## chap3-statistics

June 16, 2022

```
[1]: # import pandas library
    import pandas as pd
    # create dataframe
    sample_data = {'name': ['John', 'Alia', 'Anna', 'Monika', 'Sylwia'],
                 'gender': ['M', 'F', 'F', 'F', 'F'],
                 'communication_skill_score': [30, 40, 45, 50, 55],
                 'quantitative_skill_score': [38, 41, 42, 48, 32]}
    data = pd.DataFrame(sample_data, columns=['name',__
     # find mean of communication_skill_score column
    data['communication_skill_score'].mean(axis=0)
[1]: 44.0
[2]: # find mode of communication_skill_score column
    data['communication_skill_score'].mode()
[2]: 0
         30
    1
         40
    2
         45
    3
         50
    Name: communication_skill_score, dtype: int64
[3]: # find median of communication_skill_score column
    data['communication_skill_score'].median()
[3]: 45.0
[4]: # measuring dispersion
    column_range=data['communication_skill_score'].
     →max()-data['communication_skill_score'].min()
    print(column_range)
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[5]: # first Quartile
      q1 = data['communication_skill_score'].quantile(.25)
      # third Quartile
      q3 = data['communication_skill_score'].quantile(.75)
      # inter Quartile Ratio
      iqr = q3 - q1
      print(iqr)
     10.0
 [6]: # variance of communication_skill_score
      data['communication skill score'].var()
 [6]: 92.5
 [7]: # standard deviation of communication_skill_score
      data['communication skill score'].std()
 [7]: 9.617692030835672
 [8]: # discribe dataframe
      data.describe()
 [8]:
             communication_skill_score quantitative_skill_score
                              5.000000
                                                         5.000000
      count
                             44.000000
                                                        40.200000
     mean
                              9.617692
                                                         5.848077
      std
     min
                             30.000000
                                                        32.000000
      25%
                             40.000000
                                                        38.000000
      50%
                             45.000000
                                                        41.000000
      75%
                             50.000000
                                                        42.000000
                             55.000000
                                                        48.000000
     max
 [9]: #skewness of communication_skill_score column
      data['communication_skill_score'].skew()
 [9]: -0.5901286563843656
[10]: # kurtosis of communication_skill_score column
      data['communication_skill_score'].kurtosis()
[10]: -0.02191380569758916
[11]: # covariance between columns of dataframe
      data.cov()
```

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[11]:
                                 communication_skill_score quantitative_skill_score
                                                                                 -3.5
      communication_skill_score
                                                       92.5
                                                                                 34.2
      quantitative_skill_score
                                                       -3.5
[12]: # correlation between columns of dataframe
      data.corr(method='pearson')
[12]:
                                 communication_skill_score quantitative_skill_score
      communication_skill_score
                                                  1.000000
                                                                            -0.062228
                                                 -0.062228
                                                                             1.000000
      quantitative_skill_score
[13]: ##### Performing parametric tests #####
      import numpy as np
      from scipy.stats import ttest_1samp
      # create data
      data = np.array([63, 75, 84, 58, 52, 96, 63, 55, 76, 83])
      # find mean
      mean_value = np.mean(data)
      print("Mean: ", mean_value)
     Mean: 70.5
[14]: # perform one-sample t-test
      t_test_value, p_value = ttest_1samp(data, 68)
      print("P Value: ", p_value)
      print("t-test Value: ", t_test_value)
      # 0.05 or 5 % is significance level or alpha
      if p_value < 0.05:</pre>
          print("Hipothesis Rejected")
      else:
          print("Hipothesis Accepted")
     P Value: 0.5986851106160134
     t-test Value: 0.5454725779039431
     Hipothesis Accepted
[15]: from scipy.stats import ttest_ind
      # create numpy arrays
      data1 = np.array([63, 75, 84, 58, 52, 96, 63, 55, 76, 83])
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data2 = np.array([53, 43, 31, 113, 33, 57, 27, 23, 24, 43])

# compare samples
stat, p = ttest_ind(data1, data2)

print("p_value: ", p)

print("t_test: ", stat)

# 0.05 or 5 % is the significance level of alpha
if p < 0.05:
    print("Hypothesis Rejected")
else:
    print("Hypotesis Accepted")</pre>
```

p\_value: 0.015170931362451255
t\_test: 2.6835879913819185

Hypothesis Rejected

```
[17]: # compare results using Oneway ANOVA
stat, p = f_oneway(mumbai, chicago, london)

print("p-value:", p)

print("ANOVA:", stat)

if p < 0.05:
    print("Hypothesis Rejected")
else:
    print("Hypothesis Accepted")</pre>
```

p-value: 0.27667556390705783
ANOVA: 1.3480446381965452
Hypothesis Accepted

```
[20]: #### Performing non-parametric tests ####

## chi-square test
from scipy.stats import chi2_contingency

# average performing employee
average = [20, 16, 13, 7]

# outstanding performing employees
outstanding = [31, 40, 60, 13]

#contingency table
contingency_table = [average, outstanding]
```

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[22]: # apply test
stat, p, dof, expected = chi2_contingency(contingency_table)

print("p-value:", p)

if p < 0.05:
    print("Hypothesis Rejected")
else:
    print("Hypothesis Accepted")</pre>
```

p-value: 0.059155602774381234 Hypothesis Accepted

```
[23]: # Mann-Whitney U test

from scipy.stats import mannwhitneyu

# sample 1
data1 = [7,8,4,9,8]

# sample 2
data2 = [3,4,2,1,1]

# apply test
stat, p = mannwhitneyu(data1, data2)

print("p-value: ", p)
```

```
# 0.01 or 1 % is a significance level of alpha

if p < 0.01:
    print("Hypothesis Rejected")
else:
    print("Hypothesis Accepted")</pre>
```

p-value: 0.015333162113602824 Hypothesis Accepted

```
[24]: ## Wilcoxon test

from scipy.stats import wilcoxon

# sample1

data1 = [1, 3, 5, 7, 9]

# sample-2 after treatment
data2 = [2, 4, 6, 8, 10]

# apply
stat, p = wilcoxon(data1, data2)

print("p-value: ", p)

# 0.01 or 1 % is significance level of alpha
if p < 0.01:
    print("Hypothesis Rejected")
else:
    print("Hypothesis Accepted")</pre>
```

p-value: 0.0625 Hypothesis Accepted

```
[29]: ## Kruskal-Wallis test
from scipy.stats import kruskal

# data sample 1

x = [38, 18, 39, 83, 15, 38, 63, 1, 34, 50]

# data sample 2

y = [78, 32, 58, 59, 74, 77, 29, 77, 54, 59]

# data sample 3

z = [117, 92, 42, 79, 58, 117, 46, 114, 86, 26]
```

```
# apply kruskal-wallis test
stat, p = kruskal(x, y, z)

print("p-value: ", p)

# 0.01 or 1 % is significance level or alpha
if p < 0.01:
    print("Hypothesis Rejected")
else:
    print("Hypothesis Accepted")</pre>
```

p-value: 0.01997922369138151

Hypothesis Accepted

[]: