Question 1 Use the following data to create the Data Frame and store it in a variable named "Data" Data = pd.DataFrame({"x":[103,207,315,400,None,605,355,677,197,869], "y": [18976, 45789, None, 78964, 45683, None, 78965, None, 68546, 20015] Data Out[7]: Х у **0** 103.0 18976.0 **1** 207.0 45789.0 **2** 315.0 NaN **3** 400.0 78964.0 NaN 45683.0 **5** 605.0 NaN **6** 355.0 78965.0 **7** 677.0 NaN **8** 197.0 68546.0 **9** 869.0 20015.0 Check the presence of missing values In [10]: Data.isnull() Out[10]: У 0 False False 1 False False **2** False True **3** False False 4 True False **5** False True 6 False False **7** False True 8 False False **9** False False In [54]: Data.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 10 entries, 0 to 9 Data columns (total 2 columns): # Column Non-Null Count Dtype 9 non-null float64 Χ 7 non-null float64 У dtypes: float64(2) memory usage: 288.0 bytes Use the imputation approach to handle missing values` Droping the NaN values Data.dropna(axis=0, inplace=True) In [11]: In [13]: Out[13]: X **0** 103.0 18976.0 **1** 207.0 45789.0 **3** 400.0 78964.0 **6** 355.0 78965.0 **8** 197.0 68546.0 **9** 869.0 20015.0 data = pd.DataFrame({"x":[103,207,315,400,None,605,355,677,197,869], "y": [18976, 45789, None, 78964, 45683, None, 78965, None, 68546, 20015] data Out[19]: Χ У **0** 103.0 18976.0 **1** 207.0 45789.0 **2** 315.0 NaN **3** 400.0 78964.0 45683.0 NaN **5** 605.0 NaN **6** 355.0 78965.0 **7** 677.0 NaN **8** 197.0 68546.0 9 869.0 20015.0 • Filling the missing values with mean mean1 = int(data["x"].mean()) mean2 = int(data["y"].mean()) data["x"].fillna(mean1, inplace=True) data["y"].fillna(mean2, inplace=True) data Out[20]: У **0** 103.0 18976.0 **1** 207.0 45789.0 2 315.0 50991.0 **3** 400.0 78964.0 4 414.0 45683.0 **5** 605.0 50991.0 **6** 355.0 78965.0 677.0 50991.0 **8** 197.0 68546.0 **9** 869.0 20015.0 • Filling the missing values with mean Data_copy = pd.DataFrame({"x":[103,207,315,400,None,605,355,677,197,869], In [21]: "y": [18976, 45789, None, 78964, 45683, None, 78965, None, 68546, 20015] }) Data_copy Out[21]: Х **0** 103.0 18976.0 **1** 207.0 45789.0 **2** 315.0 NaN **3** 400.0 78964.0 NaN 45683.0 **5** 605.0 NaN **6** 355.0 78965.0 **7** 677.0 NaN **8** 197.0 68546.0 **9** 869.0 20015.0 median1 = int(data["x"].median()) median2 = int(data["y"].median()) Data_copy["x"].fillna(median1, inplace=True) Data_copy["y"].fillna(median2, inplace=True) Data_copy Out[22]: X **0** 103.0 18976.0 **1** 207.0 45789.0 **2** 315.0 50991.0 **3** 400.0 78964.0 4 377.0 45683.0 605.0 50991.0 **6** 355.0 78965.0 **7** 677.0 50991.0 **8** 197.0 68546.0 **9** 869.0 20015.0 Perform Normalization Data_copy In [25]: Out[25]: У **0** 103.0 18976.0 **1** 207.0 45789.0 **2** 315.0 50991.0 **3** 400.0 78964.0 **4** 377.0 45683.0 605.0 50991.0 **6** 355.0 78965.0 **8** 197.0 68546.0 **9** 869.0 20015.0 · Using min-max normalization with or without an inbuilt library Data_copying = Data_copy.copy() In [27]: for column in Data.columns: Data_copying[column] = (Data_copying[column] - Data_copying[column].min()) / (Data_copying[column].max() - Dat Data_copying Out[27]: Х 0.000000 0.000000 **1** 0.135770 0.446965 2 0.276762 0.533681 **3** 0.387728 0.999983 **4** 0.357702 0.445198 **5** 0.655352 0.533681 6 0.328982 1.000000 0.749347 0.533681 **8** 0.122715 0.826318 **9** 1.000000 0.017320 • Using Z score standardization with or without an inbuilt library In [28]: $z_{data} = Data_{copy}$ **for** column **in** z_data: $z_{data[column]} = (z_{data[column]} - z_{data[column]}.mean()) / z_{data[column]}.std()$ z_data Out[28]: У **0** -1.286649 -1.535415 **1** -0.851490 -0.249488 **2** -0.399593 -0.000005 **3** -0.043934 1.341554 -0.140172 -0.254572 0.813832 -0.000005 -0.232224 1.341602 1.115096 -0.000005 -0.893332 0.841917 1.918467 -1.485585 Stored the pre-processed data in the external file "preprocessed123.csv" z_data.to_csv("preprocessed123(Z-Score).csv") In [58]: Data_copying.to_csv("preprocessed123(Mim-Max).csv") Question 2 Download the dataset from the link In [44]: | social_network = pd.read_csv("Social_Network_Ads.csv") social_network Out[44]: Age EstimatedSalary **0** 19.0 19000.0 **1** 35.0 20000.0 **2** 26.0 43000.0 **3** 27.0 57000.0 4 19.0 76000.0 **395** 46.0 41000.0 **396** 51.0 23000.0 **397** 50.0 NaN 36.0 33000.0 398 36000.0 **399** 49.0 400 rows × 2 columns Check the presence of missing value social_network.isnull() In [45]: Out[45]: Age EstimatedSalary **0** False False 1 False False 2 False False 3 False False 4 False False **395** False False **396** False False **397** False True 398 False False **399** False False 400 rows × 2 columns In [46]: social_network.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 400 entries, 0 to 399 Data columns (total 2 columns): Column Non-Null Count Dtype 396 non-null float64 EstimatedSalary 396 non-null float64 dtypes: float64(2) memory usage: 6.4 KB Use the imputation approach to handle missing values Dropping the NaN values In [47]: | social_network_drop = social_network.copy() social_network_drop.dropna(axis=0, inplace=True) social_network_drop Age EstimatedSalary Out[47]: 19000.0 **0** 19.0 **1** 35.0 20000.0 **2** 26.0 43000.0 **3** 27.0 57000.0 76000.0 4 19.0 **394** 39.0 59000.0 **395** 46.0 41000.0 23000.0 **396** 51.0 **398** 36.0 33000.0 36000.0 **399** 49.0 392 rows × 2 columns • Filling the NaN values with Mean In [50]: social_network_mean = social_network.copy() mean1 = int(social_network_mean["Age"].mean()) meen2 = int(social_network_mean["EstimatedSalary"].mean()) social_network_mean["Age"].fillna(mean1, inplace=True) social_network_mean["EstimatedSalary"].fillna(mean2, inplace=True) social_network_mean Age EstimatedSalary Out[50]: **0** 19.0 19000.0 **1** 35.0 20000.0 **2** 26.0 43000.0 **3** 27.0 57000.0 4 19.0 76000.0 **395** 46.0 41000.0 23000.0 **396** 51.0 **397** 50.0 50991.0 **398** 36.0 33000.0 **399** 49.0 36000.0 400 rows × 2 columns • Filling the NaN Values with Median social_network_median = social_network.copy() In [52]: maedian1 = int(social_network_mean["Age"].median()) median2 = int(social_network_mean["EstimatedSalary"].median()) social_network_median["Age"].fillna(median1, inplace=True) social_network_median["EstimatedSalary"].fillna(median2, inplace=True) social_network_median Age EstimatedSalary Out[52]: **0** 19.0 19000.0 **1** 35.0 20000.0 **2** 26.0 43000.0 **3** 27.0 57000.0 76000.0 **4** 19.0 **395** 46.0 41000.0 **396** 51.0 23000.0 **397** 50.0 69500.0 **398** 36.0 33000.0 49.0 36000.0 399 400 rows × 2 columns **Perform Normalization** Using min-max normalization with or without an inbuilt library social_network_minmax = social_network.copy() for column in social_network_minmax: social_network_minmax[column] = (social_network_minmax[column] - social_network_minmax[column].min()) / (social_network_minmax[column] - social_network_minmax[column] - social_n social_network_minmax Age EstimatedSalary Out[55]: 0 0.023810 0.029630 **1** 0.404762 0.037037 2 0.190476 0.207407 **3** 0.214286 0.311111 4 0.023810 0.451852 **395** 0.666667 0.192593 **396** 0.785714 0.059259 **397** 0.761905 NaN 0.428571 0.133333 **399** 0.738095 0.155556 400 rows × 2 columns • Using Z score standardization with or without an inbuilt library z_social_network = social_network.copy() for column in z_social_network: $z_social_network[column] = (z_social_network[column] - z_social_network[column].mean()) \ / \ z_social_network[column] = (z_social_network[column] - z_social_network[column].mean()) \ / \ z_social_network[column] = (z_social_network[column] - z_social_network[column].mean()) \ / \ z_social_network[column] = (z_social_network[column] - z_social_network[column] - z_soc$ z_social_network Out[56]: Age EstimatedSalary -1.490638 **0** -1.779591 **1** -0.257111 -1.461305 **2** -1.113506 -0.786651 **3** -1.018351 -0.375993 -0.845317 395 0.789594 -1.373307 396 1.265369 1.170214 NaN 398 -0.161956 -1.079979 1.075059 -0.991981 400 rows × 2 columns Stored the pre-processed data in the external file "preprocessed123.csv" z_social_network.to_csv("preprocessed123(Z-scoree).csv") social_network_minmax.to_csv("preprocessed123(mim-maxx).csv")

In [1]: import pandas as pd