

Python Package Development

AI starts from here

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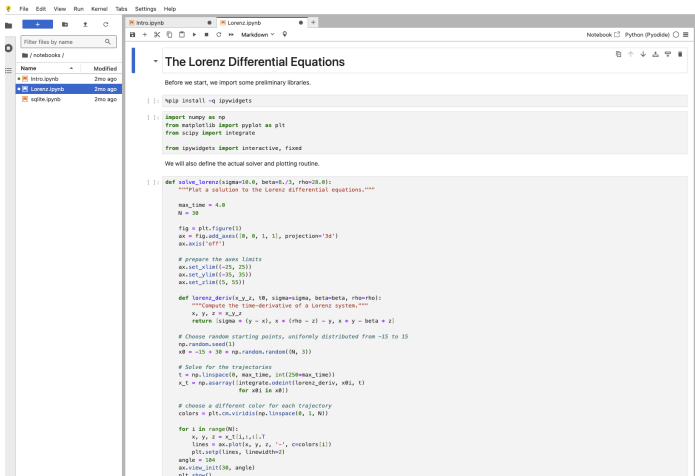
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Why we need to deliver Python Package?



The screenshot shows a Jupyter Notebook interface with a file explorer on the left and a code editor on the right. The notebook is titled "The Lorenz Differential Equations". The code in the notebook is as follows:

```
File Edit View Run Kernel Tabs Settings Help

/notebooks/
Name Modified
Intro.ipynb 2mo ago
Lorenz.ipynb 2mo ago
solve.ipynb 2mo ago

In [ ]: %pip install -q ipynbwidgets

In [ ]: import numpy as np
from matplotlib import pyplot as plt
from scipy import integrate
from ipynbwidgets import Interactive, Fixed

We will also define the actual solver and plotting routine.

In [ ]: def solve_lorenz(sigma=10.0, beta=8/3, rho=28.0):
    """Solve a solution to the Lorenz differential equations."""
    max_time = 4.0
    N = 30

    fig = plt.figure()
    ax = fig.add_subplot(0, 0, 1, 1, projection='3d')
    ax.axis('off')

    # prepare the axes limits
    ax.set_xlim((-25, 25))
    ax.set_ylim((-35, 35))
    ax.set_zlim((0, 55))

    def lorenz_deriv(x,y,z, t0, sigma=sigma, beta=beta, rho=rho):
        """Compute the time-derivative of a Lorenz system."""
        x, y, z = x,y,z
        return (sigma * (y - x), x * (rho - z) - y, x * y - beta * z)

    # Choose random starting points, uniformly distributed from -15 to 15
    np.random.seed(1)
    x0 = -15 + 30 * np.random.random(N, 3)

    # Solve for the trajectories
    t = np.linspace(0, max_time, int(100*max_time))
    x,t = np.asarray([integrate.differential_equations(lorenz_deriv, x0, t)
                      for x0 in x0])

    # choose a different color for each trajectory
    colors = plt.cm.viridis(np.linspace(0, 1, N))

    for i in range(N):
        x, y, z = x,t[i],i,t
        lines = ax.plot(x, y, z, '-', c=colors[i])
        plt.setp(lines, linewidth=2)

    angle = 284
    ax.view_init(30, angle)
    plt.show()
```

Figure: You can not deploy this to production

Why we need to deliver Python Package?

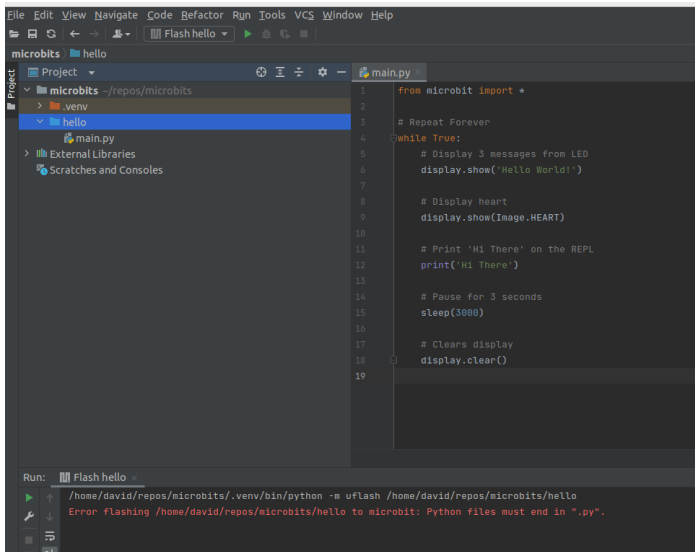


Figure: You also can not deliver this to production

Life is short



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NEW

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Color

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Pick a Color

Pick a Size

Size guide

MAKE A SELECTION

GUARANTEED SAFE & SECURE CHECKOUT





Estimated delivery to  Australia Dec 2-5

Description

Details

Shipping

Love Python? This tee is inspired by [Bruce Eckel's](#) famous quote "Life is short (You need Python)". The Chinese in the tee is the translation of the English quote "Life is short, I use Pythoan".

Python creator Guido van Rossum wore a similar tee in the picture below:



This tee is designed in the UK, printed and shipped from the US/EU.








Figure: Life is short, (I use Python/You need Python)

Do not reinvent wheels

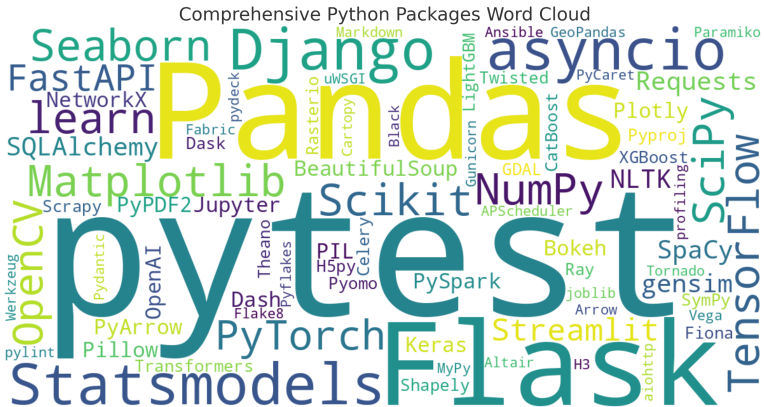


Figure: There are so many useful wheels

Virtual Environment

Why we need virtual environment?

- You will need different Python versions for different projects
- For each project, you will need different packages
- Isolates project dependencies from system-wide Python installation
- Prevents conflicts between package versions

How Python Virtual Environment Works

Core Mechanism

- Creates isolated Python environment by copying:
 - Python interpreter
 - Core libraries
 - pip package manager
- Modifies system path during activation

Directory Structure

```
myenv/
  bin/           # Scripts
  include/       # Header files
  lib/           # Libraries
  pythonX.Y/site-packages/
  pyvenv.cfg     # Configuration
```

Activation Process

- 1 Prepends venv's bin to PATH
- 2 Updates shell prompt
- 3 Changes Python path search order

Key Features

- Lightweight
- No system changes
- Easy cleanup

Understanding Dependency Hell

- **Dependency Hell:** When software dependencies conflict
- Common scenarios:
 - Project A needs Package X version 1.0
 - Project B needs Package X version 2.0
- Even worse:
Project A needs Package Y, which depends on Package X version 2.0

The hell tricks the AI

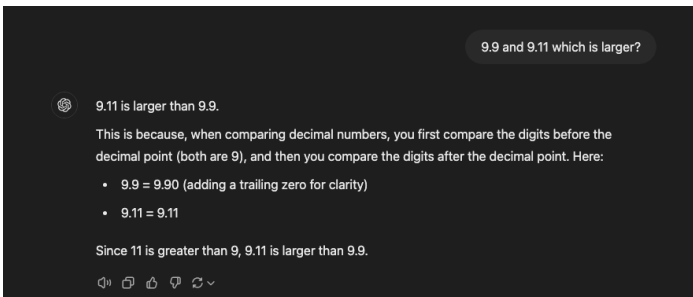


Figure: Math calculation or Version Comparison

Problematic issue calling for your solution

Core Problem

Package version resolution is essentially a graph search problem:

- Need to satisfy all version constraints simultaneously

Critical Edge Cases

■ High-Frequency Updates:

- Example: AWS CLI with near-daily releases
- Thousands of versions increase computational complexity
- Which means it will take "forever" to resolve the versions

■ Research Software Issues:

- Research packages often lack industry-standard versioning
- Inconsistent maintenance and version specifications
- Which means the resolve of different packages will fail

Dependency Management Tools

Key Management Needs

- **Python Version:** Multiple versions across projects
- **Dependencies:** Project-specific packages and versions

Evolution of Tools

- **Early Days:** pyenv (Python versions) + virtualenv (isolated environments)
- **pipenv:** Combined pip and virtualenv, introduced Pipfile
- **conda:** Handles non-Python dependencies, scientific computing focus
- **poetry:** Modern dependency resolution, simplified packaging, uses pyproject.toml

Why Different Tools?

- **conda:** Solves scientific computing needs (C/C++ dependencies)
- **poetry:** Try to address the dependency hell problem

Virtual Environment Solutions

Practical Approaches

No one-size-fits-all solution in my opinion, but here are reliable approaches:

1 Conda-based Solution

- Use when conda installation is feasible
- `conda install` or `pip install` both work
- Manage via `requirements.txt`
- Consider disk space

2 Traditional Python Setup

- Python via `brew/pyenv`
- Create env: `python -m venv venv`
- Use `pip` + `requirements.txt`

3 Docker-based Approach

- Recommended for production
- Configure in `Dockerfile`
- Manage via `requirements.txt`

4 Poetry Workflow

- Modern dependency management
- For stable, non-cutting-edge projects

1. Conda Method

Setup - Scientific Computing Choice

```
# Create environment with Python 3.9
```

```
conda create -n myenv python=3.9
```

```
# Activate environment
```

```
conda activate myenv
```

```
# Install packages
```

```
conda install numpy pandas
```

```
pip install requests # also works
```

```
# Save dependencies, I actually prefer doing this manually
```

```
pip freeze > requirements.txt
```

- **Pros:** Best for scientific packages (numpy, pandas)
- **Cons:** Large disk space, slower than pip
- **Use when:** Working with data science projects

2. Traditional venv Method

Setup - Lightweight Choice

```
# Install Python (if needed)
brew install python@3.9 # or pyenv install 3.9.0

# Create and activate environment
python -m venv venv
source venv/bin/activate # Unix
.\venv\Scripts\activate # Windows

# Install packages
pip install numpy pandas
pip freeze > requirements.txt
```

- **Pros:** Lightweight, built-in, fast
- **Cons:** Python binary management needed
- **Use when:** Simple Python projects, limited disk space

3. Docker Method

Setup - Production Choice

```
# Create Dockerfile
FROM python:3.9-slim
WORKDIR /app
# Install other binary dependencies if required
COPY requirements.txt .
RUN pip install -r requirements.txt
COPY . .

# Build and run
docker build -t myproject .
docker run myproject
```

- **Pros:** Production-ready, fully isolated, reproducible
- **Cons:** More complex setup, sometimes worse dev experience
- **Use when:** Deploying to production, team projects

4. Poetry Method

Setup - Modern Python Choice

```
# Install Poetry
curl -sSL https://install.python-poetry.org | python3 -

# Start project
poetry new myproject
cd myproject

# Install and manage packages
poetry add numpy pandas
poetry install
poetry run python script.py
```

- **Pros:** Modern dependency resolution, good package management
- **Cons:** Resolve conflict forever

Best Practices

- Create one virtual environment per project
- Always include requirements.txt
- Inside the requirements.txt, local your version
- Use version control (.gitignore for venv folder)
- Document environment setup in README.md
- Regularly update dependencies



Project Structure Overview

Basic Python Project Structure

```
my_project/  
  .git/  
  .gitignore  
  README.md  
  requirements.txt      # or pyproject.toml  
  setup.py              # if publishing package  
  my_project/           # main package directory  
    __init__.py  
    main.py  
    utils.py  
  tests/  
    __init__.py  
    test_main.py
```

Common Additional Directories

Optional but Recommended

- **docs/** - Documentation files
- **scripts/** - Utility scripts, tools
- **examples/** - Example usage
- **data/** - Project data files
- **notebooks/** - Jupyter notebooks
- **configs/** - Configuration files

Best Practices

- Keep related files together
- Separate code from data
- Clear import paths
- Follow naming conventions

Example .gitignore

Common Python Ignores

```
# Python
**pycache**/
*.py[cod]
*.so
build/
dist/
# Environment
venv/
env/
.env/
# IDE
.vscode/
.idea/
# Tests
.coverage
.pytest_cache/
# Project
data/raw/
*.log
```


Example: Docs2KG

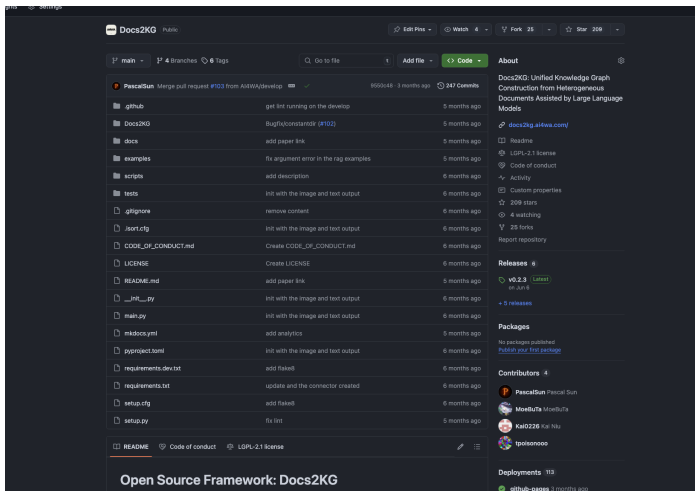


Figure: An example Python package project¹

¹<https://github.com/AI4WA/Docs2KG>



How to Execute Your Code

Understanding if `__name__ == "__main__"`

```
# file: my_package/main.py
def main():
    print("Running main function")

if __name__ == "__main__":
    main() # Only runs if file executed directly
```

Execution Methods

1 Module-style execution:

```
python -m my_package.module.script
```

2 IDE Integration:

- PyCharm: Right-click → Run 'script'
- Configurable run configurations

3 Test-Driven Development (TDD):

```
pytest tests/test_main.py
```

How to Execute your code?

```

class SpatioTemporalMetadataProcessor:
    def run(self, metadata_id: Optional[int] = None):
        # update the task status to completed
        self.api.update_task_status(task_id=task_id, status="SUCCESS")

    except Exception as e:
        logger.error(f"Error processing spatiotemporal metadata: {e}")
        logger.exception(e)
        self.api.update_task_status(task_id=task_id, status="FAILED")
        if self.metadata_id:
            self.api.update_metadata_progress(
                metadata_id=self.metadata_id,
                status="Failed",
                logs={"timestamp": str(datetime.now()), "message": str(e)},
            )
        return

if __name__ == "__main__":
    processor = SpatioTemporalMetadataProcessor()
    processor.run()
    
```

Figure: An example of how to execute your code?

Module Import Strategies

Project Structure

```
my_project/  
  my_project/  
    __init__.py  
    main.py  
    utils/  
      __init__.py  
      helper.py
```

Module Import Strategies

Relative Import

```
# in main.py
from .utils.helper import func
from ..utils import helper
```

- Complex with deep nesting
- Error-prone

Absolute Import

```
# in main.py
from my_project.utils.helper import func
from my_project.utils import helper
```

- Clear and explicit
- Easier to maintain

Import Best Practices

Relative Import Issues

- Hard to understand with deep nesting
- More prone to errors when moving files
- Can break if running scripts directly
- Confusing with multiple dot notation

Why Prefer Absolute Imports?

- Clear and explicit paths
- Easier to understand and maintain
- More reliable when refactoring
- IDE friendly (better autocompletion)

Recommendation

Always use absolute imports unless you have a specific reason not to.

Import Circular, and How to Solve It?

Problem Example

```
[python]
# a.py
from b import function_b

def function_a():
    return function_b() + 1

# b.py
from a import function_a

def function_b():
    return function_a() + 1
```


Import Circular, and How to Solve It?

Solutions:

1 Restructure Code

```
[python]  
# common.py  
def shared_logic():  
    pass
```

2 Import Inside Function

```
[python]  
def function_b():  
    from a import function_a # Local import  
    return function_a() + 1
```

3 Use Dependency Injection

```
[python]  
def function_b(func_a=None):  
    if func_a is None:  
        from a import function_a  
        func_a = function_a  
    return func_a() + 1
```

Use Class, Not Functions

We taught regarding how you write the code in function.
However, object oriented is the best practice for large projects.

Example: Functions vs Class Approach

```
[python]
# Function approach
def process_data(data):
    return data * 2

# Class approach
class DataProcessor:
    def __init__(self, data):
        self.data = data

    def process(self):
        return self.data * 2

# Usage
processor = DataProcessor(10)
result = processor.process()
```

Python @staticmethod

Static methods don't require access to class or instance state. They are utility functions that live in the class namespace.

Example: Using @staticmethod

```
[python]
class Calculator:
    @staticmethod
    def add(x, y):
        # No 'self' or 'cls' parameter needed
        return x + y

    @staticmethod
    def is_positive(num):
        return num > 0

# Usage
result = Calculator.add(5, 3)          # Returns 8
is_pos = Calculator.is_positive(5)    # Returns True
```

- No implicit first argument (no self, no cls)
- Cannot access/modify class or instance attributes
- Best for utility functions related to class purpose

Python @classmethod

Class methods receive the class itself as the first argument.
They are commonly used for alternative constructors or factory methods.

Example: Using @classmethod

```
[python]
class Date:
    def __init__(self, year, month, day):
        self.year = year
        self.month = month
        self.day = day

    @classmethod
    def from_string(cls, date_str):
        # cls is the class itself
        year, month, day = map(int, date_str.split('-'))
        return cls(year, month, day)

# Usage
date = Date.from_string('2024-11-28')
```

- Receives class as first argument (cls)
- Can access and modify class state

Code Style Tools: isort, flake8, and black

Tool Purposes & Relationships

These tools work together to ensure consistent, readable, and maintainable Python code:

- **isort**: Organizes and sorts imports systematically
- **flake8**: Enforces PEP 8 style guide and checks code quality
- **black**: Formats code with opinionated, consistent styling

isort

- Groups imports by type
- Sorts alphabetically
- Separates standard, third-party, local
- Removes unused imports

flake8

- Checks line length
- Detects syntax errors
- Finds style issues
- Reports complexity

black

- Enforces formatting
- Removes style choices
- Makes code deterministic
- Focuses on readability

Code Style

Recommended Usage Order

1. Run **isort** first to organize imports
2. Apply **black** to format the code
3. Use **flake8** to catch remaining style issues

CI/CD Integration

- Tools integrated into CI/CD pipeline
- Automated code style checks on each commit/PR
- Fails builds that don't meet standards
- Maintains consistent code quality across team

Why Loguru over Default Logger?

Default Logger

- Complex configuration
- Basic formatting
- No color support
- Manual exception handling

Loguru

- Zero configuration
- Better formatting
- Colored output
- @logger.catch decorator

```

1 # Default logger - Complex setup
2 import logging
3 logging.basicConfig(level=logging.INFO)
4 logger = logging.getLogger(__name__)
5
6 # Loguru - Simple setup
7 from loguru import logger
8

```



Introduction to CLI

What is CLI?

- Command Line Interface for program interaction
- Two main approaches:
 - Script execution: `python script.py`
 - Direct CLI: `mycli`

Basic - sys.argv

- Access arguments through `sys.argv` list
- `argv[0]`: script name, `argv[1:]`: arguments

```
1 import sys
2
3 def main():
4     print(f"Script: {sys.argv[0]}")
5     print(f"Arguments: {sys.argv[1:]}")
6
7 if __name__ == '__main__':
8     main()
9
```

Usage

```
1 $ python script.py arg1 arg2
2
```

Intermediate - argparse

- Standard library for argument parsing
- Automatic help and validation

```
1 import argparse
2
3 parser = argparse.ArgumentParser(description='Example')
4 parser.add_argument('--sum', dest='accumulate',
5                     action='store_const',
6                     const=sum, default=max,
7                     help='sum the integers')
8 parser.add_argument('integers', type=int, nargs='+')
9 args = parser.parse_args()
10
```

Usage

```
1 $ python script.py --sum 1 2 3
2
```

Advanced - Click Framework

- Modern CLI framework with decorator interface
- Direct command execution after installation

```

1 import click
2
3 @click.command()
4 @click.option('--count', default=1)
5 @click.option('--name', prompt='Your name')
6 def hello(count, name):
7     for x in range(count):
8         click.echo(f'Hello {name}!')
9

```

Usage

```

1 $ mycli --count=3 --name=John
2

```

Making CLI Executable

```
1 from setuptools import setup
2
3 setup(
4     name='mycli',
5     version='0.1',
6     py_modules=['mycli'],
7     install_requires=['click'],
8     entry_points={
9         'console_scripts': ['mycli=mycli:hello'],
10    },
11 )
12
```

Installation

```
1 $ pip install --editable .
2
```

Cli Summary

Script Execution

- `sys.argv`: Basic parsing
- `argparse`: Structured CLI
- Requires `python` command

Direct CLI

- Click: Modern interface
- Direct execution
- Professional distribution



Why Documentation?

- Documentation = Communication
- Code tells what it does, documentation tells why
- Modern approach: Documentation lives with code
- Automatic generation reduces maintenance burden

Writing Documentation in Code

Example: NumPy Style

```
def calculate_metric(data: np.ndarray) -> float:
    """Calculate performance metric from input data.

    Parameters
    -----
    data : np.ndarray
        Input time series data

    Returns
    -----
    float
        Computed metric value
    """
```

Writing Documentation in Code

Example: Google Style

```
def calculate_metric(data: np.ndarray) -> float:
    """Calculate performance metric from input data.

    Args:
        data (np.ndarray): Input time series data

    Returns:
        float: Computed metric value
    """
```

MkDocs: Modern Documentation Generator

- Why MkDocs?
 - Modern, clean appearance
 - Markdown-based (easier than RST)
 - Material theme support
 - Fast build times
- Key features:
 - Auto-navigation
 - Search functionality
 - Code highlighting
 - Math equation support
- Demo
 - <https://docs2kg.ai4wa.com/>
 - <https://openomnidocs.ai4wa.com/>

Hosting via GitHub Pages through CI/CD

Workflow

```
name: docs
on:
  push:
    branches:
      - main
jobs:
  deploy:
    runs-on: ubuntu-latest
    steps:
      - uses: actions/checkout@v3
      - name: Setup Python
        uses: actions/setup-python@v4
      - run: pip install mkdocs-material
      - run: mkdocs gh-deploy --force
```



Python Testing Overview

Testing is a crucial part of package development that ensures code reliability.

Key Testing Principles

- Write tests before or alongside code (Test-Driven Development)
- Each test should be independent and isolated
- Test both valid and invalid cases
- Keep tests simple and focused
- Aim for high test coverage

Project Structure

```
my_package/  
  src/  
    my_package/  
      __init__.py  
      module.py  
  tests/  
    __init__.py  
    test_module.py
```

pytest Basics

pytest is the most popular testing framework for Python.

Simple Test Example

```
[python]
# src/my_package/calculator.py
class Calculator:
    def add(self, x, y):
        return x + y

# tests/test_calculator.py
def test_add():
    calc = Calculator()
    assert calc.add(2, 3) == 5
    assert calc.add(-1, 1) == 0
    assert calc.add(0, 0) == 0
```

- Test files and functions should start with "test_"
- Uses simple assert statements
- Automatically discovers and runs tests

pytest Features

pytest provides powerful features for testing.

Fixtures and Parametrization

```
[python]
import pytest

@pytest.fixture
def calculator():
    return Calculator()

@pytest.mark.parametrize("x, y, expected", [
    (2, 3, 5),
    (-1, 1, 0),
    (0, 0, 0)
])
def test_add_params(calculator, x, y, expected):
    assert calculator.add(x, y) == expected
```

- Fixtures: Reusable test resources/setup
- Parametrize: Test multiple inputs easily
- Marks: Categorize and select tests

Testing Best Practices

Follow these practices for effective testing.

Example: Testing Exceptions

```
[python]
def test_division_by_zero():
    calc = Calculator()
    with pytest.raises(ValueError) as exc_info:
        calc.divide(10, 0)
    assert str(exc_info.value) == "Cannot divide by zero"

@pytest.mark.skip(reason="feature not implemented yet")
def test_future_feature():
    pass
```

- Test edge cases and exceptions
- Use meaningful test names
- One assertion per test (when possible)
- Skip or mark tests appropriately
- Keep tests fast and deterministic

Test Coverage

Monitor your test coverage to ensure thorough testing.

Using pytest-cov

```
# Install pytest-cov
pip install pytest-cov

# Run tests with coverage
pytest --cov=my_package tests/

# Generate HTML coverage report
pytest --cov=my_package --cov-report=html tests/
```

- Aim for high coverage (typically >80%)
- Coverage isn't everything - focus on quality
- Test critical paths thoroughly
- Use branch coverage for better insight
- Include coverage in CI/CD pipeline



Package Distribution Overview

Different distribution strategies serve different purposes.

Distribution Types

- Open Source: Public PyPI release
- Private Package: Git repository or private PyPI
- Service: Containerized application

Package Configuration (setup.py)

```
1 import os
2 import re
3
4 from setuptools import find_packages, setup
5
6 setup(
7     name="Docs2KG",
8     author="AI4WA",
9     author_email="admin@ai4wa.com",
10    description="Unified Knowledge Graph Construction from Heterogeneous Documents Assisted by Large Language Models",
11    long_description=long_description,
12    long_description_content_type="text/markdown",
13    version=version,
14    packages=find_packages(), # Adjust the location where setuptools looks for packages
15    include_package_data=True, # To include other types of files specified in MANIFEST.in or found in your packages
16    install_requires=read_requirements(),
17    python_requires=">=3.8", # Specify your Python version compatibility
18    classifiers=[
19        # Classifiers help users find your project by categorizing it
20        "Development Status :: 3 - Alpha",
21        "Intended Audience :: Developers",
22        "Intended Audience :: Science/Research",
23        # License: GNU LESSER GENERAL PUBLIC LICENSE
24        "License :: OSI Approved :: GNU Lesser General Public License v3 (LGPLv3)",
25        # Topic
26        "Topic :: Scientific/Engineering :: Artificial Intelligence",
27        "Programming Language :: Python :: 3.8",
28        "Programming Language :: Python :: 3.9",
29        "Programming Language :: Python :: 3.10",
30    ],
31    url="https://docs2kg.ai4wa.com",
32    project_urls={
33        "Source": "https://github.com/AI4WA/Docs2KG",
34    },
35 )
```

Figure: Example of setup.py

Publishing to PyPI

```
name: Publish package
2
3 on:
4   push:
5     tags:
6       - 'v*.*.*'
7
8 jobs:
9   publish:
10     runs-on: ubuntu-latest
11
12     steps:
13       - name: Checkout code
14         uses: actions/checkout@v2
15
16       - name: Set up Python
17         uses: actions/setup-python@v2
18         with:
19           python-version: '3.8'
20
21       - name: Install dependencies
22         run: |
23           python -m pip install --upgrade pip
24           pip install setuptools wheel twine
25
26       - name: Build package
27         run: python setup.py sdist bdist_wheel
28
29       - name: Publish package
30         uses: pypa/gh-action-pypi-publish@v1.5
31         with:
32           twine_username: __token__
33           twine_password: ${{ secrets.PYPI_API_TOKEN }}
```

Figure: CI/CD publish to pypi

- Register on PyPI (pypi.org)
- Test on test.pypi.org first
- Set up CI/CD for releases

Private Package Distribution

Distributing closed-source packages within organizations.

Install from Git

```
[bash]
# Install directly from Git
pip install git+ssh://git@github.com/org/package.git

# Or in requirements.txt
my-package @ git+ssh://git@github.com/org/package.git@v1.0.0

# Private PyPI configuration
[global]
index-url = https://private.pypi.org/simple
```

- Use private Git repositories
- Set up private PyPI server (e.g., DevPI)
- Implement access controls
- Version control with Git tags

Service-based Distribution

Packaging your code as a standalone service.

Dockerfile Example

```
[docker]
FROM python:3.9-slim

WORKDIR /app
COPY . .
RUN pip install .

EXPOSE 8000
ENTRYPOINT ["python", "-m", "my_package.server"]

# Build and run
docker build -t my-service .
docker run -p 8000:8000 my-service
```

- Containerize with Docker
- Define clear service interfaces
- Include configuration management
- Plan for scalability

Deployment Options

Different deployment strategies for different needs.

Serverless

- AWS Lambda
- Google Cloud Functions
- Azure Functions
- Event-driven architecture
- Pay-per-use model

Traditional Hosting

- Docker containers
- Kubernetes clusters
- Virtual machines
- Dedicated servers
- Continuous availability

Distribution Checklist

Essential considerations for package distribution.

Documentation

- Installation instructions
- Usage examples
- API documentation
- Deployment guides
- Troubleshooting tips

Quality Assurance

- Automated tests
- Version compatibility
- Security considerations
- Performance benchmarks
- Dependencies management



Docker GPU

Docker with GPU for Python

Docker containers can leverage GPU capabilities for ML/DL applications.

Prerequisites

- NVIDIA GPU drivers installed
- NVIDIA Container Toolkit (nvidia-docker2)
- Docker installed

Base Images

- nvidia/cuda: Basic CUDA support
- pytorch/pytorch: PyTorch with CUDA
- tensorflow/tensorflow: TensorFlow with GPU

Dockerfile Example

GPU Dockerfile

```
1 FROM nvidia/cuda:11.8.0-runtime-ubuntu22.04
2
3 WORKDIR /app
4
5 # System dependencies
6 RUN apt-get update && apt-get install -y \
7     python3-pip \
8     && rm -rf /var/lib/apt/lists/*
9
10 # Python packages
11 COPY requirements.txt .
12 RUN pip3 install --no-cache-dir -r requirements.txt
13
14 # Copy package
15 COPY . .
16 RUN pip3 install -e .
17
18 # Test GPU availability
19 CMD ["python3", "-c", "import torch; \
20     print(f'GPU Available: {torch.cuda.is_available()}', \
21     Device Count: {torch.cuda.device_count()}')"]
22
```

Dockerfile Example

Run Container

```
1 docker build -t my-gpu-app .  
2 docker run --gpus all my-gpu-app  
3
```



Why License Matters

- No license = No open source
- Affects how others can use your code
- Impacts project adoption and contributions
- Legal implications for both maintainers and users

Popular Licenses

Permissive

- MIT License
 - Simple and short
 - Allows commercial use
 - Minimal restrictions
- Apache 2.0
 - Patent protection
 - Explicit trademark rights

Copyleft

- GPL v3
 - Must share modifications
 - Strong copyleft
- LGPL
 - Library-friendly
 - Allows proprietary linking

Important Considerations

Before Choosing

- Check dependencies' licenses
- Consider commercial compatibility
- Review company policies
- Understand license obligations

Key Questions

- Do you want to allow commercial use?
- Should derivatives be open source?
- Need patent protection?
- Corporate requirements?



Homework

- Setup your package with instruction above
- and implement some modules for your project.

Q&A

Questions?

Feel free to ask anything