

# Title: Predicting and Preventing 30-Day Hospital Readmissions

Subtitle: Al-Powered Risk Scoring & Post-Discharge Care Coordination

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### **Problem Statement**

- Heading: The Challenge Hospitals Face
- 30-day readmissions cost Medicare \$26B annually, leading to financial strain.
- CMS penalizes hospitals with high readmission rates under HRRP (Hospital Readmissions Reduction Program).
- Hospitals often rely on manual review or reactive care, lacking real-time actionable insights.
- High-risk patients may not be identified before discharge, increasing chances of readmission.

### **Solution Overview**

- Heading: Al-Driven Risk Prediction & Proactive Care
- Predicts patient risk for 30-day readmission in real-time.
- Generates risk scores with actionable insights for medical staff.
- Reduces readmission costs by 15% through early interventions.
- Integrates with hospital workflow to support personalized post-discharge care.

### Input Data & Features

- Heading: Comprehensive Patient Data
- **Demographics:** Age, gender, socio-economic status
- Medical History: Previous admissions, chronic conditions, comorbidities
- Vital Signs & Lab Results: Heart rate, blood pressure, lab tests, biometrics
- Medication & Treatment: Prescriptions, adherence, surgeries
- Post-Discharge Information: Follow-ups, home care, readmission history
- External Datasets: CMS readmission data for benchmarking

### **ML** Architecture

- Heading: Predictive Model Pipeline
- Data Preprocessing:
  - Missing value imputation (median/mode)
  - Normalization and scaling of numeric features
  - Encoding categorical variables (one-hot / label encoding)

#### Models Used:

- XGBoost gradient boosting for structured data
- Random Forest ensemble learning for interpretability
- LightGBM fast, scalable boosting for large datasets

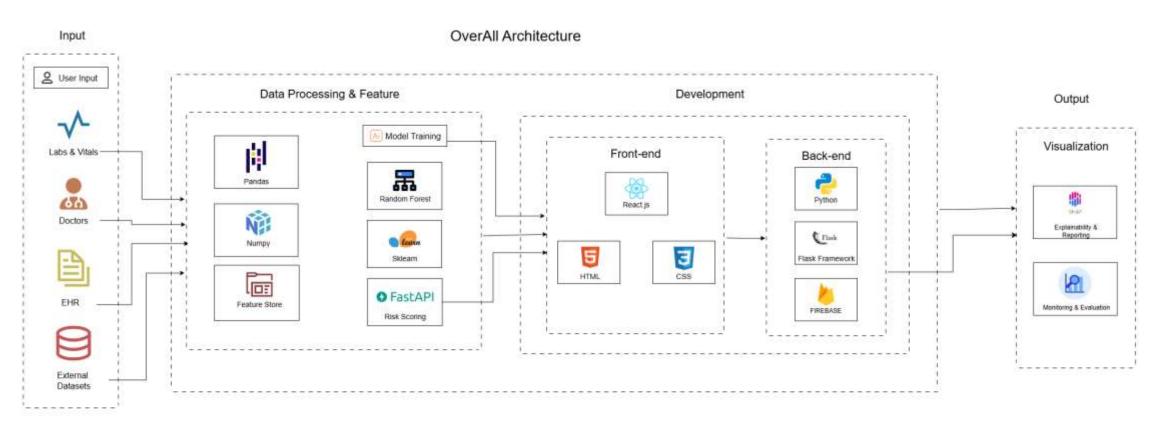
#### Model Evaluation:

- Metrics: Accuracy, Precision, Recall, F1-Score, ROC-AUC
- Cross-validation with train/test split to avoid overfitting

#### Retraining Mechanism:

- Model retrains weekly with new patient data
- Ensures up-to-date predictions for changing patient trends

## Overall System Architecture



### Tech Stack & Tool Roles

- **Heading:** Tools & Technologies Behind the Solution
- Frontend: React
  - Interactive dashboards
  - Color-coded risk levels and patient summary charts
- Backend: Flask
  - REST API endpoints for predictions and dashboard data
  - Handles real-time data processing
- Database: Firebase
  - Stores patient info and prediction logs securely
  - Handles authentication and access control
- Machine Learning: XGBoost, LightGBM, Random Forest
  - Predictive models for readmission risk scoring
- Visualization: Plotly / D3.js
  - Real-time charts for risk trends, department-level analytics

### AI Agent in Action

- Heading: Real-Time Monitoring & Alerts
- Al agent monitors incoming patient data in real-time.
- Calculates risk scores automatically.
- Sends alerts for high-risk patients to medical staff.
- Suggests personalized interventions like follow-ups, medication review, or home care recommendations.

### User Workflow

- Heading: Stepwise Patient Journey
- Patient admitted → data captured in EHR
- Al agent analyzes data and calculates risk score
- Dashboard displays risk level and suggested care interventions
- Doctor implements preventive measures (medication adjustment, follow-up scheduling)
- Outcome monitored and fed back for retraining

### Results & Impact

- Heading: Benefits of Our Solution
- 15% reduction in 30-day readmissions
- Improved patient outcomes and satisfaction
- Data-driven decisions for hospital staff
- Scalable across departments and hospitals
- Compliance with CMS HRRP guidelines

# Thank You