M.Tech Program

Advanced Industry Integrated Programs

Jointly offered by University and LTIMindTree

Python for Data Science

Knowledge partner

Implementation partner





Modules to be covered

- 1. Python Data Structures, OOPS & Modules
- 2. Python Numpy, Pandas & DS Libraries
- Scala Data Structures, OOPS & Modules
- 4. Scala DS Libraries & Spark
- 5. Factors to be considered for choosing Languages









Topics Covered



Pandas Series

- Pandas Series
- Pandas Math Functions



Pandas Functions and Methods

Data Manipulation Function

- Handling Missing and Data Types
- Data Extraction and Ranking



Advanced Pandas Techniques

Multi Index, Transpose(), Stack(), Unstack(), Melt(), Groupby()



Data Processing, Modelling, and Visualization

- Numpy Scipy Pandas Keras
- Matplotlib Seaborn Bokeh Plotly – Folium.



Pandas

Introduction to Pandas



Pandas DataFrame

- Pandas Dataframe
- Methods and Attributes



Text Data in Pandas

- Working with Text Data
- Common string methods
- The split method



Data Mining

Scrapy – Beautiful Soup

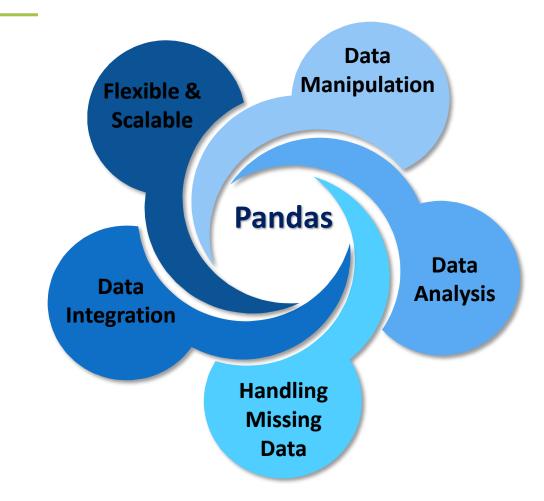






Introduction to Pandas

- Open-source library, built on top of the NumPy library.
- Used for data manipulation and analysis.
- Pandas is a powerful and versatile library that simplifies the tasks of data manipulation in Python.







Pandas Series

• A Series in Pandas is a One-Dimensional Labeled Array which comprises any datatypes such as integer, float, strings, python object and so on.

| | Series | |
|---|--------|--------------|
| а | 10 | |
| b | 20 | |
| С | 30 | Series Value |
| d | 40 | |





Python Built-In Functions

•sort_values() : Sorts a pandas DataFrame by the specified column(s).

•sort_index() : Sorts a pandas DataFrame or Series by its index.

•get() : Retrieves a value from a dictionary for a given key, with an

optional default.

•copy() : Creates a shallow copy of an object.

•apply() : Applies a function along an axis of a pandas DataFrame or Series.

map() : Applies a function to each item of a sequence or pandas Series.



Pandas Math Functions

- 1. Arithmetic: `add()`, `sub()`, `mul()`, `div()` for element-wise operations.
- **2. Aggregation:** `sum()`, `mean()`, `median()`, `std()`, `var()` for summarizing data.
- **3. Cumulative:** `cumsum()`, `cumprod()`, `cummin()`, `cummax()` for cumulative calculations.
- 4. Descriptive: `describe()`, `quantile()` for statistical summaries.
- **5. Correlation and Covariance:** `corr()`, `cov()` for relationship analysis between columns.

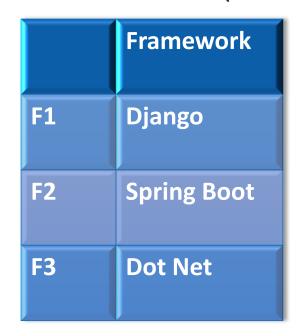


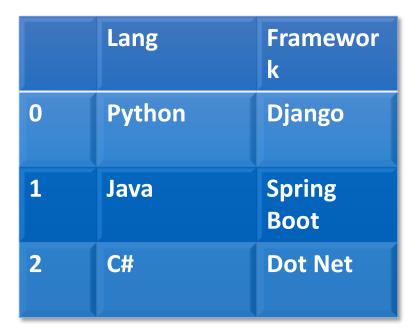
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Pandas Dataframe

• Pandas DataFrame is a two-dimensional, size-mutable, and heterogeneous tabular data structure with labeled axes (rows and columns).

| | Lang |
|----|--------|
| P1 | Python |
| P2 | Java |
| P3 | C# |





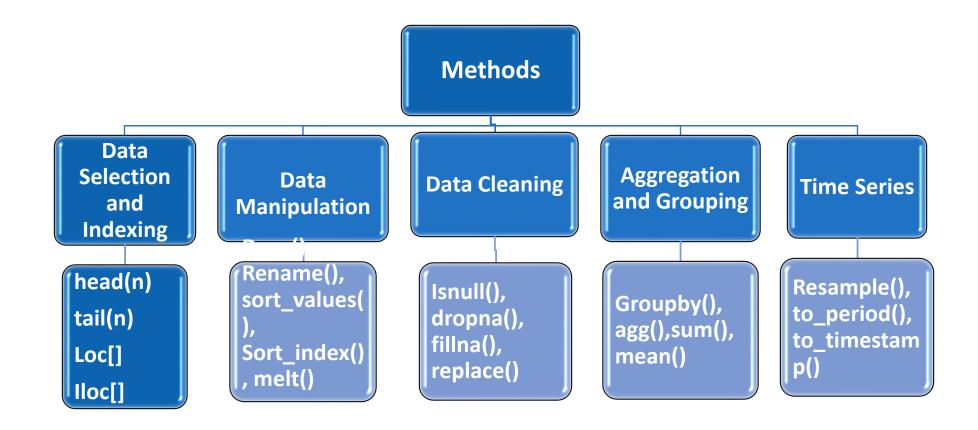
Series 1 + Series 2 = Data Frame





Methods and Attributes between Series and DataFrames

Methods







Python – Introduction to Pandas and Data Science Libraries

Methods and Attributes between Series and DataFrames

Returns a tuple **Attributes Shape** Returns the number of dimensions Returns the index ndim Index **DataFrame** Returns True if the DataFrame is empty empty columns Returns the column labels **Attribute** Size dtype Returns the number of elements Returns the data types of each column values





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Pandas Missing Value- fillna() Method

- Impute missing values using statistical measures.
- Pandas .fillna handles missing values strategically.
- Data scientists replace NaN values with .fillna.





The astype() Method

- It type cast a Pandas object (such as a DataFrame or Series) to a specified data type.
- It's particularly useful when you need to change the data type of specific columns or multiple columns simultaneously
- df.astype(dtype, copy=None, errors='raise')





Rank Series Values with the rank Method

Some of the Data Manipulation and Transformation in Pandas

Filter Data

Used to filter data in a
DataFrame.
Boolean Indexing
query Method
loc Method and iloc Method

fillna Method

fillna()

Impute missing values using statistical measures.

astype Method

astype()

Cast a Pandas object (such as a DataFrame or Series) to a specified data type.





Filtering data and methods in Dataframe

Some of the Data Manipulation and Transformation in Pandas

Rank Series Values

rank()

compute numerical data ranks along a specified axis





Data Extraction in Data Frames

Selecting Columns

- Extract specific columns using column names.
- df['column_name']

Filtering Rows

- Extract rows based on conditions using boolean indexing.
- df[df['column_name'] > value]

Locating Data

- Use loc for label-based indexing and slicing.
- df.loc[row_label, 'column_name']

Position-Based Extraction

- Use iloc for position-based indexing and slicing.
- df.iloc[row_index, column_index]

Querying Data

- Extract data using the query method with string expressions.
- df.query('column_name > value')





Working with Text Data

String Methods

df['column_name'].str.lower() for converting to lowercase.

Extracting Substrings

df['column_name'].str.extract(r'(\d+)') to extract digits.

Replacing Text

df['column name'].str.replace('old', 'new').

Splitting Text

df['column_name'].str.split(' ', expand=True)

Finding Patterns

df[df['column_name'].str.contains('pattern')]





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Common String Methods

| Method | Description | Example |
|---------------------------------------|---|---|
| str.lower() | Converts all characters to lowercase | <pre>df['column_name'].str.lower ()</pre> |
| str.upper() | Converts all characters to uppercase | <pre>df['column_name'].str.upper ()</pre> |
| str.len() | Computes the length of each string | df['column_name'].str.len() |
| str.strip() | Removes leading and trailing whitespace | df['column_name'].str.strip() |
| str.replace(pattern , replacement) | Replaces occurrences of a pattern with a replacement string | df['column_name'].str.replac e('old', 'new') |



Common String Methods – Split Method

| Method | Description | Example |
|------------------------|---|--|
| str.contains(pattern) | Checks if each string contains a pattern | <pre>df['column_name'].str.contains(' pattern')</pre> |
| str.split(separator) | Splits each string by the given separator | df['column_name'].str.split(' ') |
| str.startswith(prefix) | Checks if each string starts with the given prefix | <pre>df['column_name'].str.startswith ('prefix')</pre> |
| str.endswith(suffix) | Checks if each string ends with the given suffix | <pre>df['column_name'].str.endswith(' suffix')</pre> |
| str.extract(pattern) | Extracts substrings matching a regular expression pattern | <pre>df['column_name'].str.extract(r'(\d+)')</pre> |



Multi Index Module

- MultiIndex in pandas allows organizing data with hierarchical row and column labels.
- `pd.MultiIndex` module provides functions for creating, manipulating, and sorting MultiIndex objects.
- Functions like `from_tuples()`, `from_arrays()`, and `from_product()` create MultiIndex from different data structures.
- Methods like `get_level_values()`, `set_levels()`, and `swaplevel()` allow manipulation of MultiIndex levels.
- MultiIndex is essential for analyzing complex datasets with multiple dimensions or categories efficiently.



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Advanced Pandas Techniques



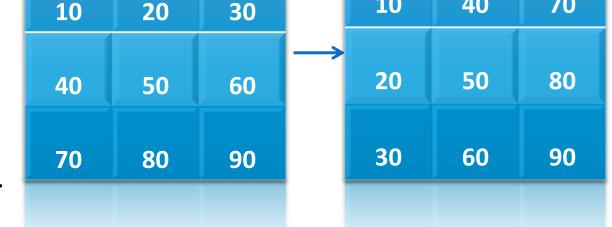


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Python - Numpy, Pandas & DS Libraries

Transpose() Method

- The **transpose** method used to swap rows and columns in a DataFrame.
- Interchanges Rows and Columns
 - Converts rows into columns and vice versa.
- Access Using .T Attribute
 - Use .T for quick DataFrame transposition.
- Useful for Data Reshaping
 - Helpful in restructuring and analyzing data.



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- **Retains Original Data Types**
 - Keeps data types intact during transposition.

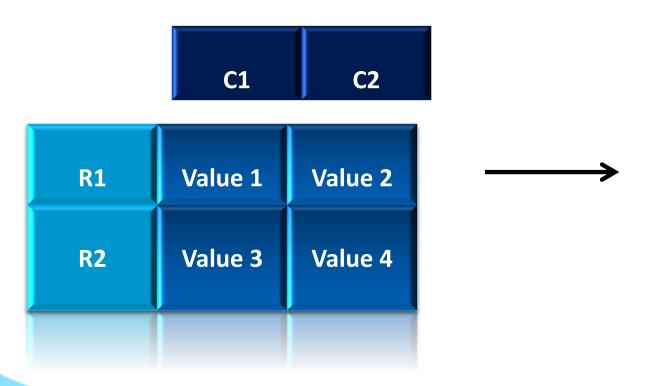


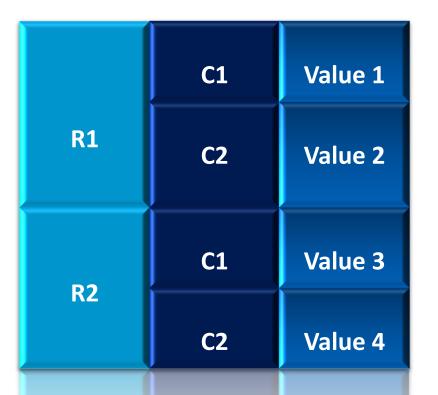
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Stack() Method

- Transform a DataFrame from a wide format to a long format.
- Easy to analyze and work with data.







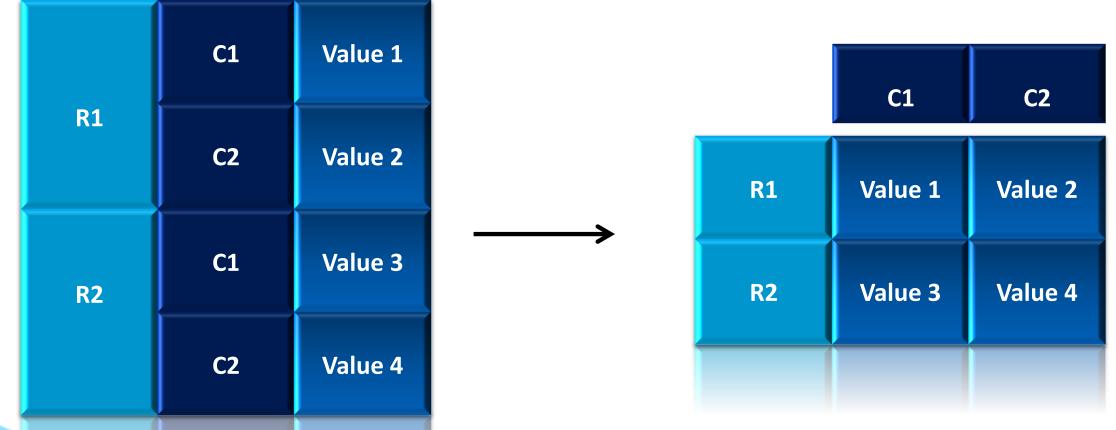


Unstack() Method

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Unstack() Method is reverse of the Stack() Method.



Melt() Method

Reshapes DataFrame from wide to long format.

| ID | NAME | MAT | PHY | СНЕМ |
|----|-------|-----|-----|------|
| 1 | Berly | 85 | 90 | 82 |
| 2 | Jessy | 90 | 95 | 92 |

| ID | NAME | SUBJ | MARKS |
|----|-------|------|-------|
| 1 | Berly | MAT | 85 |
| 2 | Jessy | MAT | 90 |
| 3 | Berly | PHY | 90 |
| 4 | Jessy | PHY | 95 |
| 5 | Berly | CHEM | 82 |
| 6 | Jessy | CHEM | 92 |





Knowledge Check

Which of the following optimizations does Pandas provide for handling large datasets?

- A. Integration with Dask for out-of-core computations.
- B. Automatically parallelizing operations across multiple CPU cores.
- C. Utilizing GPU for data processing tasks.
- D. Converting data frames to more memory-efficient structures like NumPy arrays.





Knowledge Check

Which of the following optimizations does Pandas provide for handling large datasets?

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- D. Converting data frames to more memory-efficient structures like NumPy arrays.

Ans: Integration with Dask for out-of-core computations.





Knowledge Check

Given a DataFrame df with a column 'A', which of the following statements correctly sorts the values of column 'A' in descending order and then returns the DataFrame with the sorted index?

- A. df.sort_values('A', ascending=False).sort_index()
- B. df.sort_index().sort_values('A', ascending=False)
- C. df.sort_values('A', ascending=True).sort_index(ascending=False)
- D. df.sort_values('A', ascending=False).set_index('A')



Knowledge Check

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- C. df.sort_values('A', ascending=True).sort_index(ascending=False)
- D. df.sort_values('A', ascending=False).set_index('A')

Ans: df.sort values('A', ascending=False).sort index()





Knowledge Check

When converting a DataFrame column to a specific data type using the astype method, which of the following commands correctly converts column 'D' to a categorical type?

- A. df.astype({'D': 'category'})
- B. df['D'] = df['D'].astype('category')
- C. df['D'].astype('categorical')
- D. df.convert_dtype('D', 'category')





Knowledge Check

When converting a DataFrame column to a specific data type using the astype method, which of the following commands correctly converts column 'D' to a categorical type?

- A. df.astype({'D': 'category'})
- B. df['D'] = df['D'].astype('category')
- C. df['D'].astype('categorical')
- D. df.convert_dtype('D', 'category')

Ans: df['D'] = df['D'].astype('category')





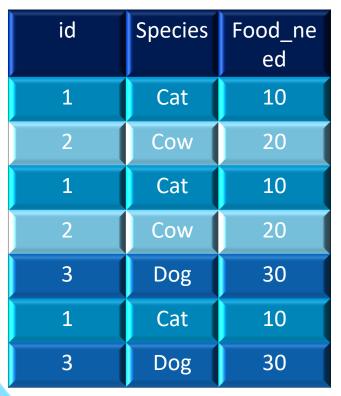
Groupby() Module

Groups DataFrame by specified column, allowing for aggregation operations on each group.





Groupby() Module



Groupby ("Id")



| id | Speci es | Food_ need | |
|----|-------------|---------------|--|
| 2 | Cow | 20 | |
| 2 | Cow | 20 | |

| id | Speci es | Food_ need |
|----|-------------|---------------|
| 3 | Dog | 30 |
| 3 | Dog | 30 |

Aggregation

| id | Spe cies | Foo d_n eed |
|----|-------------|-------------------|
| 1 | Cat | 30 |



| id | Spe cie s | Foo d_n eed |
|----|-----------------|-------------------|
| 3 | Do g | 60 |

Combine

| | id | Spec ies | Foo d_n eed |
|---|----|-------------|-------------------|
| > | 1 | Cat | 30 |
| | 2 | Cow | 40 |
| | 3 | Dog | 60 |

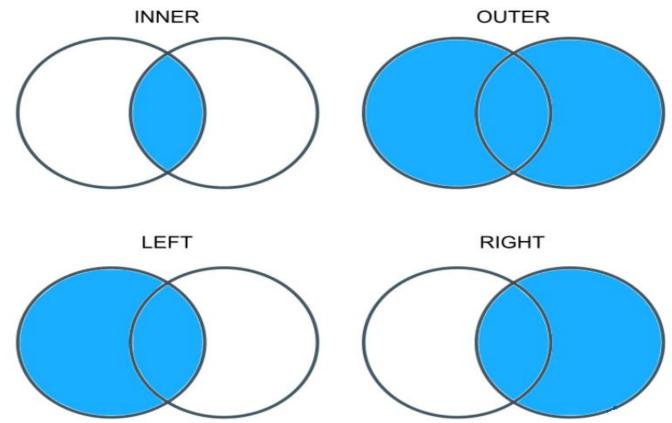




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Merging Dataframes

Combines two or more DataFrames into a single DataFrame based on a common column or index.







Merging Dataframes

Combines two or more DataFrames into a single DataFrame based on a common column or index.

Example: Unique Key combination





Merging Dataframes

Κ2

Κ0

C3

D3

Combines two or more DataFrames into a single DataFrame based on a common column or index.

```
Example: Multiple Join keys
           key1 key2
             Κ0
                   Κ0
                       Α0
                            В0
             Κ0
                   Κ1
                       Α1
                            В1
Left
                   Κ0
             Κ1
                       Α2
                            В2
                                                                        key1 key2
             Κ2
                   Κ1
                       Α3
                            В3
                                                                                  Α0
                                                                                      В0
                           pd.merge(left, right, on=["key1", "key2"])
                                                                                   A2
                                                                                      В2
                                                                          K1
                                                                               Κ0
                                                                                   A2 B2 C2 D2
            key1 key2
              Κ0
                    Κ0
                        C0
                             D0
Right
              Κ1
                    Κ0
                        C1
                             D1
              Κ1
                        C2
                    Κ0
                             D2
```





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Merging Dataframes

The how argument to merge specifies how to determine which keys are to be included in the

resulting table.

Example: A key combination does not appear in either the left or right tables → the value in join table NaN key1 key2 Κ0 Α0 В0 Left Κ0 Κ1 Α1 B1 Κ0 A2 3 Κ2 Κ1 Α3 В3 pd.merge(left, right, how="left", on=["key1", "key2"])

| | ı | key1 | key2 | C | D |
|-------|---|------|------|----|----|
| Diah+ | 0 | Κ0 | Κ0 | C0 | D0 |
| Right | 1 | Κ1 | Κ0 | C1 | D1 |
| | 2 | Κ1 | Κ0 | C2 | D2 |
| | 3 | K2 | Ka | C3 | DЗ |

| | key1 | key2 | Α | В | С | D |
|---|------|------|----|----|-----|-----|
| 0 | K0 | K0 | A0 | В0 | C0 | D0 |
| 1 | K0 | K1 | A1 | B1 | NaN | NaN |
| 2 | K1 | K0 | A2 | B2 | C1 | D1 |
| 3 | K1 | K0 | A2 | B2 | C2 | D2 |
| 4 | K2 | K1 | A3 | В3 | NaN | NaN |



Merging Dataframes

The how argument to merge specifies how to determine which keys are to be included in the resulting table.

key1 key2 Κ0 Κ0 Α0 В0 Left Κ0 Κ1 Α1 В1 Κ1 Κ0 Α2 В2 3 Κ2 Κ1 Α3 В3 Example : A key combination does not appear in either the left or right tables \rightarrow the value in join table NaN

| pd.merge | (left, ri | ght, ho | w="righ | nt". on=l | ["kev1" | . "kev | <i>1</i> 2"1) |
|-----------|-----------|-----------|----------|-----------|---------|---|---------------|
| Pallicipe | | יוי נייום | יתיי ייי | , | | , ,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,, | , — |

key1 key2 Right Κ0 Κ0 C0 D0 Κ1 Κ0 C1 D1 Κ1 Κ0 C2 D2 K2 Κ0 C3 D3

| 0 | K0 | K0 | A0 | В0 | C0 | D0 |
|---|----|----|-----|-----|----|----|
| 1 | K1 | K0 | A2 | B2 | C1 | D1 |
| 2 | K1 | K0 | A2 | B2 | C2 | D2 |
| 3 | K2 | K0 | NaN | NaN | C3 | D3 |

key1 key2





Merging Dataframes

The how argument to merge specifies how to determine which keys are to be included in the resulting table.

| | | key1 | key2 | Α | В | |
|------|---|------|------|----|----|--|
| | 0 | Κ0 | K0 | Α0 | В0 | |
| Left | 1 | Κ0 | K1 | A1 | B1 | |
| | 2 | K1 | K0 | A2 | B2 | |
| | 3 | K2 | K1 | А3 | В3 | |

Example : A key combination does not appear in either the left or right tables \rightarrow the value in join table NaN

| pd.merge(left, | right, how=' | 'outer", on: | =["key1", | "key2"]) |
|----------------|--------------|--------------|-----------|----------|

| | | key1 | key2 | C | D | |
|-------|---|------|------|----|----|--|
| D : 1 | 0 | Κ0 | K0 | C0 | D0 | |
| Right | 1 | K1 | Κ0 | C1 | D1 | |
| | 2 | Κ1 | K0 | C2 | D2 | |
| | 3 | Κ2 | Κ0 | C3 | D3 | |

| | key1 | key2 | Α | В | С | D |
|---|------|------|-----|-----|-----|-----|
| 0 | K0 | K0 | A0 | В0 | C0 | D0 |
| 1 | K0 | K1 | A1 | B1 | NaN | NaN |
| 2 | K1 | K0 | A2 | B2 | C1 | D1 |
| 3 | K1 | K0 | A2 | B2 | C2 | D2 |
| 4 | K2 | K1 | А3 | В3 | NaN | NaN |
| 5 | K2 | K0 | NaN | NaN | C3 | D3 |





Merging Dataframes

The how argument to merge specifies how to determine which keys are to be included in the resulting table.

key1 key2 Κ0 Α0 В0 Κ0 Κ1 Α1 В1 Left Κ1 Κ0 A2 В2 Κ2 Κ1 Α3 В3

Example : A key combination does not appear in either the left or right tables → the value in join table NaN

pd.merge(left, right, how="inner", on=["key1", "key2"])

key1 key2 Κ0 C0 D0 Right Κ1 Κ0 C1 D1 Κ1 Κ0 C2 D2 Κ2 Κ0 С3 D3

| | key1 | key2 | Α | В | С | D |
|---|------|------|----|----|----|----|
| 0 | K0 | K0 | A0 | В0 | C0 | D0 |
| 1 | K1 | K0 | A2 | B2 | C1 | D1 |
| 2 | K1 | K0 | A2 | В2 | C2 | D2 |



Knowledge Check

Which of the following best describes the effect of the stack() method on a DataFrame with hierarchical columns?

- A. It merges all columns into a single column.
- B. It pivots the columns to rows, removing one level of column hierarchy.
- C. It creates a new DataFrame by combining rows with similar index values.
- D. It transposes the DataFrame.





Knowledge Check

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- B. It pivots the columns to rows, removing one level of column hierarchy.
- C. It creates a new DataFrame by combining rows with similar index values.
- D. It transposes the DataFrame.

Ans: It pivots the columns to rows, removing one level of column hierarchy.





Knowledge Check

When performing a merge operation, which argument would you use to specify the type of join (e.g., inner, outer, left, right)?

- A. how
- B. method
- C. type
- D. join_type





Knowledge Check

When performing a merge operation, which argument would you use to specify the type of join (e.g., inner, outer, left, right)?

A. how

- B. method
- C. type
- D. join_type

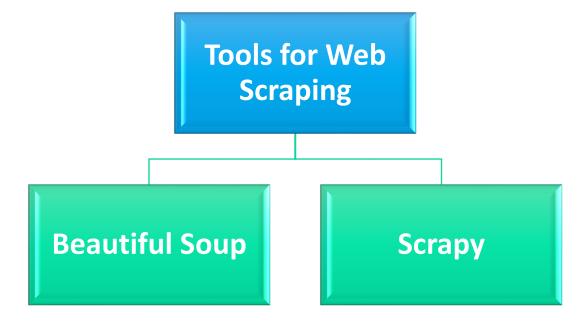
Ans: how





Data Mining(Scrapy - Beautiful Soup)

- Data mining is the process of discovering patterns and knowledge from large amounts of data.
- In the context of web scraping, it involves extracting useful information from websites.

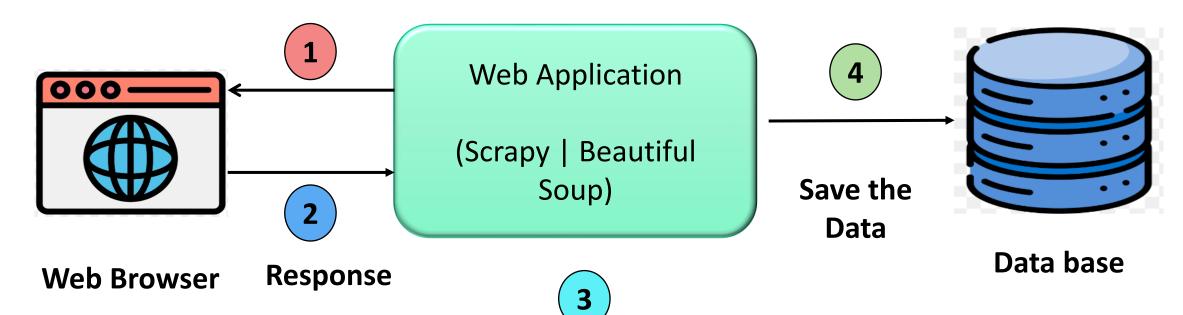






Data Mining(Scrapy – Beautiful Soup)

Request



Extract Data from Response





Data Processing and Data Modeling





Data Processing and Data Modeling

Data Processing

The series of operations on data to retrieve, transform, or classify information.

Data Modelling

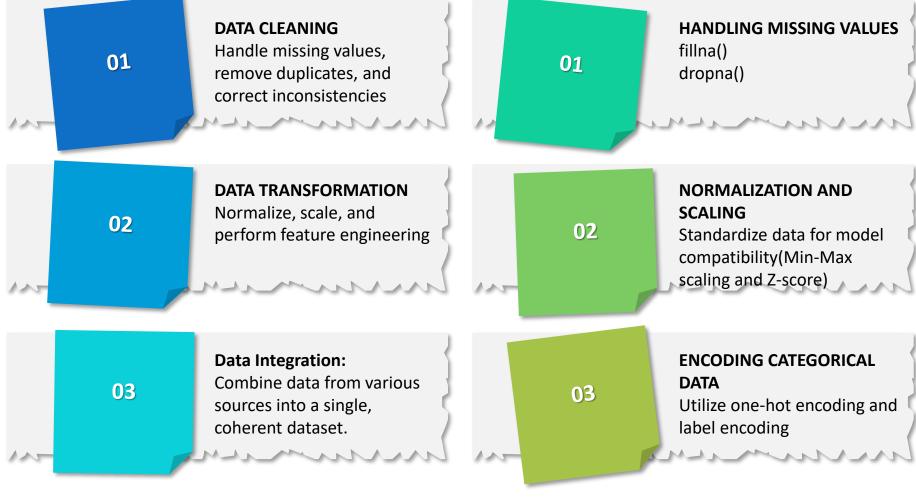
The process of creating a data model for the data to be stored in a database, representing the data structures and relationships.





DATA PROCESSING STEPS

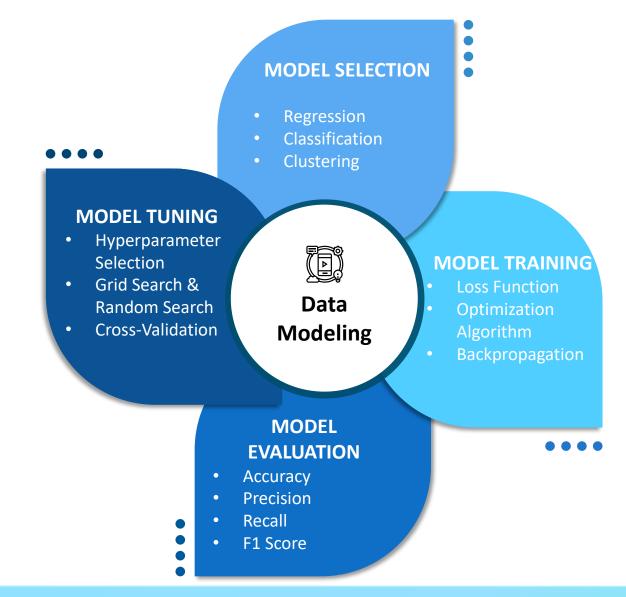
Data Processing



L&T EduTech DATA PROCESSING TECHNIQUES

LTIMindtree

Data Modelling







Data Processing and Data Modeling

- NumPy: Numerical computing in Python.
- SciPy: Scientific computing with Python.
- Pandas: Data manipulation and analysis.
- Keras: High-level neural networks API.
- Scikit-learn: Machine learning in Python.
- **PyTorch:** Deep learning research platform.
- TensorFlow: Open-source machine learning framework.
- XGBoost: Optimized gradient boosting library.





Data Visualization



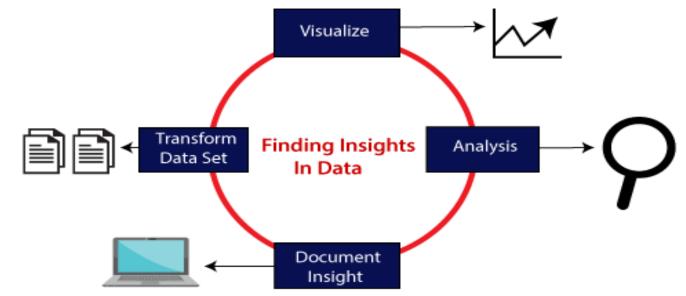


Data Visualization

Data visualization is the graphical representation of information and data using visual elements like charts, graphs, and maps.







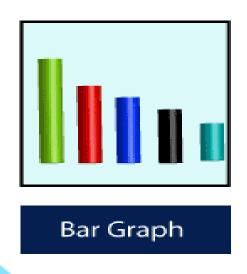


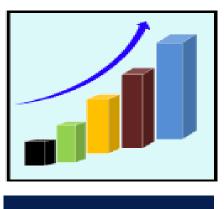


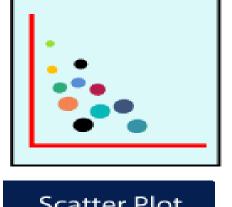
Data Visualization

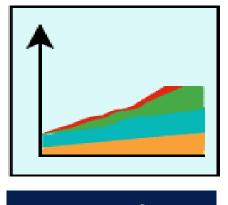
 Graphics provides an excellent approach for exploring the data, which is essential for presenting results.

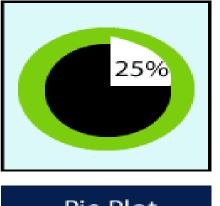
There are five key plots that are used for data visualization:











Histogram

Scatter Plot

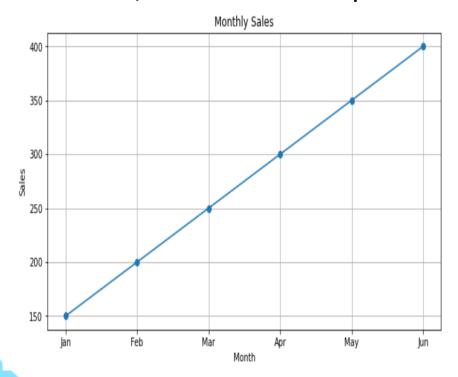
Area Plot

Pie Plot

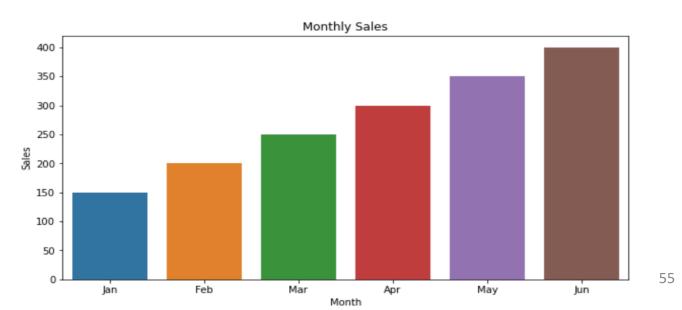


Data Visualization – Matplotlib and Seaborn

Matplotlib is a basic plotting library in Python that provides a wide range of static, animated, and interactive plots.



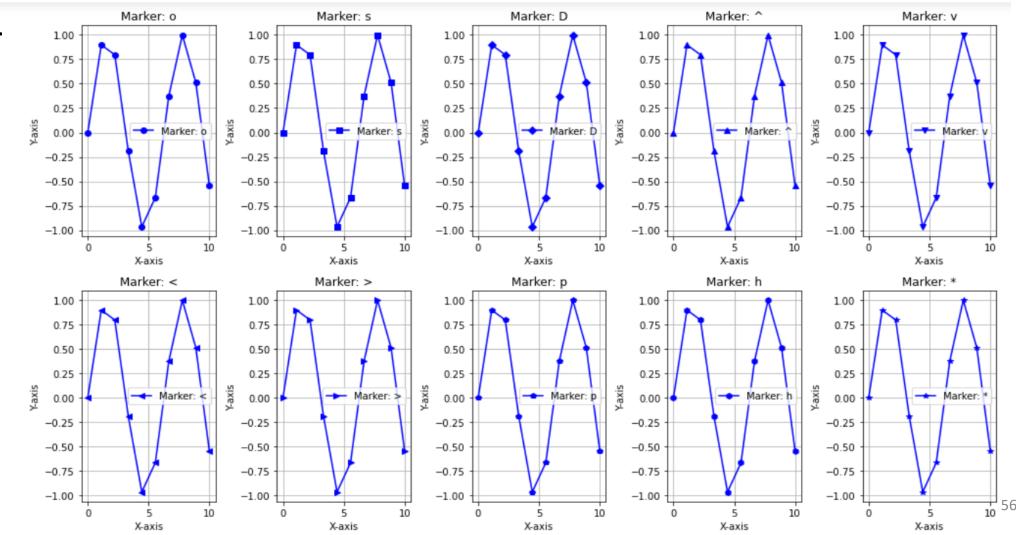
Seaborn is built on top of Matplotlib and provides a high-level interface for drawing attractive and informative statistical graphics





Data Visualization – Matplotlib and Seaborn

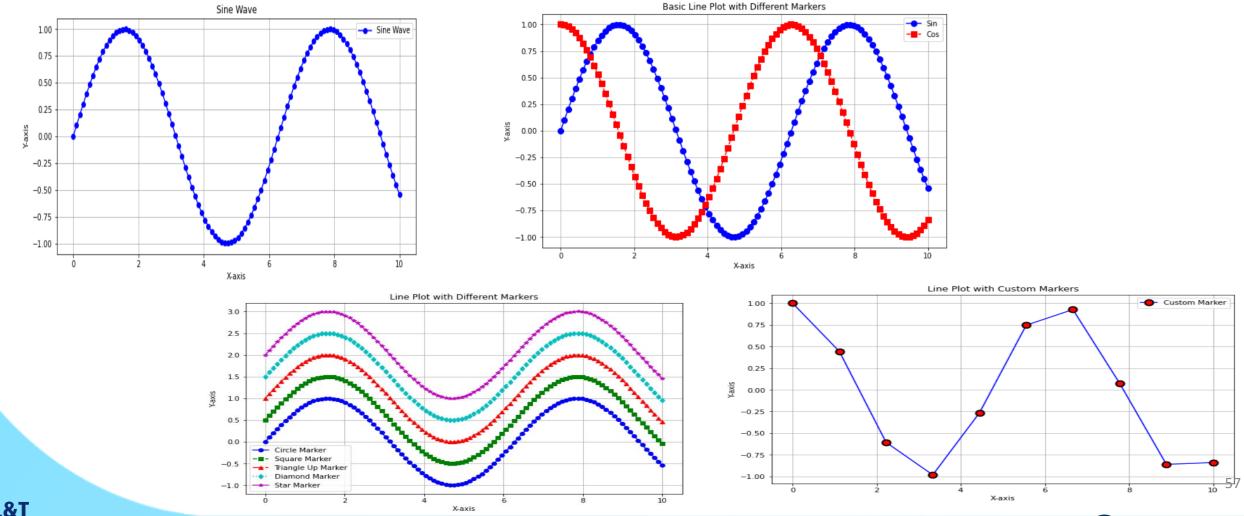
Matplotlib Line Markers





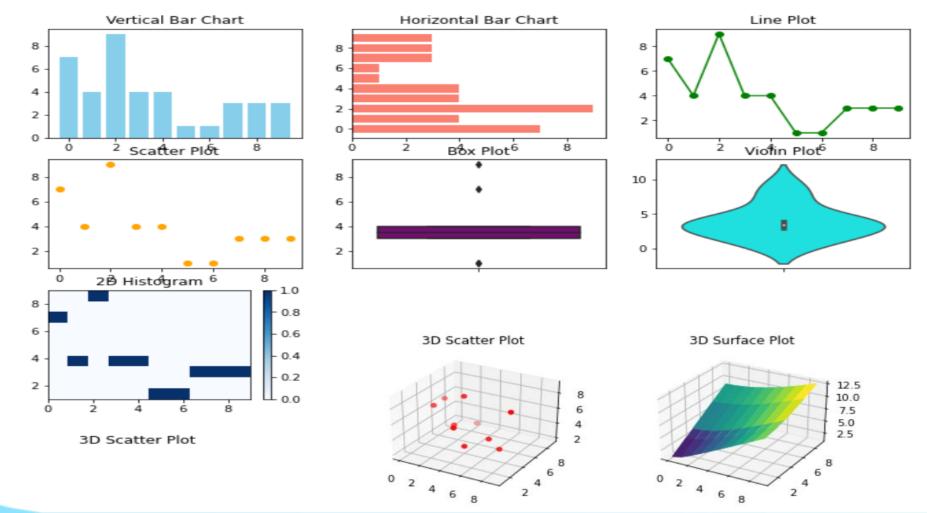
Data Visualization – Matplotlib and Seaborn

Matplotlib – Line Markers





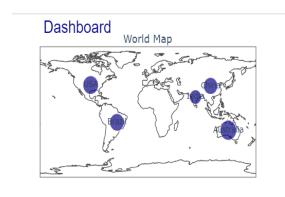
Data Visualization – Matplotlib and Seaborn

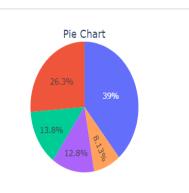


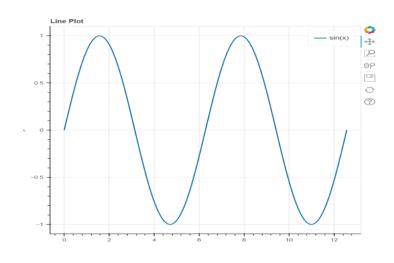


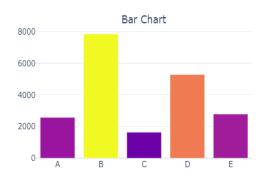


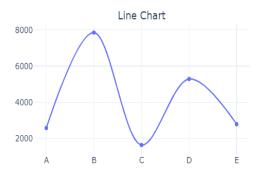
Data Visualization - Plotly, Folium, Bokeh

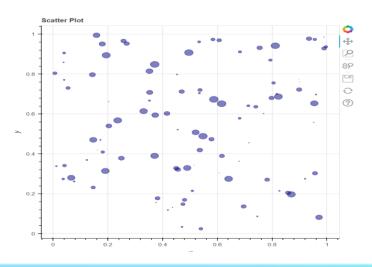








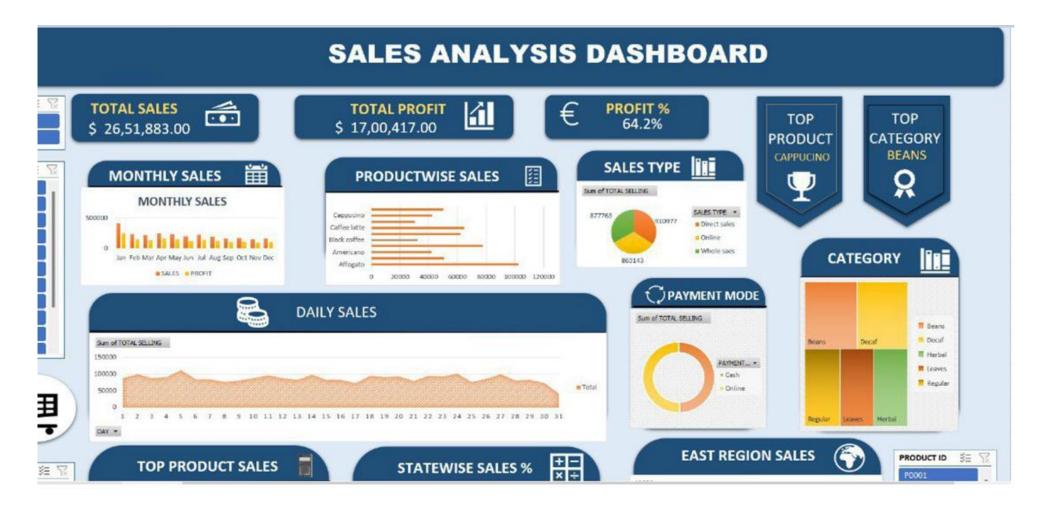








Data Visualization - Dashboard







Knowledge Check

Which method would you use to handle outliers in a DataFrame by capping them at a specified threshold?

- A. df.clip(lower=lower_threshold, upper=upper_threshold)
- B. df.limit(lower=lower_threshold, upper=upper_threshold)
- C. df.truncate(lower=lower_threshold, upper=upper_threshold)
- D. df.reduce(lower=lower_threshold, upper=upper_threshold)



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Ans: df.clip(lower=lower_threshold, upper=upper_threshold)





Knowledge Check

Which of the following statements best describes the use of the pd.get_dummies() function in data modeling?

- A. It normalizes the data to a range between 0 and 1.
- B. It transforms categorical variables into a series of binary columns.
- C. It fills missing values in the dataset.
- D. It scales numerical features to a standard normal distribution.



Knowledge Check

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Ans: It transforms categorical variables into a series of binary columns.





Knowledge Check

To create a bar chart in Plotly with grouped bars for different categories, which parameter should be used?

- A. barmode='group'
- B. barmode='stack'
- C. groupmode='bar'
- D. groupbar='mode'





Knowledge Check

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Ans: barmode='group'







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Thank Non III



