Pandas: A Python library used for data manipulation and analysis. It provides data structures and functions designed to work efficiently with structured data like tables. **DataFrame**: A two-dimensional tabular data structure with labeled axes (rows and columns) in pandas. Series: A one-dimensional array with labels. It can hold any data type (integers, strings, floats, Python objects, etc.). A Series is essentially a single column of a DataFrame. In [1]: import pandas as pd **Creating DataFrames** You can create a DataFrame from various data sources, like a Python dictionary, lists, or external data files. In [2]: |#From Lists data = [[1, 'Alice'], [2, 'Bob'], [3, 'Charlie']] df = pd.DataFrame(data, columns=['ID', 'Name']) Out[2]: ID Name 0 1 Alice 2 Bob 2 3 Charlie In [3]: #From Dictionaries data = {'ID': [1, 2, 3], 'Name': ['Alice', 'Bob', 'Charlie']} df = pd.DataFrame(data) df ID Out[3]: Name Alice 2 Bob 1 3 Charlie In [4]: You can specify an index when creating a DataFrame. This is useful if you need to align your data with a specific set of labels''' data = {'Name': ['Alice', 'Bob', 'Charlie'], 'Age': [25, 30, 35]} index = ['Person1', 'Person2', 'Person3'] df = pd.DataFrame(data, index=index) df Name Age Out[4]: Person1 Alice 25 Person2 30 Person3 Charlie In [5]: #From a Series series = pd.Series([1, 2, 3, 4], name='Column Name') df = series.to_frame() Column Name Out[5]: 0 1 3 3 '''The pd.read_csv() function reads data from a In [6]: CSV file located at the specified path and converts it into a DataFrame allowing you to work with the data in Python.''' df = pd.read_csv('C:/Users/Tarek/Desktop/archive/Iris.csv') Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species** Out[6]: Iris-setosa 0 3.5 1 2 4.9 3.0 0.2 Iris-setosa 1.4 2 1.3 Iris-setosa 3 4 3.1 1.5 0.2 Iris-setosa 4.6 4 5 5.0 3.6 Iris-setosa **145** 146 6.7 3.0 2.3 Iris-virginica **146** 147 6.3 2.5 5.0 Iris-virginica 6.5 3.0 Iris-virginica 148 149 6.2 Iris-virginica **149** 150 5.9 3.0 5.1 1.8 Iris-virginica 150 rows × 6 columns '''To determine the data type of the entire tips table In [7]: you would use the type() function:''' type(df) pandas.core.frame.DataFrame Out[7]: # Display the first few rows In [8]: df.head() Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Out[8]: **Species** 0 1 3.5 0.2 Iris-setosa 5.1 1.4 0.2 Iris-setosa 2 4.7 3.2 1.3 0.2 Iris-setosa 0.2 Iris-setosa 4.6 5 5.0 3.6 1.4 0.2 Iris-setosa # Display the last few rows In [9]: df.tail() SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Out[9]: **Species 145** 146 6.7 3.0 5.2 2.3 Iris-virginica **146** 147 5.0 6.3 2.5 Iris-virginica **147** 148 6.5 3.0 5.2 Iris-virginica **148** 149 6.2 3.4 5.4 2.3 Iris-virginica **149** 150 5.9 3.0 5.1 1.8 Iris-virginica # Summary statistics for numeric columns df.describe() Out[10]: Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm count 150.000000 150.000000 150.000000 150.000000 150.000000 75.500000 5.843333 3.054000 3.758667 1.198667 mean 43.445368 0.828066 0.433594 1.764420 0.763161 std 1.000000 4.300000 2.000000 1.000000 0.100000 min 38.250000 5.100000 2.800000 1.600000 0.300000 1.300000 50% 75.500000 5.800000 3.000000 4.350000 112.750000 6.400000 3.300000 5.100000 1.800000 max 150.000000 6.900000 2.500000 7.900000 4.400000 #Information about the DataFrame In [11]: df.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 150 entries, 0 to 149 Data columns (total 6 columns): Column Non-Null Count Dtype 0 Ιd 150 non-null int64 1 SepalLengthCm 150 non-null float64 150 non-null SepalWidthCm float64 PetalLengthCm 150 non-null float64 PetalWidthCm 150 non-null float64 150 non-null Species object dtypes: float64(4), int64(1), object(1) memory usage: 7.2+ KB In [12]: # Number of rows and columns df.shape (150, 6)Out[12]: # Unique values in a column In [13]: df['Species'].unique() array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object) Out[13]: In [14]: # Value counts for a column df['Species'].value_counts() Species Out[14]: Iris-setosa 50 Iris-versicolor 50 Iris-virginica 50 Name: count, dtype: int64 In [15]: # Checking for Missing Values df.isnull().sum() Out[15]: SepalLengthCm 0 SepalWidthCm 0 PetalLengthCm 0 PetalWidthCm 0 Species 0 dtype: int64 # Unique Counts In [16]: df.nunique() 150 Ιd Out[16]: SepalLengthCm 35 SepalWidthCm 23 43 PetalLengthCm PetalWidthCm 22 Species 3 dtype: int64 # Remove duplicate rows if exsists df.drop_duplicates() Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Out[17]: **Species** 0 1 5.1 3.5 1.4 0.2 Iris-setosa 1 2 4.9 3.0 1.4 0.2 Iris-setosa 2 3 4.7 3.2 1.3 0.2 Iris-setosa 3 4.6 3.1 1.5 0.2 Iris-setosa 5.0 4 5 3.6 1.4 0.2 Iris-setosa **145** 146 6.7 3.0 5.2 2.3 Iris-virginica 5.0 146 147 6.3 2.5 Iris-virginica **147** 148 6.5 3.0 5.2 2.0 Iris-virginica **148** 149 6.2 3.4 5.4 2.3 Iris-virginica **149** 150 5.9 3.0 5.1 1.8 Iris-virginica 150 rows × 6 columns # Using .loc[] to select rows where species is 'versicolor' and only the 'petal length (cm)' column In [18]: selected_rows_loc = df.loc[df['Species'] == 'Iris-versicolor'] selected_rows_loc.head() Out[18]: Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species 50** 51 7.0 1.4 Iris-versicolor 3.2 4.7 **51** 52 6.4 3.2 4.5 1.5 Iris-versicolor **52** 53 6.9 3.1 4.9 1.5 Iris-versicolor 1.3 Iris-versicolor **53** 54 5.5 2.3 4.0 6.5 2.8 4.6 1.5 Iris-versicolor **54** 55 # To select specific columns: In [19]: selected_rows_loc = df.loc[df['Species'] == 'Iris-versicolor', ['PetalLengthCm', 'PetalWidthCm']] selected_rows_loc.head() PetalLengthCm PetalWidthCm Out[19]: 50 4.7 1.4 51 4.5 1.5 52 4.9 1.5 53 4.0 1.3 54 4.6 1.5 # .loc[] with multiple conditions In [20]: selected_rows_loc = df.loc[(df['Species'] == 'Iris-versicolor') & ((df['PetalLengthCm'] > 4.0) | (df['PetalWidthCm'] < 1.5)),</pre> ['PetalLengthCm', 'PetalWidthCm']] selected_rows_loc.head() '''This means a row will be included if it corresponds to an 'Iris-versicolor' species and either has a petal length greater than 4.0 cm or a petal width less than 1.5 cm. The parentheses around conditions 2 and 3 ensure that the OR operation is evaluated correctly''' "This means a row will be included if it corresponds to an 'Iris-versicolor' species\nand either has a p Out[20]: etal length greater than 4.0 cm or a petal width less than 1.5 cm.\nThe parentheses around conditions 2 and 3 ensure that the OR operation is evaluated correctly" '''Boolean indexing in Pandas allows you to select rows In [21]: from a DataFrame based on the values of a boolean condition.''' # Boolean indexing to select rows where petal length is greater than 4.5 cm df[df['PetalLengthCm'] > 4.5] Out[21]: Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species** 50 51 7.0 3.2 4.7 1.4 Iris-versicolor 4.9 1.5 Iris-versicolor 52 53 6.9 3.1 54 55 6.5 2.8 4.6 Iris-versicolor 6.3 4.7 Iris-versicolor 56 57 3.3 58 6.6 2.9 4.6 Iris-versicolor **145** 146 6.7 3.0 5.2 2.3 Iris-virginica 146 147 6.3 2.5 5.0 1.9 Iris-virginica **147** 148 6.5 3.0 5.2 Iris-virginica **148** 149 Iris-virginica 6.2 3.4 5.4 2.3 **149** 150 5.9 3.0 5.1 1.8 Iris-virginica 63 rows × 6 columns # Boolean indexing with multiple conditions In [22]: df[(df['PetalLengthCm'] > 4.5) & (df['Species'] == 'Iris-versicolor')].head()Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species** Out[22]: **50** 51 7.0 3.2 4.7 1.4 Iris-versicolor **52** 53 6.9 3.1 4.9 1.5 Iris-versicolor **54** 55 6.5 2.8 4.6 1.5 Iris-versicolor 3.3 4.7 **56** 57 6.3 1.6 Iris-versicolor **58** 59 6.6 1.3 Iris-versicolor 2.9 4.6 In [23]: # Use .iloc[] to select rows/columns by integer index from the filtered DataFrame. selected_rows_iloc = df.iloc[:5, [2, 3]] # This selects the first 5 rows and columns at index 2 and 3 selected_rows_iloc SepalWidthCm PetalLengthCm Out[23]: 0 3.5 1.4 3.0 1.4 2 3.2 1.3 1.5 4 3.6 1.4 In []: