VIT-AP UNIVERSITY, ANDHRA PRADESH

CSE4005 - DWDM - LAB 6

Academic year: 2022-2023 Branch/ Class: B.Tech

Semester: Fall Date: 09/10/22
Faculty Name: Dr. Aravapalli Rama Satish School: SCOPE

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1) a.Generate 100 random numbers for the age attribute and plot the equal-width(uniform) and equal frequency histograms.

b.Add two attributes ID (1, 2, 3..., 100), Category (youth, middle_aged, Senior) for each age value. Develop user defined functions to perform sampling of the age attribute: SRSWOR, SRSWR, and stratified sampling. Use samples of size 10 and the strata "youth," "middle-aged," and "senior."

Code and output:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as pt
import random
Age_Attribute=np.random.randint(10,100,100)
Age=Age_Attribute.tolist()
print(Age_Attribute)
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as pt
import random
Age_Attribute=np.random.randint(10,100,100)
Age=Age_Attribute.tolist()
print(Age_Attribute)
[10 61 33 78 46 56 84 90 30 35 43 39 85 20 62 34 77 35 79 71 70 77 43 73
52 83 20 86 60 59 37 36 67 17 54 55 26 53 46 23 73 33 79 36 59 15 79 35
59 43 33 45 99 17 33 17 39 99 90 90 13 29 26 24 75 91 34 71 58 80 95 67
42 83 69 88 20 59 56 14 56 19 59 17 47 19 45 74 17 69 14 84 46 48 34 75
14 13 59 96]

n,bins,patches=pt.hist(Age_Attribute,edgecolor='black')
pt.show()
```

```
n,bins,patches=pt.hist(Age_Attribute,edgecolor='black')
      pt.show()
  ₽
       14
       12
       10
        8
        6
        2
        0
                20
                         40
                                   60
                                            80
                                                     100
def equalfree(x,nbin):
  return np.interp(np.linspace(0,nlen,nbin +1),
                      np.arange(nlen),
                      np.sort(x))
n,bins,patches = pt.hist(Age Attribute,equalfree(Age Attribute,20),edge
pt.show()
    def equalfree(x,nbin):
       nlen = len(x)
       return np.interp(np.linspace(0,nlen,nbin +1),
                      np.arange(nlen),
                      np.sort(x))
     n,bins,patches = pt.hist(Age_Attribute,equalfree(Age_Attribute,20),edgecolor='black')
     pt.show()
 ₽
      5
      4
      3
      2
      1
                                    80
                                           100
def SRSWR(n):
  b=[random.choice(Age_Attribute) for i in range(n)]
SRSWR(20)
```

```
def SRSWR(n):
        b=[random.choice(Age_Attribute) for i in range(n)]
        return b
      SRSWR(20)
  [20,
       34,
       75,
       13,
       33,
       20,
       33,
       23,
       13,
       20,
       91,
       43,
       34,
       24,
       46,
       34,
       59,
       35,
       14,
       90]
Category=[]
for i in range(len(Age Attribute)):
  if Age Attribute[i] > 60:
    Category.append("Senior")
  elif Age Attribute[i]>=35 and Age Attribute[i]<=50:</pre>
    Category.append("Middle-aged")
    Category.append("Youth")
print(Category)
     Category=[]
     for i in range(len(Age_Attribute)):
       if Age_Attribute[i] > 60:
        Category.append("Senior")
       elif Age_Attribute[i]>=35 and Age_Attribute[i]<=50:
         Category.append("Middle-aged")
         Category.append("Youth")
     print(Category)
  ['Youth', 'Senior', 'Youth', 'Senior', 'Middle-aged', 'Youth', 'Senior', 'Senior', 'Youth',
```

```
import pandas as pd
def Stratified():
  d=pd.DataFrame({'Age Attribute':Age Attribute, 'Category':Category})
  A=d.groupby('Category',group keys=False).apply(lambda x: x.sample(fra
c=0.40)
  print(A)
Stratified()
      import pandas as pd
  O
      def Stratified():
        d=pd.DataFrame({ Age Attribute : Age Attribute, 'Category : Category})
        A=d.groupby('Category',group_keys=False).apply(lambda x: x.sample(frac=0.40))
        print(A)
      Stratified()
        Age_Attribute
                        Category
 O
    84
                 47 Middle-aged
                 39 Middle-aged
 ₽
                  39 Middle-aged
                 46 Middle-aged
     4
                35 Middle-aged
                35 Middle-aged
                 48 Middle-aged
                 43 Middle-aged
                         Senior
                         Senior
                         Senior
                         Senior
                  90
                         Senior
                         Senior
                         Senior
                         Senior
                         Senior
     25
                          Senior
                         Senior
                         Senior
                          Youth
     0
                 10
                          Youth
                          Youth
                          Youth
                          Youth
                          Youth
     68
                          Youth
                          Youth
                          Youth
     80
                          Youth
                          Youth
     60
                          Youth
                          Youth
                          Youth
                          Youth
```

2) Develop user defined function to calculate chi-square correlation test for Nominal Data and to decide whether the two nominal attributes are independent or not.

Example Data: (Level of significance:0.05)

Apply your function all pairs of Nominal attributes and formulate a chi-square correlation matrix and label each cell as 'D' (Dependent) or 'l' (Independent).

	High School	Bachelors	Masters	Ph.d.	Total
Female	90	84	76	66	316
Male	60	64	73	83	280
Total	150	148	149	149	596

df	0.5	0.10	0.05	0.02	0.01	0.001
1	0.455	2.706	3.841	5.412	6.635	10.827
2	1.386	4.605	5.991	7.824	9.210	13.815
3	2.366	6.251	7.815	9.837	11.345	16.268
4	3.357	7.779	9.488	11.668	13.277	18.465
5	4.351	9.236	11.070	13.388	15.086	20.517

Code and output:

```
import pandas as pd
import numpy as np
def chi(observed, expected):
 chi=list()
    for k, j in observed.iterrows():
       print(k)
        print(observed[i][k])
        print(expected[i][k])
        a=observed[i][k]-expected[i][k]
       a=float(a*a/expected[i][k])
    if a<3.841:
      observed[i][k]='I'
      observed[i][k]='D'
  print(observed)
DATA={'HIGH SCHOOL':[90,60],'BACHELORS':[84,64],'MASTERS':[76,73],'PH.D':[6
6,83]}
DATA=pd.DataFrame(DATA,index=pd.Index(['Female','Male']))
DATA['TOTAL'] = DATA.sum(axis=1)
DATA.loc['TOTAL'] = DATA.sum(axis=0)
DATA2=DATA.copy()
for i in DATA2:
 for k, j in DATA.iterrows():
    if(i!='TOTAL' and k!='TOTAL'):
      DATA2[i][k]=DATA[i]['TOTAL']*DATA.loc[k]['TOTAL']/DATA['TOTAL']['TOTA
print(DATA)
print(DATA2)
chi (DATA, DATA2)
```

```
import pandas as pd
import numpy as np
def chi(observed, expected):
  chi=list()
  for i in observed:
    for k,j in observed.iterrows():
      if(i!='TOTAL' and k!='TOTAL'):
        print(i)
        print(k)
        print(observed[i][k])
        print(expected[i][k])
        a=observed[i][k]-expected[i][k]
        a=float(a*a/expected[i][k])
    if a<3.841:
      observed[i][k]='I'
      observed[i][k]='D'
  print(observed)
DATA={'HIGH SCHOOL':[90,60],'BACHELORS':[84,64],'MASTERS':[76,73],'PH.D':[66,83]}
DATA=pd.DataFrame(DATA,index=pd.Index(['Female','Male']))
DATA['TOTAL']=DATA.sum(axis=1)
DATA.loc['TOTAL']=DATA.sum(axis=0)
DATA2=DATA.copy()
for i in DATA2:
  for k,j in DATA.iterrows():
    if(i!='TOTAL' and k!='TOTAL'):
      DATA2[i][k]=DATA[i]['TOTAL']*DATA.loc[k]['TOTAL']/DATA['TOTAL']['TOTAL']
print(DATA)
print(DATA2)
chi(DATA,DATA2)
```

```
HIGH SCHOOL BACHELORS MASTERS PH.D TOTAL
Female
                     90
                                  84
                                             76
                                                   66
                                                            316
Male
                     60
                                  64
                                             73
                                                     83
                                                             280
                    150
                                 148
                                                    149
TOTAL
                                            149
                                                            596
          HIGH SCHOOL BACHELORS MASTERS PH.D
                                                          TOTAL
Female
                     79
                                 78
                                             79
                                                    79
                                                            316
Male
                     70
                                 69
                                             70
                                                    70
                                                            280
TOTAL
                    150
                                148
                                            149
                                                    149
                                                            596
HIGH SCHOOL
Female
90
HIGH SCHOOL
Male
60
70
BACHELORS
Female
84
78
BACHELORS
Male
64
MASTERS
Female
76
79
MASTERS
Male
73
70
PH.D
Female
79
PH.D
Male
70
       HIGH SCHOOL BACHELORS MASTERS PH.D TOTAL
Female 90 84 76 66 316
Male
/usr/local/lib/python 3.7/dist-packages/ipykernel\_launcher.py: 15: Setting With Copy Warning: \\
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_g">https://pandas.pydata.org/pandas-docs/stable/user_g</a>
 from ipykernel import kernelapp as app
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:1732: SettingWithCopyWarnin
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_g">https://pandas.pydata.org/pandas-docs/stable/user_g</a>
  self._setitem_single_block(indexer, value, name)
```

3) Develop user defined function to calculate the correlation coefficient for two numerical attributes and should print these two attributes are correlated (positive / negative) or not. Calculate the covariance.

Data:

age	23	23	27	27	39	41	47	49	50
%fat	9.5	26.5	7.8	17.8	31.4	25.9	27.4	27.2	31.2
age	52	54	54	56	57	58	58	60	61
%fat	34.6	42.5	28.8	33.4	30.2	34.1	32.9	41.2	25.7

Draw the boxplots for age and %fat. Draw the scatter plot on these two variables.

Code and output:

```
import pandas as pd
import numpy as np
age = [23, 23, 27, 27, 39, 41, 47, 49, 50, 52, 54, 54, 56, 57, 58, 58, 60, 61]
fat = [9.5, 26.5, 7.8, 17.8, 31.4, 25.9, 27.4, 27.2, 31.2, 34.6, 42.5, 28.8, 33.4, 3]
0.2,34.1,32.9,41.2,35.7]
data = {'age':age,'fat':fat}
df = pd.DataFrame(data)
n = len(df)
def corcoef(data, N):
  a = int(input("Select your first column number u want to correlate: "
  b = int(input("Select your second column number u want to correlate:"
  a,b = a-1,b-1
  x = data.iloc[:,a].sum()
  y = data.iloc[:,b].sum()
  x2 = (data.iloc[:,a]**2).sum()
  y2 = (data.iloc[:,b]**2).sum()
  std x = df.iloc[:,a].std()
  r = (N*xy - (x*y)) / np.sqrt((N*x2-(x**2))*2(N*y2-(y**2)))
  print("The correlation coefficient is: ",r)
  print("The covariance is: ",cov)
  if r>0:
     print("The 2 variables are negatively correlated")
corcoef(df,n)
```

```
import pandas as pd
import numpy as np
age = [23,23,27,27,39,41,47,49,50,52,54,54,56,57,58,58,60,61]
fat = [9.5,26.5,7.8,17.8,31.4,25.9,27.4,27.2,31.2,34.6,42.5,28.8,33.4,30.2,34.1,32.9,41.2,35.7]
data = {'age':age,'fat':fat}
df = pd.DataFrame(data)
n = len(df)
def corcoef(data, N):
 a = int(input("Select your first column number u want to correlate: "))
 b = int(input("Select your second column number u want to correlate:"))
 a,b = a-1,b-1
  x = data.iloc[:,a].sum()
 y = data.iloc[:,b].sum()
  xy = (data.iloc[:,a]* data.iloc[:,b]).sum()
  x2 = (data.iloc[:,a]**2).sum()
  y2 = (data.iloc[:,b]**2).sum()
  std_x = df.iloc[:,a].std()
  std_y = df.iloc[:,b].std()
  r = (N*xy - (x*y)) / np.sqrt((N*x2-(x**2))*2(N*y2-(y**2)))
 cov = r*(std_x)*(std_y)
 print("The correlation coefficient is: ",r)
 print("The covariance is: ",cov)
  if r>0:
     print("The 2 variables are positively correlated")
     print("The 2 variables are negatively correlated")
corcoef(df,n)
```

Select your first column number u want to correlate: 1
Select your second column number u want to correlate:2
The correlation coefficient is: 0.8176187964565881
The covariance is: 100.01960784313736
The 2 variables are positively correlated

```
import seaborn as sns
sns.boxplot('age',data=df)
```





