Python Dependency Management with uv: Complete Guide

Introduction: The Python Dependency Problem

Python's flexibility comes with a significant challenge: **dependency management**. Unlike languages with built-in package managers, Python's ecosystem has evolved multiple competing solutions, each with trade-offs that create friction for developers, especially beginners.

Core Challenges

Version Conflicts: Different projects require different versions of the same package, leading to "dependency hell" where satisfying one project breaks another.

Environment Isolation: Without proper isolation, installing packages globally can corrupt your system Python or cause conflicts between projects.

Slow Operations: Traditional tools like pip can be painfully slow for dependency resolution and installation, especially with complex dependency trees.

Complex Workflows: Setting up reproducible environments often requires multiple tools and commands, creating barriers for newcomers.

Traditional Tools: Strengths and Limitations

pip (Built-in Package Installer)

Purpose: Basic package installation and management

```
pip install requests numpy
pip freeze > requirements.txt
```

Limitations: No environment isolation, slow dependency resolution, poor conflict handling, no lock files for reproducible builds.

pipx (Isolated Application Installer)

Purpose: Install Python CLI tools in isolated environments

```
pipx install black
pipx install jupyter
```

Limitations: Only for applications, not libraries; doesn't solve project-level dependency management.

Poetry (Modern Dependency Manager)

Purpose: Project-centric dependency management with lock files

bash

```
poetry init
poetry add requests
poetry install
```

Limitations: Slow dependency resolution, complex for simple scripts, heavyweight for quick tasks.

Conda (Data Science Ecosystem)

Purpose: Package and environment management, especially for data science

```
bash

conda create -n myenv python=3.11

conda activate myenv

conda install numpy pandas
```

Limitations: Large installations, slow operations, mixing conda and pip can cause issues.

uv: The Modern Solution

uv is a next-generation Python package manager written in Rust that addresses most traditional pain points:

Key Advantages

- 1. **Speed**: 10-100x faster than pip for most operations
- 2. **Simplicity**: Single tool for multiple use cases
- 3. **Inline Dependencies**: Declare dependencies directly in scripts
- 4. Automatic Environment Management: Creates isolated environments automatically
- 5. Cross-platform: Works consistently across Windows, macOS, and Linux
- 6. Drop-in Replacement: Compatible with existing pip workflows

Environment Setup Guide

1. Installing uv

Windows (PowerShell):

```
powershell
powershell -c "irm https://astral.sh/uv/install.ps1 | iex"
```

macOS/Linux:

```
bash
curl -LsSf https://astral.sh/uv/install.sh | sh
```

Alternative (using pip):

```
pip install uv
```

Verify installation:

```
bash
uv --version
```

2. Python Environment Setup

Install Python versions with uv:

```
bash

# Install latest Python
uv python install

# Install specific version
uv python install 3.11
uv python install 3.12

# List available versions
uv python list
```

Create project environment:

```
bash

# Create new project
uv init my-project
cd my-project

# Or initialize in existing directory
uv init
```

3. Jupyter Notebook Setup

Install Jupyter with uv:

```
# Global installation for system-wide access
uv tool install jupyter

# Or install in project environment
uv add jupyter
```

Launch Jupyter:

```
# From anywhere (if installed as tool)
jupyter notebook

# Or from project directory
uv run jupyter notebook
```

4. VSCode/Cursor Integration

Install Python extension in VSCode/Cursor, then:

1. Select Python Interpreter:

- Open Command Palette (Ctrl+Shift+P))
- Type "Python: Select Interpreter"
- Choose the uv-managed environment (usually in (venv))

2. Integrated Terminal Setup:

```
bash
# VSCode will automatically activate the environment
# Or manually activate:
uv shell
```

3. Jupyter Integration:

• Install Jupyter extension in VSCode

- Create (.ipynb) files that automatically use your project environment
- Or use "Python: Create Jupyter Notebook" command

5. Complete Development Setup

Recommended workflow for new projects:

```
bash
# 1. Create project directory
mkdir my-ai-scripts && cd my-ai-scripts
# 2. Initialize uv project
uv init
# 3. Add common dependencies
uv add requests pandas numpy matplotlib
# 4. Add development tools
uv add --dev jupyter black ruff
# 5. Open in editor
code . # or cursor .
```

Standalone Script Tutorial

Basic Inline Dependencies

Create (fetch_weather.py):

```
# /// script
# requires-python = ">=3.8"
# dependencies = [
     "requests",
     "rich",
#
# 1
# ///
import requests
from rich.console import Console
from rich.table import Table
def get_weather(city):
    """Fetch weather data for a city."""
    url = f"https://wttr.in/{city}?format=j1"
    response = requests.get(url)
    return response.json()
def display_weather(data):
    """Display weather in a nice table."""
    console = Console()
    table = Table(title="Weather Information")
    table.add_column("Property", style="cyan")
    table.add_column("Value", style="green")
```

```
current = data['current_condition'][0]
  table.add_row("Temperature", f"{current['temp_C']}°C")
  table.add_row("Feels Like", f"{current['FeelsLikeC']}°C")
  table.add_row("Humidity", f"{current['humidity']}%")
  table.add_row("Description", current['weatherDesc'][0]['value'])

console.print(table)

if __name__ == "__main__":
  city = input("Enter city name: ")
  weather_data = get_weather(city)
  display_weather(weather_data)
```

Run the script:

```
bash
uv run fetch_weather.py
```

uv will automatically:

- 1. Create a temporary environment
- 2. Install (requests) and (rich)
- 3. Execute the script
- 4. Cache the environment for future runs

Data Analysis Script Example

Create (analyze_data.py):

```
# /// script
# requires-python = ">=3.9"
# dependencies = [
#
      "pandas",
     "matplotlib",
#
     "seaborn",
     "numpy",
#
# 1
# ///
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
from pathlib import Path
def create sample data():
    """Generate sample sales data."""
   np.random.seed(42)
    dates = pd.date_range('2024-01-01', periods=100, freq='D')
    sales = np.random.normal(1000, 200, 100) + np.sin(np.arange(100) * 0.1) * 100
    data = pd.DataFrame({
        'date': dates,
        'sales': np.maximum(sales, 0), # Ensure positive values
        'month': dates.month
```

```
}-)
    return data
def analyze sales(df):
    """Perform basic sales analysis."""
    print("Sales Analysis Report")
    print("=" * 30)
    print(f"Total Sales: ${df['sales'].sum():,.2f}")
    print(f"Average Daily Sales: ${df['sales'].mean():,.2f}")
    print(f"Best Day: {df.loc[df['sales'].idxmax(), 'date'].strftime('%Y-%m-%d')}
    # Monthly analysis
    monthly = df.groupby('month')['sales'].agg(['sum', 'mean']).round(2)
    print("\nMonthly Summary:")
    print(monthly)
def create visualizations(df):
    """Create sales visualizations."""
    fig, axes = plt.subplots(2, 2, figsize=(15, 10))
    fig.suptitle('Sales Analysis Dashboard', fontsize=16)
   # Daily sales trend
    axes[0, 0].plot(df['date'], df['sales'])
    axes[0, 0].set_title('Daily Sales Trend')
    axes[0, 0].set_xlabel('Date')
    axes[0, 0].set vlabel('Sales ($)')
```

```
# Sales distribution
   axes[0, 1].hist(df['sales'], bins=20, alpha=0.7)
   axes[0, 1].set title('Sales Distribution')
   axes[0, 1].set xlabel('Sales ($)')
   axes[0, 1].set vlabel('Frequency')
   # Monthly boxplot
   df.boxplot(column='sales', by='month', ax=axes[1, 0])
   axes[1, 0].set_title('Monthly Sales Distribution')
   # Monthly trend
   monthly_avg = df.groupby('month')['sales'].mean()
   axes[1, 1].bar(monthly_avg.index, monthly_avg.values)
   axes[1, 1].set_title('Average Monthly Sales')
   axes[1, 1].set xlabel('Month')
   axes[1, 1].set_ylabel('Average Sales ($)')
   plt.tight_layout()
   plt.savefig('sales_analysis.png', dpi=300, bbox_inches='tight')
   plt.show()
   print(f"\nVisualization saved as 'sales analysis.png'")
if name == " main ":
   # Generate or load data
   data = create sample data()
```

```
# Perform analysis
analyze_sales(data)

# Create visualizations
create_visualizations(data)
```

Run the analysis:

bash

uv run analyze_data.py

Web Scraping Script Example

Create (scrape_news.py):

```
# /// script
# requires-python = ">=3.8"
# dependencies = [
#
     "requests",
     "beautifulsoup4",
#
     "lxml",
     "rich",
#
# ]
# ///
import requests
from bs4 import BeautifulSoup
from rich.console import Console
from rich.table import Table
from rich.progress import track
from urllib.parse import urljoin
import time
def scrape_hacker_news():
    """Scrape top stories from Hacker News."""
    url = "https://news.ycombinator.com/"
    response = requests.get(url)
    soup = BeautifulSoup(response.content, 'html.parser')
    stories = []
    story_links = soup.find_all('span', class_='titleline')
```

```
for i, story in enumerate(track(story_links[:10], description="Fetching storic")
        try:
            link = story.find('a')
            if link:
                title = link.text.strip()
                href = link.get('href')
                # Handle relative URLs
                if href.startswith('item?'):
                    href = urljoin(url, href)
                stories.append({
                    'rank': i + 1,
                    'title': title[:80] + '...' if len(title) > 80 else title,
                    'url': href
                } )
            time.sleep(0.1) # Be respectful
        except Exception as e:
            print(f"Error processing story {1}: {e}")
            continue
    return stories
def display_stories(stories):
    """Display stories in a formatted table."""
    console = Console()
```

```
table = Table(title=" Top Hacker News Stories")
    table.add_column("Rank", style="cyan", width=6)
    table.add_column("Title", style="white")
    table.add_column("URL", style="blue", max_width=50)
    for story in stories:
        table.add_row(
            str(story['rank']),
            story['title'],
            story['url']
    console.print(table)
if __name__ == "__main__":
    console = Console()
   with console.status("[bold green]Scraping Hacker News..."):
        stories = scrape_hacker_news()
    if stories:
        display_stories(stories)
    else:
        console.print("[red]Failed to fetch stories[/red]")
```

Run the scraper:

```
bash
uv run scrape_news.py
```

Advanced uv Features

Project Management

Initialize with specific Python version:

```
bash
uv init --python 3.11 my-project
```

Add dependencies with version constraints:

```
uv add "requests>=2.28"
uv add "pandas~=2.0"
uv add --dev pytest black ruff
```

Lock dependencies:

```
bash
uv lock
```

Script Management

Run scripts with specific Python version:

```
bash
uv run --python 3.11 script.py
```

Install script dependencies globally:

```
uv tool install --from git+https://github.com/user/repo script-name
```

Environment Commands

```
bash

# Activate shell
uv shell

# Run commands in environment
uv run python
uv run pytest
uv run jupyter notebook

# Sync environment (install/update dependencies)
uv sync

# Export requirements
uv export --format requirements-txt > requirements.txt
```

Best Practices for AI-Generated Scripts

1. Always Use Inline Dependencies

Include the (# /// script) block in every standalone script to ensure portability.

2. Version Pin Critical Dependencies

```
python

# /// script
# dependencies = [
# "requests==2.31.0", # Pin for stability
# "pandas>=2.0,<3.0", # Allow minor updates
# ]
# ///</pre>
```

3. Handle Errors Gracefully

```
try:
    import requests
except ImportError:
    print("Run with: uv run script.py")
    exit(1)
```

4. Use Rich for Better Output

Always include (rich) for better terminal output in user-facing scripts.

5. Project Structure for Multiple Scripts

Troubleshooting Common Issues

Script Won't Run

```
# Check uv installation
uv --version

# Verify Python availability
uv python list

# Run with verbose output
uv run -v script.py
```

Dependencies Not Installing

Clear cache uv cache clean # Force reinstall uv sync --reinstall

Environment Issues

```
# Remove and recreate environment
rm -rf .venv
uv sync
```

Conclusion

uv represents a significant step forward in Python dependency management, particularly for Alassisted development workflows. Its speed, simplicity, and inline dependency features make it ideal for:

- Rapid prototyping with Al-generated scripts
- Educational environments where setup friction must be minimized
- Data science workflows requiring quick experimentation

• **Production scripts** needing reliable dependency management

By adopting uv early in your Python journey, you avoid the complexity and frustration that has historically made Python dependency management challenging, allowing you to focus on writing and running code rather than managing environments.

The inline script format is particularly powerful for AI-generated code, as it makes scripts fully self-contained and immediately executable, removing barriers between idea and implementation.