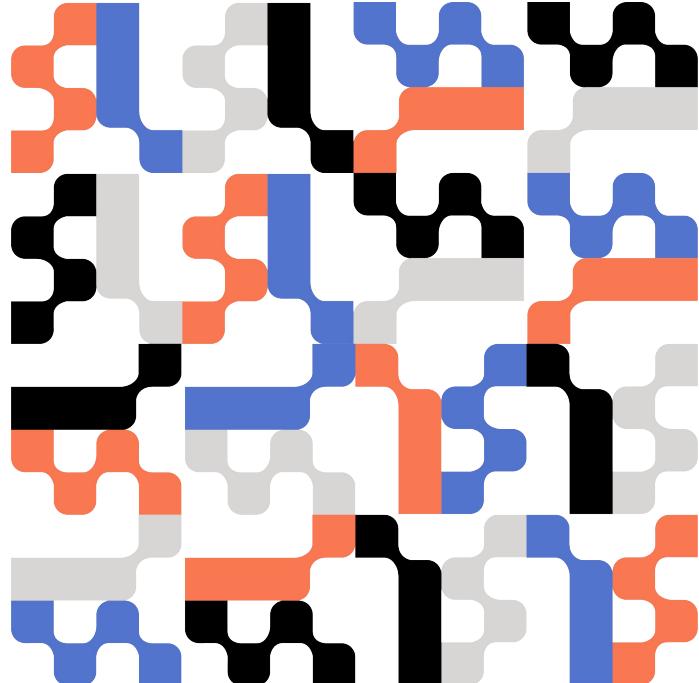


FawltyDeps

Finding **undeclared** and **unused** dependencies
in your notebooks and projects



Johan Herland – [@jherland](#)

FOSDEM 2025

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Replication crisis

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Data science

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From Wikipedia, the free encyclopedia

Not to be confused with [information science](#).

Data science is an [interdisciplinary academic field](#)^[1] that uses [statistics](#), [scientific computing](#), [scientific methods](#), [processes](#), [algorithms](#) and [systems](#) to extract or extrapolate [knowledge](#) and [insights](#) from [noisy](#), [structured](#), and [unstructured data](#).^[2]

Data science also integrates domain knowledge from the underlying application domain (e.g., natural sciences, information technology, and medicine).^[3] Data science is multifaceted and can be described as a science, a research paradigm, a research method, a discipline, a workflow, and a profession.^[4]

Data science is a "concept to unify [statistics](#), [data analysis](#), [informatics](#), and their related [methods](#)" to "understand and analyze actual [phenomena](#)" with [data](#).^[5] It uses techniques and theories drawn from many fields within the context of [mathematics](#), [statistics](#), [computer science](#), [information science](#), and [domain knowledge](#).^[6] However, data science is different from [computer science](#) and [information science](#). [Turing Award](#) winner [Jim Gray](#) imagined data science as a "fourth paradigm" of science ([empirical](#), [theoretical](#), [computational](#), and now [data-driven](#)) and asserted that "everything about science is changing because of the impact of [information technology](#)" and the [data deluge](#).^{[7][8]}

A [data scientist](#) is a professional who creates programming code and combines it with statistical knowledge to create insights from data.^[9]

[Foundations](#) (edit)



The existence of Comet NEOWISE (here depicted as a series of red dots) was discovered by analyzing astronomical survey data acquired by a space telescope, the Wide-field Infrared Survey Explorer.

Computational reproducibility of Jupyter notebooks from biomedical publications

Sheeba Samuel^{1,2*}† and Daniel Mietchen^{3,4,5†‡}

¹Heinz-Nixdorf Chair for Distributed Information Systems, Friedrich Schiller University Jena, Germany and

²Michael Stifel Center Jena, Germany and ³Ronin Institute, Montclair, New Jersey, United States and ⁴Institute for Globally Distributed Open Research and Education (IGDORE) and ⁵FIZ Karlsruhe — Leibniz Institute for Information Infrastructure, Berlin, Germany

*sheeba.samuel@uni-jena.de

†These authors contributed equally to this work

‡daniel.mietchen@ronininstitute.org

Abstract

Background Jupyter notebooks facilitate the bundling of executable code with its documentation and output in one interactive environment, and they represent a popular mechanism to document and share computational workflows, including for research publications. The reproducibility of computational aspects of research is a key component of scientific reproducibility but has not yet been assessed at scale for Jupyter notebooks associated with biomedical publications.

Approach We address computational reproducibility at two levels: (1) Using fully automated workflows, we analyzed the computational reproducibility of Jupyter notebooks associated with publications indexed in the biomedical literature repository PubMed Central. We identified such notebooks by mining the article's full text, trying to locate them on GitHub and attempting to

22,578 Python notebooks



70% had declared dependencies



46% could install dependencies



5.3% ran without errors



3.9% reproduced!



Results Out of 27,271 Jupyter notebooks from 2,660 GitHub repositories associated with 3,467 publications, 22,578 notebooks were written in Python, including 15,817 that had their dependencies declared in standard requirement files and that we attempted to re-run automatically. For 10,388 of these, all declared dependencies could be installed successfully, and we re-ran them to assess reproducibility. Of these, 1,203 notebooks ran through without any errors, including 879 that produced results identical to those reported in the original notebook, and 324 for which our results differed from the originally reported ones. Running the other notebooks resulted in exceptions.

notebooks resulted in exceptions:

reproduced in the original notebook¹ but did not run due to missing or incompatible dependencies

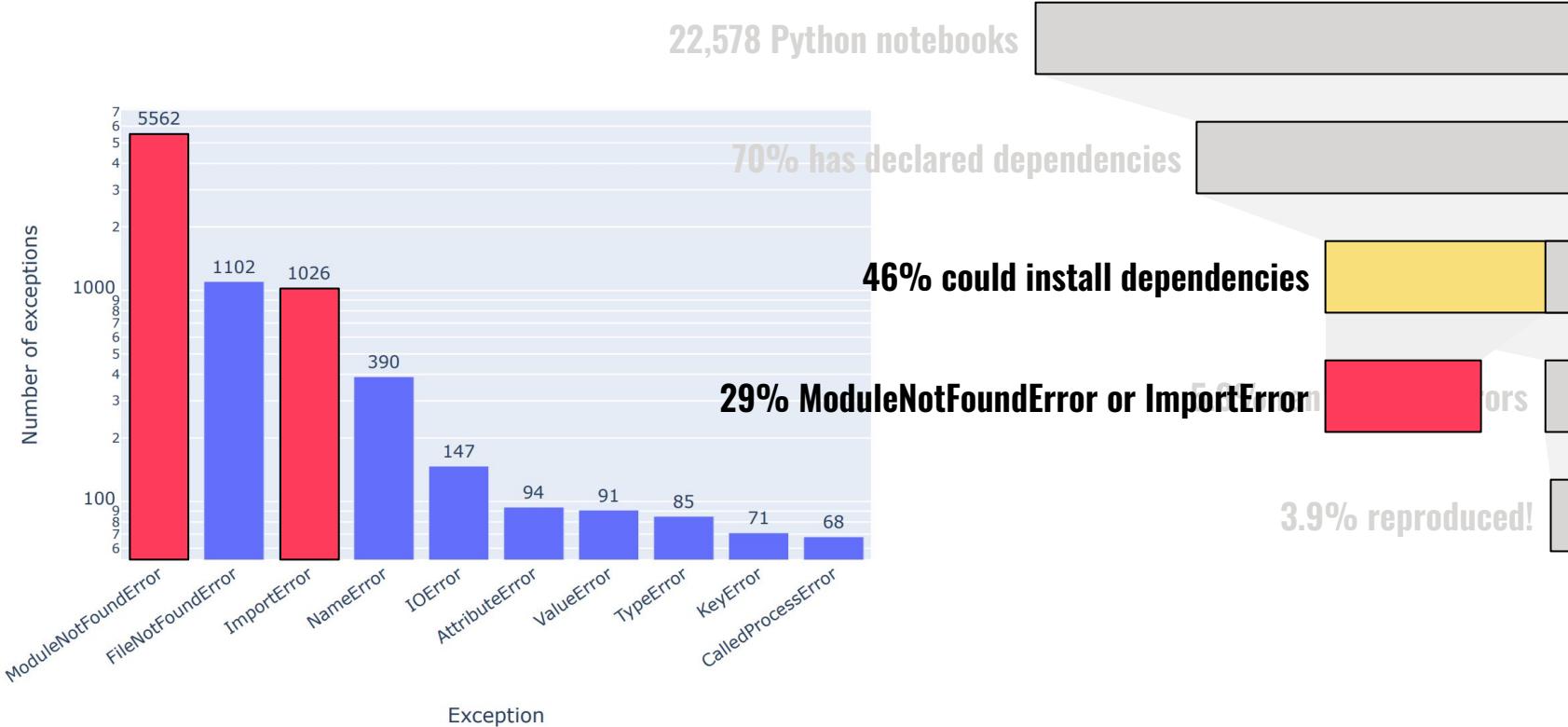


Figure 19. Exceptions occurring in Jupyter notebooks in our corpus. See Table 5 for information about the nature of these errors and potential fixes.

30% failed to declare dependencies

24% failed to install dependencies

29% failed to import dependencies

>80% fail because of missing or incorrect dependency declarations.

PAPER

Computational reproducibility of Jupyter notebooks from biomedical publications

Sheeba Samuel^{1,2*}† and Daniel Mietchen^{3,4,5†‡}

22,578 Python notebooks

70% has declared dependencies

40% could install dependencies

29% ModuleNotFoundError or ImportWarning

3.9% reproduced!

Less is More? An Empirical Study on Configuration Issues in Python PyPI Ecosystem

Yun Peng

The Chinese University of Hong Kong
Hong Kong, China
ypeng@cse.cuhk.edu.hk

Ruida Hu

Harbin Institute of Technology
Shenzhen, China
200111107@stu.hit.edu.cn

Ruke Wang

Harbin Institute of Technology
Shenzhen, China
200110930@stu.hit.edu.cn

Cuiyun Gao*

Harbin Institute of Technology
Shenzhen, China
gaocuiyun@hit.edu.cn

Shuqing Li

The Chinese University of Hong Kong
Hong Kong, China
sqli21@cse.cuhk.edu.hk

Michael R. Lyu

The Chinese University of Hong Kong
Hong Kong, China
lyu@cse.cuhk.edu.hk

ABSTRACT

Python is the most popular programming language in the world. As a source code repository, PyPI is the largest open-source library platform. Despite its diversity and popularity, there are many configuration issues in the PyPI ecosystem. In this study, we conducted a large-scale empirical study on the PyPI ecosystem to investigate the configuration issues. We found that approximately 65% of libraries successfully pass all three checks of PyPI. This indicates that approximately 30% of libraries and 51% of releases on the PyPI platform can be installed but may encounter source-level compatibility problems.

Only 5,371 (65%) libraries,

comprising 131,720 (39%) releases, successfully pass all three checks of PyPI. This indicates that approximately 30% of libraries and 51% of releases on the PyPI platform can be installed but may encounter source-level compatibility problems.

in fostering its prosperity. The accessibility and utility of Python are further amplified by the public libraries available on the Python Package Index (PyPI) platform. With over 470 thousand Python

PFRL.

Table 2: Configuration issues detected by PyCon. There may be multiple issues occurring in one release.

Category	Issue	Check	#Releases	Fatal?	Possible Reasons
Incomplete Configuration	Missing configuration files	Installation Check	251	✓	Missing required information
	Missing required libraries for setup	Installation Check	3,318	✓	
	Missing Python versions	Dependency Check	55,138	✗	
	Missing required libraries for direct imports	Import Validation	142,521	✗	
Incorrect Configuration	Dependency conflicts in setup	Installation Check	6,318	✓	Unsolvable constraints
	Incorrect Python versions	Installation Check	4,155	✓	Incorrect dependencies
	Other run-time Errors in setup	Installation Check	3,464	✓	Missing files
	Inconsistent configurations with metadata	Dependency Check	592	✗	Naming error
	Inconsistent version numbers with release dates	Dependency Check	12,018	✗	Confusing version orders
	Missing required modules for indirect imports	Import Validation	11,023	✗	
	Inconsistent modules in direct imports with installed dependencies	Import Validation	6,678	✗	Incorrect dependencies
Incorrect Code	Other run-time Errors in imports	Import Validation	8,178	✗	
	Missing source code	Dependency Check	2,588	✓	Creating placeholders
	Parsing error	Dependency Check	431	✓	Invalid syntax/encoding
	Multiple version control failure	Import Validation	15,507	✗	Incorrect dependencies

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of Hong Kong
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edu.hk

alled Dependencies:

==1.13.1
==0.26.2
py==1.21.6
ck==3.12.2
w==9.5.0Monitor as _GymMonitor
third-party libraryd utility of Python
able on the Python
n thousand Python

Dependencies and dependency declarations

code.py:

```
import numpy as np  
  
...
```

pyproject.toml

```
[project]  
name = "my_project"  
dependencies = ["numpy"]
```

setup.cfg:

```
[options]  
install_requires =  
    numpy==1.26.2
```

setup.py:

```
$ python code.py  
Traceback (most recent call last):  
  File ".../code.py", line 1, in <module>  
    import numpy as np  
ModuleNotFoundError: No module named 'numpy'
```

```
from setuptools import setup  
  
setup(  
    name="my_project",  
    install_requires=[ "numpy" ]  
)
```

Dependencies and dependency declarations

code.py:

```
import numpy as np  
  
...
```

requirements.txt:

```
numpy
```

Undeclared and unused dependencies

code.py:

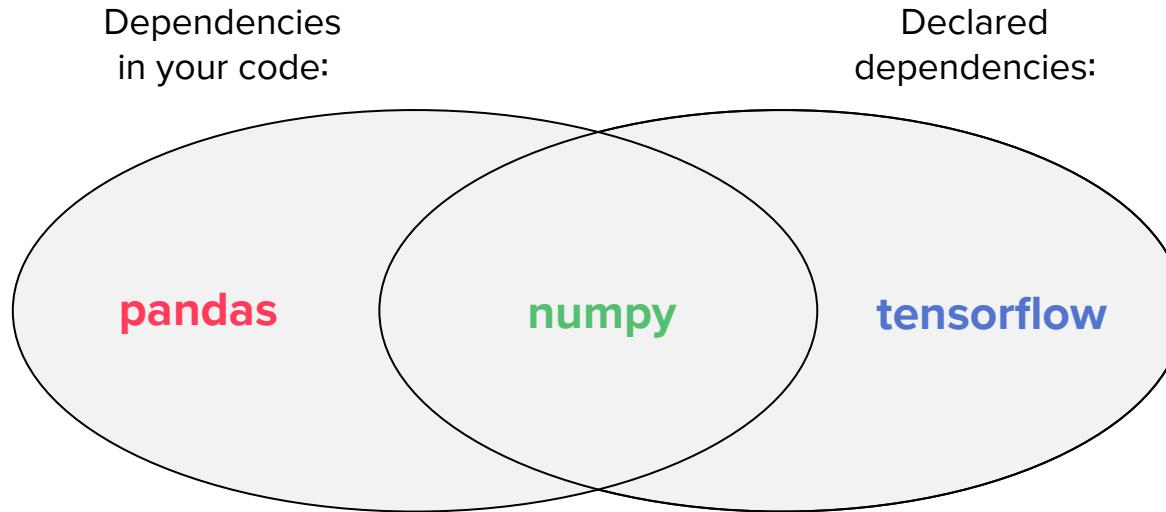
```
import numpy as np
import pandas

...
```

requirements.txt:

```
numpy
tensorflow
```

Undeclared and unused dependencies



Can we automate this analysis?



Nour El Mawass



Maria Knorps



Johan Herland



Zhihan Zhang



Richard
Bullington-McGuire



FawltYDeps

code.py:

```
import numpy as np
import pandas

...
```

requirements.txt:

```
numpy
tensorflow
```

```
$ fawltYdeps --detailed
T
These imports appear to be undeclared dependencies:
- 'pandas' imported at:
    code.py:2
T
These dependencies appear to be unused (i.e. not imported):
F
- 'tensorflow' declared in:
    requirements.txt
```

FawltYDeps

Python code and compatibility:

- + Python scripts (.py)
- + Jupyter notebooks (.ipynb)
- + Python v3.8 – v3.13
- + Linux, MacOS, and Windows

Availability:

- + via PyPI:
 - + `pip install fawltYdeps`
 - + As a dev dependency: `uv add fawltYdeps`
 - + As a stand-alone tool: `uvx fawltYdeps`
- + as a [pre-commit hook](#) or [Github Action](#)

Dependency declarations:

- + `requirements.txt`
- + `pyproject.toml` (PEP 621, Poetry, Pixi)
- + `setup.py` (simple)
- + `setup.cfg`
- + `environment.yml` (Conda)
- + `pixi.toml` (Pixi)

Configurable:

- + `fawltYdeps --help`
- + Command-line options override configuration
- + `$fawltYdeps_<directive>` in environment
- + `[tool.fawltYdeps]` in `pyproject.toml`
- + `fawltYdeps --generate-toml-config`

FawltDeps output

code.py:

```
import numpy as np
import pandas

...
```

```
$ fawltdeps --list-imports --detailed
code.py:1: numpy
code.py:2: pandas
```

requirements.txt:

```
numpy
tensorflow
```

```
$ fawltdeps --list-deps --detailed
requirements.txt: numpy
requirements.txt: tensorflow
```

```
$ fawltdeps --list-sources --detailed
```

Sources of Python code:

code.py (using ./ as base for 1st-party imports)

Sources of declared dependencies:

requirements.txt (parsed as a requirements.txt file)

FawltYDeps

code.py:

```
import numpy as np
import pandas

...
```

```
$ fawltYdeps --list
code.py:1: numpy
code.py:2: pandas
```

```
$ fawltYdeps --json
{
  "settings": { ... },
  "sources": [
    {
      "source_type": "CodeSource",
      "path": "code.py",
      "base_dir": "."
    },
    {
      "source_type": "DepsSource",
      "path": "requirements.txt",
      "parser_choice": "requirements.txt"
    }
  ],
  "imports": [
    {
      "name": "numpy",
      "source": {"path": "code.py", "lineno": 1}
    },
    {
      "name": "pandas",
      "source": {"path": "code.py", "lineno": 2}
    }
  ],
  "declared_deps": [
    {
      "name": "numpy",
      "source": {"path": "requirements.txt"}
    },
    {
      "name": "tensorflow",
      "source": {"path": "requirements.txt"}
    }
  ],
  "resolved_deps": {
    "tensorflow": {
      "package_name": "tensorflow",
      "import_names": ["tensorflow"],
      "resolved_with": "IdentityMapping",
      "debug_info": null
    },
    "numpy": {
      "package_name": "numpy",
      "import_names": ["numpy"],
      "resolved_with": "IdentityMapping",
      "debug_info": null
    }
  },
  "undeclared_deps": [
    {
      "name": "pandas",
      "references": [
        {"path": "code.py", "lineno": 2}
      ]
    }
  ],
  "unused_deps": [
    {
      "name": "tensorflow",
      "references": [
        {"path": "requirements.txt"}
      ]
    }
  ],
  "version": "0.18.0"
}
```

FawltDeps as pre-commit hook

<https://pre-commit.com>

```
$ pip install pre-commit  
[...]  
$ vim .pre-commit-config.yaml  
$ pre-commit install  
pre-commit installed at .git/hooks/pre-commit
```



.pre-commit-config.yaml:

```
repos:  
  - repo: https://github.com/tweag/FawltDeps  
    rev: v0.18.0  
    hooks:  
      - id: check-undeclared  
      - id: check-unused
```

FawltDeps as pre-commit hook

<https://pre-commit.com>

```
$ vim code.py
$ git commit -am "Add pandas"
[...]
FawltyDeps-undeclared..... Failed
- hook id: check-undeclared
- exit code: 3

These imports appear to be undeclared
dependencies:
- 'pandas' imported at:
  code.py:2

FawltyDeps-unused..(no files to check)Skipped
```



code.py:

```
import numpy as np
+import pandas
```

...

FawltDeps as pre-commit hook

<https://pre-commit.com>

```
$ vim code.py requirements.txt
$ git commit -am "Add pandas"
FawltyDeps-undeclared.....Passed
FawltyDeps-unused.....Passed
[main 6ece52e] Add pandas
  2 files changed, 2 insertions(+)
$ vim requirements.txt
$ git commit -am "Declare tensorflow dependency"
FawltyDeps-undeclared..(no files to check)Skipped
FawltyDeps-unused.....Failed
- hook id: check-unused
- exit code: 4
```

These dependencies appear to be unused (...):

- 'tensorflow' declared in:
requirements.txt



code.py:

```
import numpy as np
import pandas

...
```

requirements.txt:

```
numpy
pandas
+tensorflow
```

FawltDeps in CI (GitHub Actions)

github.com/tweag/FawltDeps-action

.github/workflows/fawltdeps.yaml:

```
name: FawltDeps

on: [push, pull_request]

jobs:
  fawltdeps:
    runs-on: ubuntu-latest
    steps:
      - name: checkout
        uses: actions/checkout@v4
      - name: lint - FawltDeps
        uses: tweag/FawltDeps-action@v0.2.0
        with:
          options: --detailed
```



GitHub Actions

FawltDeps in CI (GitHub Action)

Add undeclared use of pandas #1

Edit < Code ▾

Open j

Convers

Changes from

1 Merge pull request

import numpy as np
+ import pandas

All checks have failed

2 failing checks

FawltDeps FawltDeps

This branch Merging can be

Merge pull request

Summary

Annotations Details

fawltdeps summary

Error: 'fawltdeps --detailed' found issues:

These imports appear to be undeclared dependencies:
- 'pandas' imported at:
code.py:2

22

How to draw an Owl.

"A fun and creative guide for beginners"

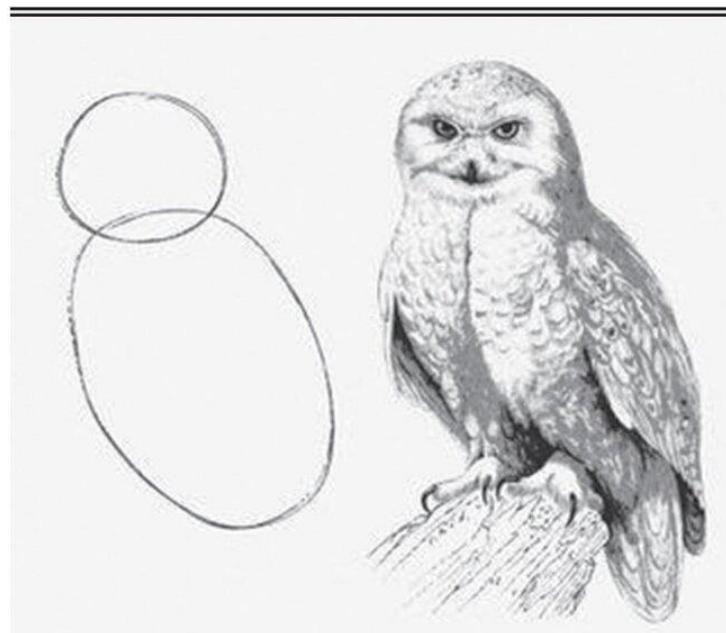


Fig 1. Draw two circles

Fig 2. Draw rest of the damn Owl

Matching imports names to package names

code.py:

```
import numpy as np  
import sklearn  
import setuptools  
import pkg_resources  
  
...
```

requirements.txt:

```
numpy  
scikit-learn  
setuptools
```

Imports are found

Dependencies are installed



Resolving dependencies into import names

- + Python environments inside the project
 - + Search the entire project?
 - + Search within the paths given to FawltDeps
 - + Python environments given with `--pyenv`
- + Python environment where FawltDeps is installed
- + Custom mapping: `[tool.fawltdeps.custom_mapping]`
 - + Takes precedence, if given
- + Auto-install from PyPI? `--install-deps`
 - + Expensive, disabled by default
- + Final resort: Identity mapping

Resolving dependencies into import names

code.py:

```
import sklearn
```

requirements.txt:

```
scikit-learn
```

Custom mapping

Python environments in project

Python environment w/FawlyDeps

Auto-install from PyPI?

Identity Mapping

scikit-learn → sklearn

scikit-learn → sklearn

scikit-learn → sklearn

scikit-learn → scikit_learn

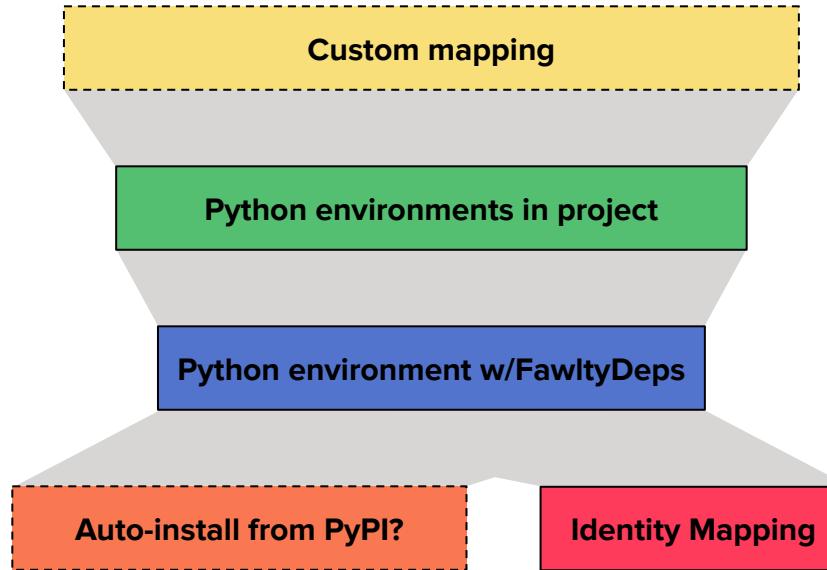
Resolving dependencies into import names

code.py:

```
import sklearn
```

requirements.txt:

```
scikit-learn
```



More complexities and advanced use cases

- + Exclude parts of your project from analysis:
 - + `--exclude tests/`
 - + `--exclude-from .gitignore`
- + Some dependencies are never meant to be imported:
 - + Tools: tox, black, flake8, ruff, pylint, mypy, etc.
 - + Type stubs
 - + Indirect imports or plugin-style package collections
 - + `--ignore-unused`
- + Conditional imports, alternative imports, e.g.: `try: import foo; except: import bar`
 - + `--ignore-undeclared`



Work in progress

- + Dynamic imports, e.g.: `importlib.import_module(<some dynamic expression>)`
- + Required dependencies vs. various degrees of optional dependencies
 - + `install-requires` vs. `extras-require` in `setup.py`
 - + `[project.dependencies]` vs `[project.optional-dependencies]` in `pyproject.toml`
 - + PEP 735: `[dependency-groups]` in `pyproject.toml`
 - + Poetry: `[tool.poetry.group.$NAME.dependencies]` in `pyproject.toml`
 - + Development dependencies: `dev-requirements.txt` vs `requirements.txt`



Summary

- + Reproducibility matters
- + Missing or incorrect dependency declarations is the greatest culprit
- + **FawltDeps** finds **undeclared** and **unused** dependencies in your project
 - + Works with Python scripts and Jupyter notebooks, across Python version and platforms
 - + Understands the many ways of declaring dependencies
 - + Works out of the box on most projects, and can be configured to suit *all* projects
 - + Can be an automated part of your workflow, either pre-commit or in your CI



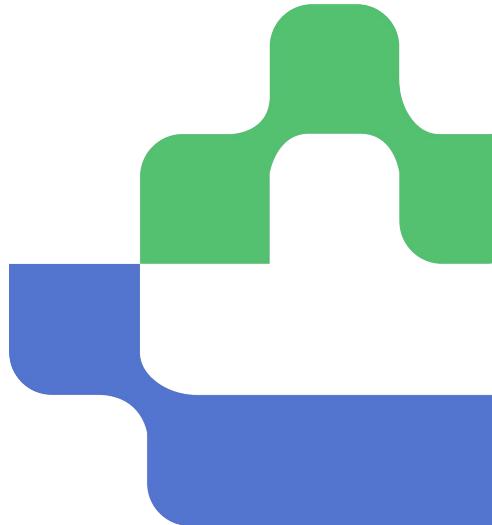
Slides

QUESTIONS?

- + FawlyDeps on GitHub
 - + FawlyDeps on PyPI
 - + FawlyDeps on Discord
-
- + Tweag blog: tweag.io/blog
 - + Tweag: tweag.io
 - + Modus Create: moduscreate.com



@jherland



THANK YOU!



MODUS CREATE



by Modus Create

