# Predicting Stroke Risk

Using individual patient data

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## Stroke Statistics

#### • Worldwide

- Second leading cause of death
- 11% of deaths (15 million people)
- 1 of 6 deaths related to cardiovascular disease
- United States
  - ~800,000 annually
  - Stroke every 40 seconds
  - Death every 3.5 minutes

#### Leading causes of death globally



https://www.who.int/news-room/fact-sheets/detail/the-top-10-causes-of-death

10

0

0

### Stroke Statistics (cont.)

- Of ~15 million annual stroke instances:
  - 1/3 result in death
  - 1/3 recover
  - 1/3 are left disabled
- A leading cause of long-term disability

- Known risk factors:
  - Cardiovascular/Health:
    - High blood pressure
    - High cholesterol
    - Obesity / diabetes
    - Age
  - Other:
    - Race
    - Where you live



### Stroke Dataset: 10 Features

- 5 Categorical:
  - 2 multiclass
    - `work\_type`, `smoking\_status`
  - 2 binary
    - `ever\_married`,
      `Residence\_type`
    - Converted to 1/0
  - 1 multiclass converted to binary
    - `gender` (removed 'Other')
    - Converted to 1/0

- 5 Numeric:
  - 3 continuous
    - `age`, `avg\_glucose\_level`,
      `bmi`
  - 2 discrete, binary
    - `hypertension`, `heart\_disease`
- Target Variable:
  - Binary (1/0)
    - stroke / no stroke
  - Extremely imbalanced
    - Stroke ~2% of dataset



#### Distributions

#### • Glucose is bimodal



- Children at lower risk
  - ~6,000 'children' in dataset
- Impacts multiple variables:
  - `age`
  - `hypertension`
  - `heart disease`
  - `ever\_married`
  - `work\_type`
  - Possibly `smoking\_status` (unknown)

### Data Preprocessing

#### • Imputing Null values

- `smoking\_status` 30% Null
  - Null  $\rightarrow$  'Other'
- `bmi` ~3% Null
  - Tried Logistic Regression
  - Landed on median
- Encoding
  - Binary features  $\rightarrow$  1 / 0
  - Multiclass  $\rightarrow$  one-hot encoded

- Transformation
  - Box-Cox age, bmi, & glucose
  - Scaled all features -1 to 1
- Balancing (~2% stroke)
  - Oversample stroke
  - Oversample using SMOTE
  - Oversample / Undersample
    - Oversample to 10% of majority using SMOTE
    - Undersample majority so stroke is 50% of majority
  - Leave imbalanced and use weights



### Model Selection / Evaluation

#### **Metrics:**

- Accuracy
  - 98%, predicting no strokes
- Recall
- Matthews Correlation
  Coefficient
- Area Under Curve (Receiver Operating Characteristic)

#### Hyperparameter Tuning:

- Grid Search CV
  - Random Search CV
- Scoring with multiple metrics
- Voting Classifier (with weighting)



### Conclusions

- Huge imbalance, not a perfect model
- Need to find a balance in the results
- More features could be helpful
- Possible inherent bias in the data
  - Could be high risk but haven't had a stroke yet





### Implementation

- Allow people a level of control over their personal healthcare
- A healthcare app:
  - Answer health questionnaire
  - Store health metrics (weight, blood pressure, health screening results, etc.)
  - Link activity apps (pulse, steps, etc.)
  - Predict risk for stroke, heart disease, and others
  - Provide personalized suggestions for lowering risk
  - A tool to discuss with your primary care physician

