Final Project – Milestone 2 Hypothesis Testing

Introduction

Dataset:

The dataset chosen for the Final Project – Milestone 2 was the same dataset used in the Final Project – Milestone 1 which was picked from Kaggle. The dataset is related to the number of deaths by risk factor, and it has various risk factors mentioned which would help in determining the cause of death rate in a particular entity which is nothing but in a particular country. The dataset contains 6468 rows of data and 32 data fields, and the data fields are entity, code, year, alcohol use, drug use, air pollution, no access to handwashing facility, smoking, outdoor air pollution, iron deficiency, etc. This dataset contains numeric as well as categorical data and is helpful in determining which risk factor in a particular country would affect the death rate.

In this assignment, the task is to answer some of the questions related to the data which will be done through inferential statistics and hypothesis testing. From the dataset the important parameters to work with and which will be helpful in answering the questions are the entity parameter which is nothing but the countries having various risk factors affecting the death rate, the year in which the risk factor affected the death rate, and all the various risk factors like the alcohol use, drug use, smoking, iron deficiency, diet low in fruits, diet low in vegetables, and many other factors.

Questions about the Data:

The dataset of number of deaths by risk factors gives us a lot of information related to the death rate taking place due to the risk factors in a particular country in the year. The questions that can be answered with respect to the dataset could be as follows.

- a. Whether the overall mean of a particular risk factor is equal to assumed value or not which can be answered and performed through one-sample t-test.
- b. Whether the mean of a particular risk factor in the entity i.e., the specific country is equal to assumed value or not which again can be solved through the one-sample t-test.
- c. Which risk factor is higher in all the entities affecting the death rate more and that can be answered through the two-sample t-test?
- d. Lastly, which country is affected more due to a particular risk factor as compared to another country? The two-sample t-test will again be helpful here in answering this question.

The data analysis and hypothesis testing of this dataset will give us more insights and help us in determining the various questions.

Data Analysis

After installing and importing all the libraries needed, the number of deaths by risk factor dataset was loaded into a variable and read. In order to know more about the dataset for further analysis, descriptive analysis was done which would give us information about the entire dataset, its attributes, and the various other parameters of the dataset. Descriptive analysis is one of the processes in the exploratory data analysis which helps in describing the data and computing the statistical values of the numeric variables of the dataset. Here, after reading the csv file, the data was described which includes the column names of the dataset that would help in knowing about the variables which would help us in analysis and testing. The starting and end records were displayed along with the summary of the entire dataset. The summary of the dataset helps in determining the statistical values of the variables which are numeric values.

Output:

1. Reading the csv file

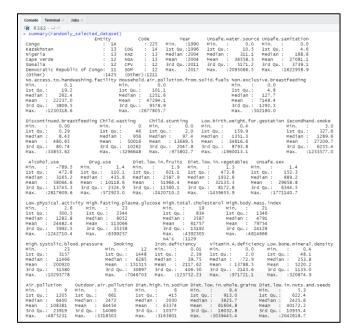
```
Jobs
  R 3.6.3 · ~/
                                                          <- read.csv("number-of-deaths-by-risk-factor.csv")
                                                                      activate_dat

Entity (
1 Afghanistan
2 Afghanistan
3 Afghanistan
5 Afghanistan
6 Afghanistan
7 Afghanistan
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12 Afghanistan
13 Afghanistan
14 Afghanistan
15 Afghanistan
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18 Afghanistan
                                                                                                                                                                                                                                                      8722.94308
8621.88492
8502.72999
7910.36627
                                                                                                                                                                                                                                                      7501 54813
                                                                                                                                                                                                                                                      7055 32209
                                                                                              11510 49271
                                                                                                                                                      8853 83887
                                                                                                                                                                                                                                                      8214 45999
                                                                                              11510.492/1
11656.61466
10973.73633
10040.77296
                                                                                                                                                      8950.64281
                                                                                                                                                                                                                                                      8331.27082
                                                                                                                                                      8410.94881
                                                                                                                                                      7673.53095
                                                                                                  8884.66772
7703.60948
6842.30998
6436.21122
                                                                                                                                                                                                                                                      6631.32202
5854.38195
5279.97858
                                                                                                                                                      6776.14823
                                                                                                                                                      5857.35361
                                                                                                                                                      4843.77111
                                                                                                                                                                                                                                                      4992.48476
                                                                                                  6023.62912
5613.05315
                                                                                                                                                      4507.04169
                                                                                                                                                                                                                                                      4725.82154
                                                                                                                                                                                                                                                      4449.08079
4474.55745
                                                                                                                                                      4166.68608
                                                                                                  5533.13574
5300.42444
5158.40086
5256.64928
                                                                                                                                                      4048.98644
                                                                                                                                                                                                                                                      4391.70827
4231.83572
                                              AFG 2015
                                                                                                                                                      3843.62273
  27 Afghanistan
28 Afghanistan
                                              AFG 2016
                                                                                                                                                      3723.50672
                                                                                                                                                                                                                                                      4126.33647
                                               AFG
                                                                                                        50.12577
51.04275
48.82450
                      Albania
         Albania ALB 1992 48.82450 30.76490 43.19028
Household.air.pollution.from.solid.fuels Non.exclusive.breastfeeding Discontinued.breastfeeding
22388.497 3221.1388 156.097553
22128.758 3150.5596 151.539851
22873.769 3331.3490 156.609194
25599.756 4477.0061 206.834451
                                                                                                                                                                                 5102.6221
5402.6604
5263.6445
5271.7718
5165.9238
                                                                                                                                                                                                                                                           233.930571
262.793340
                                                                                                  29062.619
                                                                                                                                                                                                                                                            258.134146
                                                                                                                                                                                                                                                            254.708078
251.898824
                                                                                                                                                                                  5044.3076
4566.9186
                                                                                                                                                                                                                                                                                                       36124.04
                                                                                                 28329.380
27721.641
26679.508
28325.887
28511.343
                                                                                                                                                                                                                                                            235.101472
                                                                                                                                                                                                                                                             220.52932
                                                                                                                                                                                                                                                           207.900393
254.024088
262.000727
245.140693
216.945204
```

2. Descriptive Analysis

Now, after the describing about the entire dataset, a subset of the dataset was created which had a randomly selected sample of dataset. This is done because the dataset has 6000+ entries and so rather than testing on the entire dataset, a sample of the dataset was created on which the testing can be performed. The sample_n() function was used which randomly selects some amount of data from the original dataset. Here, from the 6000+ entries, 1500 sample of data was extracted from the dataset. Along with that, the summary of this randomly selected dataset was also displayed as this was the dataset which would be used further.

Output:

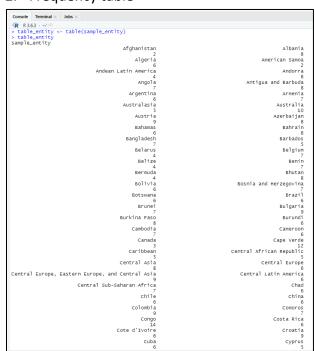


For the testing and analysis, the variables to be tested needed to be extracted for which new variables were created. Also, since the analysis needed to be done for a particular country as well, a subset of dataset was created using the subset function which extracted all the values for the particular country selected. The frequency table for the entity parameter was also displayed which helped in knowing about the count of the particular entity.

Output:

1. Creating new variables

2. Frequency table



Subset of dataset

The descriptive analysis was again performed on these new variables and on the subset of the data to get a new idea about the statistical values further which would help us in hypothesis testing.

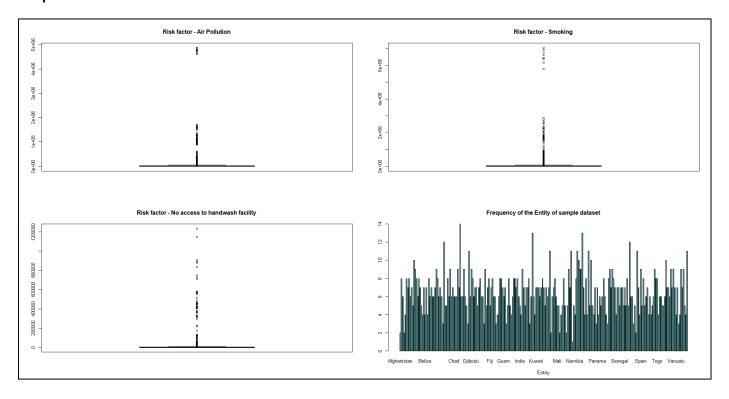
Output:

```
Console Terminal × Jobs ×
R 3.6.3 · ~/ €
> #descriptive analysis of the variables created
> mean(air_pollution)
[1] 108380.7
> summary(air_pollution)
Min. 1st Qu. Median
9 1205 6400
> mean(smoking)
[1] 151514 8
                            Median Mean 3rd Qu. Max.
6400 108381 23919 4875231
[1] 151514.8
Min. 1st Qu. Median Mean
12 1448 6285 151515
> mean(no_access_to_handwash)
[1] 22236.99
                                           Mean 3rd Qu.
                                                                     Max.
                                                    30897 7046703
> summary(no_access_to_handwash)
      Min. 1st Qu. Median Mean
0.1 19.2 262.4 22237.0
                                                       Mean
                                                                   3rd Ou.
                                                                     3809.5 1230318.6
> mean(india_entity$Alcohol.use)
[1] 437495.6
 summary(india_entity$Alcohol.use)
 Min. 1st Qu. Median Mean 3rd Qu.
292070 379993 421276 437496 517488
                                                                     мах.
> mean(australia_entity$Drug.use)
[1] 1945.317
  summary(australia_entity$Drug.use)
    Min. 1st Qu. Median Mean 3rd Qu.
1251 1737 1874 1945 2175
mean(eastasia_entity$Iron.deficiency)
                                                                     Max.
                                                                     2870
[1] 2290.679
 > summary(eastasia_entity$Iron.deficiency)
Min. 1st Qu. Median Mean 3rd Qu. Max.
389.9 632.3 1544.4 2290.7 3864.8 5856.6
```

Data Visualization

Before getting into further analysis and testing, data visualization was performed to get a better understanding of the data that was created earlier. The visualizations created were based on the variables of air pollution, smoking and no access to hand wash facility along with the frequency of the entity variable of the sample dataset. For the risk factors, boxplot was created to get a generalized idea about the data and bar plot was created for the frequency of the entity variable.

Output:



The above figure shows 4 graphs related to the dataset which explains the boxplot and bar plot for the variables of risk factors and the subset of entity data.

Hypothesis Testing

The data analysis and data visualizations gave us an idea about the dataset and now we further proceed to the hypothesis testing. Hypothesis testing is basically used to test an assumption regarding a population parameter by using a sample data. In the hypothesis testing there are two tests which are performed: one sample t-test and two sample t-test. One sample t-test is performed to test whether or not the mean of a population is equal to some value whereas two sample t-test is used to test whether or not the means of two populations are equal.

One sample t-test:

One sample t-test was performed on the new variables that were created earlier to test whether the mean of a population is equal to some assumed value or not. The risk factors that were considered for the one sample testing are, air_pollution, smoking, and no_access_to_handwash. The dataset as mentioned earlier was a sample dataset chosen at random.

The outputs of the one sample testing for the three risk factors are as below.

Output:

Air Pollution

```
Console Terminal × Jobs ×

    R 3.6.3 · ~/

> t.test(air_pollution, mu = 104500)
                                                                       #reject the alternative hypothesis
           One Sample t-test
data: air_pollution
t = 0.34139, df = 1499, p-value = 0.7329
alternative hypothesis: true mean is not equal to 104500
95 percent confidence interval:
86083.08 130678.33
sample estimates:
mean of x
108380.7
> t.test(air_pollution, mu = 2000)
                                                                    #reject the null hypothesis
           one Sample t-test
data: air_pollution
t = 9.3584, df = 1499, p-value < 2.2e-16
alternative hypothesis: true mean is not equal to 2000</pre>
95 percent confidence interval:
86083.08 130678.33
sample estimates:
mean of x
108380.7
> t.test(air_pollution, mu = 200000)
                                                                     #reject the null hypothesis
           one Sample t-test
data: air_pollution
t = -8.0599, df = 1499, p-value = 1.546e-15
alternative hypothesis: true mean is not equal to 2e+05
95 percent confidence interval:
  86083.08 130678.33
sample estimates:
mean of x
108380.7
```

2. Smoking

```
Console Terminal × Jobs ×
R 3.6.3 · ~/
> t.test(smoking, mu = 150000)
                                                #reject the alternative hypothesis
          One Sample t-test
data: smoking
t = 0.096181, df = 1499, p-value = 0.9234
alternative hypothesis: true mean is not equal to 150000
95 percent confidence interval:
120620.3 182409.3
sample estimates:
mean of x
151514.8
> t.test(smoking, mu = 100000)
                                               #reject the null hypothesis
          One Sample t-test
data: smoking t = 3.2708, df = 1499, p-value = 0.001097 alternative hypothesis: true mean is not equal to 1e+05
95 percent confidence interval:
120620.3 182409.3
sample estimates:
mean of x
151514.8
> t.test(smoking, mu = 2000000)
                                               #reject the null hypothesis
          One Sample t-test
data: smoking t = -117.36, df = 1499, p-value < 2.2e-16 alternative hypothesis: true mean is not equal to 2e+06
95 percent confidence interval:
 120620.3 182409.3
sample estimates:
mean of x
 151514.8
```

No access to handwash facility

```
Console Terminal × Jobs ×
R 3.6.3 · ~/ €
                                                                                    #reject null hypothesis
> t.test(no_access_to_handwash, mu = 25000)
            One Sample t-test
data: no_access_to_handwash
t = -1.0732, df = 1499, p-value = 0.2834
alternative hypothesis: true mean is not equal to 25000
95 percent confidence interval:
17186.89 27287.09
sample estimates:
mean of x
 22236.99
                                                                                #reject null hypothesis
> t.test(no access to handwash, mu = 13000)
           One Sample t-test
data: no_access_to_handwash
t = 3.5878, df = 1499, p-value = 0.0003442
alternative hypothesis: true mean is not equal to 13000
95 percent confidence interval:
17186.89 27287.09
sample estimates:
mean of x
 22236.99
> t.test(no_access_to_handwash, mu = 17000)
                                                                                 #reject alternative hypothesis
            One Sample t-test
data: no_access_to_handwash
t = 2.0341, df = 1499, p-value = 0.04211
alternative hypothesis: true mean is not equal to 17000
95 percent confidence interval:
17186.89 27287.09
sample estimates:
 mean of x
22236.99
```

From the outputs of the one-sample t-test, it is observed that for each of the risk factor, depending on the p-value we either reject the null hypothesis or fail to reject the null hypothesis which means that if we fail to reject the null hypothesis we conclude that the data is supporting the assumption whereas if we reject the null hypothesis we conclude that the two populations do not have the same value of parameter.

With respect to all the risk factors, depending on our mu value we determine the p-value which in turn gives us the hypothesis of rejecting the null hypothesis or not. If the p-value is greater than 0.05 which is the level of significance, we reject the alternative hypothesis and if the p-value is less than 0.05, we reject the null hypothesis. In this case, when the mu value was approx. closer to the mean value, the p-value was greater, and we rejected the alternative hypothesis and so the assumed value was same as the population mean. Also, we can conclude that changing the level of significance value from 0.05 to 0.1 does not affect the hypothesis and the conclusion remains the same. The values and analysis of each population tested can be clearly seen in the outputs displayed above.

Therefore, our first question is answered here which is to test whether the mean of population is equal to the assumed value or not.

Next analysis was tried with respect to one sample t-test with additional parameters. The additional parameter was the alternative which could be either less or greater. Basically, this parameter assumes that the actual mean value is either greater or lesser than the assumed mean value. The output obtained was as follows.

Output:

```
Console Terminal × Jobs ×
 R 3.6.3 · ~
 > #one sample testing for overall data set with additional parameters
> t.test(air_pollution, mu = 104520, alternative = "greater")
                                                                                                                                    #reject alternative hypothesis
              One Sample t-test
data: air_pollution
t = 0.33963, df = 1499, p-value = 0.3671
alternative hypothesis: true mean is greater than 104520
95 percent confidence interval:
sample estimates:
> t.test(smoking, mu = 200000, alternative = "less")
                                                                                                                                    #reject null hypothesis
              One Sample t-test
data: smoking
t = -3.0784, df = 1499, p-value = 0.001059
alternative hypothesis: true mean is less than 2e+05
95 percent confidence interval:
-Inf 177437.4
sample estimates:
mean of **
  nean of x
151514.8
> t.test(no_access_to_handwash, mu = 16000 , alternative = "less")
                                                                                                                                   #reject alternative hypothesis
             One Sample t-test
data: no_access_to_handwash
t = 2.4226, df = 1499, p-value = 0.9922
alternative hypothesis: true mean is less than 16000
95 percent confidence interval:
    _-Inf 26474.36
sample estimates:
mean of x
```

Further, the answer to the second question is solved by testing for the entity variable. Similar approach is followed here with a difference that the test is conducted for a particular entity only and not the overall sample data. Here, again, depending on the p-value we either rejected the null hypothesis or not and the level of significance here too does not change our conclusion. The output obtained for the same is as follows.

Output:

```
Console Terminal × Jobs ×
 R R3.6.3 · ·/ P>
> #one sample testing for subset of data set
> t.test(india_entity$Alcohol.use, mu = 430000)
             One Sample t-test
data: india_entity$Alcohol.use
t = 0.4534, df = 27, p-value = 0.6539
alternative hypothesis: true mean is r
95 percent confidence interval:
403574.5 471416.7
sample estimates:
mean of x
437495.6
                                                                        not equal to 430000
                                                                                                                                                                      #reject the null hypothesis
 > t.test(india_entity$Alcohol.use, mu = 200000, alternative = "greater")
              One Sample t-test
data: india_entity$Alcohol.use
t = 14.366, df = 27, p-value = 1.823e-14
alternative hypothesis: true mean is greater than 2e+05
95 percent confidence interval:
409336.6 Inf
sample estimates:
mean of x
437495.6
> t.test(australia_entity$Drug.use, mu = 2000)
                                                                                                                                                                     #reject the alternative hypothesis
              One Sample t-test
data: australia_entity$Drug.use
t = -0.66675, df = 27, p-value = 0.5106
alternative hypothesis: true mean is not equal to 2000
95 percent confidence interval:
1777.039 2113.596
> t.test(australia_entity$Drug.use, mu = 3000, alternative = "less")
                                                                                                                                                                     #reject the null hypothesis
data: australia_entity$Drug.use
t = -12.86, df = 27, p-value = 2.506e-13
alternative hypothesis: true mean is less than 3000
95 percent confidence interval:
-Inf 2085.01
sample estimates:
mean of x
1945.317
```

Two sample t-test:

Two sample t-test helped in answering the next two questions which was performed on the risk factors of the dataset and the subset of the entity parameter. Two sample t-test is performed when there are independent parameters, and we test whether the two population means are same or different. For the two-sample t-test, new variables and subset of the sample were created same as it was done for one sample testing for which the testing has to be done along with their descriptive analysis.

Output:

Descriptive analysis on the new variables.

```
Console Terminal × Jobs ×

R R3.6.3 · · / ∞

    *#descriptive analysis of the variables created
    > mean(alcohol_use)
[1] 58064.65

    *summary(alcohol_use)
    Min. 1st Qu. Median Mean 3rd Qu. Max.
    -789.5 472.8 3165.2 58064.6 13745.3 2817609.6

    * mean(drug_use)
[1] 10118.6

    *summary(drug_use)
    Min. 1st Qu. Median Mean 3rd Qu. Max.
    1.4 110.1 435.8 10118.6 2324.9 572923.0

    *mean(england_entity$Diet.low.in.fruits)
[1] 18320.67

    *summary(england_entity$Diet.low.in.fruits)
    Min. 1st Qu. Median Mean 3rd Qu. Max.
    14321 14861 18015 18321 21179 24282

    *mean(italy_entity$Diet.low.in.fruits)
[1] 12766.82

    *summary(italy_entity$Diet.low.in.fruits)
    Min. 1st Qu. Median Mean 3rd Qu. Max.
    11306 11616 12283 12767 13783 15572

    *|
```

The two-sample t-test done for the new variables created and the subset of dataset for which we obtain the answers to our questions is as below.

Output:

1. t-test on risk factors

```
Console Terminal × Jobs ×

R R 3.6.3 · ~/ 
> #two sample t-test
> 
> #t-test on risk factors
> t.test(alcohol_use,drug_use) #reject the null hypothesis

Welch Two Sample t-test

data: alcohol_use and drug_use
t = 7.9648, df = 1599.6, p-value = 3.113e-15
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
36138.68 59753.42
sample estimates:
mean of x mean of y
58064.65 10118.60
```

2. t-test on subset of dataset

```
Console Terminal × Jobs ×

R 3.63 · 
> #t-test on subset of data set
> t.test(england_entity$Diet.low.in.fruits, italy_entity$Diet.low.in.fruits) #reject the null hypothesis

Welch Two Sample t-test

data: england_entity$Diet.low.in.fruits and italy_entity$Diet.low.in.fruits
t = 7.9705, df = 35.046, p-value = 2.216e-09
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
4139.333 6968.371
sample estimates:
mean of x mean of y
18320.67 12766.82
```

From the first output, which was based on comparing the two risk factors for the overall entities, it was observed that since the p-value is less than 0.05, we reject the null hypothesis which states that the alcohol use in different entities is high than that of the drug use and that the risk factor affecting the death rate is more due to the alcohol use as compared to the drug use in different countries. The level of significance considered here was 0.05 but changing the value to 0.1 also does not change our conclusion.

The second output based on comparing a risk factor for the two entities tells us that since the p-value is less than 0.05, we reject the null hypothesis which implies that diet low in fruits, a risk factor leading to the death rate is more in the entity England as compared to the entity Italy. And so, England is affected more due to this risk factor as compared to Italy leading to the death rate factor.

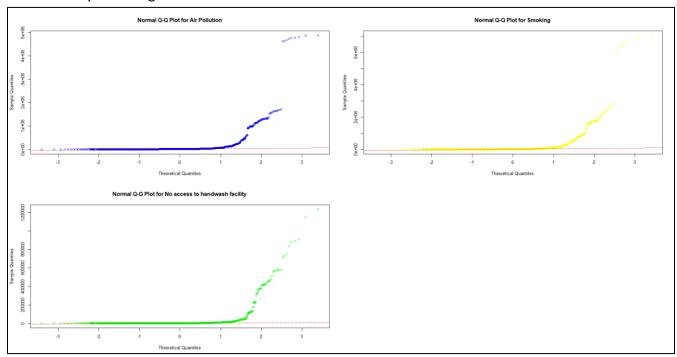
Also, it can be noted that a paired t-test cannot be performed here in this case since we reject the null hypothesis in the two-sample t-test.

Data Visualizations for Testing:

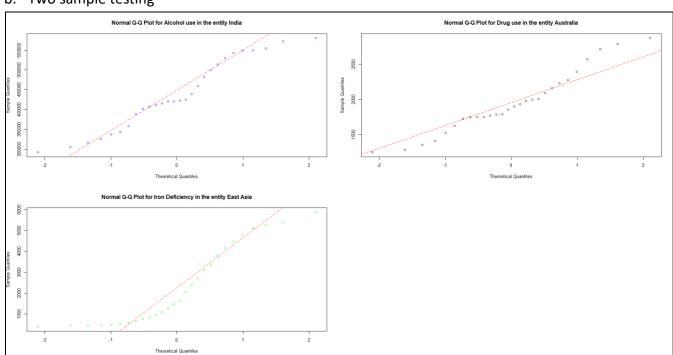
After the one-sample and two-sample t-test were conducted, visualizations were created in order to get a clear understanding about the parameters and the subset of the dataset. The visualization created was the normal Q-Q plot which is nothing but a probability plot for a graphical method to compare two probability distributions. The normal q-q plot plotted for the three variables and for the subset of the dataset is as shown below.

Output:

a. One sample testing



b. Two sample testing



Summary

Hypothesis testing is a technique to help determine whether a specific treatment has an effect on the individuals in a population. The task here was to answer some of the questions statistics and testing which would help in the analysis of the data. Descriptive analysis was performed first to get an overall idea about the dataset after which the hypothesis testing on particular attributes was performed. The summary function during the descriptive analysis helps in understanding the overall statistical values of the dataset and it returns the statistical values like the min, max, mean, median and the range. Data visualizations also performed gave insights about the dataset which further helped in analysis.

The t-test is a statistical test which is used to compare the means of two groups used in hypothesis testing to determine whether a process actually has an effect on the population of interest or whether two groups are different from one another. Therefore, by using the t-test we try to find the answers to our questions for the number of deaths by risk factors dataset. One sample and two sample tests were performed based on the requirements of the dataset and these concepts helped in understanding the analysis better. For the one sample test, the mean values were kept varying to check and understand the hypothesis. It was observed that for each of the attribute tested, the mu value was assumed and depending on the p-value we rejected the null hypothesis or not. For two-sample t-test, we would check whether the means of two populations is equal or not. Here, the two attributes were compared to check which risk factor is higher as compared to the other for the overall dataset. One another analysis that can be done with respect to the two-sample t-test is which country is most affected by a particular risk factor as compared to the other. For this analysis, England was affected more due to the risk factor of diet low in fruits as compared to in Italy.

Thus, hypothesis testing using the t-test helped in answering the questions related to the dataset and for further analysis. It gave an overall idea from both the one sample t-test and two sample t-test. The testing and analysis on the number of death rate by risk factors helps in understanding the risk factors as an individual parameter and also with respect to a particular entity whether which entity is most affected by a risk factor as compared to the other.

References

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