UNIT – 3: Python for Data Processing

3.1 Functional Python Programming  
 **3.1.1 Lambda Functions and Their Utility**

What Is Lambda Function ?

**Lambda functions** are anonymous, inline functions defined using the lambda keyword

**Syntax**

lambda arguments: expression

arguments = inputs to the function

expression = single output returned automatically

**Example**

**1 )** square = lambda x: x \*\* 2

print(square(5))

# Output: 25

**2 )** add = lambda a, b: a + b

print(add(3, 5))

# Output: 8

**Difference Between Normal Function and Lambda Function**

| **Feature** | **Normal Function** | **Lambda Function** |
| --- | --- | --- |
| **Definition** | Defined using the def keyword. | Defined using the lambda keyword. |
| **Name** | Has a name (e.g., def add():) | Anonymous (usually used without a name) |
| **Use Case** | Used for complex or reusable logic. | Used for short, one-time operations. |
| **Readability** | Easier to understand, especially for complex logic. | Less readable if overused or complex. |
| **Return Statement** | Requires an explicit return. | Implicitly returns the expression result. |

Normal Function

def add(x, y):

return x + y

print(add(5, 3))

# Output: 8

Lambda Function

add = lambda x, y: x + y

print(add(5, 3))

# Output: 8

When to Use Which?

Use **normal functions** when:

* The logic is complex.
* You need to reuse or debug the function.
* You want better readability.

Use **lambda functions** when:

* You need a quick, one-line function (e.g., with map(), filter()).
* You don't need to reuse or name the function.

**3.1.2 Map, Filter, and Reduce Functions for Efficient Data Processing**

The map() function is used when you want to apply the same operation to every item in a data set model.

* It saves you from writing a loop, and it's especially helpful when processing **large datasets** during **data cleaning or transformation**.
* Makes your code **shorter and more readable**
* Avoids **for-loops** by applying a function to all items
* Helps in **formatting**, **normalizing**, or **converting data types**
* Often used in **data pipelines** and **ETL (Extract, Transform, Load)** steps

Syntax

map(function, iterable)

function: What to do to each item

iterable: A list, tuple, etc.

Applies a function to every item in an iterable.

Example

**I want to double every number in my list.**

nums = [1, 2, 3]

squares = list(map(lambda x: x \*\* 2, nums))

# Output: [1, 4, 9]

When to use map and when to use for loop ?

map() = cleaner, shorter, good for simple transformations  
for loop = more powerful, better for complex tasks

**Example: Multiply numbers by 2**

**Using map**

numbers = [1, 2, 3, 4]

result = list(map(lambda x: x \* 2, numbers))

print(result)

# Output: [2, 4, 6, 8]

**using for loop**

numbers = [1, 2, 3, 4]

result = []

for x in numbers:

result.append(x \* 2)

print(result)

# Output: [2, 4, 6, 8]

**Why list() is used around map() ?**

If we not use list then ?

result = map(lambda x: x\*\*2, [1, 2, 3])

print(result)

**Output**  
<map object at 0x000001B7A...>

This is **not a list**, just a special object that stores results lazily (to save memory).

**Other Collection Converters Like list()**

| **Function** | **Converts to** | **Description** | **Example** |
| --- | --- | --- | --- |
| **list()** | List | Converts iterable to a list (ordered, mutable) | list(map(...)) → [1, 2, 3] |
| **tuple()** | Tuple | Converts iterable to a tuple (ordered, immutable) | tuple(filter(...)) → (2, 4, 6) |
| **set()** | Set | Converts iterable to a set (unordered, unique elements) | set([1,2,2,3]) → {1, 2, 3} |
| **dict()** | Dictionary | Converts iterable of key-value pairs to a dict | dict([('a', 1), ('b', 2)]) → {'a': 1, 'b': 2} |

**filter(function, iterable)**

The **filter()** function is used when you want to **remove items from a list or iterable** that do **not meet a specific condition**.

* It’s perfect for **cleaning data** by keeping only the valid or useful entries and ignoring the rest
* Helps **clean up dirty data**
* Removes **missing, invalid, or irrelevant entries**
* Makes downstream processing (e.g., training models) more accurate
* Keeps your code **cleaner** than using for loops

Syntax

filter(function, iterable)

function: What to do to each item

iterable: A list, tuple, etc.

Returns items from an iterable for which the function returns True.

Example

**I only want the even numbers from my list.**

nums = [1, 2, 3, 4]

evens = list(filter(lambda x: x % 2 == 0, nums))

# Output: [2, 4]

**reduce(function, iterable)**

The reduce() function is used when you want to combine all elements in a list into a single result — such as a total, product, maximum, or even a custom combination of values.

* you must import reduce() from the **functools** module.
* Takes a **function** and an **iterable** (like a list).
* The **function** must take **two arguments**.
* It applies the function to the **first two elements**, then to the **result and the next element**, and so on…
* Returns a **single final value**.

Syntax

reduce(function, iterable)

function: What to do to each item

iterable: A list, tuple, etc.

Example

#import library  
from functools import reduce

# Function that adds two numbers

def add(x, y):

return x + y

numbers = [1, 2, 3, 4, 5]

result = reduce(add, numbers)

print(result)

# Output: 15

**For Explain Its Working:**

1. add(1, 2) → 3
2. add(3, 3) → 6
3. add(6, 4) → 10
4. add(10, 5) → 15

Question : If my Array like this then what happened  
data = [True, "hello", "world"]

**3.1.3 Using These Functions in Preprocessing Tasks**

**Using map() in Preprocessing**

The map() function is used when you want to apply the same operation to every item in a data set model.

* It saves you from writing a loop, and it's especially helpful when processing **large datasets** during **data cleaning or transformation**.
* Makes your code **shorter and more readable**
* Avoids **for-loops** by applying a function to all items
* Helps in **formatting**, **normalizing**, or **converting data types**
* Often used in **data pipelines** and **ETL (Extract, Transform, Load)** steps

Example: Convert strings to integers

data = ['10', '20', '30']

converted = list(map(int, data))

# [10, 20, 30]

Example: Convert temperatures from Celsius to Fahrenheit

temps\_c = [0, 20, 30]

temps\_f = list(map(lambda c: c \* 9/5 + 32, temps\_c))

# [32.0, 68.0, 86.0]

**Using filter() in Preprocessing**

The **filter()** function is used when you want to **remove items from a list or iterable** that do **not meet a specific condition**.

* It’s perfect for **cleaning data** by keeping only the valid or useful entries and ignoring the rest
* Helps **clean up dirty data**
* Removes **missing, invalid, or irrelevant entries**
* Makes downstream processing (e.g., training models) more accurate
* Keeps your code **cleaner** than using for loops

Example

Datasets contain empty fields that you need to remove.

data = ['apple', '', 'banana', '', 'cherry']

cleaned = list(filter(lambda x: x != '', data))

# Output: ['apple', 'banana', 'cherry']

Example

You could also use bool(x) to remove any "falsy" values like '', None, or 0

data = ['apple', '', None, 'banana']

cleaned = list(filter(bool, data))

# Output: ['apple', 'banana']

Example

Keep Only Valid Emails

emails = ['user@example.com', '', 'not-an-email', 'admin@domain.com']

valid\_emails = list(filter(lambda x: '@' in x and '.' in x, emails))

# Output: ['user@example.com', 'admin@domain.com']

**Using reduce() in Preprocessing**

The reduce() function is used when you want to combine all elements in a list into a single result — such as a total, product, maximum, or even a custom combination of values

* Unlike map() and filter(), you must import reduce() from the **functools** module
* To **summarize or aggregate** data
* To **compress a dataset** into a single value
* Makes your code **cleaner** than writing explicit loops
* Very helpful for **calculating totals**, **finding extremes**, or **combining results**

Example Calculate the Sum of Values

from functools import reduce

data = [10, 20, 30]

total = reduce(lambda x, y: x + y, data)

print(total)

# Output: 60

Example Find the Maximum or Minimum

data = [45, 72, 30, 99, 60]

maximum = reduce(lambda x, y: x if x > y else y, data)

print(maximum)

# Output: 99

Combined Example

from functools import reduce

data = ['10', '20', '', '30', 'not-a-number']

# Step 1: Filter out empty and non-numeric strings

cleaned = filter(lambda x: x.isdigit(), data)

# Step 2: Convert to integers

numbers = map(int, cleaned)

# Step 3: Sum the numbers

total = reduce(lambda x, y: x + y, numbers)

print(total)

# Output: 60

**3.2 Comprehensions for Clean Code**

In data processing and analysis, it's common to loop over data to **filter**, **transform**, or **extract** values

Python offers a powerful feature called **comprehensions**, which allow you to write **concise** (**fewer lines** and **less repetition**) and **expressive (clear and easy to understand)** code, replacing traditional for loops.

There are three main types:

* **List comprehensions**
* **Dictionary comprehensions**
* **Set comprehensions**

**3.2.1 List and Dictionary Comprehensions**

**List Comprehensions :**

A list comprehension allows you to create a new list by writing an expression inside square brackets with a for loop, and optionally an if condition.

* **Quickly creates a new list** by looping through an existing iterable (like a list or range).
* **Allows filtering and transforming** items in a single line of code.

Syntax :

[expression for item in iterable if condition]

expression: what you want to do with each item (e.g. x\*\*2)

iterable: the source list or data you're looping through

condition (optional): a filter that keeps only items that match

1. Create a list of squares

squares = [x\*\*2 for x in range(5)]

# Output: [0, 1, 4, 9, 16]

2. Filter even numbers

numbers = [1, 2, 3, 4, 5, 6]

evens = [x for x in numbers if x % 2 == 0]

# Output: [2, 4, 6]

3. Clean a list of strings and lower

names = [' Alice ', 'BOB', ' Eve']

cleaned = [name.strip().lower() for name in names]

# Output: ['alice', 'bob', 'eve']

**Dictionary Comprehensions**

Dictionary comprehensions work the same way as list comprehensions, but they create key-value pairs using curly braces {}.

* **Quickly builds a dictionary** by defining how to create keys and values from an iterable.
* **Supports filtering and transformation** of key-value pairs.

Syntax

{key: value for item in iterable if condition}

key : What you want to use as the **key** in the new dictionary

value : What you want to use as the **value** in the dictionary

iterable : The data you're looping through — usually a list, tuple, or dictionary

condition : A filter to include only certain items

Examples

Create a dictionary of squares

squares = {x: x\*\*2 for x in range(5)}

# Output: {0: 0, 1: 1, 2: 4, 3: 9, 4: 16}

Filter and modify a dictionary

original = {'a': 1, 'b': 2, 'c': 3, 'd': 4}

filtered = {k: v\*10 for k, v in original.items() if v % 2 == 0}

# Output: {'b': 20, 'd': 40}

**3.2.2 Writing Compact Loops for Data Filtering and Transformation**

Comprehensions allow you to write compact (small) loops that handle both filtering and transformation in a single line — making your code cleaner and easier to maintain

What Is Clean Code?

Clean code is:

* Easy to **read** and **understand**
* Simple and **efficient**
* Helps others (or your future self) understand your logic quickly

What Are Comprehensions?

**Comprehensions** are a **one-line way** to create new lists, sets, or dictionaries from existing data by applying **filters** and **transformations**.

Example 1: Filtering Only Even Numbers

numbers = [1, 2, 3, 4, 5, 6]

even\_numbers = [n for n in numbers if n % 2 == 0]

print(even\_numbers)

#output

[2, 4, 6]

Example 2: Transforming Names to Uppercase

names = ["virat", "rohit", "pant"]

upper\_names = [name.upper() for name in names]

print(upper\_names)

#output

[‘VIRAT’, ‘ROHIT’, ‘PANT’]

Example 3: Filter Students Who Passed and Add Bonus Marks

**Problem**: Select students with marks > 40 and give them 10 bonus marks.

students = [

{"name": "Rajat", "score": 38},

{"name": "Virat", "score": 45},

{"name": "Rohit", "score": 52},

]

passed = [

{"name": s["name"], "score": s["score"] + 10}

for s in students if s["score"] > 40

]

print(passed)

#output

[{'name': ' Virat', 'score': 55}, {'name': ' Rohit ', 'score': 62}]

**3.3 Basics of Data Handling in Python**

**Pandas:**

• It is a package useful for data analysis and manipulation.

• Pandas provide an easy way to create, manipulate and wrangle the data.

• Pandas provide powerful and easy-to-use data structures, as well as the means to quickly perform operations on these structures.

Data scientists use Pandas for its following advantages:

• Easily handles missing data.

• It uses Series for one-dimensional data structure and DataFrame for multi-dimensional data structure.

• It provides an efficient way to slice the data.

• It provides a flexible way to merge, concatenate or reshape the data.

**DATA STRUCTURE IN PANDAS :**

A data structure is a way to arrange the data in such a way that so it can be accessed quickly and we can perform various operation on this data like- retrieval, deletion, modification etc.

Pandas deals with 3 data structure

1. Series

2. Data Frame

3. Panel

Series -

Series is a one-dimensional array like structure with homogeneous data, which can be used to handle and manipulate data. What makes it special is its index attribute, which has incredible functionality and is heavily mutable.

It has two parts

1. Data part (An array of actual data)

2. Associated index with data (associated array of indexes or data labels)



We can say that **Series** is a labeled one-dimensional array which can hold any type of data.

Data of **Series** is always mutable, means it can be changed.

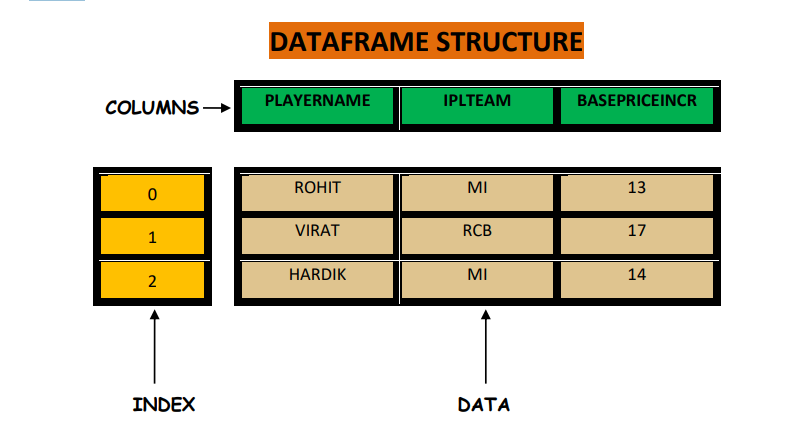
But the size of Data of **Series** is always immutable, means it cannot be changed.

DATAFRAME -

It is a two-dimensional object that is useful in representing data in the form of rows and columns.

It is similar to a spreadsheet or an SQL table. This is the most commonly used pandas object.

Once we store the data into the Dataframe, we can perform various operations that are useful in analyzing and understanding the data.



1. A Dataframe has axes (indices)- ➢ Row index (axis=0) ➢ Column index (axis=1)

2. It is similar to a spreadsheet , whose row index is called index and column index is called column name.

3. A Dataframe Size is Mutable.

4. A Dataframe Data is Mutable.