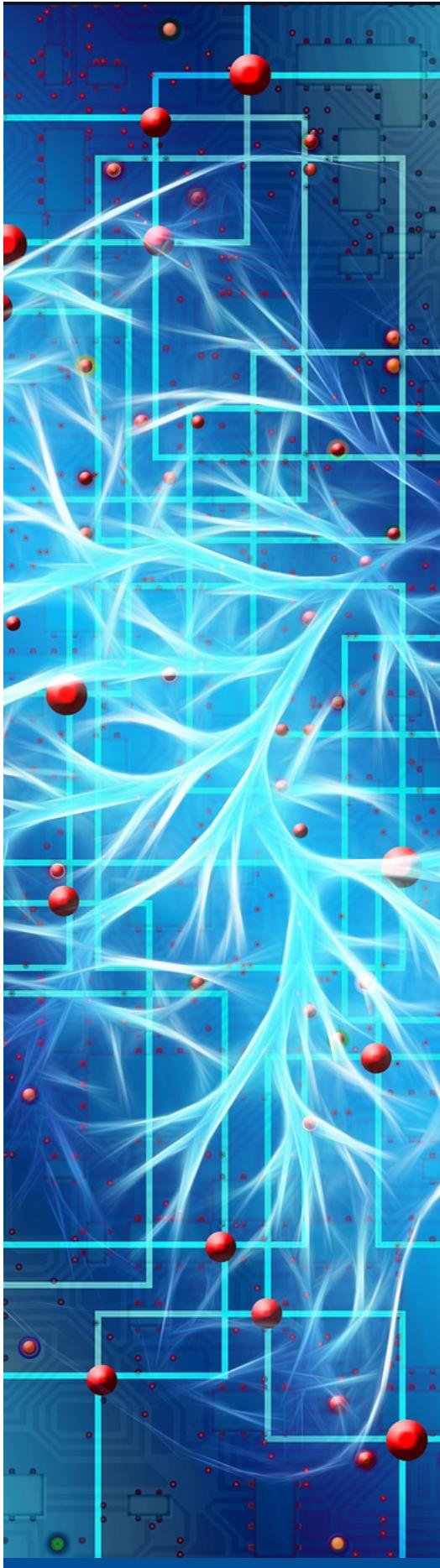


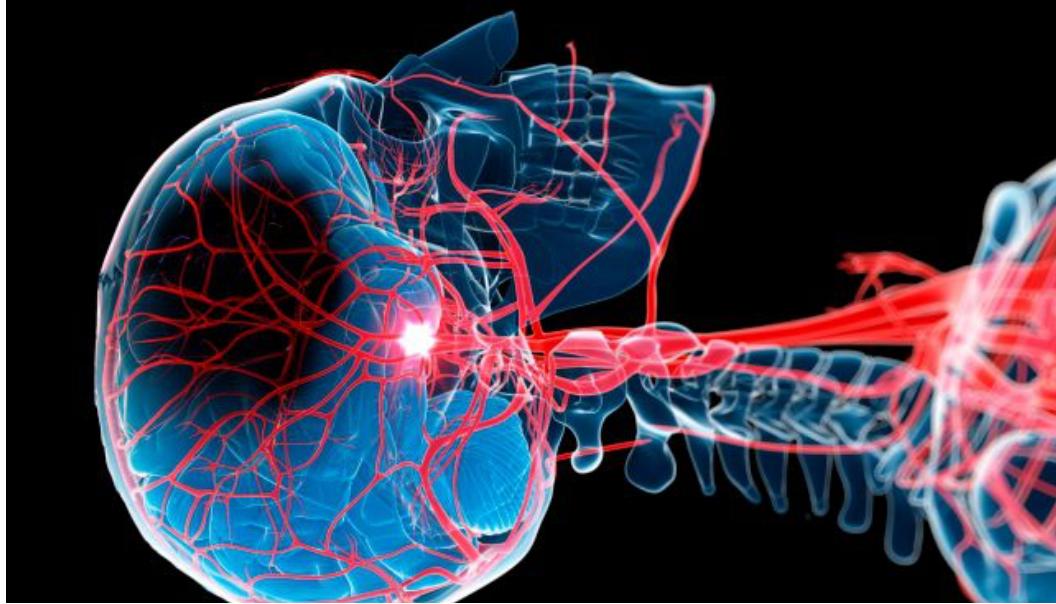
# Stroke Classification

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# Agenda

- 1 | BUSINESS UNDERSTANDING
- 2 | DATA
- 3 | BEST MODEL AND RESULTS
- 4 | RECOMMENDATIONS
- 5 | LIMITATIONS AND NEXT STEPS



# Business Understanding

## ● STAKEHOLDERS

The Mount Sinai Hospital in New York

## ● THE PROBLEM

According to the World Health Organization (WHO) stroke is the 2nd leading cause of death globally, responsible for approximately 11% of total deaths. Additionally, in the US, someone has a stroke every 40 seconds.

On top of that, strokes are a known complication of surgery.

## ● THE PROJECT

Develop a model that acts as a preliminary assessment to determine whether a person is likely to have a stroke or not during surgery using available data. The results will determine if further screening is needed.

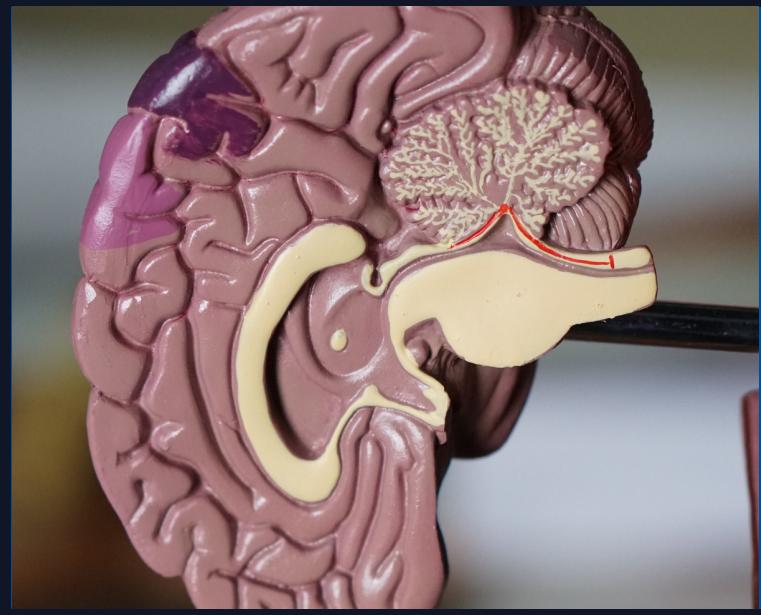
## ● THE GOAL

Introduce the model that arbitrates the best results to identify patients in need of extra screening before surgery



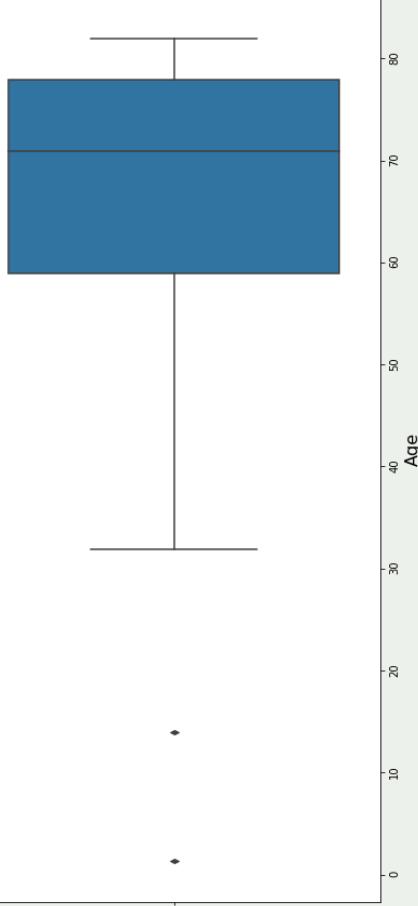
# The Data

- FEATURES
  - BMI, Smoking Status, Glucose level, Age, Gender, Ever Married, Residence Type, Workin Status
- MISSING DATA
  - BMI (Body Mass Index)
- IMBALANCE
  - 95% - No stroke
  - 5% - Stroke

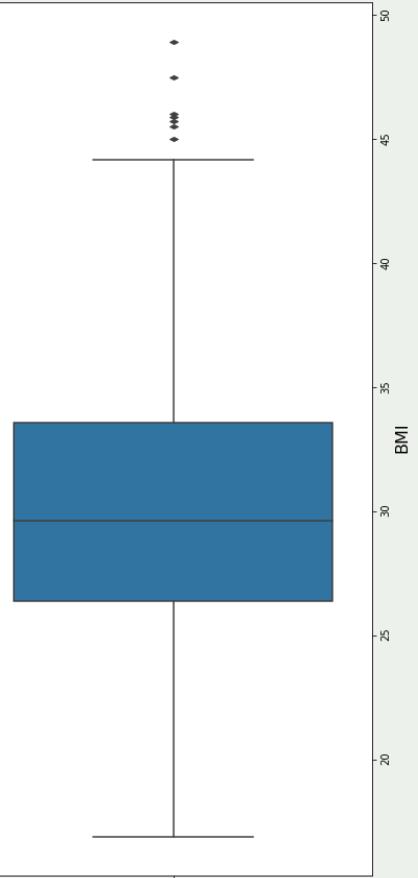


# Data Processing

Distribution of Ages Among Those With a History of a Stroke

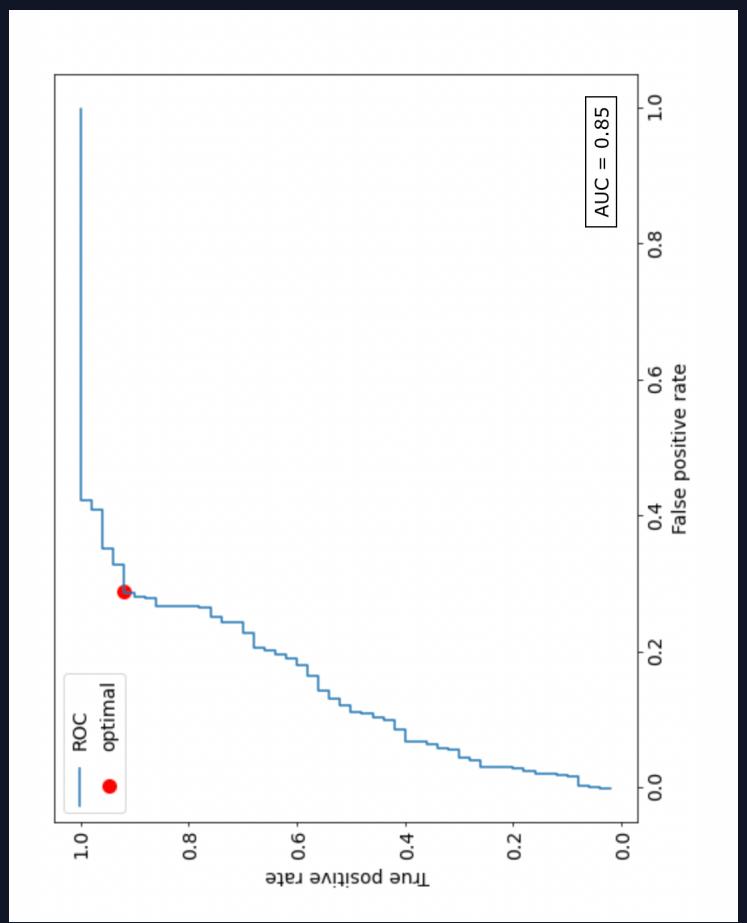
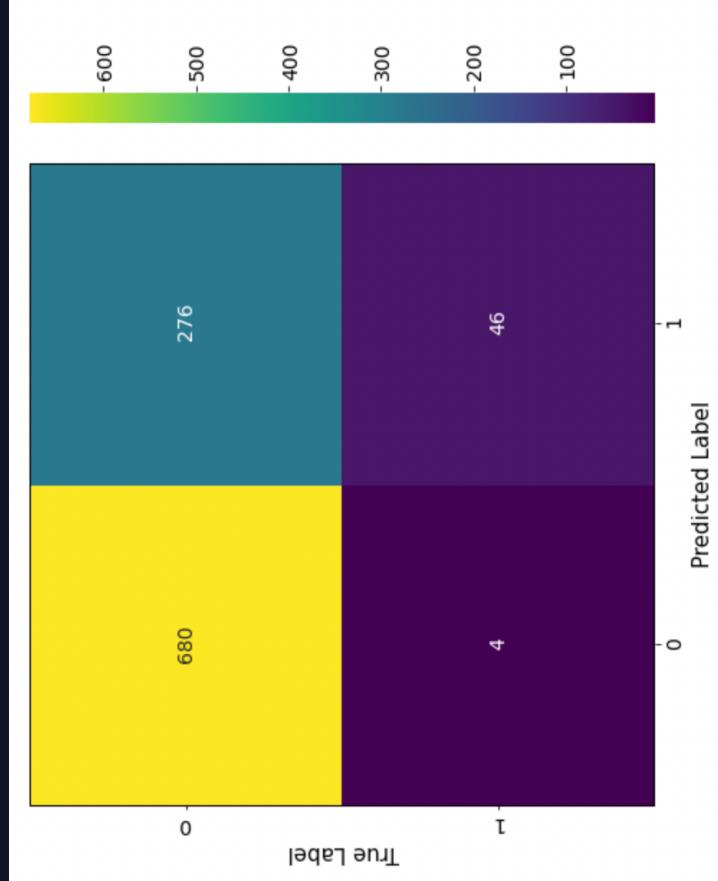


Distribution of BMI Among Those With a History of a Stroke



# Best Model

Logistic Regression - Optimal Threshold (0.446)



## ✓ CLASSIFICATION MODEL: LOGISTIC REGRESSION

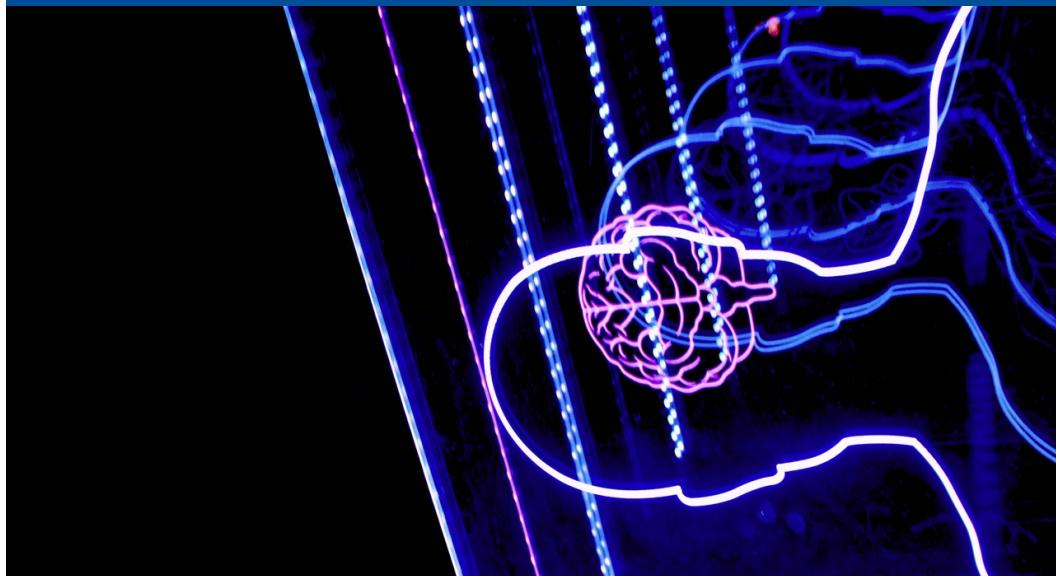
Included optimal threshold for best results

## ✓ RECALL SCORE: 92%

The main metric used to determine the accuracy of our model  
A false negative is more costly than a false positive.

## ✓ BETTER THAN BAYES?

Had best recall was our Gaussian Naive Bayes model with a recall score of 94% but a false positive rate of 0.60.



## Recommendations

### MODEL TYPE

Based on our project, we propose that logistic regression has the best classification of stroke risk and most effectively minimizes both the false negatives and false positives.

### OPTIMIZE SCREENING COSTS

By decreasing the false positives, it gives less room to the insurance company to reject claims - benefiting both the patients and the hospital.



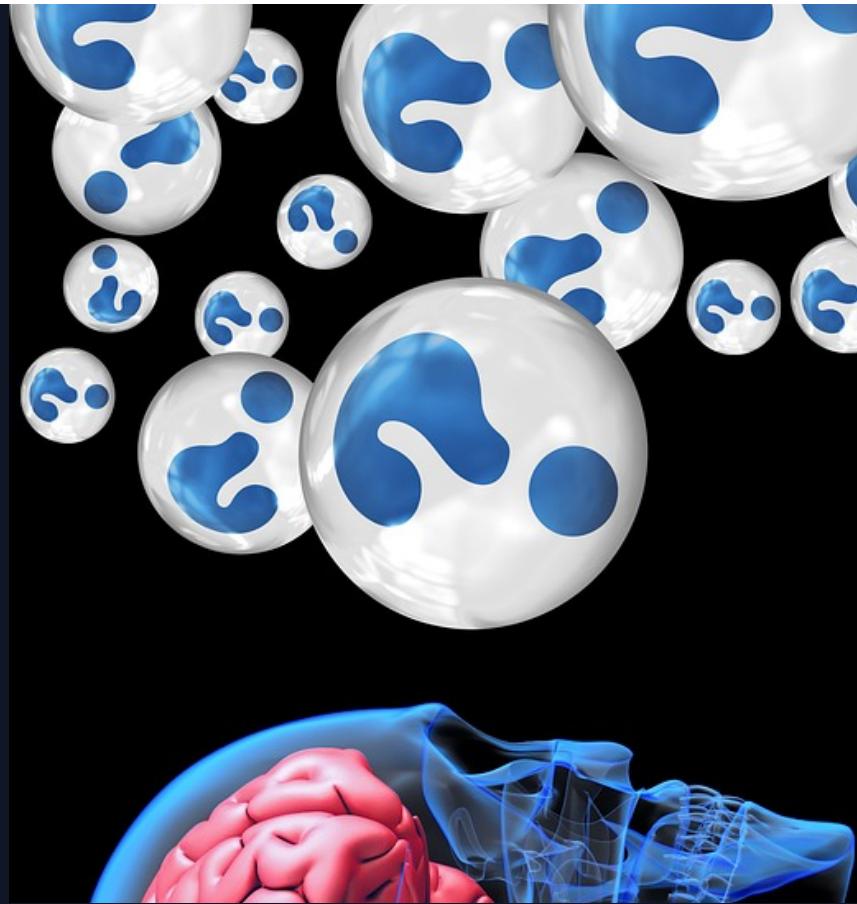
# Limitations and Next Steps

## Limitations

- Medication is not taken into account
- Missing Data
- Unknown origin of the dataset

## Next Steps

- Cholesterol
- Family history of stroke
- Number of strokes
- Race (i.e., the likelihood of stroke among different races)



# Contact US

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