



DAY 07 — Virtual Environments &

Goal : Manage Python libraries safely

1 The Problem (Why We Need Environments)

Imagine this :

- Project A needs `numpy==1.21`
- Project B needs `numpy==2.0`

If you install both globally **✗** → conflict, errors, broken projects.

2 What Is a Virtual Environment? (Simple)

A **virtual environment** is:

- An **isolated Python workspace**
- Each project gets its **own libraries**.
- Nothing affects your system Python

👉 One project = one environment

3 Creating a Virtual Environment

from your project root:

```
python -m venv venv
```

This creates:

```
venv/
```

4 Activating the Environment

On Windows

```
venv\Scripts\activate
```

On macOS / Linux

```
source venv/bin/activate
```

You'll see:

```
(venv)
```

in your terminal →  environment is active.

5 Installing Packages with `pip`

Once activated:

```
pip install numpy pandas
```

✓ Installed

✓ Safe

✓ Reproducible

Check installed packages:

```
pip list
```

6 Freezing Dependencies (VERY IMPORTANT)

This creates a list of exact versions.

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

Examples:

```
numpy==1.26.4  
pandas==2.1.0
```

Anyone can recreate your environment using:



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

 **This is mandatory for ML projects**

7 Why ML Engineers Care So Much

ML projects depend on:


- numpy
- pandas
- scikit-learn
- tensorflow / pytorch

Wrong versions =  crashes,  incompatibility.


Virtual environments prevent this.

8 What NOT to Do

 Install ML libraries globally

 Forget to activate venv

 Commit `venv/` folder to Github

 No `requirements.txt`

9 `.gitignore`

Create `.gitignore` file:

```
venv/  
__pycache__/  
*.pyc
```

This keeps your repo clean.
