Hands-on Activity 6.1 Introduction to Data Analysis and Tools

CPE311 Computational Thinking with Python

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section: CPE22S3

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ILOs

- 1. use pandas and numpy data analysis tools.
- 2. Demonstrate how to analyze data using numpy and pandas

Resources

- Personal Computer
- Jupyter Notebook
- Internet Connection

Supplementary Activities

Exercise 1

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

```
In [1]: import random
import pandas as pd

random.seed(0)
"""make a list that contains 100 elements in it,
```

```
the elements were randomized and were rounded of to 3"""
        salaries = [round(random.random()*1000000, -3) for in range(100)]
In [2]: # code for getting the table of the randomly generated numbers in the salaries list
         salariestbale = pd.Series(salaries, name = 'Salaries')
         salariestbale.index += 1 # the index should start at 1
         print(salariestbale)
        1
                844000.0
        2
               758000.0
         3
               421000.0
                259000.0
        5
                511000.0
                 . . .
        96
                917000.0
        97
               793000.0
                82000.0
        98
        99
               613000.0
        100
               486000.0
        Name: Salaries, Length: 100, dtype: float64
In [3]: # testing how the code works
        test = random.random()*1000000
         print(test)
        testR = round(test,-3)
         print(testR)
        630147.3404114728
        630000.0
```

Mean

```
In [4]: # own code
def Mmean(data):
    sum = 0
    for i in data: # Loop for adding the datas from the list passed on the function
        sum += i
        dmean = sum/len(data) # getting the mean of the list given by dividing it by its # of elements ( used the len() function)
        return dmean

print(Mmean(salaries))

585690.0
```

```
In [5]: # code from the statistics module
from statistics import mean

mean(salaries)

Out[5]: 585690.0
```

Median

```
In [6]: def Mmedian(data):
            data.sort() # sorts the data first
            median = (len(data)+1)/2 # gets the place value of the median
            if len(data)/2 == 1: # if statement if the number of salaries is odd (we can easily get the median)
                 print('list is odd')
                Imedian = median - 1 # gets the index of the median in the list
                 return data[Imedian]
            else: # else statement if the number of salaries is even
                print('list is even')
                LMedian = int(median) - 1 # getting the first number of the median (trunacating using int)
                 RMedian = int(median) # getting the second number of the median
                medianEven = (data[LMedian] + data[RMedian])/2 # gets the average number of the two number on the median
                 return medianEven
        Mmedian(salaries)
        list is even
        589000.0
Out[6]:
In [7]: from statistics import median as mdn
        mdn(salaries)
         589000.0
Out[7]:
```

Mode

```
max_freq = max(frequency.values())
    for val in frequency: # loop for getting the number with the most frequence
        if frequency[val] == max_freq: # conditional statement if the frequency of the current value is the same as the max freq
        return val # return the element with the most frequencies

mode(salaries)

Out[8]:

from statistics import mode as md
        md(salaries)

Out[9]:

4770000.0
```

Sample Variance

Sample standard deviation

```
In [12]: def SD(data):
    return variance(data)**0.5 # returns the square root of the variance
```

```
SD(salaries)
Out[12]: 265827.11382484

In [13]: from statistics import stdev stdev(salaries)
Out[13]: 265827.11382484
```

Exercise 2

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

Range

```
In [14]: data_range = max(salaries) - min(salaries)
print(data_range)

995000.0
```

Coefficient of variation interquartile range

```
In [15]:
    def Cv(data):
        return SD(data) / Mmean(data)

def Iqr(data):
    # getting the quartiles
    q1 = Mmedian(data[:len(data) // 2])
    q3 = Mmedian(data[len(data) // 2:])

# Handle equal quartile case for IQR
    iqr = 0
    if q1 != q3:
        iqr = q3 - q1

    return iqr

# Calculate CV and IQR
iqr = Iqr(salaries)
```

```
cv = Cv(salaries)
# Print the results
print("Coefficient of Variation (CV):", cv)
print("Interquartile Range (IQR):", iqr)

list is even
list is even
Coefficient of Variation (CV): 0.45386998894439035
Interquartile Range (IQR): 417500.0
```

Quartile coefficient of dispersion

```
In [16]: def QCD(data):
    return Iqr(data) / (2 * Mmedian(data))

QCD(salaries)

print('QCD:', QCD(salaries))

list is even
QCD: 0.35441426146010185
```

Exercise 3

```
In [17]: import pandas as pd
import numpy as np

diabetes_data = pd.read_csv('diabetes.csv')
    diabetes_data.index += 1 # index should start in 1
    diabetes_data #load the csv
```

| Out[17]: | | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | ВМІ | DiabetesPedigreeFunction | Age | Outcome |
|----------|-----|-------------|---------|---------------|---------------|---------|------|--------------------------|-----|---------|
| | 1 | 6 | 148 | 72 | 35 | 0 | 33.6 | 0.627 | 50 | 1 |
| | 2 | 1 | 85 | 66 | 29 | 0 | 26.6 | 0.351 | 31 | 0 |
| | 3 | 8 | 183 | 64 | 0 | 0 | 23.3 | 0.672 | 32 | 1 |
| | 4 | 1 | 89 | 66 | 23 | 94 | 28.1 | 0.167 | 21 | 0 |
| | 5 | 0 | 137 | 40 | 35 | 168 | 43.1 | 2.288 | 33 | 1 |
| | ••• | | | | | | | | | |
| | 764 | 10 | 101 | 76 | 48 | 180 | 32.9 | 0.171 | 63 | 0 |
| | 765 | 2 | 122 | 70 | 27 | 0 | 36.8 | 0.340 | 27 | 0 |
| | 766 | 5 | 121 | 72 | 23 | 112 | 26.2 | 0.245 | 30 | 0 |
| | 767 | 1 | 126 | 60 | 0 | 0 | 30.1 | 0.349 | 47 | 1 |
| | 768 | 1 | 93 | 70 | 31 | 0 | 30.4 | 0.315 | 23 | 0 |

768 rows × 9 columns

1. Identify the column names

2. Identify the data types of the data

```
In [19]: diabetes_data.dtypes
```

```
Pregnancies
                                        int64
Out[19]:
         Glucose
                                        int64
         BloodPressure
                                        int64
         SkinThickness
                                        int64
         Insulin
                                        int64
         BMI
                                      float64
         DiabetesPedigreeFunction
                                      float64
         Age
                                        int64
         Outcome
                                        int64
         dtype: object
```

3. Display the total number of records

```
In [20]: print('total number of records in diabetes.csv: ', len(diabetes_data))
total number of records in diabetes.csv: 768
```

4. Display the first 20 records

```
In [21]: diabetes_data.head(20)
```

Out[21]

| • | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | ВМІ | DiabetesPedigreeFunction | Age | Outcome |
|----|-------------|---------|---------------|---------------|---------|------|--------------------------|-----|---------|
| 1 | 6 | 148 | 72 | 35 | 0 | 33.6 | 0.627 | 50 | 1 |
| 2 | 1 | 85 | 66 | 29 | 0 | 26.6 | 0.351 | 31 | 0 |
| 3 | 8 | 183 | 64 | 0 | 0 | 23.3 | 0.672 | 32 | 1 |
| 4 | 1 | 89 | 66 | 23 | 94 | 28.1 | 0.167 | 21 | 0 |
| 5 | 0 | 137 | 40 | 35 | 168 | 43.1 | 2.288 | 33 | 1 |
| 6 | 5 | 116 | 74 | 0 | 0 | 25.6 | 0.201 | 30 | 0 |
| 7 | 3 | 78 | 50 | 32 | 88 | 31.0 | 0.248 | 26 | 1 |
| 8 | 10 | 115 | 0 | 0 | 0 | 35.3 | 0.134 | 29 | 0 |
| 9 | 2 | 197 | 70 | 45 | 543 | 30.5 | 0.158 | 53 | 1 |
| 10 | 8 | 125 | 96 | 0 | 0 | 0.0 | 0.232 | 54 | 1 |
| 11 | 4 | 110 | 92 | 0 | 0 | 37.6 | 0.191 | 30 | 0 |
| 12 | 10 | 168 | 74 | 0 | 0 | 38.0 | 0.537 | 34 | 1 |
| 13 | 10 | 139 | 80 | 0 | 0 | 27.1 | 1.441 | 57 | 0 |
| 14 | 1 | 189 | 60 | 23 | 846 | 30.1 | 0.398 | 59 | 1 |
| 15 | 5 | 166 | 72 | 19 | 175 | 25.8 | 0.587 | 51 | 1 |
| 16 | 7 | 100 | 0 | 0 | 0 | 30.0 | 0.484 | 32 | 1 |
| 17 | 0 | 118 | 84 | 47 | 230 | 45.8 | 0.551 | 31 | 1 |
| 18 | 7 | 107 | 74 | 0 | 0 | 29.6 | 0.254 | 31 | 1 |
| 19 | 1 | 103 | 30 | 38 | 83 | 43.3 | 0.183 | 33 | 0 |
| 20 | 1 | 115 | 70 | 30 | 96 | 34.6 | 0.529 | 32 | 1 |

5. Display the last 20 records

In [22]: diabetes_data.tail(20)

Out[22]

| • | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | вмі | DiabetesPedigreeFunction | Age | Outcome |
|----|-------------|---------|---------------|---------------|---------|------|--------------------------|-----|---------|
| 74 | 9 3 | 187 | 70 | 22 | 200 | 36.4 | 0.408 | 36 | 1 |
| 75 | 0 6 | 162 | 62 | 0 | 0 | 24.3 | 0.178 | 50 | 1 |
| 75 | 1 4 | 136 | 70 | 0 | 0 | 31.2 | 1.182 | 22 | 1 |
| 75 | 2 1 | 121 | 78 | 39 | 74 | 39.0 | 0.261 | 28 | 0 |
| 75 | 3 | 108 | 62 | 24 | 0 | 26.0 | 0.223 | 25 | 0 |
| 75 | 4 0 | 181 | 88 | 44 | 510 | 43.3 | 0.222 | 26 | 1 |
| 75 | 5 8 | 154 | 78 | 32 | 0 | 32.4 | 0.443 | 45 | 1 |
| 75 | 6 1 | 128 | 88 | 39 | 110 | 36.5 | 1.057 | 37 | 1 |
| 75 | 7 7 | 137 | 90 | 41 | 0 | 32.0 | 0.391 | 39 | 0 |
| 75 | 8 0 | 123 | 72 | 0 | 0 | 36.3 | 0.258 | 52 | 1 |
| 75 | 9 1 | 106 | 76 | 0 | 0 | 37.5 | 0.197 | 26 | 0 |
| 76 | 0 6 | 190 | 92 | 0 | 0 | 35.5 | 0.278 | 66 | 1 |
| 76 | 1 2 | 88 | 58 | 26 | 16 | 28.4 | 0.766 | 22 | 0 |
| 76 | 2 9 | 170 | 74 | 31 | 0 | 44.0 | 0.403 | 43 | 1 |
| 76 | 3 9 | 89 | 62 | 0 | 0 | 22.5 | 0.142 | 33 | 0 |
| 76 | 4 10 | 101 | 76 | 48 | 180 | 32.9 | 0.171 | 63 | 0 |
| 76 | 5 2 | 122 | 70 | 27 | 0 | 36.8 | 0.340 | 27 | 0 |
| 76 | 6 5 | 121 | 72 | 23 | 112 | 26.2 | 0.245 | 30 | 0 |
| 76 | 7 1 | 126 | 60 | 0 | 0 | 30.1 | 0.349 | 47 | 1 |
| 76 | 8 1 | 93 | 70 | 31 | 0 | 30.4 | 0.315 | 23 | 0 |

6. Change Outcome column to Diagnosis

| Out[23]: | | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | вмі | DiabetesPedigreeFunction | Age | Diagnosis |
|----------|-----|-------------|---------|---------------|---------------|---------|------|--------------------------|-----|-----------|
| | 1 | 6 | 148 | 72 | 35 | 0 | 33.6 | 0.627 | 50 | 1 |
| | 2 | 1 | 85 | 66 | 29 | 0 | 26.6 | 0.351 | 31 | 0 |
| | 3 | 8 | 183 | 64 | 0 | 0 | 23.3 | 0.672 | 32 | 1 |
| | 4 | 1 | 89 | 66 | 23 | 94 | 28.1 | 0.167 | 21 | 0 |
| | 5 | 0 | 137 | 40 | 35 | 168 | 43.1 | 2.288 | 33 | 1 |
| | ••• | | | | | | | | | |
| | 764 | 10 | 101 | 76 | 48 | 180 | 32.9 | 0.171 | 63 | 0 |
| | 765 | 2 | 122 | 70 | 27 | 0 | 36.8 | 0.340 | 27 | 0 |
| | 766 | 5 | 121 | 72 | 23 | 112 | 26.2 | 0.245 | 30 | 0 |
| | 767 | 1 | 126 | 60 | 0 | 0 | 30.1 | 0.349 | 47 | 1 |
| | 768 | 1 | 93 | 70 | 31 | 0 | 30.4 | 0.315 | 23 | 0 |

768 rows × 9 columns

7. Create a new column Classification that display "Diabetes" if the value of outcome is 1, otherwise "No Diabetes"

| 4]: | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | ВМІ | DiabetesPedigreeFunction | Age | Diagnosis | Classification |
|-----|-------------|---------|---------------|---------------|---------|------|--------------------------|-----|-----------|----------------|
| 1 | 1 6 | 148 | 72 | 35 | 0 | 33.6 | 0.627 | 50 | 1 | Diabetes |
| 2 | 2 1 | 85 | 66 | 29 | 0 | 26.6 | 0.351 | 31 | 0 | No Diabetes |
| 3 | 8 | 183 | 64 | 0 | 0 | 23.3 | 0.672 | 32 | 1 | Diabetes |
| 4 | 4 1 | 89 | 66 | 23 | 94 | 28.1 | 0.167 | 21 | 0 | No Diabetes |
| į | 5 0 | 137 | 40 | 35 | 168 | 43.1 | 2.288 | 33 | 1 | Diabetes |
| •• | •• | | | | | | | | | |
| 764 | 4 10 | 101 | 76 | 48 | 180 | 32.9 | 0.171 | 63 | 0 | No Diabetes |
| 765 | 5 2 | 122 | 70 | 27 | 0 | 36.8 | 0.340 | 27 | 0 | No Diabetes |
| 766 | 6 5 | 121 | 72 | 23 | 112 | 26.2 | 0.245 | 30 | 0 | No Diabetes |
| 767 | 7 1 | 126 | 60 | 0 | 0 | 30.1 | 0.349 | 47 | 1 | Diabetes |
| 768 | 8 1 | 93 | 70 | 31 | 0 | 30.4 | 0.315 | 23 | 0 | No Diabetes |

768 rows × 10 columns

8. Create a new dataframe "withDiabetes" that gathers data with diabetes

```
In [25]: withDiabetesDF = pd.DataFrame(diabetes_data[diabetes_data['Diagnosis'] == 1])
    print(withDiabetesDF)
```

| | Pregnancies | Glucose B | loodPre | ssure S | Skir | Thickness | Insulir | n BMI | . \ |
|-----|--------------|-------------|---------|---------|------|------------|---------|-------|-----|
| 1 | 6 | 148 | | 72 | | 35 | 6 | 33.6 | ; |
| 3 | 8 | 183 | | 64 | | 0 | 6 | 23.3 | 3 |
| 5 | 0 | 137 | | 40 | | 35 | 168 | 43.1 | - |
| 7 | 3 | 78 | | 50 | | 32 | 88 | 31.0 |) |
| 9 | 2 | 197 | | 70 | | 45 | 543 | 30.5 | , |
| • • | • • • | • • • | | • • • | | • • • | • • • | | |
| 756 | 1 | 128 | | 88 | | 39 | 116 | 36.5 | , |
| 758 | 0 | 123 | | 72 | | 0 | 6 | 36.3 | } |
| 760 | 6 | 190 | | 92 | | 0 | 6 | 35.5 | , |
| 762 | 9 | 170 | | 74 | | 31 | 6 | 44.6 |) |
| 767 | 1 | 126 | | 60 | | 0 | 6 | 30.1 | - |
| | | | | | | | | | |
| | DiabetesPedi | greeFunctio | n Age | Diagnos | sis | Classifica | tion | | |
| 1 | | 0.62 | 7 50 | | 1 | Diab | etes | | |
| 3 | | 0.67 | 2 32 | | 1 | Diab | etes | | |
| 5 | | 2.28 | 8 33 | | 1 | Diab | etes | | |
| 7 | | 0.24 | 8 26 | | 1 | Diab | etes | | |
| 9 | | 0.15 | 8 53 | | 1 | Diab | etes | | |
| | | | | | | | • • • | | |
| 756 | | 1.05 | 7 37 | | 1 | Diab | etes | | |
| 758 | | 0.25 | 8 52 | | 1 | Diab | etes | | |
| 760 | | 0.27 | 8 66 | | 1 | Diab | etes | | |
| 762 | | 0.40 | 3 43 | | 1 | Diab | etes | | |
| 767 | | 0.34 | 9 47 | | 1 | Diab | etes | | |
| | | | | | | | | | |

[268 rows x 10 columns]

9. Create a new dataframe "noDiabetes" thats gathers data with no diabetes

```
In [26]: NoDiabetesDF = pd.DataFrame(diabetes_data[diabetes_data['Diagnosis'] == 0])
print(NoDiabetesDF)
```

| | | | | | | _ | | |
|-----|--------------------------|---------|----------|-------|-----------------|---------|------|---|
| | Pregnancies | Glucose | BloodPre | ssure | SkinThickness | Insulin | BMI | , |
| 2 | 1 | 85 | | 66 | 29 | 0 | 26.6 | |
| 4 | 1 | 89 | | 66 | 23 | 94 | 28.1 | |
| 6 | 5 | 116 | | 74 | 0 | 0 | 25.6 | |
| 8 | 10 | 115 | | 0 | 0 | 0 | 35.3 | |
| 11 | 4 | 110 | | 92 | 0 | 0 | 37.6 | |
| | • • • | • • • | | | | • • • | | |
| 763 | 9 | 89 | | 62 | 0 | 0 | 22.5 | |
| 764 | 10 | 101 | | 76 | 48 | 180 | 32.9 | |
| 765 | 2 | 122 | | 70 | 27 | 0 | 36.8 | |
| 766 | 5 | 121 | | 72 | 23 | 112 | 26.2 | |
| 768 | 1 | 93 | | 70 | 31 | 0 | 30.4 | |
| | | | | | | | | |
| | DiabetesPedigreeFunction | | | Diagn | osis Classifica | tion | | |

| | DiabetesPedigreeFunction | Age | Diagnosis | Classification |
|-----|--------------------------|-----|-----------|----------------|
| 2 | 0.351 | 31 | 0 | No Diabetes |
| 4 | 0.167 | 21 | 0 | No Diabetes |
| 6 | 0.201 | 30 | 0 | No Diabetes |
| 8 | 0.134 | 29 | 0 | No Diabetes |
| 11 | 0.191 | 30 | 0 | No Diabetes |
| | ••• | | | • • • |
| 763 | 0.142 | 33 | 0 | No Diabetes |
| 764 | 0.171 | 63 | 0 | No Diabetes |
| 765 | 0.340 | 27 | 0 | No Diabetes |
| 766 | 0.245 | 30 | 0 | No Diabetes |
| 768 | 0.315 | 23 | 0 | No Diabetes |

[500 rows x 10 columns]

10. Create a new dataframe "Pedia" that gathers data with age 0 to 19

```
In [27]: PediaDF = pd.DataFrame(diabetes_data[diabetes_data['Age'] <= 19])
print(PediaDF)</pre>
```

Empty DataFrame

Columns: [Pregnancies, Glucose, BloodPressure, SkinThickness, Insulin, BMI, DiabetesPedigreeFunction, Age, Diagnosis, Classification]

Index: []

11. Create a new dataframe "Adult" that gathers data with age greater than 19

```
In [28]: AdultDF = pd.DataFrame(diabetes_data[diabetes_data['Age'] > 19])
```

| prir | nt(AdultDF) | | | | | | | | | | |
|------|----------------|-------------|----------|-------|-------|-------|--------|--------|----|-------|---|
| | Pregnancies | Glucose | BloodPre | ssure | Skin | Thick | ness | Insuli | n | BMI | \ |
| 1 | 6 | 148 | | 72 | | | 35 | | 0 | 33.6 | |
| 2 | 1 | 85 | | 66 | | | 29 | | 0 | 26.6 | |
| 3 | 8 | 183 | | 64 | | | 0 | | 0 | 23.3 | |
| 4 | 1 | 89 | | 66 | | | 23 | 9 | 4 | 28.1 | |
| 5 | 0 | 137 | | 40 | | | 35 | 16 | 8 | 43.1 | |
| • • | • • • | • • • | | • • • | | | • • • | | • | • • • | |
| 764 | 10 | 101 | | 76 | | | 48 | 18 | 30 | 32.9 | |
| 765 | 2 | 122 | | 70 | | | 27 | | 0 | 36.8 | |
| 766 | 5 | 121 | | 72 | | | 23 | 11 | .2 | 26.2 | |
| 767 | 1 | 126 | | 60 | | | 0 | | 0 | 30.1 | |
| 768 | 1 | 93 | | 70 | | | 31 | | 0 | 30.4 | |
| | DiabetesPedi | lgreeFuncti | on Age | Diagn | osis | Class | ificat | ion | | | |
| 1 | | 0.6 | _ | Ü | 1 | | Diabe | | | | |
| 2 | | 0.3 | 51 31 | | 0 | No | Diabe | tes | | | |
| 3 | | 0.6 | 72 32 | | 1 | | Diabe | tes | | | |
| 4 | | 0.1 | 67 21 | | 0 | No | Diabe | tes | | | |
| 5 | | 2.2 | 88 33 | | 1 | | Diabe | tes | | | |
| • • | | | | | • • • | | | • • • | | | |
| 764 | | 0.1 | 71 63 | | 0 | No | Diabe | tes | | | |
| 765 | | 0.3 | 40 27 | | 0 | No | Diabe | tes | | | |
| 766 | | 0.2 | 45 30 | | 0 | No | Diabe | tes | | | |
| 767 | | 0.3 | 49 47 | | 1 | | Diabe | tes | | | |
| 768 | | 0.3 | 15 23 | | 0 | No | Diabe | tes | | | |
| [768 | 3 rows x 10 cc | olumns] | | | | | | | | | |

12. Use numpy to get the average age and glucose value.

```
In [29]: glucose_ave = np.mean(diabetes_data['Glucose'])
    print(f'average of glucose: {glucose_ave}')
    average of glucose: 120.89453125
```

13. Use numpy to get the median age and glucose value.

```
In [30]: age_median = np.median(diabetes_data['Age'])
print(f'median of age column: {age_median}')
```

```
glucose_median = np.median(diabetes_data['Glucose'])
print(f'median of glucose column: {glucose_median}')

median of age column: 29.0
median of glucose column: 117.0
```

14. Use numpy to get the middle values of glucose and age.

```
In [31]: glucose_median = np.median(diabetes_data['Glucose'])
    print(f'median of glucose column: {glucose_median}')

    age_median = np.median(diabetes_data['Age'])
    print(f'median of age column: {age_median}')

    median of glucose column: 117.0
    median of age column: 29.0
```

15. Use numpy to get the standard deviation of the skinthickness

```
In [32]: skinThickness_std = np.std(diabetes_data['SkinThickness'])
print(f'Standard deviation of skin thickness: {skinThickness_std}')

Standard deviation of skin thickness: 15.941828626496978
```

Conclusion

In this Hands-On activity we reviewed multiple topics (the mean,median,mode,etc.) from our past subjects/courses like math in the modern world from our first year, and Statistics and Probability from our Senior High School level, but in this activity we have used it, or applied those topics in coding and data science, in this activity I have concluded that the statistical tools that we have learned from school are all contained in the python library, those are the numPy and the statistics module,I also observed that the data set that we have used in this case the diabetes.csv file corresponds to a dictionary in python where the column headers are like the keys in the dictionary and the datas under those column are the values and lastly, I have also concluded that we can use our stats in the field of data science.