

What Drives Nurse Turnover?

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Project Introduction

Nurse turnover and shortage has been a challenge in healthcare industry particularly in hospital care. Although researchers have been studying the intensity of nurse turnover through Likert scale questions in order to better understand the factors that potentially lead to the issue, the nurse turnover rate still remains high nowadays. Our research aims to not only study the reasons that cause nurse turnover but also the reasons for nursing shortage. Thus, questions such as “what drives nurse turnover and what drives nurse shortage” await to be answered.

Detail

Data

This is a survey data that was conducted by Regis professor and student from nursing and data science department.

- Demographic data
- Basic and current job data (job changes and nurse traveler)

Analysis Methods

- Chi-Square: using statistical measurements to find out relationship among categorical variables and continuous variables.
- EDA: Creating graphs to slice and dice demographic and job data from different perspectives.
- Logistic Regression: Predicting whether a nurse will change jobs in the next six months based on the predictors.

Analysis Process

I. Data Cleaning

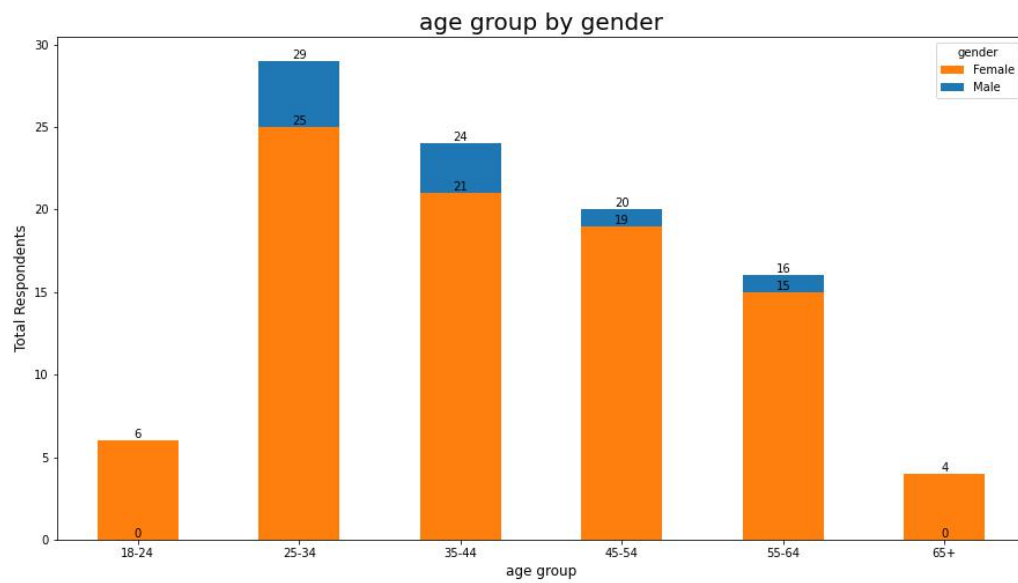
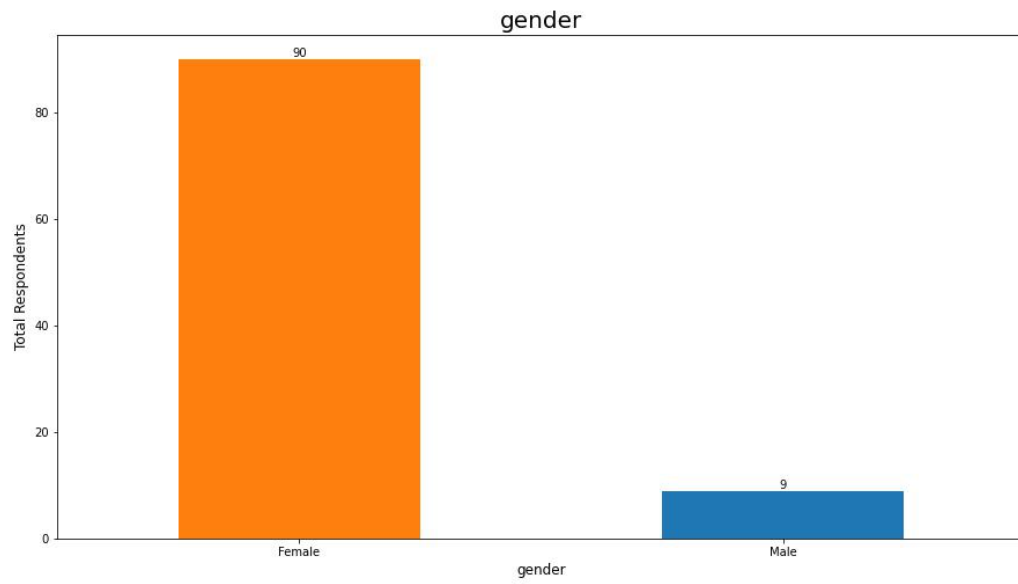
Since this project is the extended study of the last research, data structure is already reformatted to a regular table with column, row, and header. Variables are selected for studying what drives nurse turnover, which only include demographics and job history data. All null answers in plan to change job question as well as the retired respondents were dropped. Nulls in other fields include "is_nurse_traveler", "cur_annual_range", "cur_total_hour_work" were also removed. Finally, observations that have data quality issues were also dropped. Quality issues in open ended questions were fixed where groups were assigned to each observation. After data cleaning, there were 99 total observations remain in the dataset.

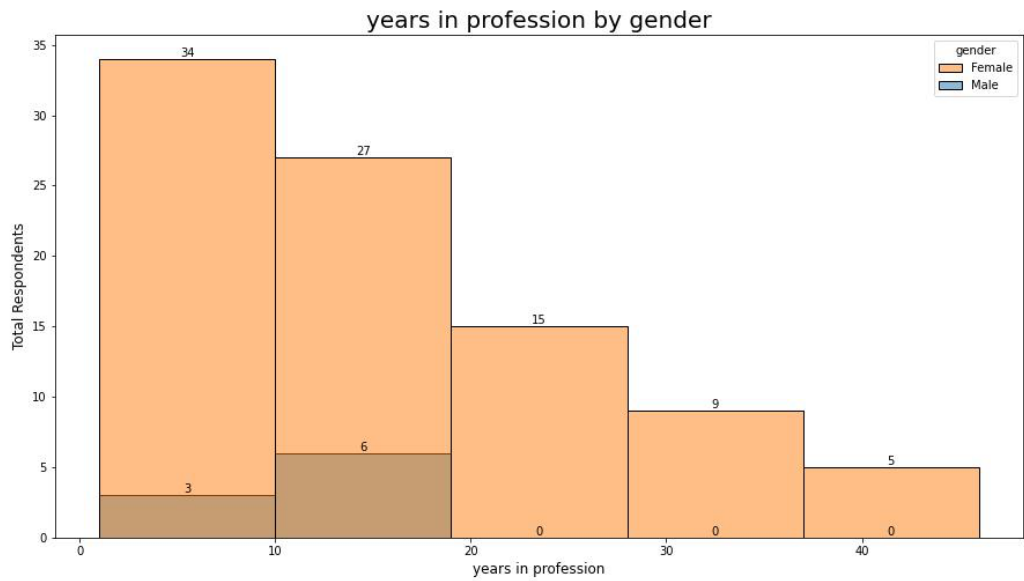
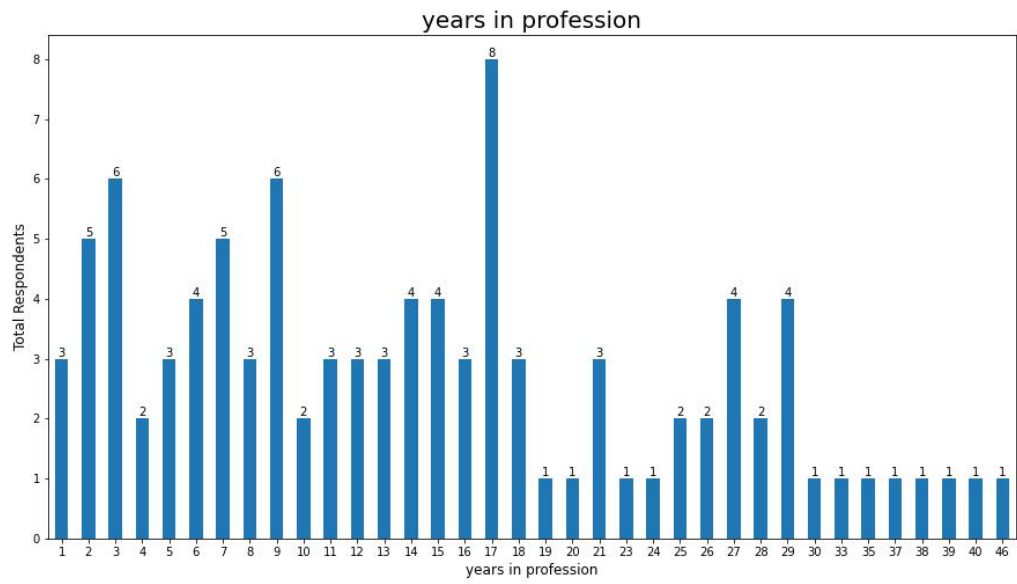
II. EDA

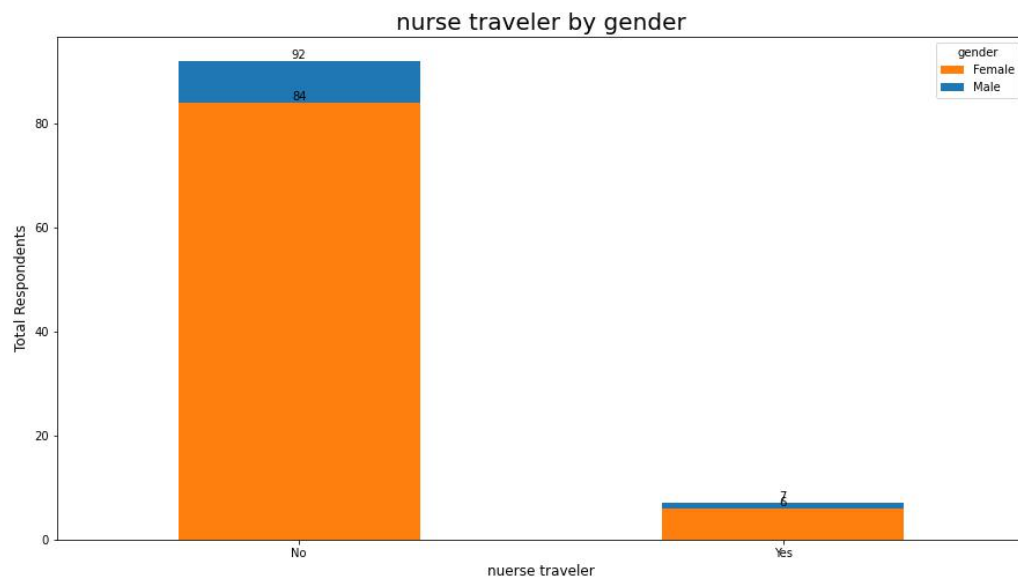
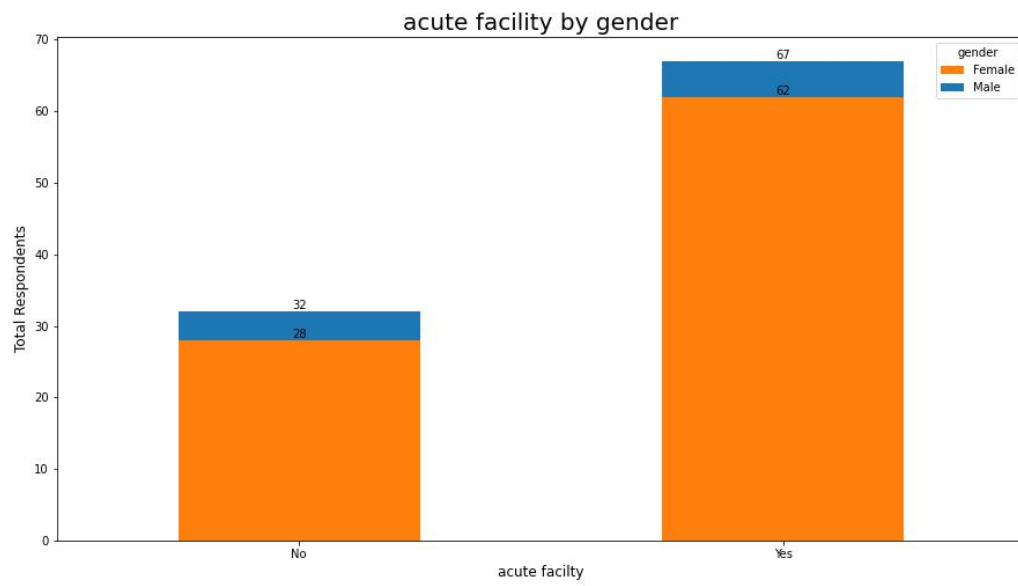
EDA focuses on learning the data, which includes demographic and job changes information. Interesting findings are 1) only around 10% of nurses are planning to change jobs in the next six

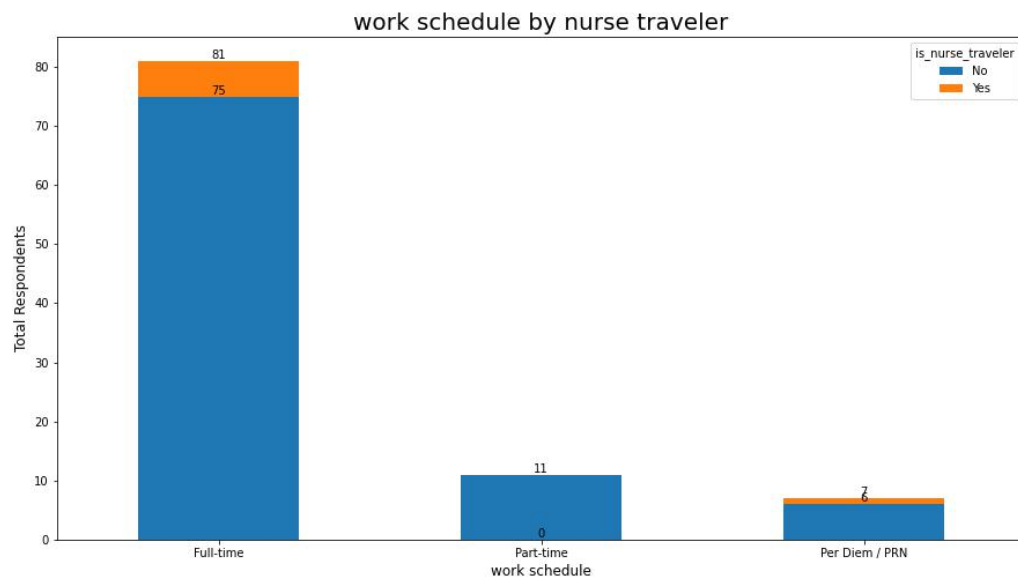
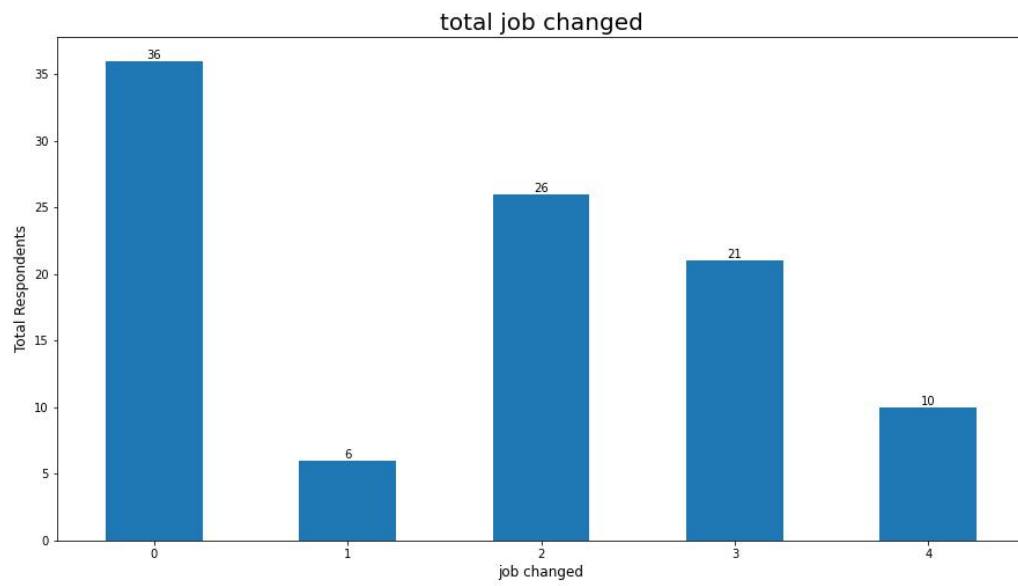
months, whereas approximately 65% nurses said no and 25% of them answered “maybe”. 2) for those who work in acute care hospitals and said yes to change jobs, 57% of them are from critical unit. 3) for the seven nurse travelers (7/99), three answered maybe, two for no and two for yes, which is 43%, 29%, and 29% respectively. 4) 80% of nurses who said yes (8/10) to change job have less than 10 years’ experience, which accounts for 22% of population in this group (<10 years in profession). 5) in the group of less than 20 years and greater than 10 years in profession, almost half of population answered maybe in job change question. Moreover, more than 60% of respondents who have 28-36 years’ experience also considered to change their jobs with the answer of maybe. 6) 10/44 respondents (23%) who work between 20-40 hours are part time nurses and 41/81 full time nurses works an average of 41-50 hours a week.

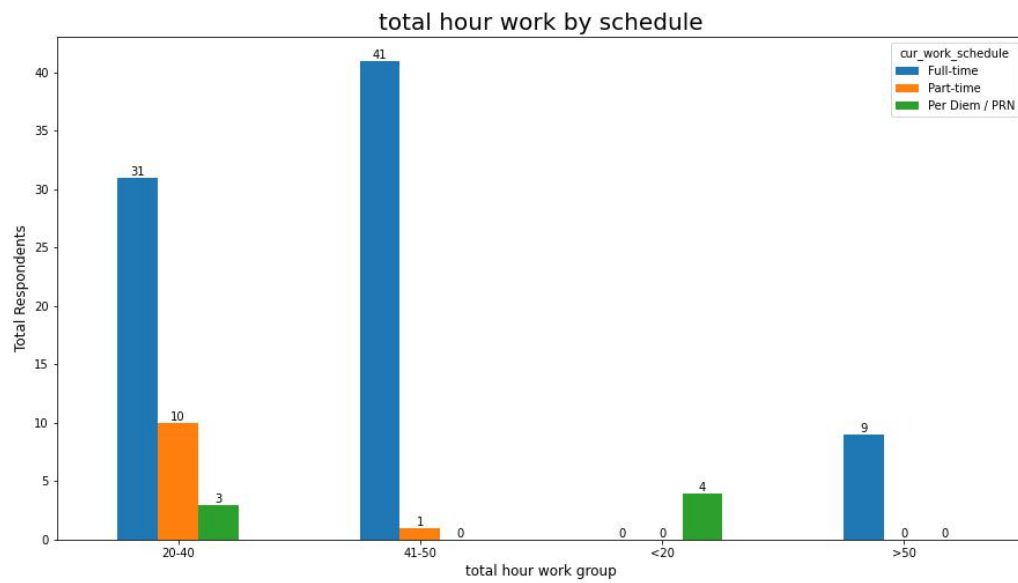
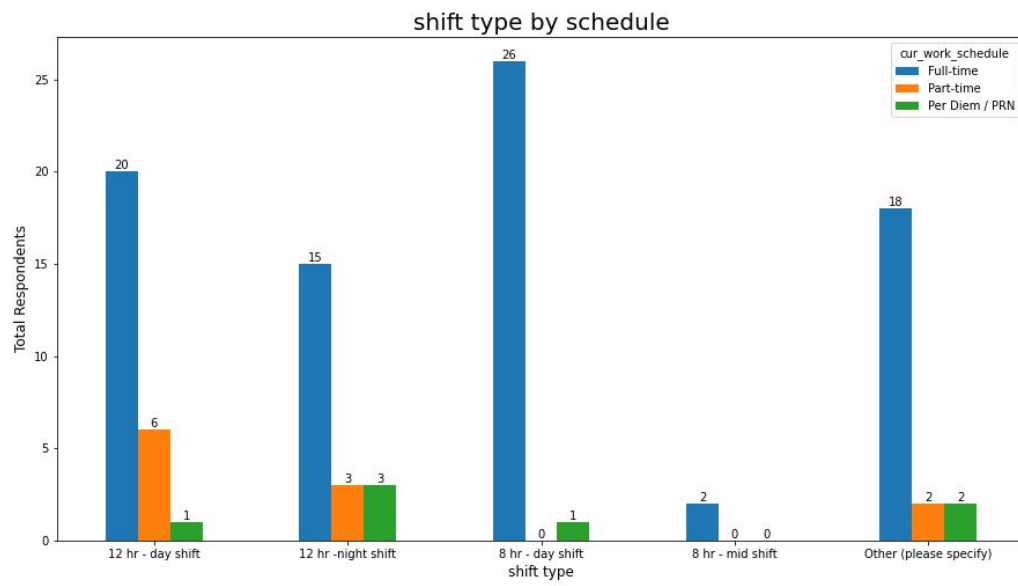
As for the two continuous variables – years in profession and total job changed, there’s no correlation between them. It looks like the two variables are right skewed in distribution.

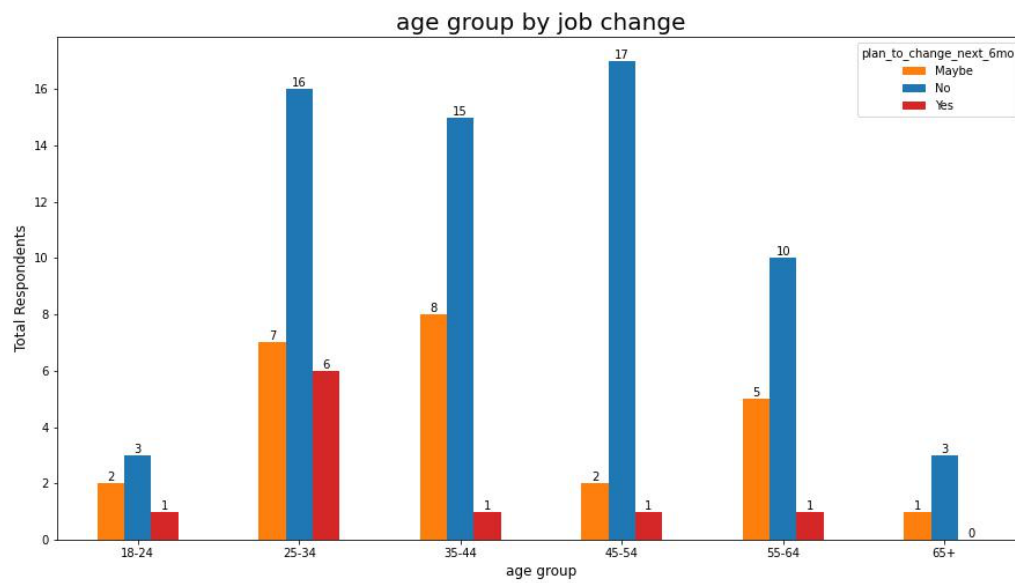
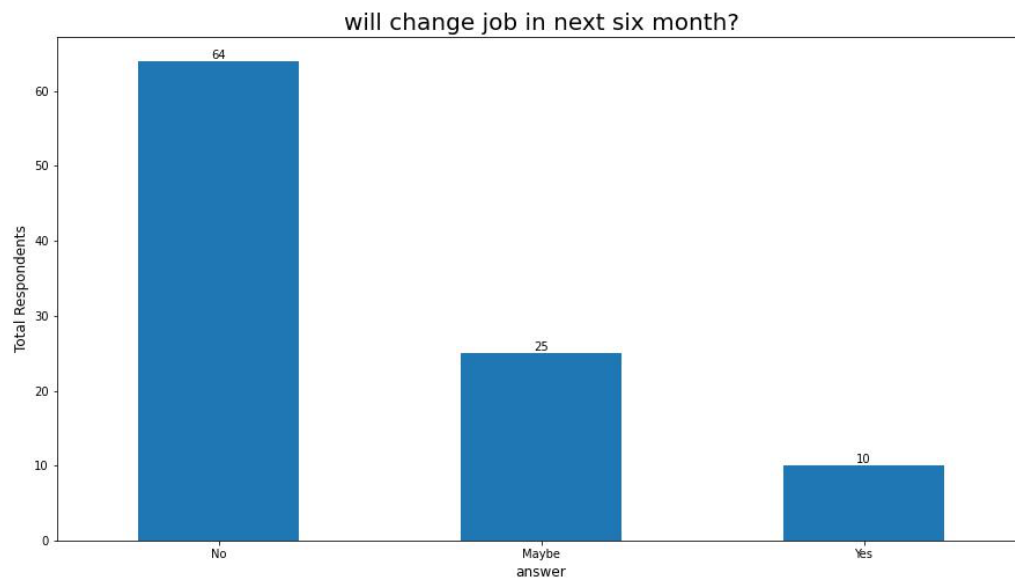


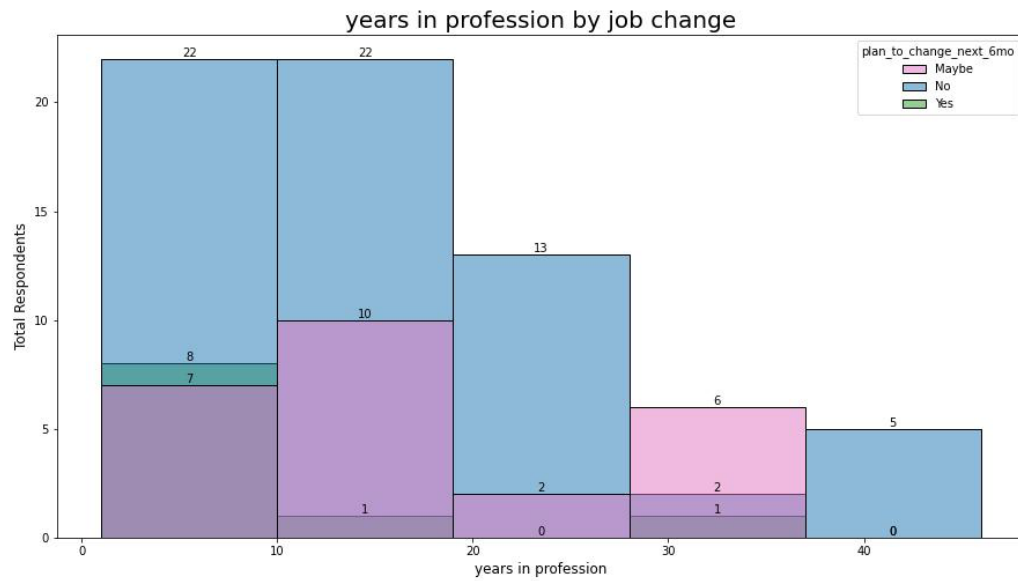
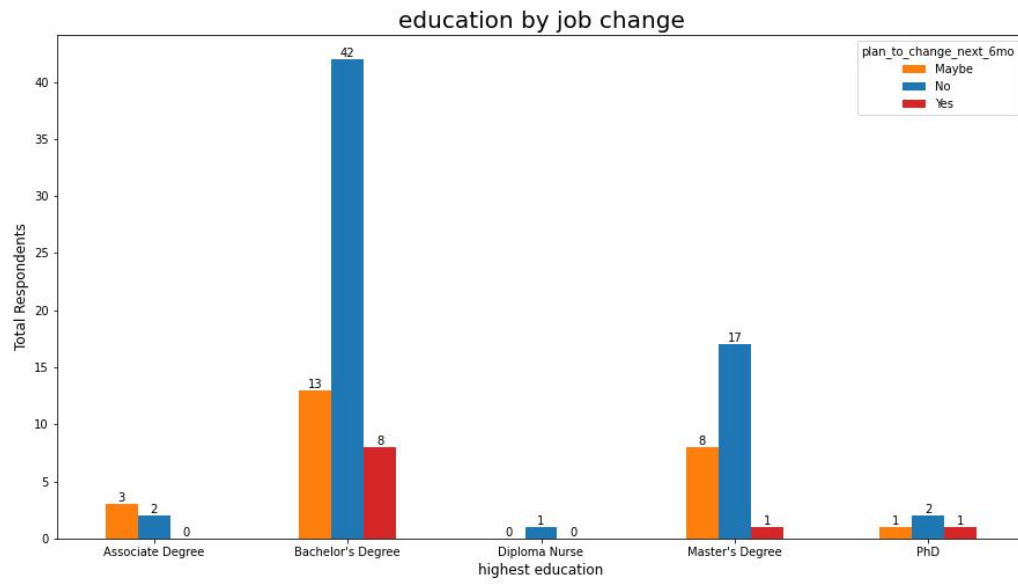


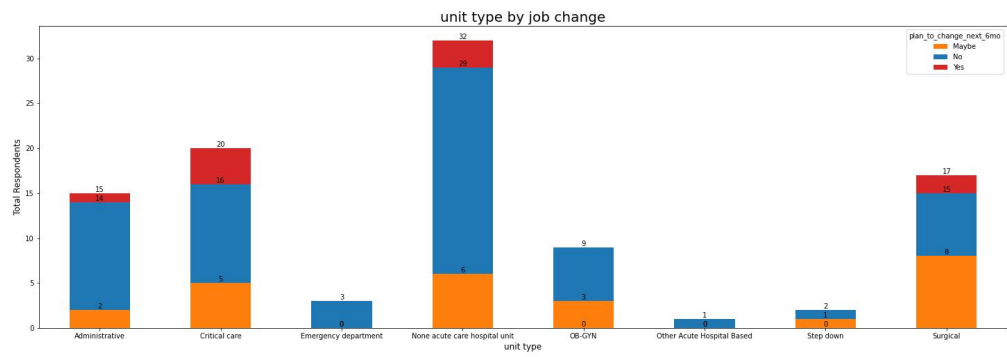
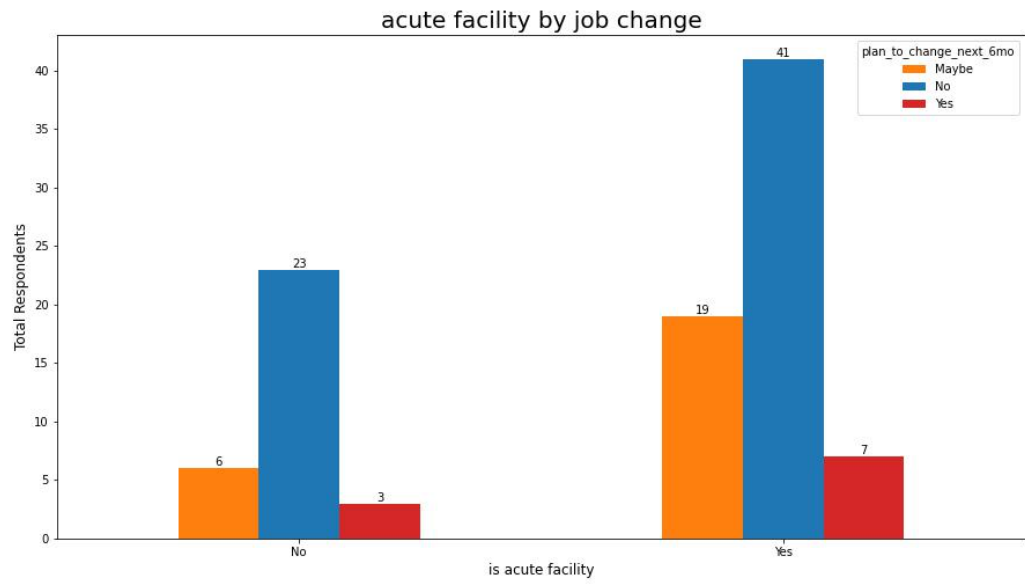


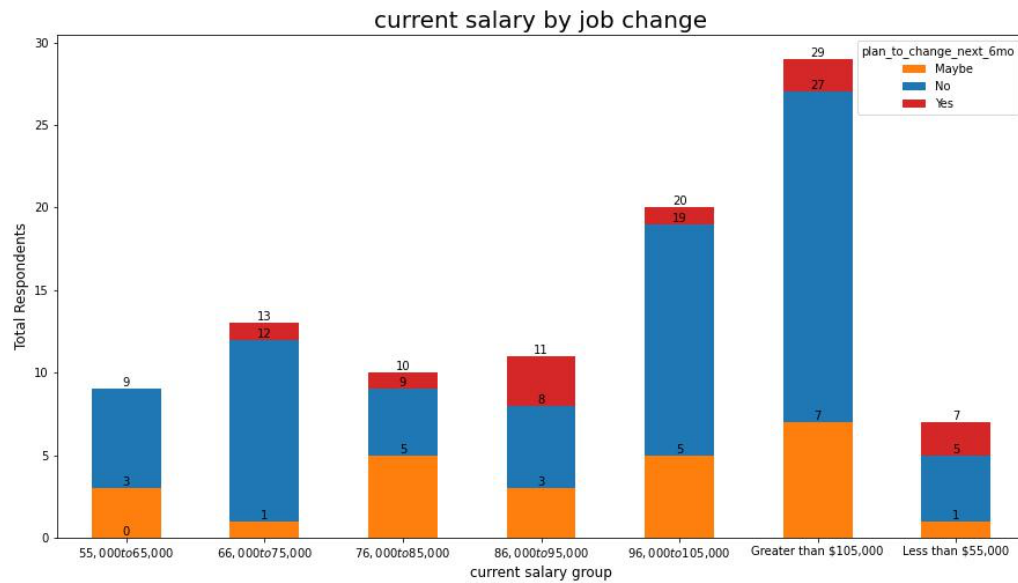
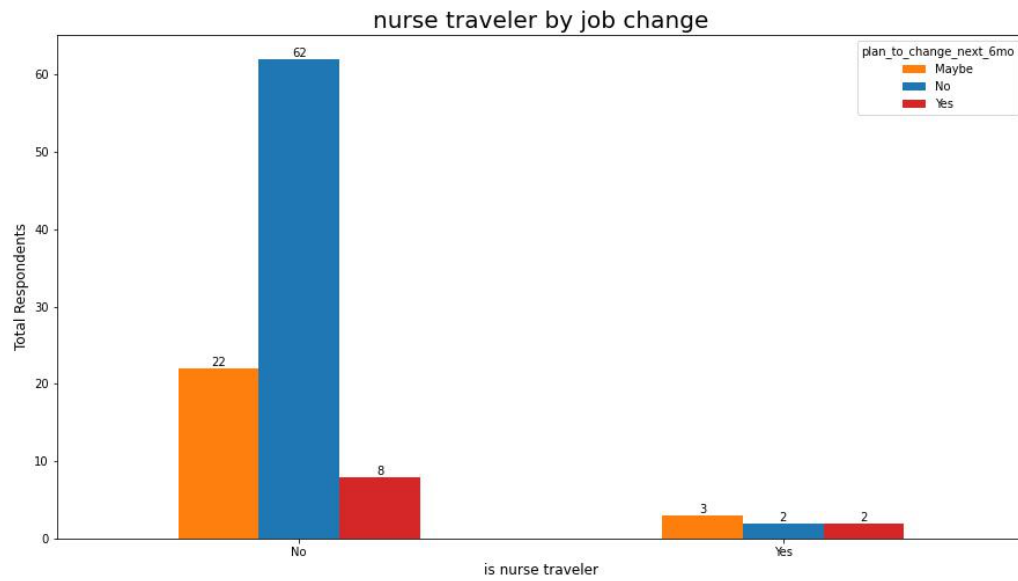




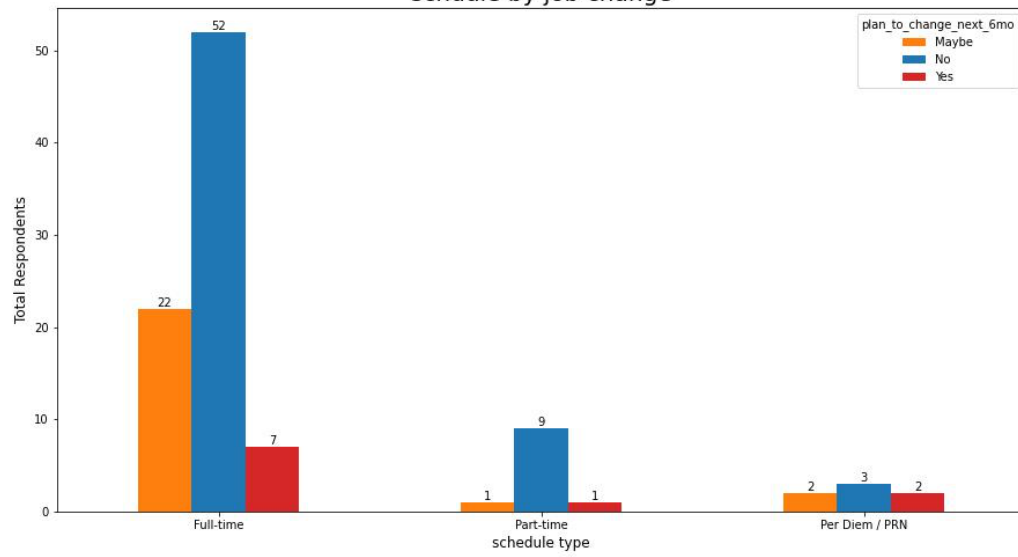




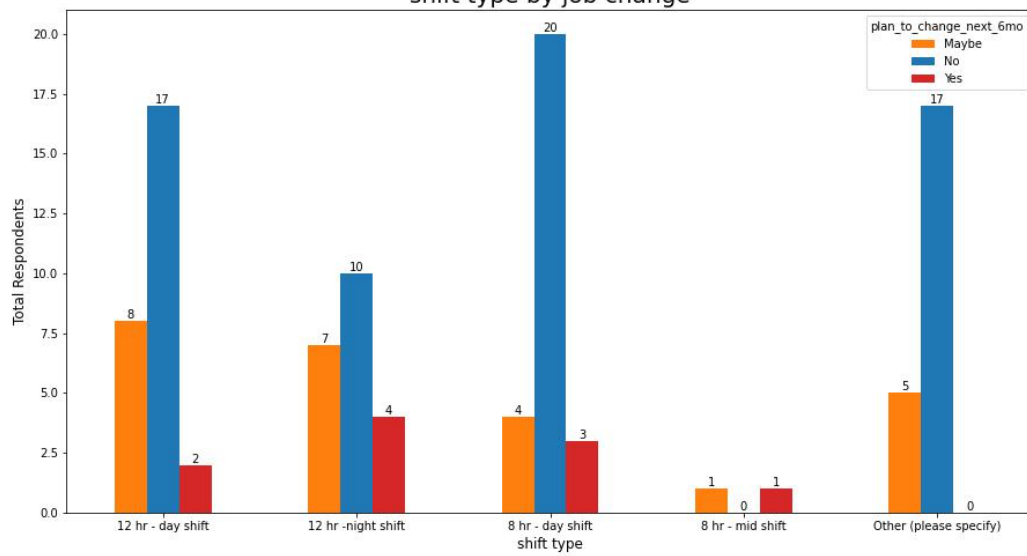


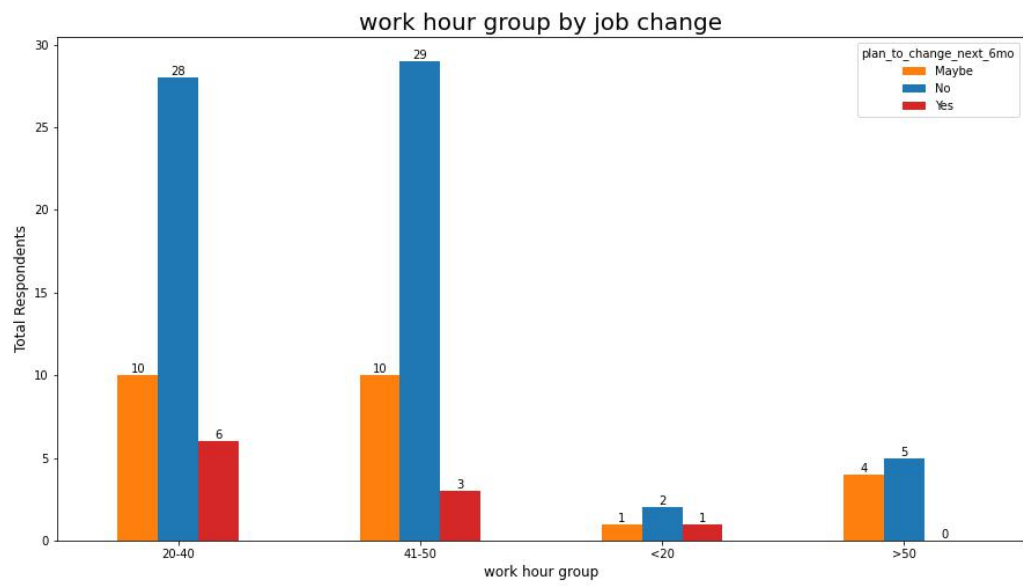


schdule by job change

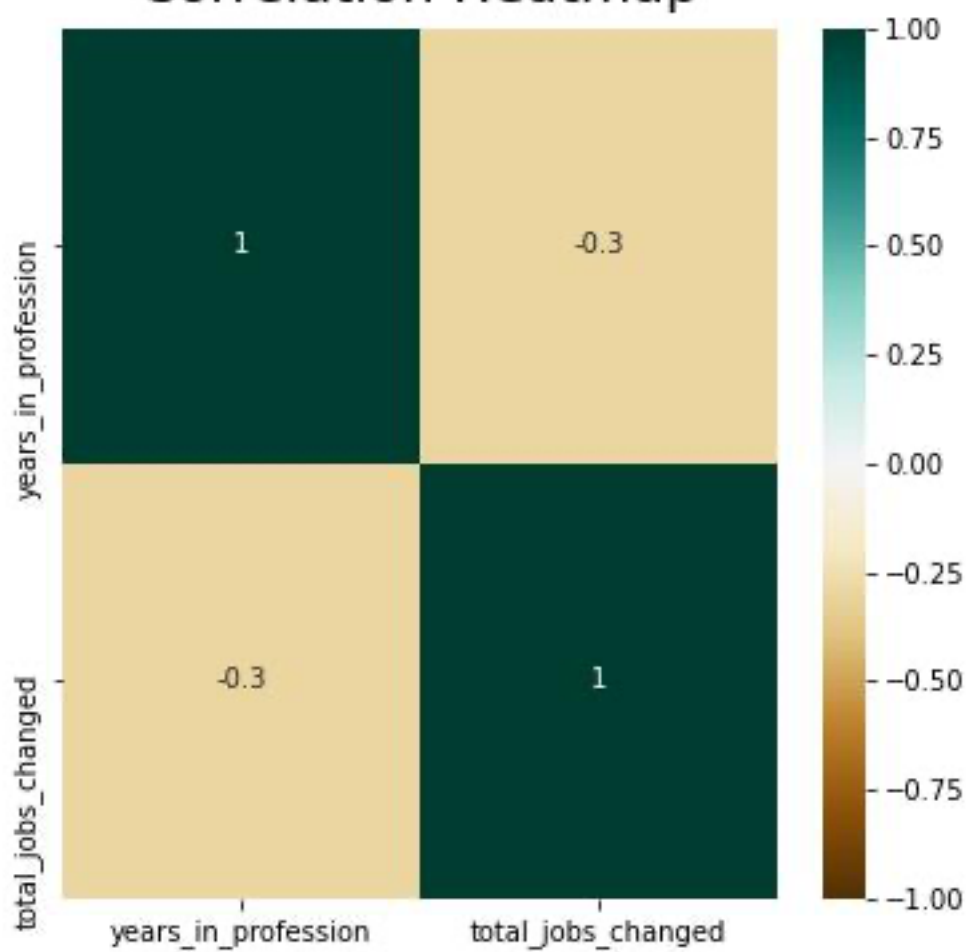


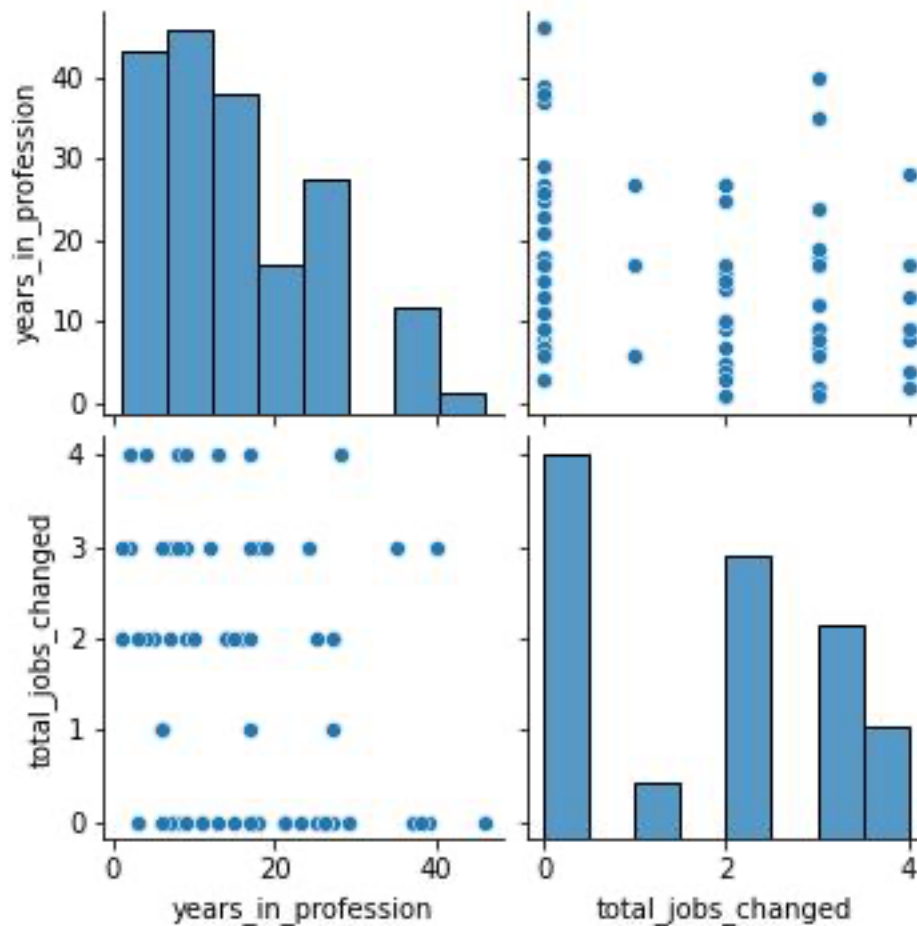
shift type by job change





Correlation Heatmap





III. Chi-Square

In this section, each categorical variable is paired with the rest categorical variables to test for independence.

H0: two variables are independent meaning no relationship.

H1: two variables are dependent meaning there's a relationship.

The results below are significant due to the fact of $p < \alpha$ (0.05). It explains change jobs is related to licensed state type, length of current license level, or current shift type.

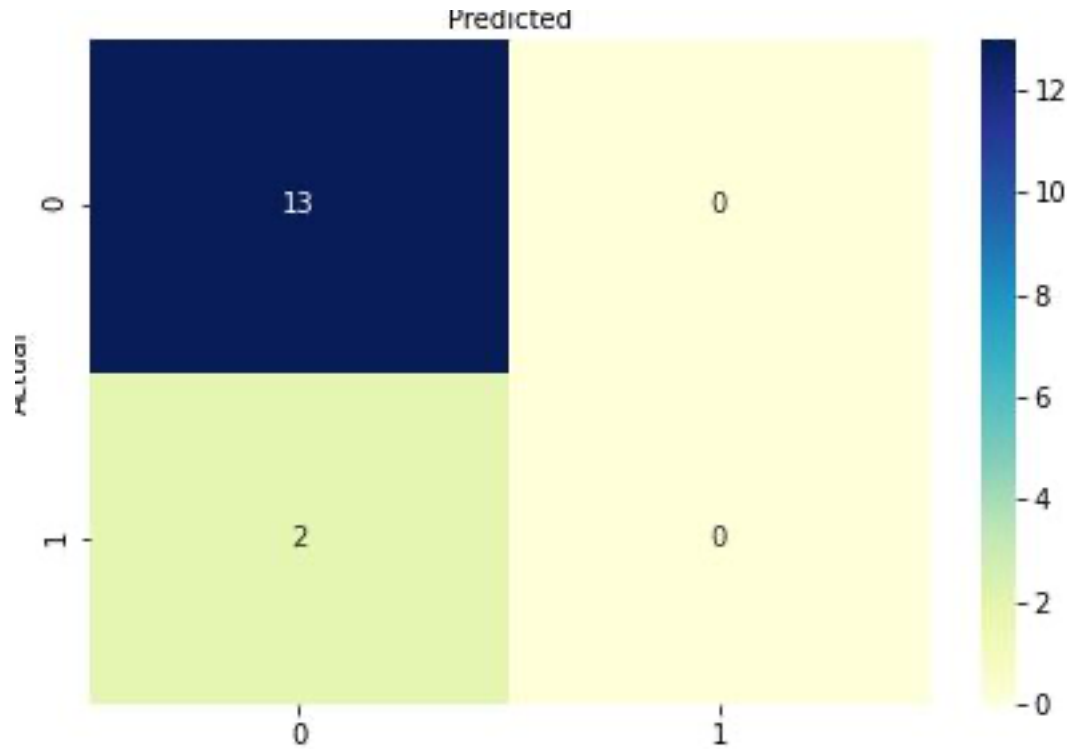
('plan_to_change_next_6mo', 'cur_licensed_state_type', 0.014844509365393704)

('plan_to_change_next_6mo', 'len_of_cur_license_level', 0.03264350167884819)

('plan_to_change_next_6mo', 'cur_shift_type', 0.01794285288127181)

IV. Predictive

There were only 77 observations in the predictive model where 15 are test dataset and 62 are training dataset. The result shows 13/15 no answers were predicted correctly as no and 2/15 no answers were predicted as yes. There was zero yes in the testing data.



Summary

Logistic regression was not successful due to the imbalanced output data and small sample size. In this case, the model failed to predict the answer of job change in the next six months. Therefore, a conclusion of what drives nurse turnover cannot be made based on the current results. However, this model can still be used when sample size grows. Moreover, decision tree can be built to predict three outcomes: yes, no, or maybe if the study scope changes. Although the predictive model did not work, some interesting finding via EDA can still explain some of the situations.