

# **Stroke Prediction System using Logistic Regression**

Final Project Documentation

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## **Acknowledgement**

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# 1. Project Overview

## 1.1 Introduction

Stroke is a serious medical condition that can lead to long-term disability or death. It occurs when the blood supply to part of the brain is cut off. Early prediction and prevention are critical in reducing stroke-related mortality.

This system leverages **Logistic Regression** and a web interface built with **Flask** to allow users to input personal and medical information and instantly receive a risk prediction. If a stroke risk is detected, a warning is issued along with helpful resources.

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## 1.2 Scope and Objective

- **Scope:** Predict the likelihood of a stroke based on user input and display educational content if risk is high.
  - **Objective:**
    - Perform real-time risk prediction using a trained model.
    - Educate users with information like FAST symptoms.
    - Provide immediate alerts (video, modal popup) for high-risk cases.
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## 1.3 Modules and Description

### User Module

- **Input Form:** Enter data such as age, gender, health conditions, etc.
- **Validation:** Client-side validations to ensure input integrity.
- **Prediction:** Display Likely / Not Likely result with visual aids.

### **Model Input Features:**

1. Gender
  2. Age
  3. Hypertension
  4. Heart Disease
  5. Ever Married
  6. Work Type
  7. Residence Type
  8. Average Glucose Level
  9. BMI
  10. Smoking Status
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## **1.4 Existing vs Proposed System**

Existing System	Proposed System
Manual risk analysis	ML-based automated prediction
No real-time warning	Modal + Video alert for high-risk
Lack of user guidance	FAST image + stroke education built-in
Desktop only	Mobile responsive web app

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## **2. Project Lifecycle**

### **2.1 Model Used: Waterfall Model**

- Requirements Analysis

- Design
  - Implementation
  - Testing
  - Maintenance
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## **3. Project Design**

### **3.1 Architecture**

- Frontend: HTML, CSS, Bootstrap 5
  - Backend: Python (Flask)
  - ML Model: Logistic Regression + SMOTE
  - Data Flow: User Input → Preprocessor → Model → Result View
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## **4. Implementation**

### **4.1 Technologies Used**

- Python 3.9
- Flask Framework
- Scikit-learn, Pandas, Numpy
- HTML5, CSS3, Bootstrap
- Joblib (for model export)
- SMOTE (to handle class imbalance)

## 4.2 Frontend Highlights

- Client-side validation for age, glucose, BMI
- Responsive design with Bootstrap
- Video background for "Likely"
- FAST image + modal education section

## 4.3 Preprocessing

- Categorical columns encoded using OneHotEncoder & OrdinalEncoder
- Numerical columns passed directly
- Final input: 18 feature columns
- Preprocessing object saved with model

## 4.4 Model Building

- Multiple models tested: Logistic Regression, Random Forest, XGBoost, LightGBM
- Logistic Regression with SMOTE gave highest **recall**

Final model stored as dictionary:

```
python
CopyEdit
{
    'model': lr_with_smote,
    'preprocessor': column_transformer,
    'encoded_cols': [...],
    'numeric_cols': [...]
}
```

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## 4.5 Flask Integration

- Routes:
    - / → Input form
    - POST / → Accept form, transform, predict
  - Output:
    - Likely / Not Likely with probability
    - Custom visuals for each case
- 

## 5. Testing

### 5.1 Levels of Testing

- Unit Testing: Each route, function
- Integration Testing: Model + Web together
- System Testing: Frontend to backend flow
- User Acceptance Testing: Usability tested with users

### 5.2 Validation Criteria

- Age must be >10 and <110
- All fields are mandatory
- Inputs are range-validated and sanitized

### 5.3 Test Case Example

Test	Input	Expected Output
Age Validation	Age = 5	Error: Age must be >10

Stroke Case

Smoker + Heart  
Disease

Prediction: Likely + Warning Modal

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## 6. Advantages & Limitations

### Advantages

- Accurate ML-based stroke risk prediction
- Real-time alerts and symptom guidance
- Easy, mobile-friendly web access
- High recall from SMOTE balancing

### Limitations

- Prediction based on limited features
  - Not a replacement for clinical diagnosis
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## 7. Conclusion

This project successfully integrates machine learning with a user-friendly web interface to predict stroke risk. By combining clinical features with logistic regression and strong UI/UX, it serves as a powerful tool for early intervention and public awareness.