



Stroke Awareness Federal Campaign





Overview



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USING OUR MODEL



Our GOAL is to use MACHINE LEARNING to develop a model that can help vulnerable patients ASSESS their own risk of a stroke in lieu of traditional healthcare resources



INFERENTIAL MODELING

**Of the patients who
were admitted to the
hospital for a stroke,
what factors did they
have in common?**

The background of the slide is a complex, abstract molecular structure. It features numerous spheres in shades of blue, purple, and red, connected by thin, metallic-looking rods. The structure is dense and multi-layered, with some parts in sharp focus and others blurred, creating a sense of depth. The overall color palette is cool, dominated by blues and purples, with the red spheres providing a focal point.

Data Modeling

PATIENT ATTRIBUTES

This slides depicts the patient attributes we will be focused on in analyzing ties between the likelihood of stroke.





Machine Learning Model

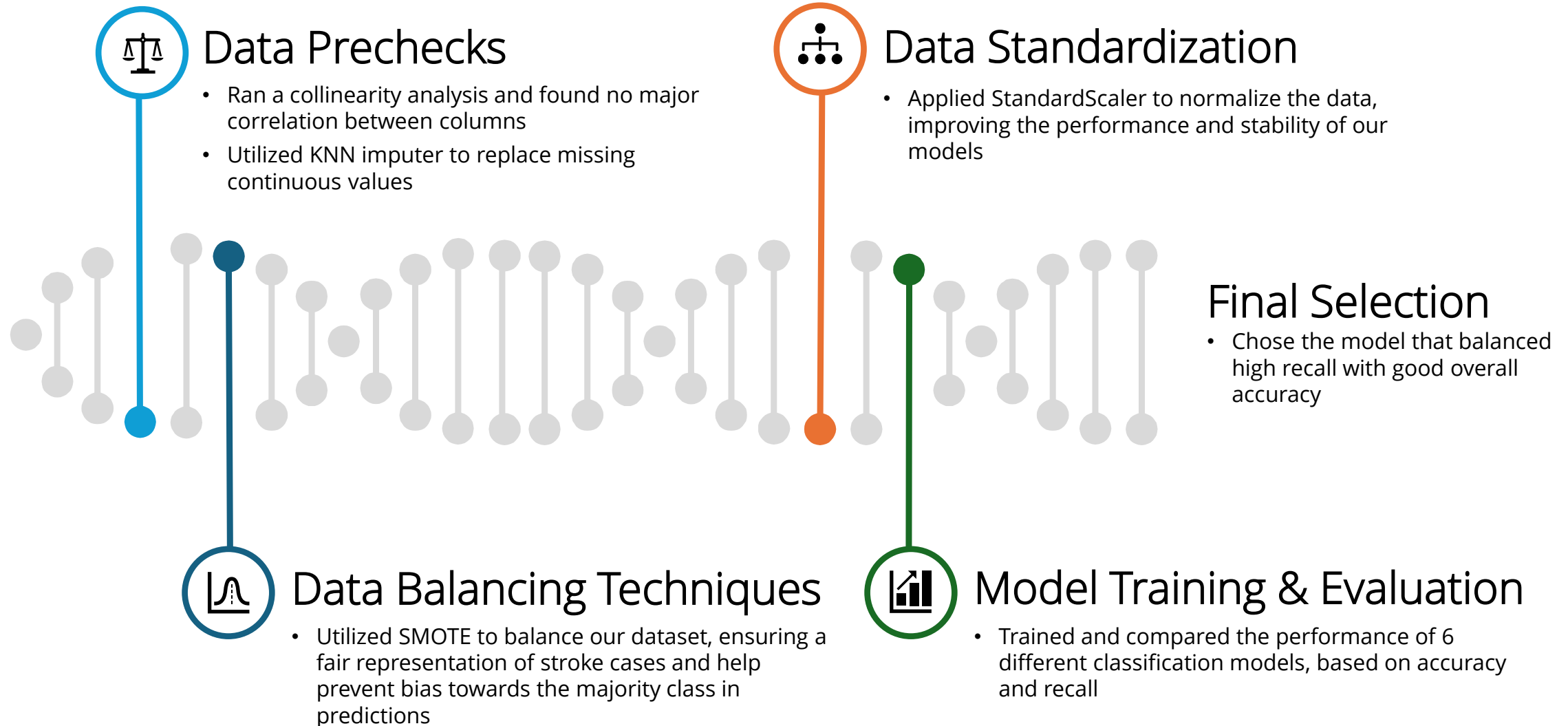
Methodologies and Tools Used

- Our data has 5,109 cases, and of those 249 (**around 5%**) patients were positive for a stroke
- We used **Synthetic Minority Oversampling Techniques** (SMOTE) to address the class imbalance
- There was **no significant collinearity** within the features (<70%)
- We compared the performance of 6 different models, focusing on **recall** as our top metric



Data Modeling Process for Stroke Prediction

Our goal is to predict stroke occurrences accurately, aiding in preventative health measures.



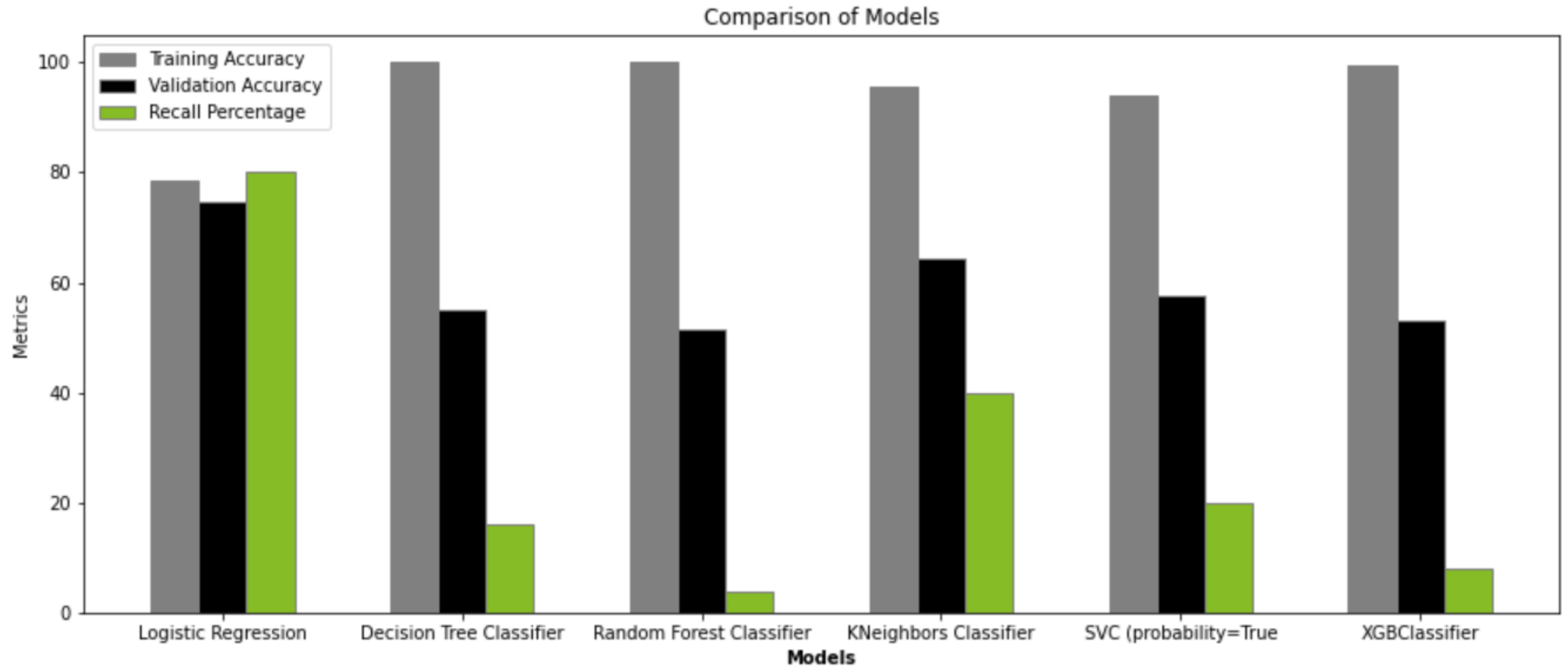


Evaluating the Model



Model Performance of Six Classification Models

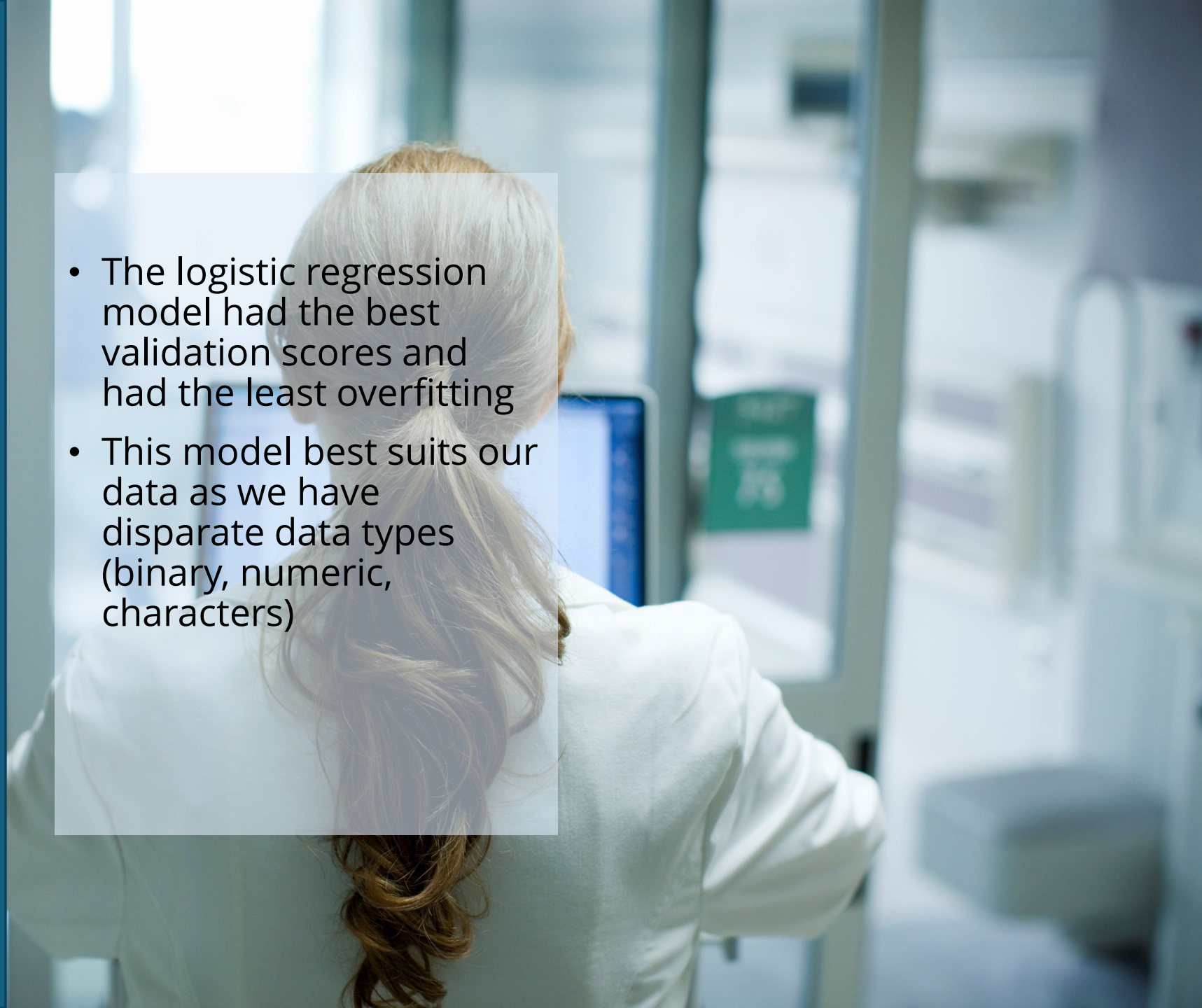
Here we compared the training set accuracy, testing set accuracy, and recall. The logistic regression model had the most balanced performance and highest accuracy in predicting stroke cases (recall).





Logistic Regression Model

- The logistic regression model had the best validation scores and had the least overfitting
- This model best suits our data as we have disparate data types (binary, numeric, characters)

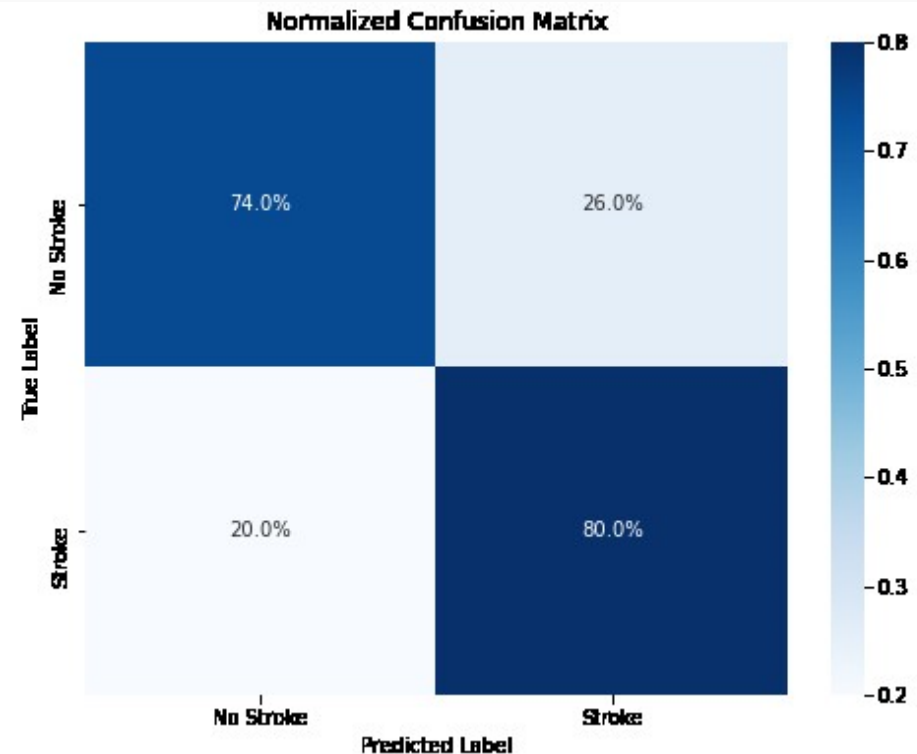


Accuracy

Our logistic regression model good had scores for **accuracy (78.5% training, 76.9% validation)** and **recall (80%)**

To improve accuracy, we included:

- the addition of balanced class weights to allow the model to function without being thrown off by outliers
- setting a higher recall value



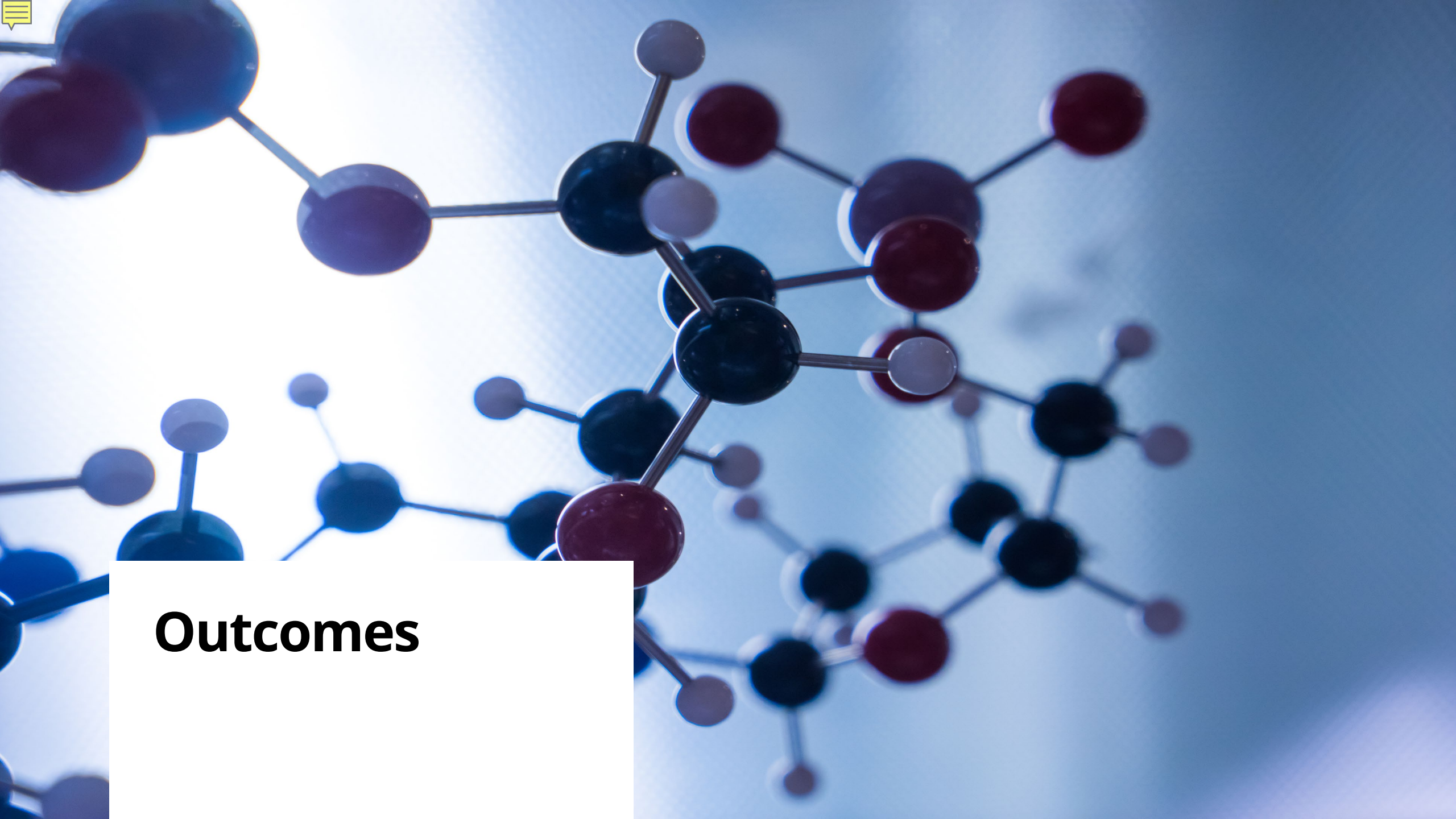


Interpretation



Purpose of this model:

Using the results, we can bring awareness to people with limited access to healthcare that have certain symptoms or are identified as ‘high risk’ by their doctors

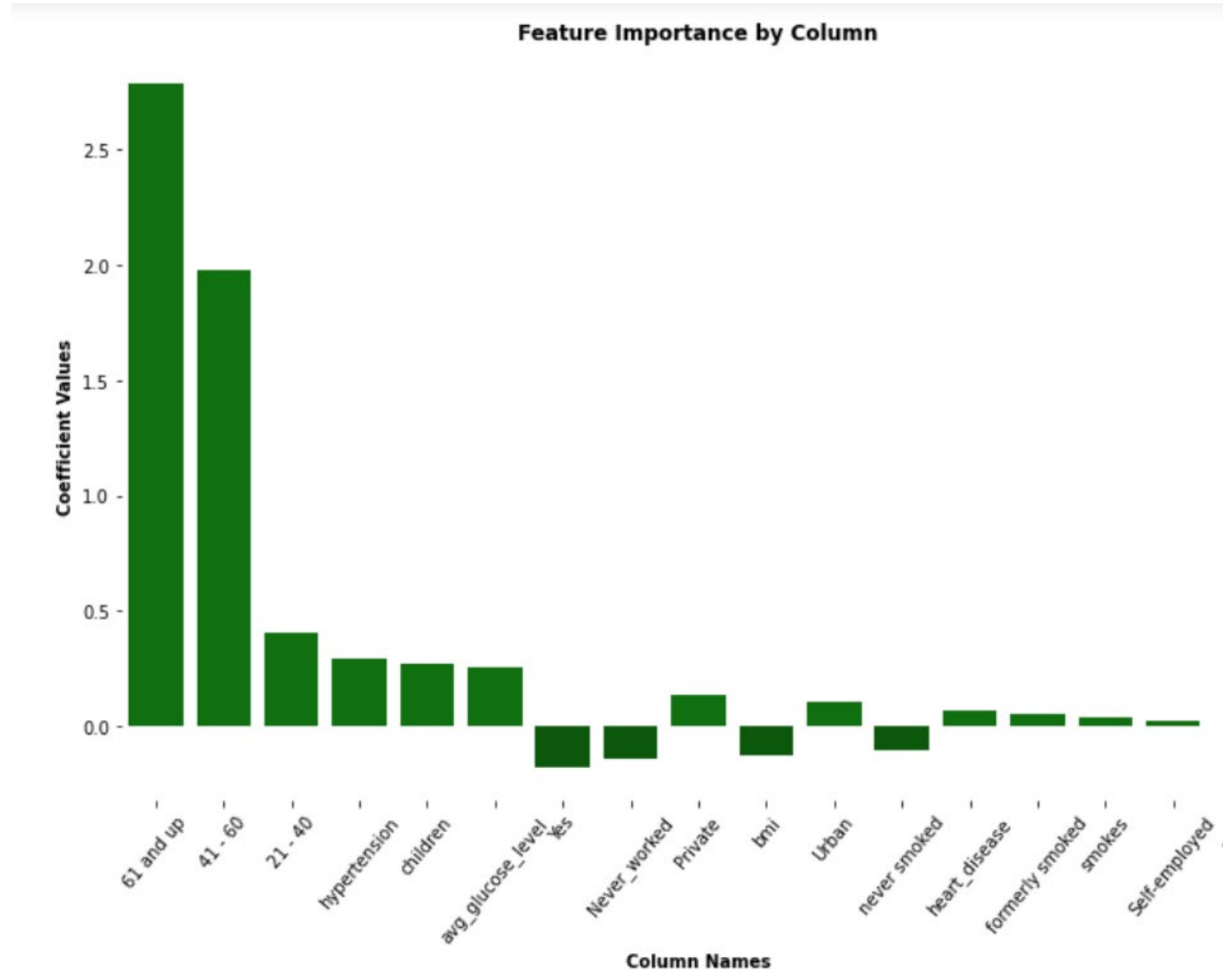


Outcomes

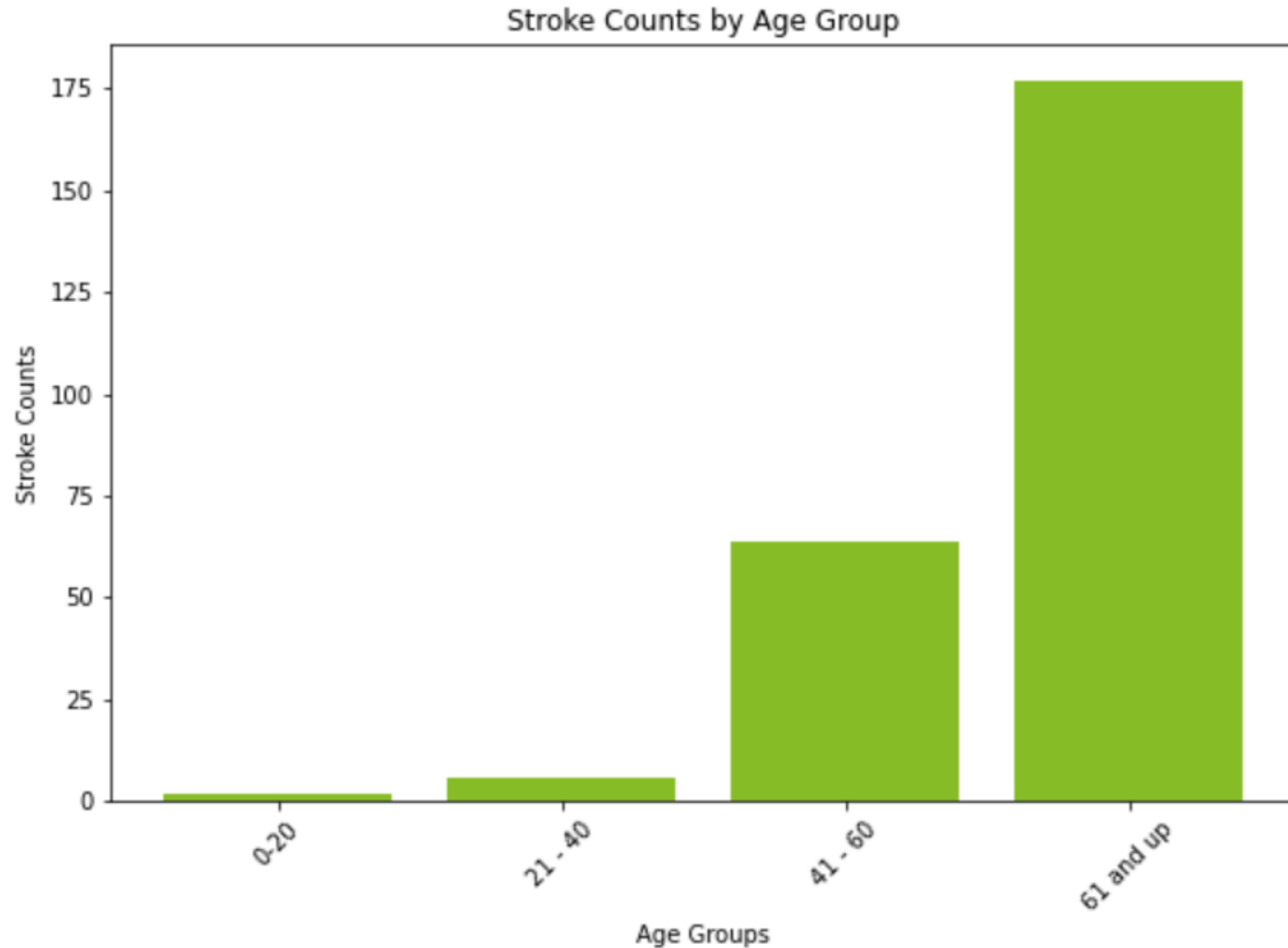
Feature Importance

Top 3 features

- Age (above the age of 41)
- Hypertension
- Children



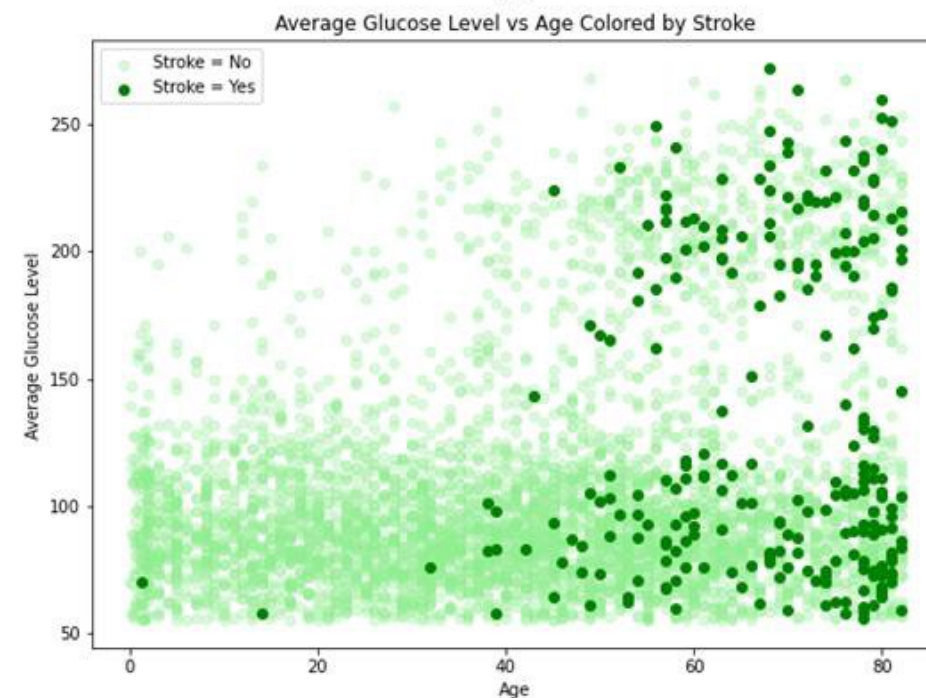
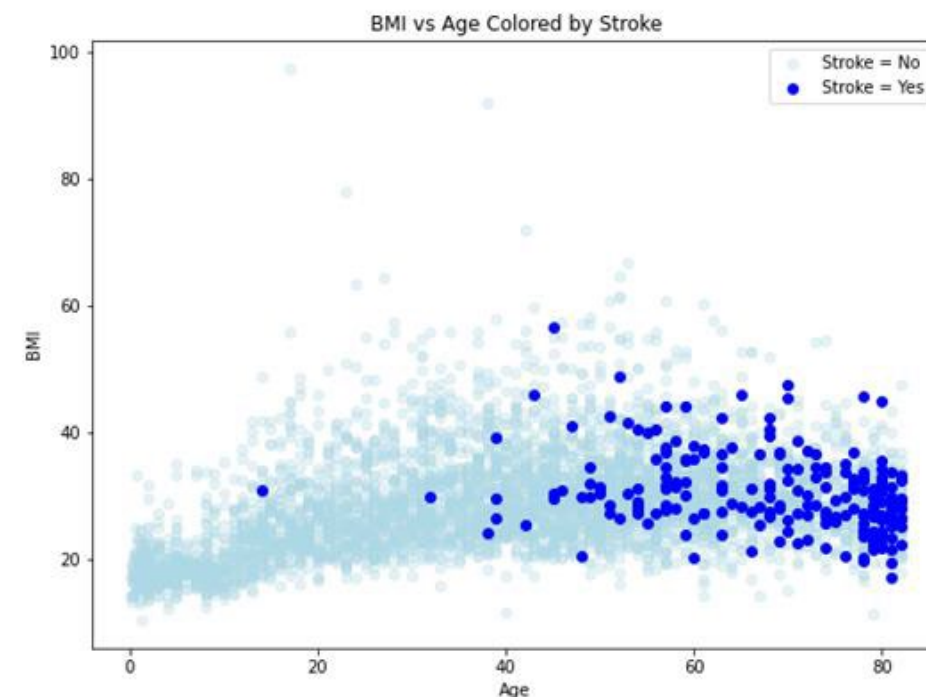
Age Distribution for Patients



Age Factor Analysis

Here we looked at some numeric features, and we decided to do a deeper on age in relation to strokes.

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- BMI and average glucose levels show a correlation with stroke incidence.
 - Older age groups tend to have a higher incidence of stroke
 - The visualizations highlight the importance of monitoring BMI and glucose levels as part of stroke risk assessment.
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Outreach



Recommendations

Communicate with Physicians and Primary care doctors to educate their patients above the age of 40 on stroke awareness

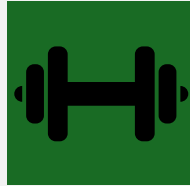
Build a public dashboard on stroke awareness and the high-risk factors based on our feature importances



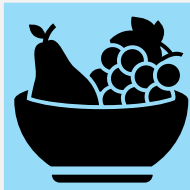
Future Work



What factors can decrease the risk of strokes



Exercise/Activity Level



Diet and Supplements

Thank you

Thank you for listening to our proposal. Please do not hesitate to reach out to any one of us regarding any comments or queries you may have.

Connect with the Team



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