

hands-on-activity-6-1

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1 Hands-on Activity 6.1 Introduction to Data Analysis and Tools

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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.

2 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.

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

```
[1]: [844000.0,
      758000.0,
      421000.0,
      259000.0,
      511000.0,
      405000.0,
      784000.0,
      303000.0,
      477000.0,
      583000.0,
      908000.0,
      505000.0,
      282000.0,
      756000.0,
      618000.0,
      251000.0,
      910000.0,
      983000.0,
```

810000.0,
902000.0,
310000.0,
730000.0,
899000.0,
684000.0,
472000.0,
101000.0,
434000.0,
611000.0,
913000.0,
967000.0,
477000.0,
865000.0,
260000.0,
805000.0,
549000.0,
14000.0,
720000.0,
399000.0,
825000.0,
668000.0,
1000.0,
494000.0,
868000.0,
244000.0,
325000.0,
870000.0,
191000.0,
568000.0,
239000.0,
968000.0,
803000.0,
448000.0,
80000.0,
320000.0,
508000.0,
933000.0,
109000.0,
551000.0,
707000.0,
547000.0,
814000.0,
540000.0,
964000.0,
603000.0,
588000.0,

```
445000.0,  
596000.0,  
385000.0,  
576000.0,  
290000.0,  
189000.0,  
187000.0,  
613000.0,  
657000.0,  
477000.0,  
90000.0,  
758000.0,  
877000.0,  
923000.0,  
842000.0,  
898000.0,  
923000.0,  
541000.0,  
391000.0,  
705000.0,  
276000.0,  
812000.0,  
849000.0,  
895000.0,  
590000.0,  
950000.0,  
580000.0,  
451000.0,  
660000.0,  
996000.0,  
917000.0,  
793000.0,  
82000.0,  
613000.0,  
486000.0]
```

#Mean

```
[2]: def mean(): # Define a function called mean  
      total = 0  
      for salary in salaries: # Iterate  
          total += salary  
      salary_mean = total/len(salaries) # Calculating the mean by dividing the  
      ↪total by the number of salaries  
      print("Mean salary:", salary_mean) # Display
```

```
[3]: mean() # Calling the function
```

Mean salary: 585690.0

#Median

```
[4]: def median():
      sort = sorted(salaries) # Sort the salaries
      n = len(sort) # Find the total number of salaries
      m = n // 2 # Calculate the middle

      if n % 2: # If odd, the median is the middle salary
          median = sort[m]
          print("Median salary is odd:", median)
      else: # If even, calculate the median by averaging the two middle salaries
          median = (sort[m - 1] + sort[m]) / 2
          print("Median salary is even:", median)
```

```
[5]: median() # Calling the function
```

Median salary is even: 589000.0

3 Mode

```
[6]: def mode(salaries):
      salary_count = {}
      # Checking if the salary is already inside the dictionary
      # if not, set its count to 0
      # If it exists, increase its count by 1
      for num in salaries:
          salary_count[num] = salary_count.get(num, 0) + 1

      max_count = max(salary_count.values()) # Find the maximum count of among
      ↪all salaries
      mode_salaries = ', '.join(str(salary) for salary, count in salary_count.
      ↪items() if count == max_count) # Creating a list to store the salaries that
      ↪found with maximum counts
      print("The mode:", mode_salaries)
```

```
[7]: mode(salaries) # Calling the function
```

The mode: 477000.0

#Sample Variance

```
[10]: def sv():
      mean = sum(salaries)/len(salaries) # Calculate the mean (average) of the
      ↪salaries.
      sum_ = 0 # store the sum of squared differences from the mean.
      for salary in salaries:
```

```

        sum_ += (salary - mean) ** 2 # Add the squared difference between each
↪salary, mean to the sum.

```

```

sample_var = sum_/(len(salaries) - 1) # Calculate the sample variance
print("Sample Variance: ", sample_var)

```

```
[11]: sv()
```

Sample Variance: 70664054444.44444

#Sample Standard Deviation

```
[12]: def ssd():
    mean = sum(salaries)/len(salaries) # Calculate the mean (average) of the
↪salaries.
    sum_ = 0 # store the sum of squared differences from the mean.
    for salary in salaries:
        sum_ += (salary - mean) ** 2 # Add the squared difference between each
↪salary, mean to the sum.

    sample_var = sum_/(len(salaries) - 1)
    std_deviation = sample_var ** 0.5 # Calculate the standard deviation
    print("Sample Standard Deviation:", std_deviation)

```

```
[13]: ssd()
```

Sample Standard Deviation: 265827.11382484

4 Code from the statistics module

```
[14]: from statistics import mean

mean(salaries)
```

```
[14]: 585690.0
```

```
[15]: from statistics import median as mediann

mediann(salaries)
```

```
[15]: 589000.0
```

```
[16]: from statistics import mode as mde

mde(salaries)
```

```
[16]: 477000.0
```

```
[17]: from statistics import stdev

stdev(salaries)
```

[17]: 265827.11382484

#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

#Range

```
[18]: def rangee():
    data_range = max(salaries) - min(salaries)
    print("range: ", data_range)

rangee()
```

range: 995000.0

#Coefficient of variation Interquartile range

```
[26]: def cv(salaries): # Function to calculate Coefficient of Variation (CV)
    return ssd(salaries) / mean(salaries)

def ir(salaries): # Function to Interquartile Range (CV)
    sorted_salaries = sorted(salaries)
    n = len(sorted_salaries)
    q1 = median(sorted_salaries[:n // 2])
    q3 = median(sorted_salaries[n // 2:])
    ir = q3 - q1
    return ir

cv_result = cv(salaries)
ir_result = ir(salaries)

print("Coefficient of Variation (CV):", cv_result)
print("Interquartile Range (IQR):", ir_result)
```

Coefficient of Variation (CV): 0.45386998894439035

Interquartile Range (IQR): 417500.0

#Quartile coefficient of dispersion

```
[36]: def qcd(salaries):
    return ir(salaries)/(2*median(salaries))
qcd_result = qcd(salaries)

print("Quartile coefficient of dispersion: ", qcd_result)
```

Quartile coefficient of dispersion: 0.35441426146010185

#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 DiagnosisIn [: • 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.

```
[67]: import pandas
import numpy

diab_data = pandas.read_csv('diabetes.csv')
diab_data
```

```
[67]:
```

| | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | BMI | \ |
|-----|-------------|---------|---------------|---------------|---------|------|---|
| 0 | 6 | 148 | 72 | 35 | 0 | 33.6 | |
| 1 | 1 | 85 | 66 | 29 | 0 | 26.6 | |
| 2 | 8 | 183 | 64 | 0 | 0 | 23.3 | |
| 3 | 1 | 89 | 66 | 23 | 94 | 28.1 | |
| 4 | 0 | 137 | 40 | 35 | 168 | 43.1 | |
| .. | ... | ... | ... | ... | ... | ... | |
| 763 | 10 | 101 | 76 | 48 | 180 | 32.9 | |
| 764 | 2 | 122 | 70 | 27 | 0 | 36.8 | |
| 765 | 5 | 121 | 72 | 23 | 112 | 26.2 | |
| 766 | 1 | 126 | 60 | 0 | 0 | 30.1 | |
| 767 | 1 | 93 | 70 | 31 | 0 | 30.4 | |

| | DiabetesPedigreeFunction | Age | Outcome |
|-----|--------------------------|-----|---------|
| 0 | 0.627 | 50 | 1 |
| 1 | 0.351 | 31 | 0 |
| 2 | 0.672 | 32 | 1 |
| 3 | 0.167 | 21 | 0 |
| 4 | 2.288 | 33 | 1 |
| .. | ... | ... | ... |
| 763 | 0.171 | 63 | 0 |
| 764 | 0.340 | 27 | 0 |
| 765 | 0.245 | 30 | 0 |
| 766 | 0.349 | 47 | 1 |
| 767 | 0.315 | 23 | 0 |

[768 rows x 9 columns]

```
#Identify the column names
```

```
[68]: diab_data.columns
```

```
[68]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',  
         'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],  
        dtype='object')
```

```
#Identify the data types of the data
```

```
[69]: diab_data.dtypes
```

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

5 Display the total number of records

```
[70]: print("total records of diabetes: ", len(diab_data))
```

```
total records of diabetes: 768
```

6 Display the first 20 records

```
[71]: diab_data.head(20)
```

```
[71]:
```

| | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | BMI | \ |
|----|-------------|---------|---------------|---------------|---------|------|---|
| 0 | 6 | 148 | 72 | 35 | 0 | 33.6 | |
| 1 | 1 | 85 | 66 | 29 | 0 | 26.6 | |
| 2 | 8 | 183 | 64 | 0 | 0 | 23.3 | |
| 3 | 1 | 89 | 66 | 23 | 94 | 28.1 | |
| 4 | 0 | 137 | 40 | 35 | 168 | 43.1 | |
| 5 | 5 | 116 | 74 | 0 | 0 | 25.6 | |
| 6 | 3 | 78 | 50 | 32 | 88 | 31.0 | |
| 7 | 10 | 115 | 0 | 0 | 0 | 35.3 | |
| 8 | 2 | 197 | 70 | 45 | 543 | 30.5 | |
| 9 | 8 | 125 | 96 | 0 | 0 | 0.0 | |
| 10 | 4 | 110 | 92 | 0 | 0 | 37.6 | |
| 11 | 10 | 168 | 74 | 0 | 0 | 38.0 | |

| | | | | | | |
|----|----|-----|----|----|-----|------|
| 12 | 10 | 139 | 80 | 0 | 0 | 27.1 |
| 13 | 1 | 189 | 60 | 23 | 846 | 30.1 |
| 14 | 5 | 166 | 72 | 19 | 175 | 25.8 |
| 15 | 7 | 100 | 0 | 0 | 0 | 30.0 |
| 16 | 0 | 118 | 84 | 47 | 230 | 45.8 |
| 17 | 7 | 107 | 74 | 0 | 0 | 29.6 |
| 18 | 1 | 103 | 30 | 38 | 83 | 43.3 |
| 19 | 1 | 115 | 70 | 30 | 96 | 34.6 |

| | DiabetesPedigreeFunction | Age | Outcome |
|----|--------------------------|-----|---------|
| 0 | 0.627 | 50 | 1 |
| 1 | 0.351 | 31 | 0 |
| 2 | 0.672 | 32 | 1 |
| 3 | 0.167 | 21 | 0 |
| 4 | 2.288 | 33 | 1 |
| 5 | 0.201 | 30 | 0 |
| 6 | 0.248 | 26 | 1 |
| 7 | 0.134 | 29 | 0 |
| 8 | 0.158 | 53 | 1 |
| 9 | 0.232 | 54 | 1 |
| 10 | 0.191 | 30 | 0 |
| 11 | 0.537 | 34 | 1 |
| 12 | 1.441 | 57 | 0 |
| 13 | 0.398 | 59 | 1 |
| 14 | 0.587 | 51 | 1 |
| 15 | 0.484 | 32 | 1 |
| 16 | 0.551 | 31 | 1 |
| 17 | 0.254 | 31 | 1 |
| 18 | 0.183 | 33 | 0 |
| 19 | 0.529 | 32 | 1 |

#Display the last 20 records

```
[72]: diab_data.tail(20)
```

```
[72]:
```

| | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | BMI | \ |
|-----|-------------|---------|---------------|---------------|---------|------|---|
| 748 | 3 | 187 | 70 | 22 | 200 | 36.4 | |
| 749 | 6 | 162 | 62 | 0 | 0 | 24.3 | |
| 750 | 4 | 136 | 70 | 0 | 0 | 31.2 | |
| 751 | 1 | 121 | 78 | 39 | 74 | 39.0 | |
| 752 | 3 | 108 | 62 | 24 | 0 | 26.0 | |
| 753 | 0 | 181 | 88 | 44 | 510 | 43.3 | |
| 754 | 8 | 154 | 78 | 32 | 0 | 32.4 | |
| 755 | 1 | 128 | 88 | 39 | 110 | 36.5 | |
| 756 | 7 | 137 | 90 | 41 | 0 | 32.0 | |
| 757 | 0 | 123 | 72 | 0 | 0 | 36.3 | |
| 758 | 1 | 106 | 76 | 0 | 0 | 37.5 | |

| | | | | | | |
|-----|----|-----|----|----|-----|------|
| 759 | 6 | 190 | 92 | 0 | 0 | 35.5 |
| 760 | 2 | 88 | 58 | 26 | 16 | 28.4 |
| 761 | 9 | 170 | 74 | 31 | 0 | 44.0 |
| 762 | 9 | 89 | 62 | 0 | 0 | 22.5 |
| 763 | 10 | 101 | 76 | 48 | 180 | 32.9 |
| 764 | 2 | 122 | 70 | 27 | 0 | 36.8 |
| 765 | 5 | 121 | 72 | 23 | 112 | 26.2 |
| 766 | 1 | 126 | 60 | 0 | 0 | 30.1 |
| 767 | 1 | 93 | 70 | 31 | 0 | 30.4 |

| | DiabetesPedigreeFunction | Age | Outcome |
|-----|--------------------------|-----|---------|
| 748 | 0.408 | 36 | 1 |
| 749 | 0.178 | 50 | 1 |
| 750 | 1.182 | 22 | 1 |
| 751 | 0.261 | 28 | 0 |
| 752 | 0.223 | 25 | 0 |
| 753 | 0.222 | 26 | 1 |
| 754 | 0.443 | 45 | 1 |
| 755 | 1.057 | 37 | 1 |
| 756 | 0.391 | 39 | 0 |
| 757 | 0.258 | 52 | 1 |
| 758 | 0.197 | 26 | 0 |
| 759 | 0.278 | 66 | 1 |
| 760 | 0.766 | 22 | 0 |
| 761 | 0.403 | 43 | 1 |
| 762 | 0.142 | 33 | 0 |
| 763 | 0.171 | 63 | 0 |
| 764 | 0.340 | 27 | 0 |
| 765 | 0.245 | 30 | 0 |
| 766 | 0.349 | 47 | 1 |
| 767 | 0.315 | 23 | 0 |

7 Change the Outcome column to DiagnosisIn []:

```
[73]: diab_data.rename(columns = {'Outcome': 'Diagnosis'}, inplace = True)
      diab_data
```

```
[73]:
```

| | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | BMI | \ |
|-----|-------------|---------|---------------|---------------|---------|------|---|
| 0 | 6 | 148 | 72 | 35 | 0 | 33.6 | |
| 1 | 1 | 85 | 66 | 29 | 0 | 26.6 | |
| 2 | 8 | 183 | 64 | 0 | 0 | 23.3 | |
| 3 | 1 | 89 | 66 | 23 | 94 | 28.1 | |
| 4 | 0 | 137 | 40 | 35 | 168 | 43.1 | |
| .. | ... | ... | ... | ... | ... | ... | |
| 763 | 10 | 101 | 76 | 48 | 180 | 32.9 | |
| 764 | 2 | 122 | 70 | 27 | 0 | 36.8 | |

| | | | | | | |
|-----|---|-----|----|----|-----|------|
| 765 | 5 | 121 | 72 | 23 | 112 | 26.2 |
| 766 | 1 | 126 | 60 | 0 | 0 | 30.1 |
| 767 | 1 | 93 | 70 | 31 | 0 | 30.4 |

| | DiabetesPedigreeFunction | Age | Diagnosis |
|-----|--------------------------|-----|-----------|
| 0 | 0.627 | 50 | 1 |
| 1 | 0.351 | 31 | 0 |
| 2 | 0.672 | 32 | 1 |
| 3 | 0.167 | 21 | 0 |
| 4 | 2.288 | 33 | 1 |
| .. | ... | ... | ... |
| 763 | 0.171 | 63 | 0 |
| 764 | 0.340 | 27 | 0 |
| 765 | 0.245 | 30 | 0 |
| 766 | 0.349 | 47 | 1 |
| 767 | 0.315 | 23 | 0 |

[768 rows x 9 columns]

#Create a new column Classification that display “Diabetes” if the value of outcome is 1 , otherwise “No Diabetes”

```
[74]: diab_data["Classification"] = numpy.where(diab_data["Diagnosis"] == 1,
↪      ↪ "Diabetes", "No Diabetes")
diab_data
```

```
[74]:      Pregnancies  Glucose  BloodPressure  SkinThickness  Insulin   BMI   \
0              6      148            72           35          0  33.6
1              1       85            66           29          0  26.6
2              8      183            64            0          0  23.3
3              1       89            66           23         94  28.1
4              0      137            40           35        168  43.1
..          ...      ...            ...           ...          ...  ...
763            10      101            76           48        180  32.9
764              2      122            70           27          0  36.8
765              5      121            72           23        112  26.2
766              1      126            60            0          0  30.1
767              1       93            70           31          0  30.4
```

| | DiabetesPedigreeFunction | Age | Diagnosis | Classification |
|-----|--------------------------|-----|-----------|----------------|
| 0 | 0.627 | 50 | 1 | Diabetes |
| 1 | 0.351 | 31 | 0 | No Diabetes |
| 2 | 0.672 | 32 | 1 | Diabetes |
| 3 | 0.167 | 21 | 0 | No Diabetes |
| 4 | 2.288 | 33 | 1 | Diabetes |
| .. | ... | ... | ... | ... |
| 763 | 0.171 | 63 | 0 | No Diabetes |

| | | | | |
|-----|-------|----|---|-------------|
| 764 | 0.340 | 27 | 0 | No Diabetes |
| 765 | 0.245 | 30 | 0 | No Diabetes |
| 766 | 0.349 | 47 | 1 | Diabetes |
| 767 | 0.315 | 23 | 0 | No Diabetes |

[768 rows x 10 columns]

#Create a new dataframe “withDiabetes” that gathers data with diabetes

```
[78]: diab_data = pandas.DataFrame(diab_data)
withDiabetes = diab_data[diab_data['Diagnosis'] == 1].copy()

withDiabetes
```

```
[78]:      Pregnancies  Glucose  BloodPressure  SkinThickness  Insulin   BMI  \
0              6      148             72           35         0  33.6
2              8      183             64           0         0  23.3
4              0      137             40           35        168  43.1
6              3       78             50           32         88  31.0
8              2      197             70           45        543  30.5
..          ...    ...             ...           ...         ...  ...
755            1      128             88           39        110  36.5
757            0      123             72           0         0  36.3
759            6      190             92           0         0  35.5
761            9      170             74           31         0  44.0
766            1      126             60           0         0  30.1
```

| | DiabetesPedigreeFunction | Age | Diagnosis | Classification |
|-----|--------------------------|-----|-----------|----------------|
| 0 | 0.627 | 50 | 1 | Diabetes |
| 2 | 0.672 | 32 | 1 | Diabetes |
| 4 | 2.288 | 33 | 1 | Diabetes |
| 6 | 0.248 | 26 | 1 | Diabetes |
| 8 | 0.158 | 53 | 1 | Diabetes |
| .. | | | ... | ... |
| 755 | 1.057 | 37 | 1 | Diabetes |
| 757 | 0.258 | 52 | 1 | Diabetes |
| 759 | 0.278 | 66 | 1 | Diabetes |
| 761 | 0.403 | 43 | 1 | Diabetes |
| 766 | 0.349 | 47 | 1 | Diabetes |

[268 rows x 10 columns]

8 Create a new dataframe “noDiabetes” thats gathers data with no diabetes

```
[84]: diab_data = pandas.DataFrame(diab_data)
Nodiab = diab_data[diab_data['Diagnosis'] == 0].copy()

Nodiab
```

```
[84]:      Pregnancies  Glucose  BloodPressure  SkinThickness  Insulin   BMI  \
1              1        85             66           29         0  26.6
3              1        89             66           23        94  28.1
5              5       116             74            0         0  25.6
7             10       115              0            0         0  35.3
10             4       110             92            0         0  37.6
..           ...      ...             ...           ...      ...  ...
762            9        89             62            0         0  22.5
763           10       101             76           48       180  32.9
764            2       122             70           27         0  36.8
765            5       121             72           23       112  26.2
767            1        93             70           31         0  30.4
```

```
      DiabetesPedigreeFunction  Age  Diagnosis  Classification
1              0.351    31         0    No Diabetes
3              0.167    21         0    No Diabetes
5              0.201    30         0    No Diabetes
7              0.134    29         0    No Diabetes
10             0.191    30         0    No Diabetes
..           ...  ...             ...           ...
762             0.142    33         0    No Diabetes
763             0.171    63         0    No Diabetes
764             0.340    27         0    No Diabetes
765             0.245    30         0    No Diabetes
767             0.315    23         0    No Diabetes
```

[500 rows x 10 columns]

#Create a new dataframe “Pedia” that gathers data with age 0 to 19

```
[85]: diab_data = pandas.DataFrame(diab_data)
pedia = diab_data[diab_data['Age'] <= 19].copy()

pedia
```

```
[85]: Empty DataFrame
Columns: [Pregnancies, Glucose, BloodPressure, SkinThickness, Insulin, BMI,
DiabetesPedigreeFunction, Age, Diagnosis, Classification]
Index: []
```

#Create a new dataframe “Adult” that gathers data with age greater than 19

```
[87]: diab_data = pandas.DataFrame(diab_data)
Adult = diab_data[diab_data['Age'] > 19].copy()
```

Adult

```
[87]:      Pregnancies  Glucose  BloodPressure  SkinThickness  Insulin   BMI  \
0              6     148             72           35         0  33.6
1              1      85             66           29         0  26.6
2              8     183             64            0         0  23.3
3              1      89             66           23        94  28.1
4              0     137             40           35       168  43.1
..          ...    ...             ...           ...         ...  ...
763            10     101             76           48       180  32.9
764             2     122             70           27         0  36.8
765             5     121             72           23       112  26.2
766             1     126             60            0         0  30.1
767             1      93             70           31         0  30.4
```

```
      DiabetesPedigreeFunction  Age  Diagnosis  Classification
0              0.627     50           1      Diabetes
1              0.351     31           0      No Diabetes
2              0.672     32           1      Diabetes
3              0.167     21           0      No Diabetes
4              2.288     33           1      Diabetes
..          ...    ...             ...           ...
763            0.171     63           0      No Diabetes
764            0.340     27           0      No Diabetes
765            0.245     30           0      No Diabetes
766            0.349     47           1      Diabetes
767            0.315     23           0      No Diabetes
```

[768 rows x 10 columns]

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

```
[88]: mean_age = numpy.mean(diab_data['Age'])
mean_glucose = numpy.mean(diab_data['Glucose'])

print("Average Age:", mean_age)
print("Average Glucose Value:", mean_glucose)
```

Average Age: 33.240885416666664

Average Glucose Value: 120.89453125

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

```
[89]: median_age = numpy.median(diab_data['Age'])
      median_glucose = numpy.median(diab_data['Glucose'])

      print("Median Age:", median_age)
      print("Median Glucose Value:", median_glucose)
```

Average Age: 29.0

Average Glucose Value: 117.0

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

```
[ ]: median_age = numpy.median(diab_data['Age'])
      median_glucose = numpy.median(diab_data['Glucose'])

      print("Median Age:", median_age)
      print("Median Glucose Value:", median_glucose)
```

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

```
[91]: skinthickness_std = numpy.std(diab_data['SkinThickness'])

      print("Standard deviation of the skinthickness: ", skinthickness_std)
```

Standard deviation of the skinthickness: 15.941828626496939

#6.4 Conclusion

In this HOA, I think i got a little hang of it now at coding because we tried doing things without using special tools or bringing in extra stuff. We looked at our own code and compared it to what's already built into Python. We also practiced bringing in data from csv and using helpful tools like numpy and pandas. Using these tools made our work quicker and simpler, especially when we had clear instructions. So basically, we learned some tricks and figured out how to use built-in features more effectively.