PYTHON

Overview of Python

- Why Python?
- Comparison with R, MATLAB, and other languages
- Applications in Machine Learning, AI, and Big Data

2. Installing Python and Setting Up the Environment

- Downloading and installing Python (Windows, Mac, Linux)
- Setting up Anaconda for Data Science
- Using pip and conda for package management
- Understanding virtual environments (venv, conda env)

3. Jupyter Notebook, Anaconda, Google Colab, VS Code

- Introduction to Jupyter Notebook
- Creating and running notebooks
- Magic commands and markdown cells
- Google Colab: Cloud-based Python environment
- Importing datasets from Google Drive
- VS Code and Python extensions for development
- Terminal vs. IDE-based execution

4. Introduction to Python Syntax and Basic Constructs

- Writing and executing Python scripts
- Understanding indentation and comments
- Python's interactive shell (python, IPython)

5. Variables, Data Types, and Operators

- Declaring and using variables
- Numeric types: int, float, complex
- Text data: str
- Boolean values: True, False
- Type conversion (int(), float(), str())
- Operators:
- Arithmetic: +, -, *, /, %, **
- Comparison: ==, !=, <, >, <=, >=
- Logical: and, or, not
- Assignment: =, +=, -=, *=, /=
- Membership operators: in, not in

6. Conditional Statements (if, else, elif)

- Writing conditional statements
- Nested if statements
- Ternary operators

7. Loops (for, while, break, continue)

- Using for loops with sequences (lists, tuples, strings)
- Using while loops for repetitive execution
- Loop control statements: break, continue, pass

8. Lists, Tuples, and Dictionaries

- Creating lists and tuples
- Accessing elements using indexing and slicing
- Adding, updating, and removing elements
- Iterating over lists and tuples
- Understanding dictionary key-value pairs

• Adding, updating, and deleting dictionary entries

9. Creating and Manipulating Lists

- List methods: append(), extend(), insert(), pop(), remove()
- Sorting and reversing lists (sort(), reverse())
- List comprehension for data manipulation

10. Dictionary Operations and Key-Value Pairs

- Using keys(), values(), items()
- Dictionary comprehension for quick transformations

11. Sets and String Manipulation

- Creating and using sets
- Set operations: Union, Intersection, Difference
- Removing duplicates from lists using sets

12. String Operations

- String formatting (format(), f-strings)
- Splitting, joining, and replacing strings
- Regular expressions using the re module

13. Defining Functions and Lambda Functions

- Defining functions using def
- Function arguments (positional, keyword, default)
- Returning values from functions
- Using *args and **kwargs
- Anonymous functions using lambda

14. Working with Modules and Packages (math, os, sys)

- Importing and using built-in modules (math, os, sys)
- Creating and importing user-defined modules
- Installing external libraries using pip

15. Exception Handling (try-except blocks)

- Handling runtime errors using try-except
- Using finally for cleanup operations
- Raising custom exceptions with raise

16. Iterators, Generators, and Decorators

- Understanding iter() and next()
- Creating custom iterators
- Using yield to create generators
- Understanding Python decorators
- Implementing function decorators (@staticmethod, @classmethod)

17. Reading and Writing Text Files (open(), read(), write())

- Opening files in different modes (r, w, a, rb, wb)
- Reading and writing text files
- Using with statement for safe file handling

18. Working with CSV Files (csv module)

- Reading CSV files using csv.reader()
- Writing CSV files using csv.writer()
- Using Pandas for CSV file manipulation (read_csv(), to_csv())

19. JSON Data Handling (json module)

- Understanding JSON format
- Reading JSON data from files (json.load())
- Writing JSON data (json.dump())
- Converting Python objects to JSON (json.dumps())

DJANGO

1. Introduction to Django

- What is Django?
- Features of Django
- MVC vs MVT Architecture
- Installation and Setup
- Creating a Diango Project
- Understanding manage.py and Django Project Structure
- Running the Django Development Server

2. Django Basics

- Creating a Django App
- Understanding settings.py, urls.py, views.py, models.py
- Working with Django Shell
- Understanding Django Requests and Responses
- URL Routing and Regular Expressions

3. Django Models & Database Handling

- Understanding Django Models
- Creating and Managing Models
- Migrations (makemigrations and migrate)
- Working with SQLite, PostgreSQL, MySQL
- CRUD Operations with Django ORM
- Querying the Database (filter(), get(), exclude(), Q objects)
- Foreign Key and Many-to-Many Relationships
- Using Django Admin Panel

Real-Time Project 1: Blog Application

- User authentication (login/logout/signup)
- Create, update, delete blog posts
- Category and tags for posts
- Comment system
- Admin panel customization

4. Django Views & Templates

- Function-Based Views (FBVs)
- Class-Based Views (CBVs)
- Template Rendering
- Template Inheritance
- Static Files (CSS, JS, Images)
- Template Filters and Tags
- Form Handling in Django
- Django Form Validation

Real-Time Project 2: Portfolio Website

- Dynamic portfolio sections
- Contact form with email functionality
- Admin-controlled content update

5. Django Forms & ModelForms

• Django Forms Basics

- ModelForms for CRUD Operations
- Form Validation
- Handling File Uploads (Images, PDFs)
- Customizing Form Fields and Widgets

Real-Time Project 3: Online Job Portal

- Employer and Job Seeker Registration
- Job Listings & Applications
- Resume Upload
- Email Notifications

6. Django Authentication & Authorization

- Django User Model
- User Registration & Login System
- Password Reset and Change
- User Roles and Permissions
- Custom User Model
- Social Authentication (Google, Facebook)

Real-Time Project 4: E-commerce Website (Part 1)

- User Authentication (Signup/Login)
- Profile Management

7. Django REST Framework (DRF) - API Development

- What is Django REST Framework?
- Serializers
- API Views
- Function-Based vs. Class-Based API Views
- Authentication & Permissions (JWT, Token Authentication)
- Pagination, Filtering, and Throttling
- Testing APIs with Postman

Real-Time Project 5: Task Management System (REST API)

- User Authentication
- CRUD APIs for Tasks
- Assigning Tasks to Users
- Status Updates

8. Django Middleware & Signals

- Custom Middleware in Django
- Built-in Middleware Functions
- Django Signals Pre-save & Post-save

9. Django Caching & Performance Optimization

- Using Memcached & Redis
- Database Indexing & Query Optimization
- Asynchronous Tasks with Celery

10. Django Deployment & DevOps

• Preparing Django Project for Deployment

- Configuring Gunicorn & Nginx
- Deploying on AWS, Heroku, DigitalOcean
- CI/CD Pipelines with GitHub Actions
- Dockerizing a Django Project

Real-Time Project 6: E-commerce Website (Part 2)

- Shopping Cart System
- Order & Payment Integration
- Admin Dashboard
- Deployment on AWS/Heroku

NUMPY

Overview of NumPy

- 1.1 What is NumPy
- 1.2 Importance of NumPy in scientific computing
- 1.3 Key features: speed, efficiency, and functionality
- 1.4 Comparison between Python lists and NumPy arrays
- 1.5 Applications in data science, machine learning, and numerical analysis

Installing and Importing NumPy

- 2.1 Installation using pip and conda
- 2.2 Importing NumPy with alias (import numpy as np)
- 2.3 Verifying the NumPy version (np.__version__)

Creating NumPy Arrays

- 3.1 Array Creation from Existing Data
- 3.1.1 Creating arrays from Python lists or tuples using np.array()
- 3.1.2 Specifying the data type using the dtype parameter
- 3.2 Using Built-in Functions to Create Arrays
- 3.2.1 np.zeros() Arrays filled with zeros
- 3.2.2 np.ones() Arrays filled with ones
- 3.2.3 np.empty() Uninitialized arrays
- 3.2.4 np.full() Arrays filled with a specific value
- 3.2.5 np.arange() Arrays with a range of values
- 3.2.6 np.linspace() Arrays with evenly spaced values
- 3.2.7 np.logspace() Arrays with logarithmically spaced values

Creating Random Arrays

- 4.1 Using np.random for random number generation
- 4.2 Generating random integers, floats, and arrays (np.random.rand(), np.random.randint())
- 4.3 Setting seeds for reproducibility (np.random.seed())

Array Properties

- 5.1 Checking the shape, size, and dimensions of an array
- 5.2 Array data types and dtype

Arithmetic Operations

- 6.1 Element-wise addition, subtraction, multiplication, and division
- 6.2 Broadcasting in NumPy arrays

Mathematical Functions

- 7.1 Sum, mean, median, min, max, and standard deviation
- 7.2 Element-wise functions like np.exp(), np.sqrt(), and np.log()

Comparisons

8.1 Element-wise comparisons (>, <, ==)

8.2 Logical operations (np.any(), np.all())

Introduction to Universal Functions (ufuncs)

- 9.1 What are ufuncs
- 9.2 Importance of ufuncs in NumPy for efficient computations

Commonly Used Universal Functions

- 10.1 Trigonometric functions: np.sin(), np.cos(), np.tan()
- 10.2 Rounding functions: np.round(), np.floor(), np.ceil()
- 10.3 Aggregate functions: np.sum(), np.mean(), np.prod()

Customizing Universal Functions

- 11.1 Using out parameter for in-place operations
- 11.2 Applying universal functions on multidimensional arrays

Indexing and Slicing Arrays

- 12.1 Selecting and Retrieving Data
- 12.2 Slicing
- 12.3 Advanced Indexing

Iterating Over NumPy Arrays

- 13.1 Iterating Over 1D Arrays
- 13.2 Iterating Over 2D and Higher-dimensional Arrays
- 13.3 Iterating with Enumerations

Array Shape Manipulation

- 14.1 Reshaping Arrays
- 14.2 Transposing and Swapping Axes
- 14.3 Expanding and Squeezing Dimensions

Combining and Splitting Arrays

- 15.1 Stacking Arrays
- 15.2 Splitting Arrays

Copies and Views

- 16.1 Understanding Memory Sharing in NumPy
- 16.2 Creating Copies
- 16.3 Practical Examples

Advanced Indexing Techniques

- 17.1 Indexing with Boolean Arrays
- 17.2 Indexing with Arrays of Indices

Broadcasting in NumPy

- 18.1 Introduction to Broadcasting
- 18.2 Practical Examples

Linear Algebra with NumPy

- 19.1 Matrix Operations
- 19.2 Advanced Linear Algebra

Random Numbers and Simulation

- 20.1 Generating Random Numbers
- 20.2 Seeding Random Generators

Saving and Loading Data

- 21.1 Saving Arrays
- 21.2 Loading Arrays

Performance Optimization with NumPy

- 22.1 Vectorization
- 22.2 Memory Efficiency

PANDAS

Introduction to Pandas

- 1. What is Pandas?
- 1.1 Overview of the Pandas library
- 1.2 Importance of Pandas in data analysis and machine learning
- 1.3 Key components of Pandas: Series and DataFrame
- 2. Setting Up the Environment
- 2.1 Installing Pandas using pip or conda
- 2.2 Installing Anaconda and Jupyter Notebook
- 2.3 Importing Pandas and understanding the pd alias
- 3. First Steps with Pandas
- 3.1 Creating Series and DataFrames manually
- 3.2 Loading data from various formats: CSV, Excel, JSON, HTML, SQL, Pickle
- 3.3 Exploring datasets with head(), tail(), info(), and describe()

Pandas Basics

- 4. Data Structures
- 4.1 Series: One-dimensional labeled arrays
- 4.2 DataFrame: Two-dimensional labeled data structures
- 4.3 Index objects: Labels for rows and columns
- 5. Basic DataFrame Operations
- 5.1 Accessing rows and columns using loc and iloc
- 5.2 Modifying data: Adding and removing rows or columns
- 5.3 Transposing data using .T
- 6. Data Selection and Slicing
- 6.1 Selecting rows and columns by labels or positions
- 6.2 Conditional selection using Boolean indexing
- 7. Data Types in Pandas
- 7.1 Identifying data types with dtypes
- 7.2 Converting data types using astype()
- 7.3 Handling categorical data

Data Cleaning and Preprocessing

- 8. Handling Missing Values
- 8.1 Identifying missing data with isnull() and notnull()
- 8.2 Dropping missing values using dropna()
- 8.3 Filling missing values with fillna() (mean, median, or specific values)
- 9. Renaming Columns and Rows
- 9.1 Using rename() for labels
- 9.2 Renaming directly
- 10. Removing Duplicates
- 10.1 Identifying duplicates with duplicated()
- 10.2 Dropping duplicates using drop duplicates()
- 11. Replacing Data
- 11.1 Using replace() to substitute values

Data Transformation

- 12. Sorting Data
- 12.1 Sorting by index using sort_index()
- 12.2 Sorting by values using sort_values()

- 13. Merging, Joining, and Concatenation
- 13.1 Combining datasets using merge() for SQL-like joins
- 13.2 Using concat() for stacking datasets vertically or horizontally
- 13.3 Using join() for index-based joining
- 14. GroupBy Operations
- 14.1 Aggregation functions (sum, mean, count, etc.)
- 14.2 Grouping data based on multiple columns
- 14.3 Custom aggregation functions
- 15. Pivot Tables
- 15.1 Creating pivot tables for multi-dimensional aggregation
- 15.2 Difference between pivot and groupby
- 16. Melt and Stack
- 16.1 Using melt() to convert wide to long format
- 16.2 Transforming hierarchical indices using stack() and unstack()

Advanced Techniques

- 17. Apply and Map Functions
- 17.1 Element-wise operations with apply(), map(), and applymap()
- 18. Crosstab
- 18.1 Creating summary tables of categorical data using pd.crosstab()
- 19. Cut and Binning
- 19.1 Creating intervals for numerical data using cut()
- 19.2 Equal-width vs. custom-width bins
- 20. Datetime Operations
- 20.1 Converting columns to datetime using to_datetime()
- 20.2 Extracting date components (year, month, day, hour, etc.)
- 20.3 Performing date arithmetic and filtering
- 21. Handling Large Datasets
- 21.1 Optimizing memory usage
- 21.2 Processing large datasets in chunks using iterators

Data Analysis and Visualization

- 22. Exploratory Data Analysis (EDA)
- 22.1 Performing summary statistics
- 22.2 Identifying trends and patterns in data
- 23. Data Visualization
- 23.1 Plotting using Pandas: Line, bar, histogram, scatter, and box plots
- 23.2 Integrating Pandas with Matplotlib and Seaborn
- 24. Boolean Indexing
- 24.1 Advanced conditional filtering
- 24.2 Combining multiple conditions using logical operators

Performance Optimization

- 25. Vectorized Operations
- 25.1 Using NumPy under the hood for faster computations
- 26. Working with Sparse Data
- 26.1 Reducing memory usage with sparse matrices
- 27. Profiling and Debugging
- 27.1 Profiling performance with %timeit
- 27.2 Debugging slow code

Real-World Applications

- 28. Case Studies
- 28.1 Analysing real-world datasets: Titanic dataset, financial data, sales data
- 29. Integration with Other Libraries
- 29.1 Using Pandas with Scikit-learn for machine learning preprocessing
- 29.2 Exporting cleaned data to formats like Excel and SQL

MATPLOT LIB AND SEABORN

Module 1: Introduction to Data Visualization

- 1.1 Overview of Data Visualization
 - Importance of visualization in data analysis
 - Types of data visualization (static vs dynamic)
 - Key principles of effective data visualization
 - Tools for data visualization in Python
- 1.2 Introduction to Matplotlib
 - Overview of Matplotlib and its history
 - Installing and setting up Matplotlib
 - Anatomy of a Matplotlib plot (Figure, Axes, Axis)
 - Basic syntax and structure of a Matplotlib plot

Module 2: Basic Plotting with Matplotlib

- 2.1 Basic Plotting Concepts
 - Line plots
 - Scatter plots
 - Bar plots
 - Histogram plots
 - Pie charts
- 2.2 Matplotlib Components
 - Figure and Axes
 - Plot, Subplot, and Grid
 - Axis, ticks, and labels
 - Titles, labels, and legends
- 2.3 Customizing Plots
 - Customizing colors, markers, and line styles
 - Adjusting axis limits and scales (linear, logarithmic)
 - Adding gridlines
 - Annotating plots (text and arrows)
- 2.4 Advanced Plotting
 - Creating subplots and multi-plot grids
 - 3D plotting (3D scatter, surface plots)
 - Saving and exporting plots (PNG, PDF, SVG)
 - Practical Exercise:
 - Create multiple plots (line, scatter, bar, histogram) with customization and annotation.

Module 3: Introduction to Seaborn

- 3.1 Overview of Seaborn
 - Difference between Matplotlib and Seaborn
 - Installing Seaborn
 - Introduction to Seaborn's high-level interface

- Seaborn's built-in themes and color palettes
- 3.2 Seaborn Basics
 - Creating simple plots with Seaborn (line, scatter, bar, etc.)
 - Understanding Seaborn's syntax and default styling
 - Visualizing categorical vs continuous data with Seaborn

Module 4: Advanced Data Visualizations with Seaborn

- 4.1 Seaborn's Advanced Plot Types
 - Pair plots (pairwise relationships)
 - Heatmaps (correlation matrix, clustered heatmap)
 - Box plots, violin plots, and distribution plots
 - Swarm plots and strip plots
 - Facet grids (subplots based on categories)
- 4.2 Customization in Seaborn
 - Customizing color palettes
 - Customizing plot styles and themes
 - Adjusting axes and legends in Seaborn plots
 - Working with multiple categorical variables
- 4.3 Integration of Seaborn with Pandas
 - Visualizing data directly from Pandas DataFrames
 - Handling missing data in visualizations
 - Aggregating data for Seaborn plots

Module 5: Comparing Matplotlib and Seaborn

- 5.1 When to Use Matplotlib vs Seaborn
 - Strengths of Matplotlib
 - Strengths of Seaborn
 - Performance and flexibility considerations
 - Combining Matplotlib and Seaborn for customized visualizations
- 5.2 Case Study: Visualization with Real Data
 - Using both Matplotlib and Seaborn to visualize a real-world dataset (e.g., sales data, weather data, or a dataset from Kaggle)
 - Visualizing trends, distributions, and relationships in the data
 - Customizing the visualization for publication-quality plots

Module 6: Interactive Visualization (Optional)

- 6.1 Introduction to Interactive Visualization
 - Importance of interactive visualizations
 - Using Plotly for interactive plotting
 - Integrating Matplotlib/Seaborn with Plotly
- 6.2 Basic Interactive Plots
 - Creating interactive line, scatter, and bar plots
 - Adding interactivity (zooming, hovering, etc.)
 - Saving interactive plots as HTML