

Exploratory Data Analysis – D207  
Western Governor's University  
Performance Assessment  
Matthew Morgan  
Student ID: 010471280  
9/27/2022

A1. For this assessment I am using the medical data set. The purpose of this assessment is to answer a question regarding patient readmissions. I chose to answer the following question, "Is there a relationship between the initial\_days column and the ReAdmis column?"

A2. This data set points out that an external organization penalizes hospitals for excessive readmissions. This analysis of the data is the first step towards reducing patient readmissions for the hospital. The question I am answering can help identify a specific variable which leads to patients having a higher chance for readmission than others. This can then be used in the future to start enacting change to reduce readmissions. Ultimately, if readmissions can be reduced that will lead to less penalty payouts as well as better overall outcomes for patients.

A3. Because my question is very specifically about the relationship between the initial\_days and ReAdmis columns, I will only be using those two columns for my analysis.

B1. To analyze my question, I ran a two-sample t-test. The code (Bowne-Anderson et al., n.d.) is as follows:

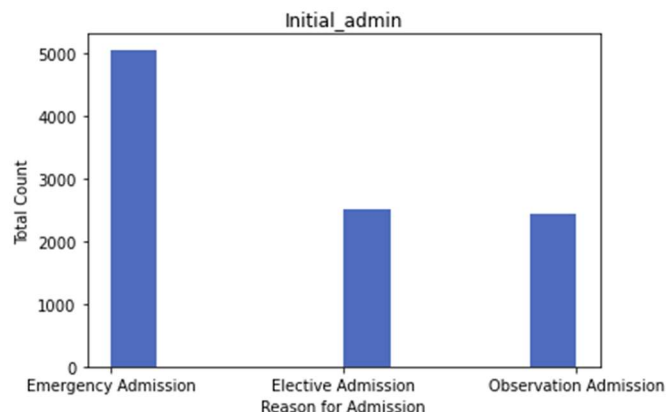
```
stats.ttest_ind(df['ReAdmis_numeric'], df['Initial_days'])
```

B2. The output was:

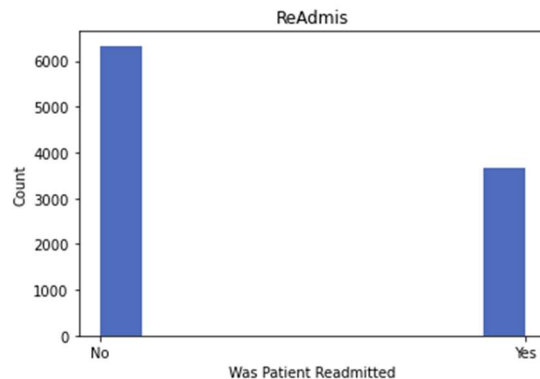
```
Ttest_indResult(statistic=-129.54592813419822, pvalue=0.0)
```

B3. I chose a t-test as I have one numeric and one categorical variable to compare. The initial\_days column being numeric data and the ReAdmis being categorical.

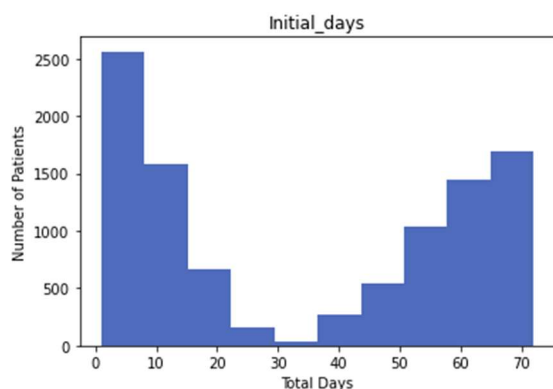
C1. Below are the distributions for four required variables used in the analysis for this report. They include two categorical (Initial\_admin, and ReAdmis) variables as well as two continuous (Initial\_days, Children) variables.



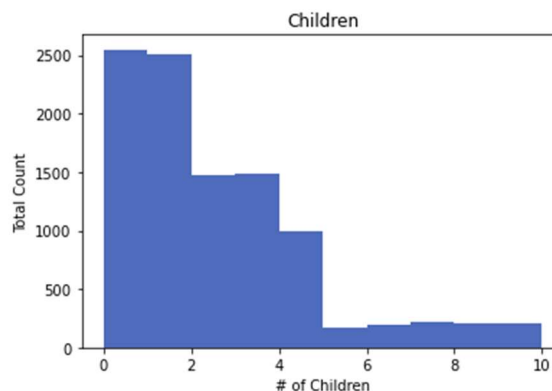
Initial\_admin\_numeric Stats  
 min: 0  
 25th Quantile: 0.0  
 50th Quantile: 0.0  
 75th Quantile: 1.0  
 max: 2  
 mean: 0.7376  
 median: 0.0  
 mode: 0  
 Std: 0.8251147322840162  
 skew: 0.5191601076816872  
 kurtosis: -1.3392723170631167



ReAdmis\_numeric Stats  
 min: 0  
 25th Quantile: 0.0  
 50th Quantile: 0.0  
 75th Quantile: 1.0  
 max: 1  
 mean: 0.3669  
 median: 0.0  
 mode: 0  
 Std: 0.48198300878982964  
 skew: 0.5524121095443897  
 kurtosis: -1.695179937226946



Initial Days Stats  
 min: 1.001980919  
 25th Quantile: 7.896214698  
 50th Quantile: 35.83624435  
 75th Quantile: 61.16102  
 max: 71.98149  
 mean: 34.45529926595239  
 median: 35.83624435  
 mode: 63.54432  
 Std: 26.30934131161786  
 skew: 0.07028608266045329  
 kurtosis: -1.7545246170896873

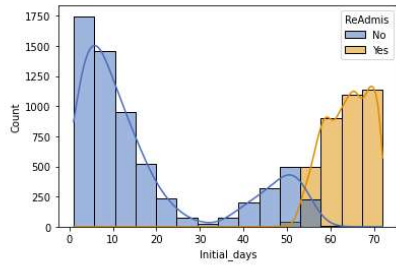


Children Stats  
 min: 0  
 25th Quantile: 0.0  
 50th Quantile: 1.0  
 75th Quantile: 3.0  
 max: 10  
 mean: 2.0972  
 median: 1.0  
 mode: 0  
 Std: 2.16365900779899  
 skew: 1.4480126219332756  
 kurtosis: 2.076321273332364

D1. Included below is the python code (Bowne-Anderson et al., n.d.) used to produce 3 bivariate distributions comparing Initial\_days, Children, and Initial\_admin to ReAdmis.

```
In [28]: sns.histplot(data=df, x="Initial_days", hue="ReAdmis", kde=True)
```

```
Out[28]: <AxesSubplot:xlabel='Initial_days', ylabel='Count'>
```



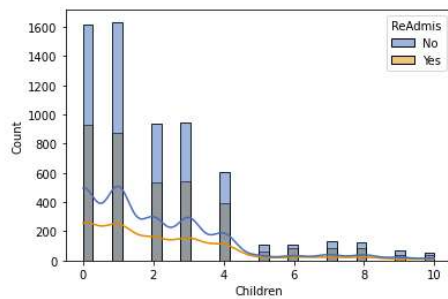
```
In [37]: cor = df['Initial_days'].corr(df['ReAdmis_numeric'])
```

```
print(cor)
```

```
0.8508616016470936
```

```
In [29]: sns.histplot(data=df, x="Children", hue="ReAdmis", kde=True)
```

```
Out[29]: <AxesSubplot:xlabel='Children', ylabel='Count'>
```

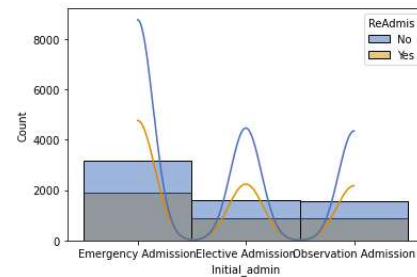


```
In [38]: cor = df['Children'].corr(df['ReAdmis_numeric'])
print(cor)
```

```
0.0235315217234477
```

```
In [30]: sns.histplot(data=df, x="Initial_admin", hue="ReAdmis", kde=True)
```

```
Out[30]: <AxesSubplot:xlabel='Initial_admin', ylabel='Count'>
```



```
In [39]: cor = df['Initial_admin_numeric'].corr(df['ReAdmis_numeric'])
```

```
print(cor)
```

```
-0.018170281715283412
```

E1. Based on the results of the t-test, the p-value between Initial\_days and ReAdmis is 0. This shows that we can reject the null hypothesis. Based on the question being asked, "Is there a relationship between the Initial\_days column and the ReAdmis column?" the alternative and null hypothesis would be as follows:

Ha: There is a relationship between the Initial\_days and the ReAdmis column.

Ho: There is no relationship between Initial\_days and the ReAdmis column.

Again, because the p-value of our t-test with Initial\_days vs ReAdmis is  $< 0.05$  we have sufficient evidence to reject our null hypothesis.

E2. I only tested for a correlation between readmissions and one other variable. In the exploratory data analysis phase I imagine we'd want to try and find all the variables that have a relationship with readmissions before proceeding.

E3. Because Initial\_days has a strong relationship with patient readmissions the next course of action would be to find other variables that are related to readmissions as well. One possible avenue is to also look at variables or trends for patients who have long initial hospital stays. If we can modify something about patients who have long hospital stays then maybe we can lower patient readmissions.

F. <https://wgu.hosted.panopto.com/Panopto/Pages/Viewer.aspx?id=e039cfe9-f5d6-428f-a3c8-af1e01650db5>

G. Bowne-Anderson, H., Scouwenaars, F., Matsui, Maggie., Ramirez M., Alexander A., (n.d.) *D207 – Exploratory Data Analysis* [OC]. Datacamp.  
<https://app.datacamp.com/learn/custom-tracks/custom-d207-exploratory-data-analysis>

H. No sources quoted, paraphrased or summarized.