

# Introduction to Python

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Md. Aminul Islam Shazid

# Outline

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- 4 Getting Started with Python

# Introduction and History

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# What is Python?

- High-level, interpreted, general-purpose programming language
- Emphasizes readability and simplicity
- Python philosophy: "Readability counts"

# The Zen of Python

Key principles:

- Beautiful is better than ugly
- Simple is better than complex
- Readability counts
- There should be one, and preferably only one, obvious way to do it

Run `import this` to view all of it.

# History of Python

- Created by Guido van Rossum in 1991
- Influences: ABC, Modula-3, Unix shell scripting
- Designed for scripting, prototyping, and general programming

# Original Goals and Evolution

- Original goals: ease of use, readability, “batteries included”
- Evolution:
  - From scripting to web development, data science, AI
  - Strong ecosystem and community growth

# Language Features

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# Basic Language Features

- Interpreted, dynamically typed
- High-level, cross-platform
- Large standard library

# Interpreters and Implementations

- CPython (default, reference implementation)
- PyPy (JIT compilation)
- Cython, Numba (performance optimization)
- MicroPython (embedded)
- Jython (Java), IronPython (.NET)
- Pyodide (CPython ported to WebAssembly), Brython (Python interpreter written in JavaScript)

# Interpreted vs Compiled

- Interpreted: executes directly, easier debugging, slower
- Compiled: precompiled to machine code, faster, less flexible
- While Python is primarily an interpreted language, one can use both approaches via JIT or AOT compilers

# High-Performance Compilers and Tools

- PyPy, Cython, Numba for speed
- GPU-accelerated workloads supported via Polars, cuDF, OneAPI, ROCm
- Distributed computing via Dask, Ray, PySpark, Modin

# Interoperability with Other Languages

- C, C++, Fortran extensions
- Java (via Jython), R (via rpy2)
- Integration for high-performance or legacy code

# Object-Oriented Programming

- Classes, inheritance, polymorphism
- Encapsulation of data and behavior
- Widely used in web frameworks, GUI apps, simulations

# Functional Programming

- Functions are treated as first-class objects
- `map()`, `filter()`, `reduce()` and other higher order functions
- Comprehensions and generators

# Why Learn Python?

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# Why Learn Python

- Simplicity & readability → low entry barrier
- Versatility:
  - Web development: Django, Flask
  - Backend & APIs: FastAPI, Flask
  - Desktop applications: GUI apps, utilities
  - Mobile development: Kivy, BeeWare (smaller adoption)
  - Automation & scripting: one-off scripts, data cleaning, DevOps
- Community & ecosystem → global adoption, extensive libraries

# Python in Action — Famous Apps & Services

- Web: Instagram, Pinterest, Reddit, Spotify, Netflix, YouTube, Quora
- Desktop: Dropbox, BitTorrent, Calibre, Anki, Blender (Python scripting)

# Why Learn Python for Data Science?

- Core libraries: NumPy, Pandas, Polars
- ML/AI frameworks: Scikit-learn, TensorFlow, PyTorch
- Visualization: Matplotlib, Seaborn, Plotly
- Big Data & cloud: PySpark, Dask
- Dominant language for AI research and production

# PyPI and the Library Ecosystem

- Over 500k community made packages
- Domains: web dev, scientific computing, ML/AI, automation
- Easy to install and manage via `pip` and other package managers

# Solving the Two-Language Problem

- Rapid prototyping in Python
- Production optimization via Cython, Numba, Rust, or C++ integration
- Combines speed of compiled languages with ease of Python

# Current State of Python

- Widespread adoption: academia, enterprise, startups
- Strong ecosystem of libraries and tools
- Significant performance improvements in Python 3.13 as well as introduction of an experimental Just-In-Time (JIT) compiler

# Future Prospects and Innovation

- Continued performance improvements (Python 3.14+)
- Wider adoption of GPU acceleration and distributed computing
- Rust integration, WebAssembly, cloud-native tools
- Expansion into embedded and edge devices
- Growth in AI, data science, scientific computing, and high-performance applications

# Python vs Other Languages

- Python vs C: simplicity vs raw performance
- Python vs R: general-purpose vs domain-specific (statistics)
- Complementary use cases; Python bridges prototyping & production



# Getting Started with Python

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# Getting Started with Python

- Easiest option: coding on Kaggle or Google Colab
- To code on local machine:
  - For Windows and MacOS, download Python from [python.org](https://python.org)
  - On most Linux distributions, Python is already installed
  - Once Python is installed, need an IDE
    - VSCode with Python extension recommended
    - Other options include PyCharm, Spyder

Another easy option is to code using [Thonny](#), it is an IDE purpose built for begginers learning Python, comes with Python bundled.