Pandas Pandas stand for Panel Data and it is the core library for data manupulation and data analysis. It consists of single and multi dimentional data structures for data manupulation. **Pandas Data Structure** 1. Single Dimentional is called as Series Object 2. Multi-dimentional is called as Data Frame **Series Object** Series Object is one dimentional labelled object. import pandas as pd s1 = pd.Series([1,2,3,4,5])Out[1]: 0 3 4 5 dtype: int64 # In the above output 0-1 is teh index number and 1-5 are the value stores at corrsponding indices type(s1) Out[3]: pandas.core.series.Series **Changing Index** In [4]: import pandas as pd #invoking pandas library is not required every time for a new set of codes. s1 = pd.Series([1,2,3,4,5],index = ['a','b','c','d','e'])Out[4]: a b 3 С d 4 5 dtype: int64 Creating series object fron 'Dictionary' pd.Series({'a':10,'b':20,'c':30}) 10 Out[5]: a 20 b dtype: int64 Changing index position $pd.Series({'a':10,'b':20,'c':30}, index = ['b','c','d','a'])$ 20.0 Out[6]: b d NaN a 10.0 dtype: float64 **Extracting Individual Elements** #Extracting Single element s1 = pd.Series([1,2,3,4,5,6,7,8,9])s1[3] Out[7]: 4 In [8]: #Extracting a Sequence of elements s1[:4] # This will extract first 4 elements, from index 0 to 3. Out[8]: 0 1 2 3 dtype: int64 In [9]: #Extracting Elements from back s1[-3:] # This will extract the last 3 elements Out[9]: 6 dtype: int64 **Basic Maths operation on Series** # Adding a scaler value to Series Elements s1 + 5 Out[10]: 0 6 8 3 4 10 5 11 6 12 13 7 8 14 dtype: int64 # Adding two Series object s1 = pd.Series([1,2,3,4,5,6,7,8,9])s2 = pd.Series([10,20,30,40,50,60,70,80,90])s1 + s2 Out[11]: 0 11 22 2 33 3 44 4 55 5 66 6 77 88 99 dtype: int64 # Multiplication of two Series object s1 * s2 Out[12]: 0 90 160 4 250 360 490 640 8 810 dtype: int64 # Division of two Series object s1/s2 0.1 0.1 2 0.1 0.1 0.1 0.1 0.1 6 0.1 0.1 dtype: float64 **Data Frames** Data Frame is a two dimentional labelled data structure. It comprises of row and columns. In [14]: import pandas as pd pd.DataFrame({"Name":['Shila','Munni','Dhanno'],"Marks":[75,67,87]}) Out[14]: Name Marks 0 Shila 75 67 Munni 87 **2** Dhanno **Data Frame in-built functions** # head() - 1st five rows # tail() - last 5 rows # shape() - number of rows and colums # describe() - General information about the data set iris = pd.read_csv('iris.csv') In [16]: iris.head() Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Species 0 0.2 Iris-setosa 4.9 3.0 0.2 Iris-setosa 2 4.7 3.2 0.2 Iris-setosa 4.6 0.2 Iris-setosa 5.0 3.6 1.4 0.2 Iris-setosa iris.head(10) #firts 10 rows Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm 0 0.2 Iris-setosa 2 4.9 3.0 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 4 5 5.0 3.6 1.4 0.2 Iris-setosa 5 6 5.4 3.9 1.7 0.4 Iris-setosa 7 6 4.6 3.4 1.4 0.3 Iris-setosa 7 5.0 3.4 1.5 0.2 Iris-setosa 8 9 4.4 2.9 1.4 0.2 Iris-setosa 4.9 3.1 1.5 0.1 Iris-setosa In [18]: iris.tail() $Id \quad SepalLengthCm \quad SepalWidthCm \quad PetalLengthCm \quad PetalWidthCm \quad \\$ Out[18]: **Species** 145 146 6.7 3.0 5.2 2.3 Iris-virginica 146 6.3 5.0 1.9 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 5.9 **149** 150 3.0 5.1 1.8 Iris-virginica In [19]: iris.tail(10) $Id \quad SepalLengthCm \quad SepalWidthCm \quad PetalLengthCm \quad PetalWidthCm \quad$ Out[19]: **Species 140** 141 2.4 Iris-virginica 6.7 3.1 5.6 141 2.3 Iris-virginica 2.7 142 143 5.8 5.1 1.9 Iris-virginica 2.3 Iris-virginica 143 6.8 5.9 2.5 Iris-virginica 144 145 6.7 3.3 5.7 145 6.7 3.0 2.3 Iris-virginica 146 147 6.3 2.5 5.0 1.9 Iris-virginica **147** 148 6.5 3.0 2.0 Iris-virginica 6.2 148 149 3.4 5.4 2.3 Iris-virginica 150 5.9 149 3.0 1.8 Iris-virginica iris.shape (150, 6)iris.describe() **PetalWidthCm** Id SepalLengthCm SepalWidthCm PetalLengthCm count 150.000000 150.000000 150.000000 150.000000 150.000000 3.054000 3.758667 1.198667 75.500000 5.843333 mean 0.433594 43.445368 0.828066 1.764420 0.763161 std 1.000000 2.000000 min 4.300000 1.000000 0.100000 1.600000 25% 38.250000 5.100000 2.800000 0.300000 75.500000 5.800000 3.000000 4.350000 1.300000 5.100000 112.750000 6.400000 3.300000 1.800000 7.900000 6.900000 max 150.000000 4.400000 2.500000 .iloc[] iris.iloc[0:3,0:2] Id SepalLengthCm 4.9 4.7 iris.iloc[1:3,1:3] SepalLengthCm SepalWidthCm 1 4.9 3.0 3.2 In [24]: iris.iloc[5:11,2:] #rows 5 to 10 and column 2 to last Out[24]: SepalWidthCm PetalLengthCm PetalWidthCm 5 1.7 0.4 Iris-setosa 3.4 1.4 0.3 Iris-setosa 3.4 1.5 0.2 Iris-setosa 1.4 0.2 Iris-setosa 1.5 0.1 Iris-setosa 10 0.2 Iris-setosa .loc[] iris.loc[1:5,("SepalWidthCm","Species")] SepalWidthCm **Species** 1 3.0 Iris-setosa 3.2 Iris-setosa 3 3.1 Iris-setosa 3.6 Iris-setosa 5 3.9 Iris-setosa **Dropping Columns** iris.head() Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm 0.2 Iris-setosa 0.2 Iris-setosa 4.6 3.1 0.2 Iris-setosa 1.4 0.2 Iris-setosa 5.0 3.6 iris.drop('PetalLengthCm', axis = 1) #axis = Id SepalLengthCm SepalWidthCm PetalWidthCm **Species** 0 5.1 3.5 Iris-setosa 1 3.0 0.2 Iris-setosa 2 3 4.7 3.2 0.2 Iris-setosa 4.6 3.1 Iris-setosa 5 5.0 3.6 Iris-setosa **145** 146 6.7 3.0 2.3 Iris-virginica **146** 147 6.3 2.5 1.9 Iris-virginica **147** 148 6.5 3.0 2.0 Iris-virginica **148** 149 6.2 3.4 2.3 Iris-virginica **149** 150 5.9 3.0 1.8 Iris-virginica 150 rows × 5 columns In [28]: iris.drop(1,axis = 0) #axis = 0 is used for rows Out[28]: Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species** 0 5.1 3.5 1.4 0.2 Iris-setosa 2 4.7 3.2 1.3 0.2 Iris-setosa 3 4.6 3.1 1.5 0.2 Iris-setosa 5.0 1.4 3.6 0.2 Iris-setosa 5 5.4 3.9 1.7 0.4 Iris-setosa **145** 146 6.7 3.0 5.2 2.3 Iris-virginica **146** 147 2.5 5.0 1.9 Iris-virginica 6.3 **147** 148 6.5 3.0 5.2 2.0 Iris-virginica 5.4 **148** 149 6.2 2.3 Iris-virginica 3.4 **149** 150 5.9 3.0 5.1 1.8 Iris-virginica 149 rows × 6 columns iris.head() Id SepalLengthCm SepalWidthCm PetalLengthCm **Species** 0 1 5.1 3.5 1.4 0.2 Iris-setosa 1.4 4.9 3.0 0.2 Iris-setosa 4.7 3.2 1.3 0.2 Iris-setosa 4.6 3.1 0.2 Iris-setosa **4** 5 5.0 3.6 1.4 0.2 Iris-setosa Drop more than one rows iris.drop(["SepalLengthCm", "SepalWidthCm", "PetalLengthCm"], axis = 1) Out[30]: Id PetalWidthCm **Species** 0.2 Iris-setosa 1 2 0.2 Iris-setosa 2 3 0.2 Iris-setosa 0.2 Iris-setosa 5 0.2 Iris-setosa **145** 146 2.3 Iris-virginica **146** 147 1.9 Iris-virginica **147** 148 2.0 Iris-virginica **148** 149 Iris-virginica **149** 150 1.8 Iris-virginica 150 rows × 3 columns Drop more than one column iris.head() Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species** 0.2 Iris-setosa 0 5.1 3.5 1.4 1 2 4.9 3.0 1.4 0.2 Iris-setosa 2 3 4.7 3.2 1.3 0.2 Iris-setosa 4.6 3.1 1.5 0.2 Iris-setosa 5.0 3.6 1.4 0.2 Iris-setosa iris.drop([2,3,4],axis = 0)Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species** 0 1 5.1 3.5 1.4 0.2 Iris-setosa 4.9 3.0 1.4 0.2 Iris-setosa 6 5.4 3.9 1.7 0.4 Iris-setosa 4.6 3.4 1.4 0.3 Iris-setosa 7 5.0 3.4 1.5 0.2 Iris-setosa **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 2.0 Iris-virginica **148** 149 6.2 5.4 2.3 Iris-virginica 3.4 **149** 150 5.9 3.0 5.1 1.8 Iris-virginica 147 rows × 6 columns Mean, Median, Maximum & Minimum In [34]: iris.head() Out[34]: Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species** 0 5.1 3.5 1.4 0.2 Iris-setosa 1 4.9 3.0 2 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 0.2 Iris-setosa 5 5.0 3.6 1.4 # Minimum iris.min() Out[37]: Id 1 SepalLengthCm 4.3 SepalWidthCm 2.0 1.0 PetalLengthCm PetalWidthCm 0.1 Species Iris-setosa dtype: object # Maximum iris.max() 150 Out[38]: Id 7.9 SepalLengthCm SepalWidthCm 4.4 PetalLengthCm 6.9 PetalWidthCm 2.5 Iris-virginica Species dtype: object # Median iris.median() Out[39]: Id 75.50 SepalLengthCm 5.80 3.00 SepalWidthCm 4.35 PetalLengthCm PetalWidthCm 1.30 dtype: float64 In [46]: # Mean iris.mean() 75.500000 Out[46]: Id SepalLengthCm 5.843333 SepalWidthCm 3.054000 PetalLengthCm 3.758667 1.198667 PetalWidthCm dtype: float64 **Apply Function** In [47]: iris.head() Out[47]: Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Species **0** 1 0.2 Iris-setosa 5.1 3.5 1.4 4.9 3.0 0.2 Iris-setosa 2 3 4.7 3.2 1.3 0.2 Iris-setosa 0.2 Iris-setosa 4 5 5.0 1.4 0.2 Iris-setosa 3.6 In [48]: def half(s): #'half is the user defined function' return s*0.5 iris[['SepalLengthCm','SepalWidthCm']].apply(half) Out[48]: SepalLengthCm SepalWidthCm 2.55 1.75 2.45 1.50 2 2.35 1.60 3 2.30 1.55 2.50 1.80 4 145 3.35 1.50 146 1.25 147 3.25 1.50 148 3.10 1.70 149 2.95 1.50 150 rows × 2 columns value_counts() - value of each category in a coulmn iris['Species'].value_counts() Out[50]: Iris-versicolor Iris-virginica Iris-setosa Name: Species, dtype: int64 sort_values() - in increasing order iris.head() Out[51]: Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species** 0 1 5.1 3.5 1.4 0.2 Iris-setosa 0.2 Iris-setosa 2 3 4.7 3.2 1.3 0.2 Iris-setosa 4.6 3.1 1.5 0.2 Iris-setosa **4** 5 5.0 3.6 1.4 0.2 Iris-setosa In [54]: view = iris.sort_values(by = 'SepalLengthCm') view.head(10) Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species 13** 14 0.1 Iris-setosa 4.3 3.0 1.1 **42** 43 4.4 3.2 1.3 0.2 Iris-setosa **38** 39 4.4 3.0 1.3 0.2 Iris-setosa **8** 9 2.9 4.4 1.4 0.2 Iris-setosa **41** 42 4.5 2.3 1.3 0.3 Iris-setosa **22** 23 4.6 3.6 1.0 0.2 Iris-setosa **3** 4 4.6 3.1 1.5 0.2 Iris-setosa **6** 7 4.6 3.4 1.4 0.3 Iris-setosa **47** 48 4.6 3.2 1.4 0.2 Iris-setosa **2** 3 4.7 3.2 1.3 0.2 Iris-setosa