Today's Agenda • Types of Data Analytics Python Libraries for DA · Codes and Examples of various statistical measurements Data Visualization • Outlier Detection In [ ]: 4 Types of Data Analytics **VALUE Prescriptive** Defines future actions - i.e., "What to do next?" **Diagnostic** Tells What's likely to Based on current data happen? analytics, predefined future plans, goals, Descriptive Based on historical Automated RCA and objectives **Root Cause Analysis** data, and assumes a Based on Live Data, static business Tells what's Advanced algorithms Explains "why" things plans/models happening in real to test potential are happening time outcomes of each **Helps Business** decision and decisions to be Helps trouble shoot Accurate & Handy for recommends the best automated using issues Operations course of action algorithms. management Easy to Visualize Complexity © Arun Kottolli Credits - Image from Internet In [ ]: **Python tools for Data Analysis** NumPy matplotlib IP [y]: IPython
Interactive Computing Credits - Image from Internet Other libraries OpenCV Pandas Profiling GeoPandas BioPandas Plotly Statsmodels Statistics Tensorflow Pytorch Dash Flask **Basic Statistics**  Mean → Average of all the data values Sum of all data values divided by total number of data values Median 

The value separating the higher half from the lower half of the data Mode → The value that appears most frequently in the data set Standard deviation → Used to measure of the amount of variation or dispersion of set of values Low standard deviation → All values very close to mean High standard deviation → All values are far from the mean import the necessary packages In [1]: import pandas as pd import numpy as np from collections import Counter Mean In [2]: def calculate\_mean(data\_values): return sum(data\_values) / len(data\_values) In [3]: # show example x = [5, 6, 1, -10, 4, 8, 10]mean value = calculate\_mean(data\_values=x) print(mean value) 3.4285714285714284 Median In [4]: 5 / 2 Out[4]: 2.5 In [5]: int(5 / 2)Out[5]: 2 5 // 2 In [6]: Out[6]: 2 Procedure -· First sort the values If the total number of values is odd Take the middle value If the total number of values is even Take the two middle values Find the average of those two middle values In [7]: def calculate\_median(data\_values): sorted values = sorted(data values) mid\_index = len(data\_values) // 2 # odd case if len(data values) % 2 != 0: median = sorted\_values[mid\_index] # even case else: mid\_index\_l = mid\_index - 1 mini\_data = [sorted\_values[mid\_index\_1], sorted\_values[mid\_index]] median = calculate\_mean(mini\_data) return median In [8]: | # show example x = [5, 6, 1, -10, 4, 8, 10, 9, 100, 32]median\_value = calculate\_median(data\_values=x) print(median\_value) 7.0 Mode In [9]: d = [1, 1, 2, 3, 4, 4, 5, 6, 7, 8, 9, 8]print(Counter(d)) print( list( Counter(d).keys() ) ) print( list( Counter(d).values() ) ) Counter({1: 2, 4: 2, 8: 2, 2: 1, 3: 1, 5: 1, 6: 1, 7: 1, 9: 1}) [1, 2, 3, 4, 5, 6, 7, 8, 9][2, 1, 1, 2, 1, 1, 1, 2, 1] In [10]: # show dictionary example d = [1, 2, 3, 4, 5, 6, 7, 9, 8]Counter(d).items() # in case repetition is similar, take the the minimum value from the repetition (key - original) Out[10]: dict\_items([(1, 1), (2, 1), (3, 1), (4, 1), (5, 1), (6, 1), (7, 1), (9, 1), (8, 1)]) In [11]: | def calculate\_mode(data\_values): data\_counter = Counter(data\_values) max\_freq = max(list(data\_counter.values())) if max\_freq == 1: return "Mode doesn't exist" mode = [i for i, j in data\_counter.items() if j == max\_freq] return min(mode) # show example In [12]: x = [1, 1, 2, 3, 4, 4, 4, 4, 5, 6, 7, 8, 8, 9, 8]mode\_value = calculate\_mode(data\_values=x) print(mode\_value) Standard deviation Formula  $\sigma = \sqrt{rac{\sum (x_i - \mu)^2}{N}} 
ightarrow i = 1, 2, 3, \ldots n$ where •  $\sigma$  = Standard deviation •  $x_i$  = each data value •  $\mu$  = Mean • N = Total size of the data In [13]: def calculate\_stddev(data\_values): return np.std(a=data\_values) In [14]: # show example  $x_{list} = [1, 1, 2, 3, 4, 4, 5, 6, 7]$ std\_v = calculate\_stddev(data\_values=x\_list) print(std\_v) 2.0 In [ ]: **Data visualization** In [15]: from matplotlib import pyplot as plt **Line Plot** In [16]: x = [1, 2, 3, 4, 5, 6, 7, 9, 10]y = [3, 5, 2, 7, 4, 3, 8, 6, 9]# show example plt.figure(figsize=(10, 4)) plt.plot(x, y) plt.show() 7 6 5 4 3 2 **Scatter Plot** In [17]: x = [1, 2, 3, 4, 5, 6, 7, 9, 10]y = [3, 5, 2, 7, 4, 3, 8, 6, 9]# show example plt.figure(figsize=(10, 4)) plt.scatter(x, y) plt.show() 9 8 7 6 5 4 3 Line and Scatter together In [18]: x = [1, 2, 3, 4, 5, 6, 7, 9, 10]y = [3, 5, 2, 7, 4, 3, 8, 6, 9]# show example plt.figure(figsize=(10, 4)) plt.plot(x, y, 'o-', color='red') plt.show() 9 8 7 6 5 4 3 10 Handwritten plots In [19]: # x = [1, 2, 3, 4, 5, 6, 7, 9, 10]# y = [3, 5, 2, 7, 4, 3, 8, 6, 9]# # show example # plt.figure(figsize=(10, 4)) # plt.xkcd() # plt.plot(x, y, 'o-r') # plt.show() Read data In [20]: df = pd.read\_csv('students\_hw.csv') df.head() Out[20]: Height(Inches) Weight(Pounds) 0 65.78 112.99 136.49 1 71.52 2 69.40 153.03 3 68.22 142.34 67.79 144.30 **Heights** → **Mean**, **Median**, **Mode** In [21]: x\_list = df['Height(Inches)'].to\_list() y\_list = [0 for i in range(len(x\_list))] In [22]: print(x\_list) [65.78, 71.52, 69.4, 68.22, 67.79, 68.7, 69.8, 70.01, 67.9, 66.78, 66.49, 67.62, 68.3, 67.12, 68.28, 71.09, 66.46, 68.65, 71.23, 67.13, 67.83, 68.88, 63.48, 68.42, 67.63, 67.21, 70.84, 67.49, 66.53, 65. 44, 69.52, 65.81, 67.82, 70.6, 71.8, 69.21, 66.8, 67.66, 67.81, 64.05, 68.57, 65.18, 69.66, 67.97, 6 5.98, 68.67, 66.88, 67.7, 69.82, 69.09, 69.91, 67.33, 70.27, 69.1, 65.38, 70.18, 70.41, 66.54, 66.36, 67.54, 66.5, 69.0, 68.3, 67.01, 70.81, 68.22, 69.06, 67.73, 67.22, 67.37, 65.27, 70.84, 69.92, 64.29, 68.25, 66.36, 68.36, 65.48, 69.72, 67.73, 68.64, 66.78, 70.05, 66.28, 69.2, 69.13, 67.36, 70.09, 70.1 8, 68.23, 68.13, 70.24, 71.49, 69.2, 70.06, 70.56, 66.29, 63.43, 66.77, 68.89, 64.87, 67.09, 68.35, 6 5.61, 67.76, 68.02, 67.66, 66.31, 69.44, 63.84, 67.72, 70.05, 70.19, 65.95, 70.01, 68.61, 68.81, 69.7 6, 65.46, 68.83, 65.8, 67.21, 69.42, 68.94, 67.94, 65.63, 66.5, 67.93, 68.89, 70.24, 68.27, 71.23, 6 9.1, 64.4, 71.1, 68.22, 65.92, 67.44, 73.9, 69.98, 69.52, 65.18, 68.01, 68.34, 65.18, 68.26, 68.57, 6 4.5, 68.71, 68.89, 69.54, 67.4, 66.48, 66.01, 72.44, 64.13, 70.98, 67.5, 72.02, 65.31, 67.08, 64.39, 69.37, 68.38, 65.31, 67.14, 68.39, 66.29, 67.19, 65.99, 69.43, 67.97, 67.76, 65.28, 73.83, 66.81, 66. 89, 65.74, 65.98, 66.58, 67.11, 65.87, 66.78, 68.74, 66.23, 65.96, 68.58, 66.59, 66.97, 68.08, 70.19, 65.52, 67.46, 67.41, 69.66, 65.8, 66.11, 68.24, 68.02, 71.39] In [23]: mean val = calculate mean(data values=x list) median\_val = calculate\_median(data\_values=x\_list) mode val = calculate\_mode(data\_values=x\_list) print(mean\_val) print(median\_val) print(mode\_val) 67.9497999999998 67.935 65.18 In [24]: plt.figure(figsize=(15, 3)) plt.yticks([]) plt.scatter(x\_list, y\_list, label='Data Values') plt.axvline(x=mean\_val, ymin=0.3, ymax=0.7, ls='--', color='red', label='Mean') plt.axvline(x=median\_val, ymin=0.3, ymax=0.7, ls='--', color='orange', label='Median') plt.axvline(x=mode\_val, ymin=0.3, ymax=0.7, ls='--', color='green', label='Mode') plt.legend() plt.show() --- Mean --- Median --- Mode Data Values 72 In [ ]: Weights → Mean, Median, Mode In [25]: x\_list = df['Weight(Pounds)'].to\_list() y\_list = [0 for i in range(len(x\_list))] In [26]: mean val = calculate mean(data values=x list) median val = calculate median(data values=x list) mode\_val = calculate\_mode(data\_values=x\_list) print(mean val) print(median val) print(mode\_val) 127.2219500000001 127.875 123.49 In [27]: plt.figure(figsize=(15, 3)) plt.yticks([]) plt.scatter(x\_list, y\_list, label='Data Values') plt.axvline(x=mean\_val, ymin=0.3, ymax=0.7, ls='--', color='red', label='Mean') plt.axvline(x=median val, ymin=0.3, ymax=0.7, ls='--', color='orange', label='Median') plt.axvline(x=mode\_val, ymin=0.3, ymax=0.7, ls='--', color='green', label='Mode') plt.legend() plt.show() --- Mean Median Mode Data Values 160 100 110 120 130 150 In [ ]: **Outliers** · An outlier is a data point that differs significantly from other data values In [28]: x = [1, 2, 3, 4, 5, 6, 7, 9, 10, 50, 5, 6, 9, 4, 7]y = [3, 5, 2, 7, 4, 3, 8, 6, 9, 65, 6, 8, 3, 6, 9]How can we detect outliers? 1. By graphing scatter plot box plot 2. By calculating z\_score values • if z\_score value is  $> 3 \rightarrow$  reject • if z score value is  $< -3 \rightarrow$  reject Formula  $z=rac{(x_i-\mu)}{\sigma}
ightarrow i=1,2,3\dots n$ where •  $\mu$  = Mean •  $\sigma$  = Standard deviation •  $x_i$  = each data value 1 - a) scatter plot In [29]: # size (10, 4) plt.figure(figsize=(10, 4)) plt.scatter(x, y) plt.show() 60 50 40 30 20 10 20 50 1 - b) box plot In [30]: # size (10, 4) - x plt.figure(figsize=(10, 4)) plt.boxplot(x) plt.show() 50 40 30 20 10 In [31]: # size (10, 4) - y plt.figure(figsize=(10, 4)) plt.boxplot(y) plt.show() 60 50 40 30 20 10 2 - zscore method In [32]: def calculate\_zscore(data\_values): mean vals = calculate mean(data values) # standard deviation std dev = calculate stddev(data values=data values) # applying the formula for all the values zscore = [(i - mean\_vals)/std\_dev for i in data\_values] return zscore In [33]: | z\_x = calculate\_zscore(data\_values=x) print(z x) [-0.6631473670024701, -0.5751189554534697, -0.4870905439044692, -0.3990621323554687, -0.3110337208064]683, -0.2230053092574678, -0.13497689770846735, 0.041079925389533554, 0.12910833693853402, 3.65024479 88985525, -0.3110337208064683, -0.2230053092574678, 0.041079925389533554, -0.3990621323554687, -0.1348985525, -0.3110337208064683, -0.2230053092574678, 0.041079925389533554, -0.3990621323554687, -0.1348985525, -0.3990621323554687, -0.1348985525, -0.3990621323554687, -0.1348985525, -0.3990621323554687, -0.1348985625, -0.3990621323554687, -0.1348985625, -0.3990621323554687, -0.1348985625, -0.1348985665, -0.134898565, -0.134898565, -0.134898565, -0.134898565, -0.134898565, -0.134898565, -0.134896565, -0.13489665, -0.13489665, -0.13489665, -0.13489665, -0.13489665, -0.13489665, -0.13489665, -0.13489665, -0.134896665, -0.13489665, -0.13489665, -0.134896665, -0.134896665, -0.134896665, -0.134896665, -0.134896665, -0.134896665, -0.134896665, -0.13489666565, -0.134896665, -0.134896665, -0.134896665, -0.134896665, -0.134896665, -0.13489666565, -0.134896666565, -0.1348966665, -0.1348966665, -0.1348966665, -0.1348966665, -0.13489666665, -0.13489697689770846735] In [34]: def get\_outlier\_indices(data\_values): # z\_score values of all the data values z\_score = calculate\_zscore(data\_values=data\_values) # get the index of the outlier # whose value is > 3 # whose value is < -3 outlier\_indices = [i for i, j in enumerate(z\_score) if j > 3 or j < -3] return outlier\_indices In [35]: x\_index = get\_outlier\_indices(data\_values=x) print(x\_index) [9] y\_index = get\_outlier\_indices(data\_values=y) In [36]: print(y\_index) [9] In [37]:  $z_x = x[x_index[0]]$ In [38]: z\_x Out[38]: 50 In [39]: heights = df['Height(Inches)'].to\_list() weights = df['Weight(Pounds)'].to list() In [40]: out\_heights = get\_outlier\_indices(data\_values=heights) print(out\_heights) [138, 174] In [41]: heights[138] Out[41]: 73.9 In [42]: heights[174] Out[42]: 73.83 In [43]: out\_weights = get\_outlier\_indices(data\_values=weights) print(out\_weights) [] In [ ]:

