

Statistics Assignment 4
Submitted by: Amit Kumar
21/10/2020

In [2]: *# Problem Statement 1:
Is gender independent of education level? A random sample of 395 people were surveyed and each person was asked
to report the highest education level they obtained. The data that resulted from the survey is summarized in the
following table:*

#	High School	-	Bachelors	-	Masters	-	Ph.d.	-	Total
# Female	60	-	54	-	46	-	41	-	201
# Male	40	-	44	-	53	-	57	-	194
# Total	100	-	98	-	99	-	98	-	395

*# Question:
Are gender and education level dependent at 5% level of significance? In other words, given the data collected
above, is there a relationship between the gender of an individual and the level of education that they have obtained?*

In [1]: *# Chi-Square test of independence
H0 : Null Hypothesis: The two categorical variables are independent.
H1: Alternative Hypothesis: The two categorical variables are dependent.*

```
import numpy as np
import pandas as pd
import scipy.stats as stats

male = [40,44,53,57]
female = [60,54,46,41]
High_school=[60,40]
Bachelors = [54,44]
Masters = [46,53]
Phd = [41,57]

marks = male+female
print(marks)
sex=['M','M','M','M','F','F','F','F']
education =['High_school','Bachelors','Masters','Ph.d','High_school','Bachelors','Masters','Ph.d']
DF=pd.DataFrame({"Education":education,"Marks":marks,"Sex":sex})
DF
print(DF)
```

```
[40, 44, 53, 57, 60, 54, 46, 41]
```

	Education	Marks	Sex
0	High_school	40	M
1	Bachelors	44	M
2	Masters	53	M
3	Ph.d	57	M
4	High_school	60	F
5	Bachelors	54	F
6	Masters	46	F
7	Ph.d	41	F

```
In [2]: cross_tab = pd.crosstab([DF.Sex,DF.Marks],DF.Education,margins=True)
cross_tab
```

Out[2]:

		Education	Bachelors	High_school	Masters	Ph.d	All
Sex	Marks						
F	41		0	0	0	1	1
	46		0	0	1	0	1
	54		1	0	0	0	1
	60		0	1	0	0	1
M	40		0	1	0	0	1
	44		1	0	0	0	1
	53		0	0	1	0	1
	57		0	0	0	1	1
All			2	2	2	2	8

```
In [3]: DF1 = pd.crosstab(DF.Sex, DF.Education,DF.Marks, aggfunc="sum",margins=True)
DF1
```

Out[3]:

		Education	Bachelors	High_school	Masters	Ph.d	All
Sex							
F			54	60	46	41	201
M			44	40	53	57	194
All			98	100	99	98	395

```
In [4]: DF1.columns = ["Bachelors","High_School","Masters","Ph.d.","Genderwise_total"]
DF1.index = ["Female","Male","Combined"]
DF1
```

Out[4]:

	Bachelors	High_School	Masters	Ph.d.	Genderwise_total
Female	54	60	46	41	201
Male	44	40	53	57	194
Combined	98	100	99	98	395

```
In [5]: # Creating a table exlcuding the total for later use
DF2 = DF1.iloc[0:2,0:4]
DF2
```

Out[5]:

	Bachelors	High_School	Masters	Ph.d.
Female	54	60	46	41
Male	44	40	53	57

In [6]: *# For a test of independence, we use the same chi-squared formula that we used for the goodness-of-fit test.*
The main difference is we have to calculate the expected counts of each cell in a 2-dimensional table instead of
a 1-dimensional table. To get the expected count for a cell, multiply the row total for that cell by the column
total for that cell and then divide by the total number of observations. We can quickly get the expected counts
for all cells in the table by taking the row totals and column totals of the table, performing an outer product
on them with the np.outer() function and dividing by the number of observations:

```
DF3=np.outer(DF1["Genderwise_total"][0:2],DF1.loc["Combined"][0:4]) / 395.0
DF3 = pd.DataFrame(DF3)
DF3.columns = ["Bachelors", "High_School", "Masters", "Ph.d."]
DF3.index = ["Female", "Male"]
DF3
```

Out[6]:

	Bachelors	High_School	Masters	Ph.d.
Female	49.868354	50.886076	50.377215	49.868354
Male	48.131646	49.113924	48.622785	48.131646

In [7]: *# Now we will calculate the chisquare statistic, critical value and p value.*
We called the .sum() twice, once to get the column sum and second time to add the sum, returning the sum of entire
2D table

```
chi_squared_stat = (((DF3-DF2)**2)/DF3).sum().sum()
print(chi_squared_stat)
```

8.006066246262538

In [9]: *#Find the critical value for 95% confidence and degree of freedom (df) is 3*
cvalue = stats.chi2.ppf(q = 0.95,df= 3)
print("Critical value")
print(cvalue)

Critical value
7.814727903251179

In [10]: *# Find the p-value*
p_value = 1 - stats.chi2.cdf(x=chi_squared_stat,df=3)
print("P value")
print(p_value)

P value
0.04588650089174717

In [11]: *# Use stats.chi2_contingency() function to conduct a test of independence automatically given a frequency table*
of observed counts:
 result = stats.chi2_contingency(observed= DF2)
 print(result)
 print('-'*115)
 print('The output shows the chi-square statistic = 8, the p-value as 0.045 and the degrees of freedom as 3')
 print('The critical value with 3 degree of freedom is 7.815. Since 8.006 > 7.815, therefore we reject the null hypothesis and conclude that the education level depends on gender at a 5% level of significance.')

```
(8.006066246262538, 0.045886500891747214, 3, array([[49.86835443, 50.88607595, 50.37721519, 49.86835443],
           [48.13164557, 49.11392405, 48.62278481, 48.13164557]]))
```


 The output shows the chi-square statistic = 8, the p-value as 0.045 and the degrees of freedom as 3
 The critical value with 3 degree of freedom is 7.815. Since 8.006 > 7.815, therefore we reject the null hypothesis and conclude that the education level depends on gender at a 5% level of significance.

In [12]: *# Problem Statement 2:*
Using the following data, perform a oneway analysis of variance using $\alpha=.05$. Write up the results in APA format.
[Group1: 51, 45, 33, 45, 67] [Group2: 23, 43, 23, 43, 45] [Group3: 56, 76, 74, 87, 56]

```
In [16]: #The analysis of variance or ANOVA is a statistical inference test that lets you compare multiple groups at the same
# time. The one-way ANOVA tests whether the mean of some numeric variable differs across the levels of one categorical
# variable. It essentially answers the question: do any of the group means differ from one another?

#The scipy library has a function for carrying out one-way ANOVA tests called
scipy.stats.f_oneway()
import scipy.stats as stats
Group1 = [51, 45, 33, 45, 67]
Group2 = [23, 43, 23, 43, 45]
Group3 = [56, 76, 74, 87, 56]
# Perform the ANOVA
statistic, pvalue = stats.f_oneway(Group1, Group2, Group3)
print("F Statistic value {}, p-value {}".format(statistic, pvalue))
if pvalue < 0.05:
    print('True')
else:
    print('False')
print("-"*115)
print("The test result suggests the groups have different same sample means in
this example, since the p-value is significant at a 99% confidence level. Here
the p-value returned is 0.00305 which is < 0.05")
```

```
F Statistic value 9.747205503009463 , p-value 0.0030597541434430556
True
```

```
-----
The test result suggests the groups have different same sample means in this
example, since the p-value is significant at a 99% confidence level. Here the p-
value returned is 0.00305 which is < 0.05
```

```
In [ ]: # Problem Statement 3:
# Calculate F Test for given 10, 20, 30, 40, 50 and 5, 10, 15, 20, 25. For 10, 20, 30, 40, 50:
```

```
In [14]: stats.f_oneway([10, 20, 30, 40, 50],[5,10,15, 20, 25])

Group1 = [10, 20, 30, 40, 50]
Group2 = [5,10,15, 20, 25]
mean_1 = np.mean(Group1)
mean_2 = np.mean(Group2)
grp1_sub_mean1 = []
grp2_sub_mean2 = []
add1 = 0
add2 = 0
for items in Group1:
    add1 += (items - mean_1)**2
for items in Group2:
    add2 += (items - mean_2)**2
var1 = add1/(len(Group1)-1)
var2 = add2/(len(Group2)-1)

F_Test = var1/var2
print("F Test for given 10, 20, 30, 40, 50 and 5,10,15, 20, 25 is : ",F_Test)
```

F Test for given 10, 20, 30, 40, 50 and 5,10,15, 20, 25 is : 4.0