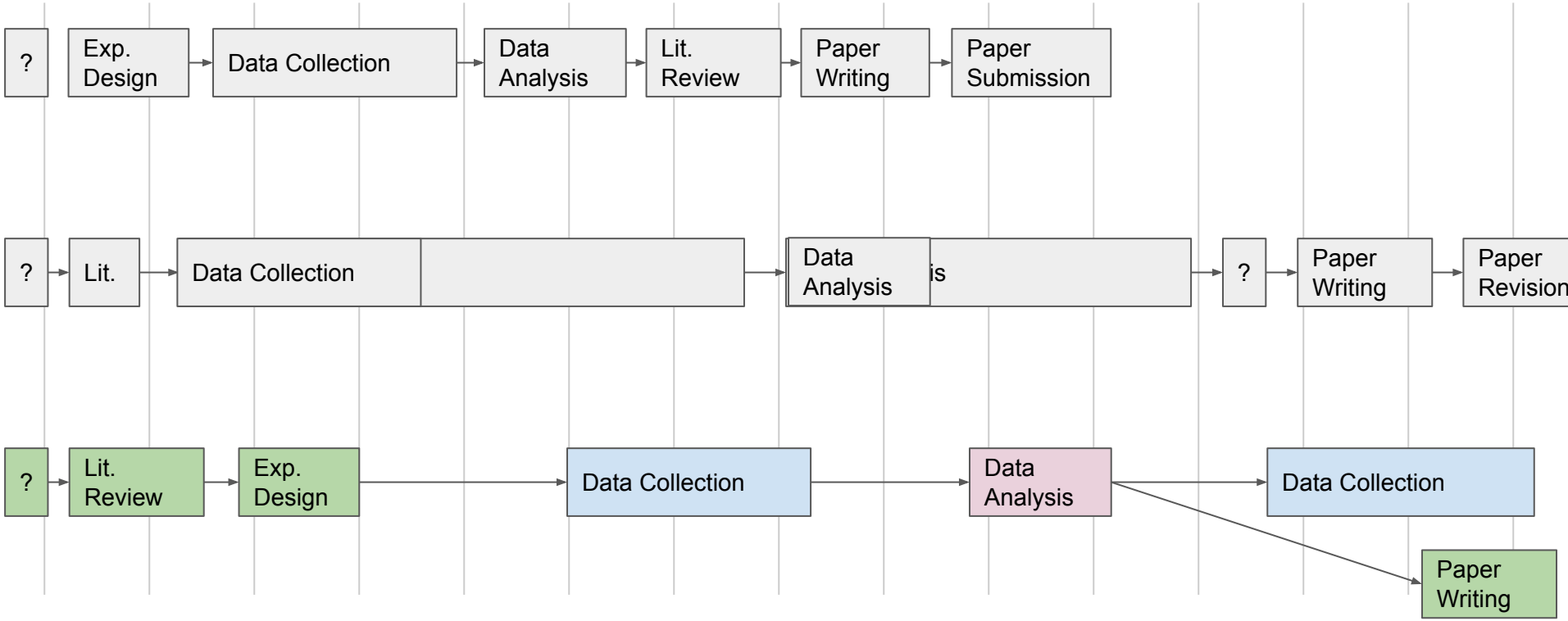
Catherine Winchester

COMMENT
24 AUG 2017
Nature

A long journey to reproducible results

Replicating our work took four years and 100,000 worms but brought surprising discoveries, explain Gordon J. Lithgow, Monica Driscoll and Patrick Phillips

Wastes in of Lean Product Development
(Effects of Waterfalls and Siloed Knowledge)



DevOps is...

- Management principles adapted from the Lean community to software deployment: an application of the scientific method, queuing theory, and systems design to productivity.
- A set of software tools to help a team work toward constant, usable learning and build while maintaining a productive environment.
- The recognition that software development is an integral part of an organization.
- the observation that much wasted effort is produced by short-sightedness within individual steps in a pipeline.
- An observation that all steps contribute toward the achievement of some goal.

Version Control: Essential Tool #1

- Takes snapshots of a folder, creating a timeline of changes to that folder.
- Has tools for copying that timeline and making alternative timelines.
- Asks users to describe, in their own words, the point of each snapshot.

Version Control System

Code Hosting Website & Social Network

Code Hosting Tool

Scientific DevOps:
Designing Reproducible Data Analysis Pipelines with Containerized Workflow Managers

Nicholas Del Grosso
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These Slides Available at:
github.com/nickdelgrosso/devops_talk_euroscipy2019

The Three “Ways” of DevOps

1. Build Smooth Forward Flow
2. Increase Feedback
3. Foster an Environment of Continuous Improvement

Building Smooth, Forward Flow

The First Way

Data ------> Figures ----> Reader

Ambiguity ----------> Certainty

- MyProject
  - README.txt

README.txt

My Project
This project makes histograms of my awesome data!

My Project
This project makes histograms of my awesome data!

- MyProject
  - README.md

README.md

```
# My Project

This project makes histograms of my awesome data!

## Run the Analysis

```python
import pandas as pd
df = pd.read_csv("mydata1.csv")
fig = df.plot.hist()
fig.savefig("myresults.png")
```
```

My Project

This project makes histograms of my awesome data!

Run the Analysis

```
import pandas as pd
df = pd.read_csv("mydata1.csv")
fig = df.plot.hist()
fig.savefig("myresults.png")
```

```
cat README.md | codedown python | python
```

- MyProject
  - README.md
  - get_results.py

get_results.py

```
import pandas

read_csv
plot_histogram
savefig
```

My Project

This project makes histograms of my awesome data!

Run the Analysis

```
python get_results.py
```

- MyProject
  - README.md
  - get_results.py

My Project

This project makes histograms of my awesome data!

Install the Dependencies

```
pip install pandas
pip install numpy
pip install matplotlib
```

Run the Analysis

```
python get_results.py
```

- MyProject
  - README.md
  - get_results.py
  - install_requirements.sh

install_requirements.sh

```
pip install pandas
pip install numpy
pip install matplotlib
```

My Project

This project makes histograms of my awesome data!

Install the Dependencies

```
bash install_requirements.sh
```

Run the Analysis

```
python get_results.py
```

- MyProject
  - README.md
  - get_results.py
  - requirements.txt

requirements.txt

```
pandas
matplotlib
numpy
```

My Project

This project makes histograms of my awesome data!

Install the Dependencies

```
pip install -r requirements.txt
```

Run the Analysis

```
python get_results.py
```

- MyProject
  - README.md
  - get_results.py
  - requirements.txt
  - requirements.lock

requirements.lock

```
cycler==0.10.0
kiwisolver==1.1.0
matplotlib==3.1.1
numpy==1.17.1
pandas==0.25.1
pyparsing==2.4.2
python-dateutil==2.8.0
pytz==2019.2
six==1.12.0
```

My Project

This project makes histograms of my awesome data!

Install the Dependencies

```
pip install -r requirements.lock
```

Run the Analysis

```
python get_results.py
```

Developers

Save the current versions of the packages:

```
pip freeze > requirements.lock
```

- MyProject
  - README.md
  - get_results.py
  - requirements.txt
  - requirements.lock

- (Virtual Environments)
  - myproject
    - bin
      - python
      - activate

My Project

This project makes histograms of my awesome data!

Install the Dependencies

```
conda create --name myproject python=3.7
conda activate myproject
pip install -r requirements.lock
```

Run the Analysis

```
conda activate myproject
python get_results.py
```

Developers

Save the current versions of the packages:

```
pip freeze > requirements.lock
```

- MyProject
  - README.md
  - get_results.py
  - environment.yml
  - requirements.lock

My Project

This project makes histograms of my awesome data!

Install the Dependencies

```
conda create --name myproject python=3.7
conda activate myproject
conda install --file environment.yml
```

Run the Analysis

```
conda activate myproject
python get_results.py
```

Developers

Save the current versions of the packages:

```
conda list --export  > requirements.lock
```

- MyProject
  - README.md
  - get_results.py
  - Pipfile
  - Pipfile.lock

My Project

This project makes histograms of my awesome data!

Install the Dependencies

```
pip install pipenv
pipenv install
```

Run the Analysis

```
pipenv shell
python get_results
```

```
cd MyProject
pipenv --python 3.7
pipenv install pandas
```

- MyProject
  - README.md
  - get_results.py
  - Pipfile
  - Pipfile.lock

My Project

This project makes histograms of my awesome data!

Install the Dependencies

```
pip install pipenv
pipenv install
```

Run the Analysis

```
pipenv run get_results.py
```

```
cd MyProject
pipenv --python 3.7
pipenv install pandas
pipenv lock
```

- MyProject
  - README.md
  - get_results.py
  - pyproject.toml
  - poetry.lock

pyproject.toml

```
[tool.poetry]
name = "MyProject"
version = "0.1.0"
description = ""
authors = ["Nicholas A. Del Grosso
<delgrosso.nick@gmail.com>"]

[tool.poetry.dependencies]
python = "^3.6"
pandas = "^0.25.1"

[tool.poetry.dev-dependencies]
pytest = "^3.0"
```

My Project

This project makes histograms of my awesome data!

Install the Dependencies

```
pip install poetry
poetry install
```

Run the Analysis

```
poetry shell
python get_results.py
```

```
poetry MyProject
cd MyProject
pipenv add pandas
```

- MyProject
  - README.md
  - get_results.py
  - pyproject.toml
  - poetry.lock

pyproject.toml

```
[tool.poetry]
name = "MyProject"
version = "0.1.0"
description = ""
authors = ["Nicholas A. Del Grosso
<delgrosso.nick@gmail.com>"]

[tool.poetry.dependencies]
python = "^3.7"
pandas = "^0.25.1"

[tool.poetry.dev-dependencies]
pytest = "^3.0"
```

My Project

This project makes histograms of my awesome data!

Install the Dependencies

```
pip install poetry
poetry install
```

Run the Analysis

```
poetry run python get_results.py
```

- MyProject
  - README.md
  - get_results.py
  - pyproject.toml
  - poetry.lock

pyproject.toml

```
[tool.poetry]
name = "MyProject"
version = "0.1.0"
description = ""
authors = ["Nicholas A. Del Grosso
<delgrosso.nick@gmail.com>"]

[tool.poetry.dependencies]
python = "^3.7"
pandas = "^0.25.1"

[tool.poetry.dev-dependencies]
pytest = "^3.0"
```

My Project

This project makes histograms of my awesome data!

Install the Dependencies

(Runs on Ubuntu 16.4. I haven't tested it on Windows.)

```
pip install poetry
poetry install
```

Run the Analysis

```
poetry run python get_results.py
```

- MyProject
  - README.md
  - get_results.py
  - pyproject.toml
  - poetry.lock
  - Singularity

Dockerfile

```
FROM python:3.7-slim

WORKDIR /MyProject

COPY . /MyProject

RUN pip install poetry
RUN poetry install

CMD ['bash']
```

Singularity Recipe

```
Bootstrap: docker
From: python:3.7-slim

%files
  ./pyproject.toml .
  ./poetry.lock .

%post
  pip install poetry
  poetry install
```

- MyProject
  - README.md
  - get_results.py
  - pyproject.toml
  - poetry.lock
  - Singularity
  - myproject.simg

My Project

This project makes histograms of my awesome data!

Install the Dependencies

1. install singularity

Run the Analysis

```
singularity shell myproject.simg
python scripts/get_results.py
```

Building the Singularity Image

```
sudo singularity build myproject.simg Singularity
```

- MyProject
  - README.md
  - get_results.py
  - pyproject.toml
  - poetry.lock
  - Singularity
  - myproject.simg

My Project

This project makes histograms of my awesome data!

Install the Dependencies

1. install singularity

Run the Analysis

singularity run myproject.simg

```
...

%runscript
  python get_results.py
```

- MyProject
  - README.md
  - get_results.py
  - pyproject.toml
  - poetry.lock
  - Singularity
  - myproject.simg

My Project

This project makes histograms of my awesome data!

Install the Dependencies

1. install singularity
2. Get the data.

Run the Analysis

```
singularity run myproject.simg
```

- MyProject
  - README.md
  - get_results.py
  - pyproject.toml
  - poetry.lock
  - Singularity
  - myproject.simg

- /data
  - raw
    - d1.csv
    - d2.csv

RENKU

Collaborative Data Science

Talk to Chandrasekhar Ramakrishnan and Rok Roškar for more info, here at the conference!

GIN

Modern Research Data Management for Neuroscience

Quilt

Manage data like code

Quilt versions and deploys data

- MyProject
  - README.md
  - get_results.py
  - pyproject.toml
  - poetry.lock
  - Singularity
  - myproject.simg

- /data
  - raw
    - d1.csv
    - d2.csv

My Project

This project makes histograms of my awesome data!

Install the Dependencies

1. install singularity

Run the Analysis

```
singularity run -B DataFolder:/data myproject.simg
```

- MyProject
  - README.md
  - get_results.py
  - pyproject.toml
  - poetry.lock
  - Singularity
  - myproject.simg

- /data
  - raw
    - d1.csv
    - d2.csv

My Project

This project makes histograms of my awesome data!

Install the Dependencies

1. install singularity

Run the Analysis

singularity run myproject.simg

```
...

%files
  DataFolder /data
```

- MyProject
  - **README.md**
  - **pyproject.toml**
  - **poetry.lock**
  - **Singularity**
  - **myproject.simg**
  - scripts
    - get_results.py

- /data
  - raw
    - d1.csv
    - d2.csv

My Project

This project makes histograms of my awesome data!

Install the Dependencies

1. install singularity

Run the Analysis

```
singularity run myproject.simg
```

```
...

%files
  DataFolder /data
```

Shrinking Your Batch Size

The First Way: Part 2

```
raw_files = glob("data/raw/data_*.csv")
processed_files = []

for raw_file in raw_files:
    session = path.basename(path.splitext(raw_file)[0]).split('_')[-1]

    # Preprocess Data
    df = pd.read_csv(raw_file)
    df2 = do_process(df)
    os.makedirs("data/processed", exist_ok=True)
    process_file = f"data/processed/data_{session}.h5"
    df2.to_hdf(processed_file, '/')
    processed_files.append(processed_file)

    # Make histograms
    fig = df2["Variable"].plot.hist()
    os.makedirs("results/hists", exist_ok=True)
    fig.savefig(f"results/hists/hist_{session}.svg")

# Make Figure1
df_all = pd.concat([pd.read_hdf(f, '/') for f in processed_files])
os.makedirs("results", exist_ok=True)
df_all.plot(x='Time', y='Happiness').savefig("results/figure1.svg")
```

get_results.py

```
raw_files = glob("data/raw/data_*.csv")
processed_files = []

for raw_file in raw_files:
    session = path.basename(path.splitext(raw_file)[0]).split('_')[-1]

    # Preprocess Data
    df = pd.read_csv(raw_file)
    df2 = do_process(df)
    os.makedirs("data/processed", exist_ok=True)
    process_file = f"data/processed/data_{session}.h5"
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# Make Figure1
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os.makedirs(“results”, exist_ok=True)
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```

get_results.py

```
# Preprocess Data
for raw_file in glob("data/raw/data_*.csv"):
    session = path.basename(path.splitext(raw_file)[0]).split('_')[-1]
    df = pd.read_csv(raw_file)
    df2 = do_process(df)
    os.makedirs("data/processed", exist_ok=True)
    processed_file = f"data/processed/data_{session}.h5"
    df2.to_hdf(processed_file, '/')

# Make Histograms
for processed_file in glob("data/processed/data_*.h5"):
    session = path.basename(path.splitext(raw_file)[0]).split('_')[-1]
    df = pd.read_hdf(processed_file, '/')
    fig = df["Variable"].plot.hist()
    os.makedirs("results/hists", exist_ok=True)
    fig_file = f"results/hists/hist_{session}.svg"
    fig.savefig(fig_file)

# Make Figure 1
files = glob("data/processed/data_*.h5")
df_all = pd.concat([pd.read_hdf(f, '/') for f in  files])
fig1 = df_all.plot(x='Time', y='Happiness')
os.makedirs("results", exist_ok=True)
fig1.savefig("results/figure1.svg")
```

get_results.py

```
# Preprocess Data
inputs =  [f"/data/raw/data_{session}.csv" for session in sessions]
outputs = [f"/data/processed/data_{session}.h5" for session in sessions]
os.makedirs("/data/processed", exist_ok=True)
for input, output in zip(inputs, outputs):
    df = pd.read_csv(input)
    df2 = do_process(df)
    df2.to_hdf(output, '/')

# Make Histograms
inputs =  [f"/data/processed/data_{session}.h5" for session in sessions]
outputs = [f"results/hists/hist_{session}.svg" for session in sessions]
os.makedirs("results/hists", exist_ok=True)
for input, output in zip(inputs, outputs):
    df = pd.read_hdf(input, '/')
    fig = df["Variable"].plot.hist()
    fig.savefig(output)

# Make Figure1
inputs =  [f"/data/processed/data_{session}.h5" for session in sessions]
output =  "results/figure1.svg"
os.makedirs("results", exist_ok=True)
df_all = pd.concat([pd.read_hdf(input, '/') for input in inputs])
df_all.plot(x='Time', y='Happiness').savefig(output)
```

get_results.py

```
sessions = [1, 2, 3]

rule process_data:
    input:  "/data/raw/data_{session}.csv"
    output: "/data/processed/data_{session}.h5"
    run:
        df = pd.read_csv(input)
        df2 = do_process(df)
        df2.to_hdf(output, '/')

rule build_histograms:
    input:  "/data/processed/data_{session}.h5"
    output: "results/hists/hist_{session}.svg"
    run:
        df = pd.read_hdf(input, '/')
        fig = df["Variable"].plot.hist()
        fig.savefig(output)

rule figure1:
    input: expand("/data/processed/data_{session}.h5", session=sessions)
    output: "results/figure1.svg"
    run:
        df_all = pd.concat([pd.read_hdf(input, '/') for input in inputs])
        df_all.plot(x='Time', y='Happiness').savefig(output)
```

Snakefile

snakemake

```
pip install snakemake
snakemake figure1
```

```
sessions = [1, 2, 3]

rule process_data:
    input:  "/data/raw/data_{session}.csv"
    output: "data/processed/data_{session}.h5"
    script:  “scripts/process_data.py”

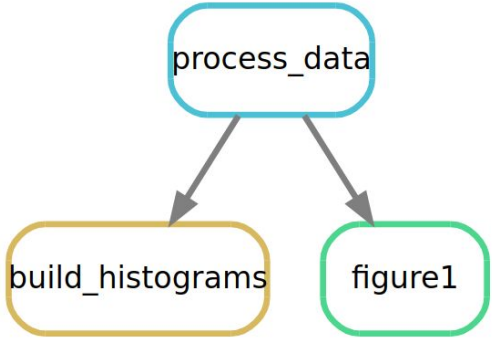
rule build_histograms:
    input:  "/data/processed/data_{session}.h5"
    output: "results/hists/hist_{session}.svg"
    script:  “scripts/plot_histogram.py”

rule figure1:
    input: expand("/data/processed/data_{session}.h5", session=sessions)
    output: "results/figure1.svg"
    script:  “scripts/plot_figure1.py”
```

scripts/process_data.py

```
import pandas as pd

filename = snakemake.input[0]
df = pd.read_csv(filename)
df2 = do_process(df)
df2.write_csv(snakemake.output[0])
```



- MyProject
  - README.md
  - requirements.txt
  - Singularity
  - myproject.simg
  - .gitignore
  - **Snakefile**
  - scripts
    - process_data.py
    - plot_histogram.py
    - plot_figure1.py
  - results
- /data
  - raw
    - data_1.csv
    - data_2.csv

My Project

This project makes histograms from my data!

Before You Begin

1. install singularity

Running the Analysis Script

Run the Singularity App, connecting its /data folder to the folder you downloaded the data to.

```
singularity run myproject.simg
```

```
%runscript
  snakemake figure1
```

Of Courses, It Can Get Much More Complex!

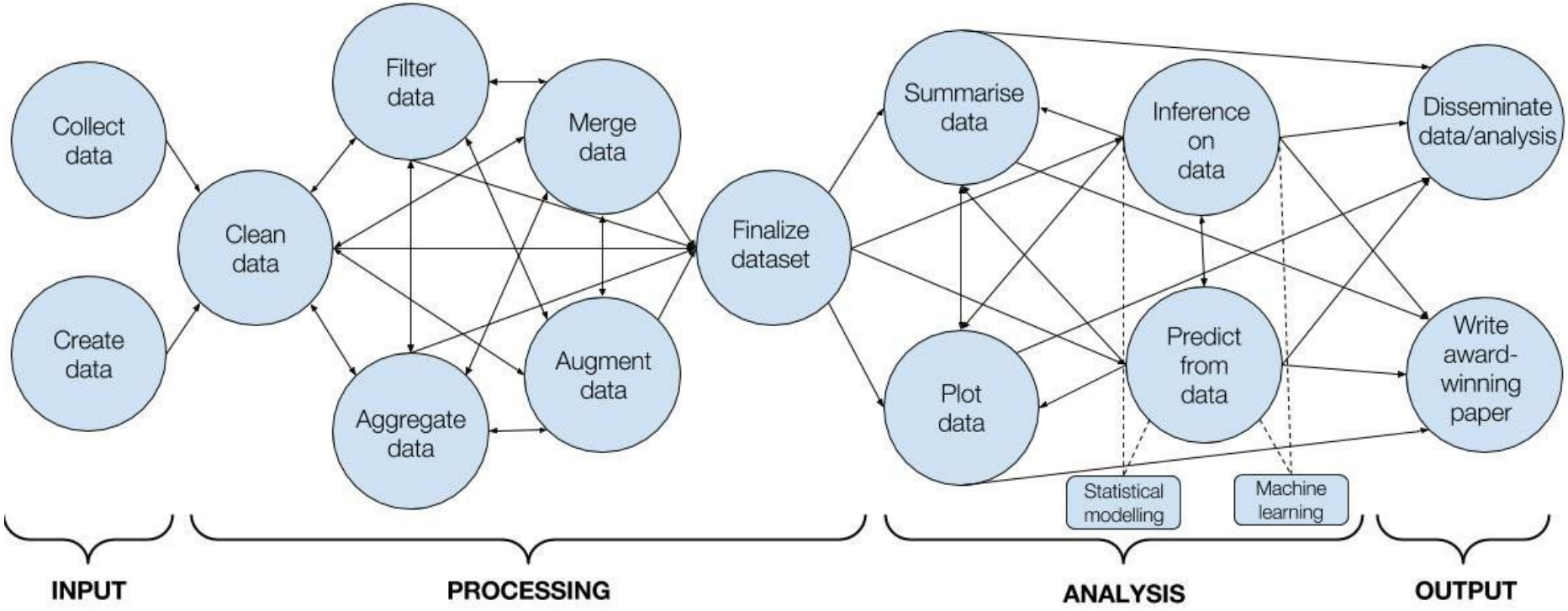


Image: "Data Science as a DAG" by Soubhik Barari

```
def task_dot():
    return {
        'file_dep': ['requests.models.deps'],
        'targets': ['requests.models.dot'],
        'actions': [module_to_dot],
    }
```

```
rule process_data:
    input:  "data/raw/data_{session}.csv"
    output: "data/processed/data_{session}.h5"
    script:  process_data.py
```

```
dag = DAG('tutorial')

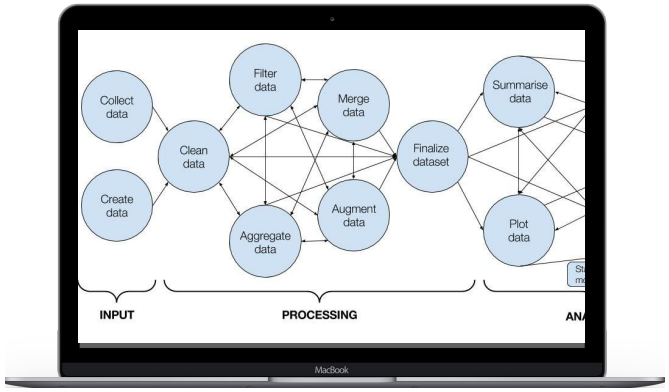
t1 = BashOperator(ttask_id='print_date',
    bash_command='date',  dag=dag)

t2 = BashOperator(task_id='sleep',
    bash_command='sleep 5', dag=dag)

t2.set_upstream(t1)
```

Because Workflow Managers Build Task Graphs, They Can Do:

- Partial Processing
- Parallel Processing
- Multi-Language Gluing
- Auto-Retries
- Remote Data Connections
- Workflow Monitoring
- Workflow Scheduling
- ...and much more!



SCIENTIFIC FILESYSTEM

```
%apphelp figure1
  Figure 1 from my paper,
showing how time spent with
Python increases your
happiness!

%apprun figure1
  snakemake figure1

%appinstall notebook
  pip install jupyterlab

%appinstall notebook
  cd notebooks
  jupyter lab

%apprun overview
  snakemake all -n --rulegraph
```

My Project

This project makes histograms from my data!

Before You Begin

1. install singularity

Exploring the Container

```
singularity myproject.simg help
singularity myproject.simg apps
```

Get Figure 1

```
singularity help myproject.simg  --app figure1
singularity run myproject.simg  --app figure1
```

Get a Graphical Overview of the Pipeline

```
singularity run myproject.simg  --app overview
```

Interact with the Analysis Notebooks

```
singularity run myproject.simg  --app notebook
```

Building Smooth, Forward Flow

The First Way

Readme Files as the human entry and first goal statement.

Version Control Systems to continually work towards goals

Package Managers to download versioned software and reduce version regressions

Environment Managers for isolating the Python interpreter

Container Systems for isolating everything.

Workflow Managers to break pipelines into smaller steps and separate out file overhead.

DataJoint™ is a free, open-source framework for programming scientific databases and computational data pipelines.

Sign up for a free tutorial database.

Getting Quick Feedback

The Second Way

Automated Testing to get feedback on the code's functionality
(PyTest, Hypothesis, Pytest-BDD)

Test-Driven Development to set goals, prioritize work, and maintain testable code.

Continuous Integration to ensure constant testing off the developer's machine.
(Travis CI, Jenkins, Circle CI)

Pair Programming to get extremely rapid feedback.

Interactive Coding Environments to get immediate results for each line of code.

A Final Tool: The Jupyter Notebook and Binder

jupyter Lorenz Differential Equations (autosaved)

File Edit View Insert Cell Kernel Help Python 3

Code Cell Toolbar: None

Exploring the Lorenz System

In this Notebook we explore the Lorenz system of differential equations:

$$\dot{x} = \sigma(y - x)$$
$$\dot{y} = \rho x - y - xz$$
$$\dot{z} = -\beta z + xy$$

This is one of the classic systems in non-linear differential equations. It exhibits a range of complex behaviors as the parameters (σ, β, ρ) are varied, including what are known as *chaotic solutions*. The system was originally developed as a simplified mathematical model for atmospheric convection in 1963.

In [7]:
```
interact(Lorenz, N=fixed(10), angle=(0.,360.),
         σ=(0.0,50.0),β=(0.,5), ρ=(0.0,50.0))
```

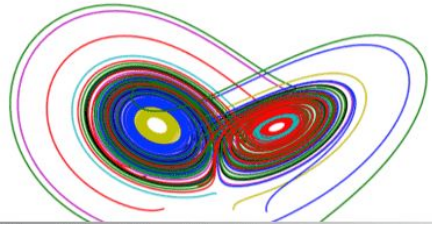
angle 308.2

max_time 12

σ 10

β 2.6

ρ 28



jupyter Welcome to P

File Edit View Insert Cell

jupyter

Welcome to the

This Notebook Server wa

WARNING

Don't rely on this serv

Your server is hosted thar

Run some Python

To run the code below:

1. Click on the cell to se
2. Press SHIFT+ENTER

A full tutorial for using the

In [ ]: %matplotlib inline

```
import pandas as pd
import numpy as np
import matplotlib
```

binder (beta)

Turn a GitHub repo into a collection of interactive notebooks

a repository full of Jupyter notebooks? With Binder, open those notebooks in an executable environment, making your code immediately reproducible by anyone, anywhere.

d launch a repository

o or URL

ithub.com/marius311/CMBLensing.jl

tag, or commit

Path to a notebook file (optional)

nb/jointmax_128example/jointmax_128example.ip

File

launch

share this URL:

/v2/gh/marius311/CMBLensing.jl/master?filepath=nb%2Fjointmax_128example%2Fjointmax_128example.ipynb

paste into your README to get badge: launch binder

Building

Maintaining an Environment of Continuous Improvement

The Third Way

Opportunistic Refactoring

Pair Programming

Chaos Monkey Testing

Blameless Post-Mortems

Opportunistic Refactoring

Training Groups

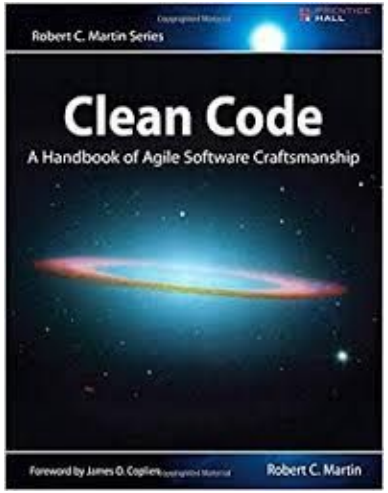
Notebook Tutorials

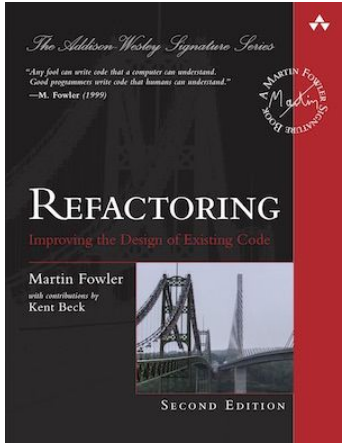
Teaching Cohorts

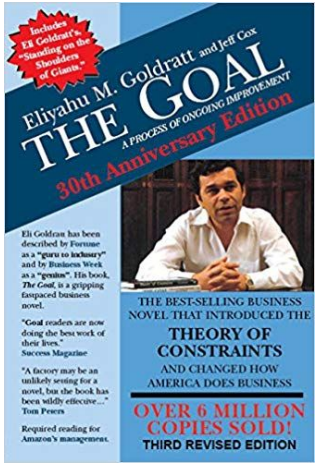
Improvement Katas

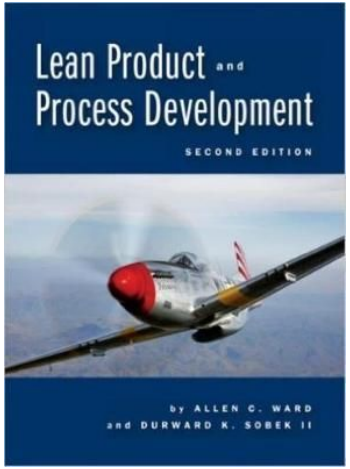
Refactoring-to-Teach

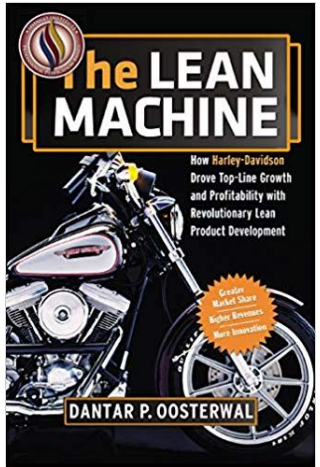
Books Recommendations: DevOps, Coding Practices, and Lean Management











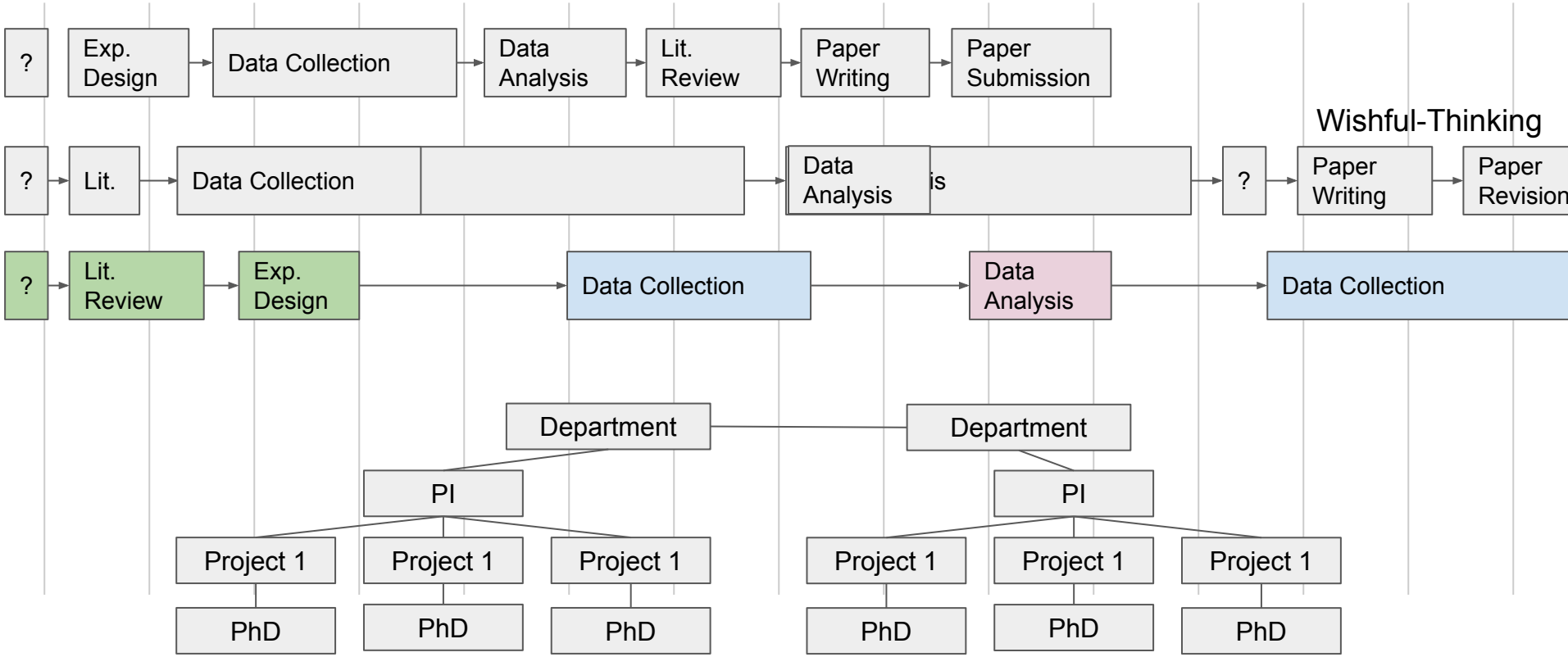
<u>These Slides Available at:</u>
github.com/nickdelgrosso/devops_talk_euroscipy2019

Thank you for your attention!

nickdelgrosso

nick-del-grosso

Wastes in of Lean Product Development
(Effects of Waterfalls and Siloed Knowledge)



Further Directions: ResearchOps and Continuous Science: Obliterating Batched Push Experimental Pipelines using Lean Product Development Models