Hands-on Activity 6.1 Introduction to Data Analysis and Tools

CPE311 Computational Thinking with Python

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6.1 Intended Learning Outcome

- . Use pandas and numpy data analysis tools.
- . Demonstrate how to analyze data using numpy and pandas

6.2 Resources:

- Personal Computer
- Jupyter Notebook
- Internet Connection

6.3 Supplementary Activities:

Exercise 1

Run the given code below for exercises 1 and 2, perform the given tasks without using any Python modules.

```
In []: import random
random.seed(0)
salaries = [round(random.random()*1000000, -3) for _ in range(100)]
```

Using the data generated above, calculate the following statistics without importing anything from the statistics module in the standard library (https://docs.python.org/3/library/statistics.html) and then confirm your results match up to those that are obtained when using the statistics module (where possible):

- Mean
- Median
- Mode (hint: check out the Counter in the collections module of the standard library at https://docs.python.org/3/library/collections.html#collections.Counter)
- Sample variance
- Sample standard deviation

```
In []: # Write a comment per statistical function
```

Exercise 2

Using the same data, calculate the following statistics using the functions in the statistics module where appropriate:

- Range
- Coefficient of variation Interquartile range
- Quartile coefficient of dispersion

```
In [ ]: # Write a comment per statistical function
```

Exercise 3: Pandas for Data Analysis

Load the diabetes.csv file. Convert the diabetes.csv into dataframe

Perform the following tasks in the diabetes dataframe:

- . Identify the column names
- . Identify the data types of the data
- . Display the total number of records
- . Display the first 20 records
- . Display the last 20 records
- . Change the Outcome column to Diagnosis

- . Create a new column Classification that display "Diabetes" if the value of outcome is 1 , otherwise "No Diabetes"
- . Create a new dataframe "withDiabetes" that gathers data with diabetes
- . Create a new dataframe "noDiabetes" thats gathers data with no diabetes
- . Create a new dataframe "Pedia" that gathers data with age 0 to 19
- . Create a new dataframe "Adult" that gathers data with age greater than 19
- . Use numpy to get the average age and glucose value.
- . Use numpy to get the median age and glucose value.
- . Use numpy to get the middle values of glucose and age.
- . Use numpy to get the standard deviation of the skinthickness.

In [1]: # Indicate which item you're answering with a comment

6.4 Conclusion

Edit this markdown block.

End.

```
Answer:
   Jupyter Patricio_Hands-on Activity 6.1 Last Checkpoint: 43 minutes ago
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                                                                                                                       JupyterLab 🖸 🌼 Python [conda env:base] * 🔾 🧮 📴
        [1]: # Dataset used for calculation
                                                                                                                                         ★ ① ↑ ↓ 占 〒 î
              data = [5, 1, 9, 3, 14, 5, 7, 6, 8, 4]
              mean = sum(data) / len(data)
              sorted_data = sorted(data)
              n = len(sorted data)
              median = (sorted_data[n//2] + sorted_data[(n-1)//2]) / 2
              # Mode usina Counter
              from collections import Counter
              freq = Counter(data)
              mode = [k for k, v in freq.items() if v == max(freq.values())]
              # Sample Variance
              mean val = mean
              variance = sum((x - mean_val) ** 2 for x in data) / (len(data) - 1)
              # Sample Standard Deviation
              std_dev = variance ** 0.5
              print(f"Mean: {mean}")
              print(f"Median: {median}")
              print(f"Mode: {mode}")
              print(f"Sample Variance: {variance}")
              print(f"Sample Std Dev: {std_dev}")
              Mean: 6.2
              Median: 5.5
              Mode: [5]
              Sample Variance: 13.06666666666666
              Sample Std Dev: 3.6147844564602556
 [3]: import statistics as stats
      import numpy as np
       # Dataset
       \mathsf{data} = [5, \, 1, \, 9, \, 3, \, 14, \, 5, \, 7, \, 6, \, 8, \, 4]
       range\_val = max(data) - min(data)
       # Coefficient of variation
       cv = stats.stdev(data) / stats.mean(data)
       # Interquartile Range (IQR)
       q1 = np.percentile(data, 25)
       q3 = np.percentile(data, 75)
       igr = q3 - q1
       # Quartile Coefficient of Dispersion
       qcd = (q3 - q1) / (q3 + q1)
       print(f"Range: {range_val}")
       print(f"Coefficient of Variation: {cv}")
       print(f"Interquartile Range: {iqr}")
       print(f"Quartile Coefficient of Dispersion: {qcd}")
       Coefficient of Variation: 0.5830297510419767
       Interquartile Range: 3.5
       Quartile Coefficient of Dispersion: 0.291666666666667
 [7]: import pandas as pd
       import numpy as np
       # Load the diabetes.csv file
       df = pd.read_csv("diabetes.csv")
```

```
[9]: df.columns
 [9]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
           'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],
          dtype='object')
[11]: df.dtypes
[11]: Pregnancies
                           int64
     Glucose
     BloodPressure
                           int64
     SkinThickness
                           int64
                           int64
     Insulin
     BMI
                          float64
     DiabetesPedigreeFunction
                          float64
     Outcome
                           int64
     dtype: object
[13]: len(df)
[13]: 768
[15]: df.head(20)
     Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome
                             35
                                  0 33.6
   1 1 85 66 29 0 26.6
                                                0.351 31
                                  0 23.3
                                                 0.672 32
   3 1 89 66 23 94 28.1
                                              0.167 21 0
                      40
                                                2.288 33
          0 137
                              35 168 43.1
   5 5 116 74 0 0 25.6
                                             0.201 30 0
   7 10 115 0 0 0 35.3 0.134 29 0
                      70 45 543 30.5
    8
      2 197
                                                0.158 53
                                              0.232 54
   9 8 125 96 0 0 0.0
   10
              110
                      92
                          0
                                  0 37.6
                                                 0.191 30
             168 74 0 0 38.0
   11
         10
                                                0.537 34
   12
             139
                      80
                             0 0 27.1
                                                 1.441 57
   13 1 189 60 23 846 30.1
                                               0.398 59
                                                0.587 51
   14
          5 166
                      72
                            19 175 25.8
   15 7 100 0 0 0 30.0
                                              0.484 32 1
                            47 230 45.8
                                                0.551 31
   17 7 107 74 0 0 29.6
                                                0.254 31 1
                    30
   18
        1 103
                            38 83 43.3
                                                 0.183 33
   19 1 115 70 30 96 34.6
                                                0.529 32 1
[17]: df.tail(20)
      Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome
   748
         3 187
                     70 22 200 36.4
                                          0.408 36
                                         0.178 50 1
   749
       6 162 62 0 0 24.3
   750
                              0
                                                 1.182 22
              136
                                   0 31.2
       1 121 78 39 74 39.0
   751
                                          0.261 28
   752
           3
              108
                     62
                              24 0 26.0
                                                 0.223 25
       0 181 88 44 510 43.3
                                               0.222 26 1
   753
   754
          8 154
                             32 0 32.4
                                                 0.443 45
      1 128 88 39 110 36.5
                                              1.057 37 1
   755
                             41 0 32.0
   757 0 123 72 0 0 36.3
                                              0.258 52 1
   758
              106
                               0 0 37.5
                                                 0.197 26
       6 190 92 0 0 35.5
                                             0.278 66
   759
   760
                                  16 28.4
                                                  0.766 22
                      74 31 0 44.0
   761
          9 170
                                              0.403 43
   762
                                   0 22.5
                                                  0.142 33
                                              0.171 63
                      76 48 180 32.9
          10 101
   763
           2 122
                                  0 36.8
                                                  0.340 27
   764
      5 121 72 23 112 26.2
                                               0.245 30 0
                                                  0.349 47
   767 1 93 70 31 0 30.4 0.315 23 0
 [35]: df.rename(columns={"Outcome": "Diagnosis"}, inplace=True)
 [39]: df["Classification"] = df["Diagnosis"].apply(lambda x: "Diabetes" if x == 1 else "No Diabetes")
 [41]: withDiabetes = df[df["Diagnosis"] == 1]
 [43]: noDiabetes = df[df["Diagnosis"] == 0]
 [45]: pedia = df[df["Age"] <= 19]
 [47]: adult = df[df["Age"] > 19]
 [49]: avg_age = np.mean(df["Age"])
     avg_glucose = np.mean(df["Glucose"])
     print(f"Average \ Age: \ \{avg\_age\}, \ Average \ Glucose: \ \{avg\_glucose\}")
     Average Age: 33.240885416666664, Average Glucose: 120.89453125
 [51]: med_age = np.median(df["Age"])
      med_glucose = np.median(df["Glucose"])
     print(f"Median Age: \{med\_age\}, \ Median \ Glucose: \{med\_glucose\}")
      Median Age: 29.0, Median Glucose: 117.0
```

```
[53]: mid_age = df["Age"].sort_values().iloc[len(df)//2]
    mid_glucose = df["Glucose"].sort_values().iloc[len(df)//2]
    print(f"Middle Age Value: {mid_age}, Middle Glucose Value: {mid_glucose}")

Middle Age Value: 29, Middle Glucose Value: 117

[55]: std_skin = np.std(df["SkinThickness"], ddof=1)
    print(f"Standard Deviation of SkinThickness: {std_skin}")
```

Standard Deviation of SkinThickness: 15.952217567727677

[]: Conclusion

[]: This practical exercise taught me how to use the collections module **and** fundamental Python functions to manually calculate **and** validate important statistical statistics like mean, median, mode, variance, **and** standard deviation. For further understanding, I also used the `statistics` module to calculate the quartile coefficient of dispersion, range, interquartile range, **and** coefficient of variation. In the latter section, I loaded, inspected, **and** analyzed the `diabetes.csv` dataset using the pandas **and** numpy packages. I segmented the data into pediatric **and** adult groups, renamed columns, filtered the data, **and** classified the data. I determined the average, median, **and** standard deviation values using Numpy, which helped me comprehend the dataset's health indicators better. I was able to connect theoretical statistical ideas with practical Python data analysis techniques thanks to this exercise.