An introduction to our data and goals.

Some analysis of the medical features of o..

Some statistics from the lifestyles feature..

# **Stroke Prediction**

#### Project Introduction

For this project the dataset selected was comprised of some key health and lifestyle data from over 5,000 patients including if each patient suffered a stroke. With this dataset, it was determined to pursue two tasks and two questions. The two tasks were to visually analyze the dataset itself for trends, and secondly to create a predictive machine learning model which would analyze a patients health and lifestyle variables to predict if they were likely to suffer a stroke. Using machine learning, the data was tested with many different types of models to compare the performance. It was determined that logist Regression offered the best results.

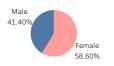
A small disclaimer that the none of the team members who assembled this project are medically trained, and our dataset was of course limited in scope somewhat. We were however, able to find some interesting correlations from the data, and create an accurate predictive model from ..

Total Cases

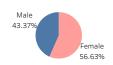
Stroke Status	
Non-Stroke	4,861
Stroke	249

Data Source..

Patient Gender Breakdown



Stroke Cases Gender Breakdown



## Logistic Regression Prediction Results

	Predicted	
Actual:	Stroke	No Stroke
Stroke	44	6
No Stroke	7	43

#### Feature

- Age, Hypertension, Heart Disease, Average Glucose Level
- Training 332 points, Testing 84 points.

Model Prediction Accuracy: 84.32324%

Model F1 Score: 82.0

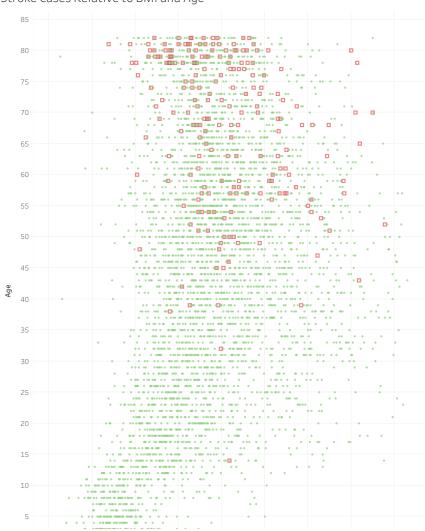
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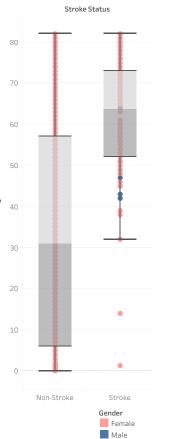
Some tatistics from the lifestyles feature.

# **Medical Data**

## Stroke Cases Relative to BMI and Age

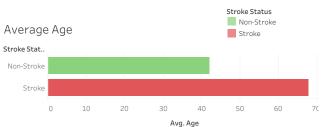


#### Stroke Status Box Plot



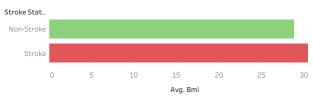
#### Age and BMI Correlate Strongly

The scatter plot to the left shows the strong correlation between age and BMI in regards to stroke outcomes. Each red square is a stroke



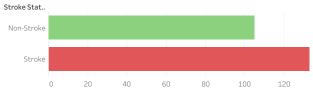
When examing the data around patient age in the dataset, it stands out from both the analysis and the predictive models that age was the strongest contributing risk factor for having a stroke

### Average BMI



In contrast to age, the data shows that BMI did not correlate strongly as a stroke predictor. The difference in average BMI between stroke and non-stroke outcomes was just under 2.

## Average Glucose Level

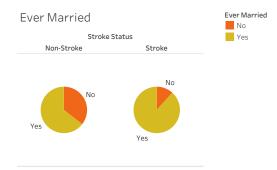


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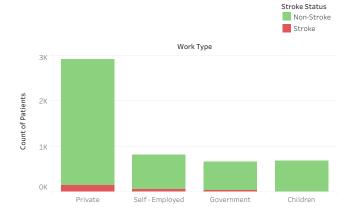
Some statistics from the lifestyles feature..

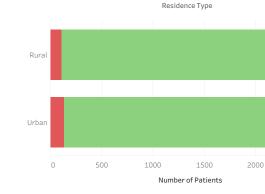
# Lifestyle Data



Included with our patient data were not only medical features, but some key features for each patient about different lifestyle factors. While we did try to create predictive models that included these factors we found they seemed to reduce the accuracy of the model and eventually did not include them in the final model. We did analyze these lifestyle factors to see if there were any interesting correlations or trends to stroke risk. Smoking Status, Type of Work, and Residence Type all seemed to have absolutely no correlation to stroke risk.

We found that a noticeably higher percentage of the patients who suffered strokes were married than the patients who did not have a stroke. However, this was found to not be predictive and likely just a characteristic of the dataset used. The data included some patients who were children which likely skewed these marriage statistics this way.





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