

Data Science Summer School 2025 Python for Data Science

ds3.ai - Dang Ngoc Huy



This course will be quite different



Python is an opensource, general-purpose, scripting language



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Open-Source

- Developed and maintained by a global community
- Free to use for anyone, anywhere
- Constantly improved and updated
- Transparent and collaborative development



Python is an open-source, general-purpose, scripting language

General-Purpose

- You can automate a lot of repetitive tasks with Python.
- Python isn't limited to data science, but it's very popular with data scientists!



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Scripting

- No strict definition, but:
- Think of it as a sequence of instructions or commands
- Automates a task or process
- Like a pipeline: Inputs → Processing → Outputs



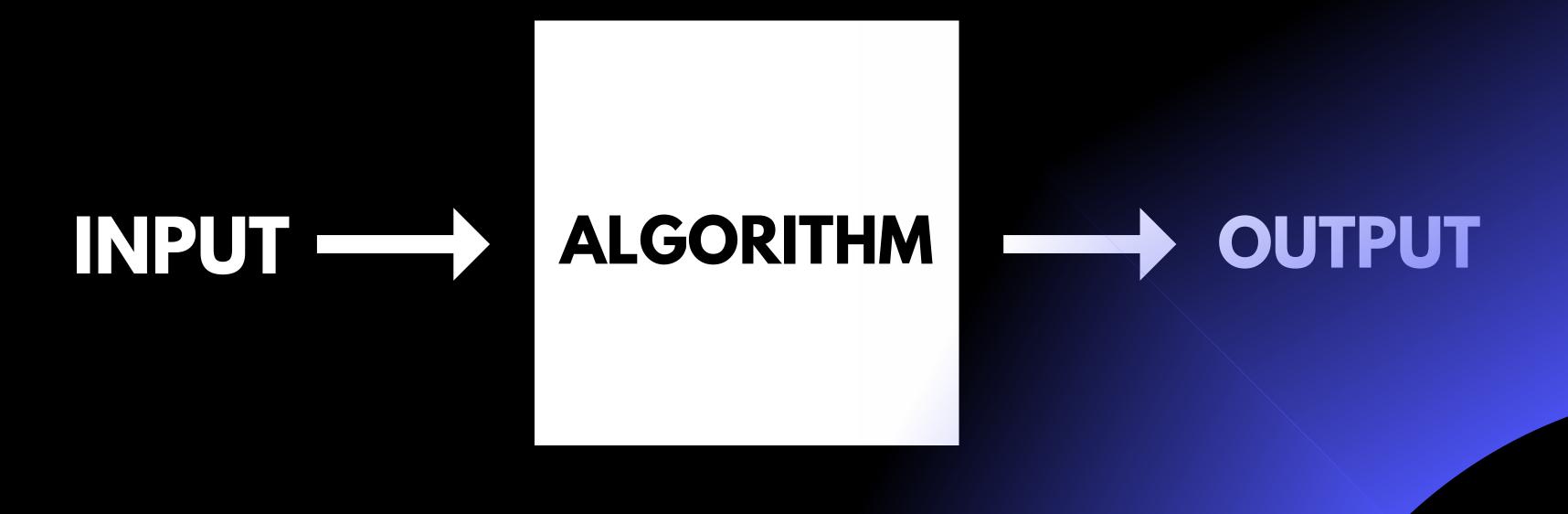
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Language

- Python is a programming language, not a ready-made app
- It follows grammar and syntax rules like any language
- You write commands that the computer interprets and runs
- Because of its flexibility, you can build almost anything with Python!



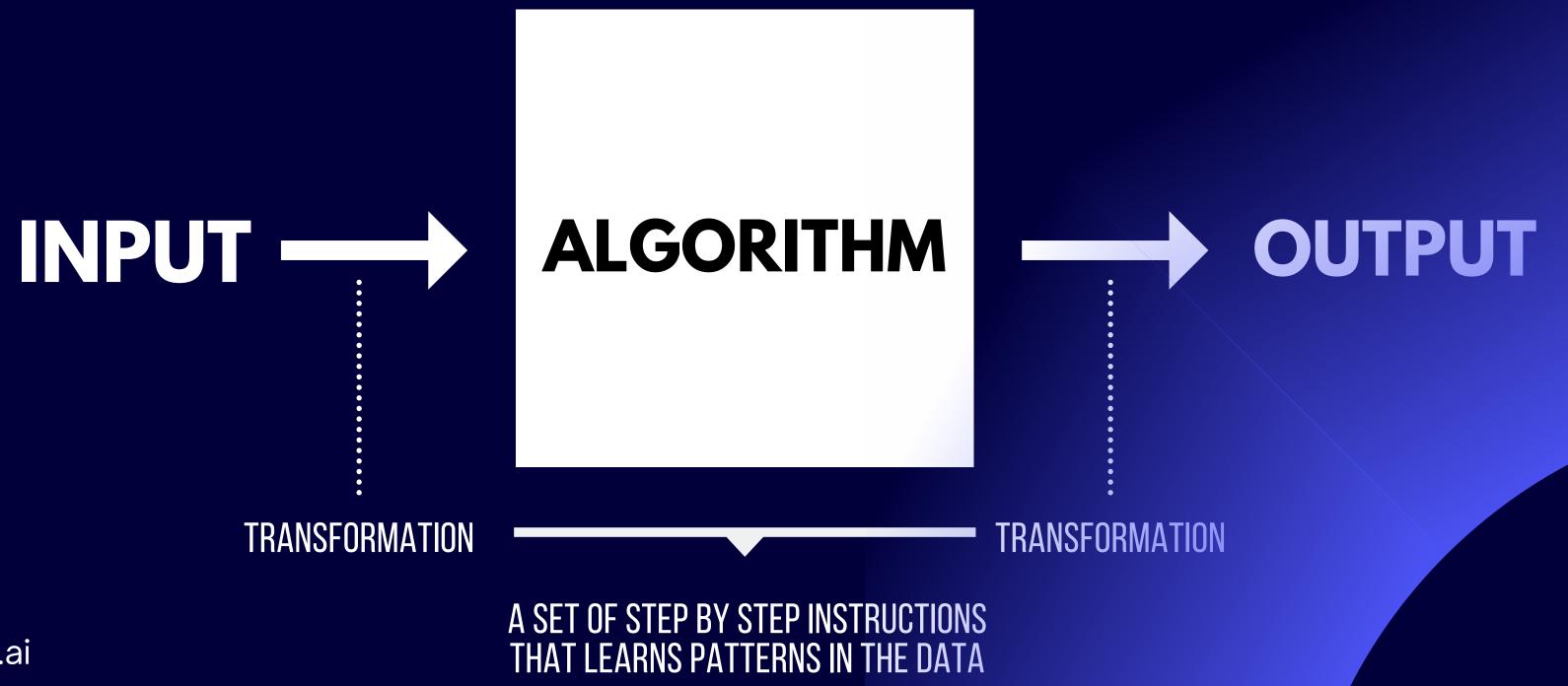
Hertie School Data Science Lab How does scripting work?



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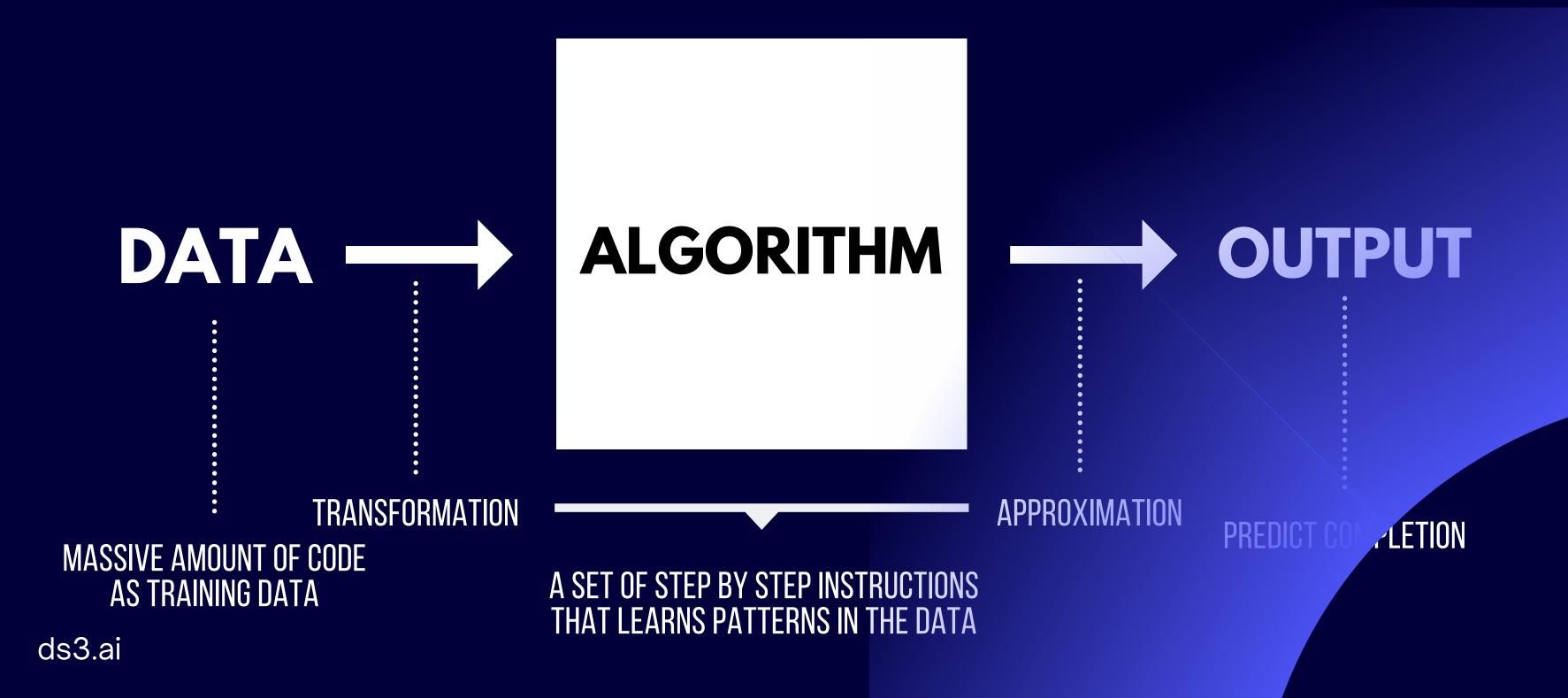
How does scripting work?



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How does Al work?





For Researchers: What Can We Use Python For?

Python is great for automating any repetitive computer-based task.

Here are some common research applications:

- Data Collection: e.g., Web scraping, API access
- Data Cleaning & Analysis: Handle messy datasets, transform data for insights
- Data Visualization: Create plots, dashboards, and visual reports
- Machine Learning: Train models to detect patterns or make predictions
- Deep Learning: Work with neural networks for advanced AI tasks

If it's repetitive and done on a computer — Python can probably automate it.



For Engineers: How Is Research Different from Development?

Aspect	Research	Development
	Scripting, prototyping,	Building full
Usage	interactive work	applications
Goal	Quick experimentation, testing ideas	Production-ready systems
Priorities	Ease of use, speed of iteration	Portability, resource efficiency, deployment readiness

In research, we care about flexibility and speed. In development, we optimize for stability and performance.



Code Editors & IDEs

Jupyter Notebook / JupyterLab

→ Interactive, visual, great for exploration and documentation

VS Code

→ Lightweight, supports extensions (Python, Jupyter, Git, etc.)

Cursor (Al-powered)

→ Built on VS Code, integrates AI to help write, edit, and explain code

Pro Tip: Choose your tools based on your project stage:

→ Use Colab/Jupyter to explore, VS Code/Cursor to refine, and GitHub to collaborate



Cloud-Based Tools

Google Colab

- → Free cloud-hosted Jupyter notebooks
- → No setup needed, GPU/TPU support, great for sharing

Deepnote

- → Real-time collaboration on notebooks (like Google Docs + Jupyter)
- → Clean UI, great for teaching or teamwork

Kaggle Notebooks

→ Online environment with direct access to datasets and competitions



Package & Environment Management

Conda

→ Popular for managing Python/R environments and dependencies

Mamba

→ Fast alternative to Conda

pip + venv

→ Native Python tools, lightweight and simple for small projects



Version Control & Reproducibility

Git + GitHub

→ Track changes, collaborate, and share code

DVC (Data Version Control)

→ Version control for datasets and ML models

ReproZip / Binder

→ Tools to reproduce and share research environments



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R VS. PYTHON – THE ETERNAL DEBATE

A question that comes up often is why the MDS programme focuses on 2 programming languages, when Python is clearly leading the pack as the default language in machine learning, deep learning and many other data and devops workflow. There are a few reasons:

R is still the superior language for building statistical model with a robust ecosystem of packages maintained by scientists and statisticians. This is an area that is still quite lacking in Python;

Python is the desired language for industry roles, but if you want to work for government agencies, research foundations, think tank, international organisations or remain in academia, R (and Excel) is still the go-to tool in many of these institutions. Being proficient in both gives you the best exposure to the job market and the ability to work in & manage cross-language teams;

Being able to code effectively in multiple languages is a sign of good technical talent to potential employers.



Is knowing how to code still useful?

In the age of AI, what's the point of learning how to code when AI can do a much better and faster job than you?

KNOWING HOW TO USE PYTHON EMPOWERS YOU TO UNDERSTAND, BE IN CONTROL, AND BE ABLE BUILD BETTER WITH AI



Is knowing how to code still useful?

Coding alone is no longer the star

What's More Important Now:

Problem Solving:

- → Understanding the why behind the code
- → Framing meaningful questions

Critical Thinking:

- → Choosing the right method, not just using the latest tool
- → Interpreting results in real-world context

Communication:

- → Explaining insights clearly to non-technical audiences
- → Storytelling with data



Is knowing how to code still useful?

From "Coder" to "Scientist"

Being a scientist means:

- Asking the right questions
- Designing experiments
- Making sense of complexity
- Using code as your instrument not your identity

It's the science in data science that sets you apart.



The Wrong Approach

- Binge-watch tutorials
- Read tons of docs
- Take notes...
- Then forget everything a week later



The Right Way: Learn by Doing — With Purpose

Step 1: Start with a specific goal

Learning without a goal = wandering without a map.

Examples:

- "Build a sales forecast system using Python and Machine Learning"
- "Get a job as a Python developer in 6" months"
- "Automate my reports at work using Python scripts"

Ask yourself: Why are you learning to code?



The Right Way: Learn by Doing — With Purpose

Step 2: Create a Smart Roadmap

- Don't learn randomly
- Do research on how to achieve your goal
- Use tools like NotebookLM to:
 - Collect relevant materials
 - Ask focused questions
 - Build a personal learning path
 - Stay organized
- Use a Python roadmap to keep track of what other things you can learn about the language



The Right Way: Learn by Doing — With Purpose

Step 3: Execute and track Your Progress

- Mark what you've done each day
- Log resources used
- Know what's next
- See your growth visually (charts, streaks, checklists)



The Right Way: Learn by Doing — With Purpose

Step 4: Challenge Yourself, Every Day

- Use AI (like ChatGPT) to generate coding challenges
- Apply what you learn immediately
- Write code everywhere, from your coding notebooks to your terminal
- Fail fast, make mistakes and learn deeply:
 Every error = a new brain connection and new lesson



The Right Way: Learn by Doing — With Purpose

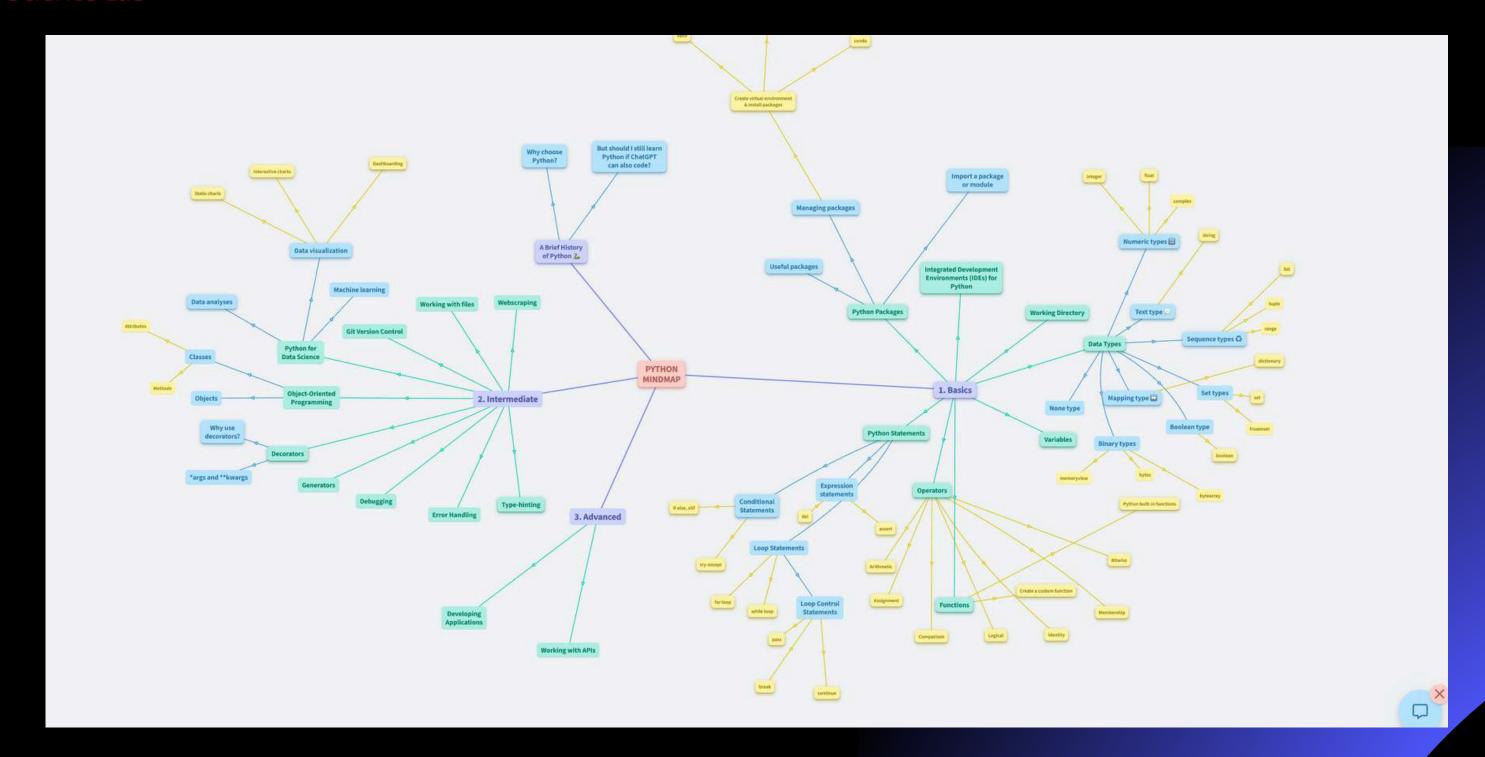
Step 5: Reinforce the cycle of Learn → Practice → Apply → Review

- Don't just learn facts build mental connections
- Reflect: how does this concept relate to what you already know?
- Build a personal "knowledge map"
- Teach others or write your own summary



Code is a tool, not the goal.
Use it to build, solve, explore. That's how you'll remember it for life

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Python Roadmap



Thankyou