

Of course. I've created a separate, final document for the web app creation process.

Here is the complete documentation for building and testing the model, now including the final step of creating the web app interface.

Final Documentation: Building and Deploying the EMSync Severity Prediction Model

Introduction

- **Goal:** This document provides a complete walkthrough for creating, testing, and deploying a machine learning model that predicts patient severity (Critical, Moderate, Low Urgency).
 - **Final Model:** The final result is an XGBoost Classifier with a **71% Balanced Accuracy**.
 - **Prerequisites:**
 - Anaconda installed on your system.
 - The following MIMIC-IV-ED data files (as .csv.gz) in your project folder: edstays.csv.gz, triage.csv.gz, vitalsign.csv.gz.
-

Part 1: Building the Machine Learning Model

Step 1: Environment Setup

1. **Open the Anaconda Prompt.**
2. **Create the environment:**

```
Bash  
conda create --name emsync_env python=3.9
```
3. **Activate the environment:**

```
Bash
```

```
conda activate emsync_env
```

4. Install required libraries:

```
Bash
```

```
pip install pandas scikit-learn xgboost matplotlib seaborn jupyterlab pyarrow
```

Step 2: Data Preprocessing & Cleaning

1. Create preprocess.py and check_nan_irregulars.py with the code from our previous discussions.
2. Run the scripts in order to generate cleaned_dataset.csv:

```
Bash
```

```
python preprocess.py
```

```
python check_nan_irregulars.py
```

Step 3: Training the Final Model

1. Create the train_model.py script with the final, corrected code we developed.
2. Run the training pipeline. This will generate the best model file (e.g., best_model_20250912_153322.pkl) and the feature importance chart.

```
Bash
```

```
python train_model.py
```

3. **Secure the model:** Rename the best model file for permanent use.

```
Bash
```

```
ren best_model_20250912_153322.pkl xgboost_baseline_71bal_acc.pkl
```

Part 2: Deploying the Model with a Web App

Step 4: Install Streamlit

1. In your active Anaconda Prompt (emsync_env), install the web app library:

```
Bash
pip install streamlit
```

Step 5: Create the Web App Script

1. In your project folder, create a new file named app.py.
2. Copy and paste the entire code block below into this file.

```
Python
# app.py
import streamlit as st
import joblib
import pandas as pd
import numpy as np

# --- Page Configuration ---
st.set_page_config(
    page_title="EMsync Severity Predictor",
    page_icon="🚑",
    layout="wide"
)

# --- Load Model and Functions ---
def add_engineered_features(df):
    df = df.copy()
    sbp_safe = df["sbp"].replace(0, np.nan)
    df["shock_index"] = df["heartrate"] / sbp_safe
    df["pulse_pressure"] = df["sbp"] - df["dbp"]
    df["hypoxia_flag"] = (df["o2sat"] < 90).astype(int)
    df["fever_flag"] = (df["temperature"] >= 38).astype(int)
    return df

model_filename = 'xgboost_baseline_71bal_acc.pkl'
try:
    model = joblib.load(model_filename)
except FileNotFoundError:
    st.error(f"Model file '{model_filename}' not found. Please make sure it's in the same folder.")
    st.stop()
```

```

# --- Streamlit App Interface ---
st.title('EMsync: Real-Time Severity Prediction')
st.write("This tool uses a trained XGBoost model to predict patient severity based on their initial vital signs. Adjust the sliders in the sidebar to match the patient's vitals and click 'Predict'.")

# --- Input Vitals via Sidebar ---
st.sidebar.header('Patient Vitals Input')
temperature = st.sidebar.slider('Temperature (°C)', 35.0, 42.0, 36.8, 0.1)
heartrate = st.sidebar.slider('Heart Rate (bpm)', 40, 200, 80)
resprate = st.sidebar.slider('Respiration Rate (breaths/min)', 10, 40, 18)
o2sat = st.sidebar.slider('Oxygen Saturation (%)', 80, 100, 98)
sbp = st.sidebar.slider('Systolic Blood Pressure (mmHg)', 70, 180, 120)
dbp = st.sidebar.slider('Diastolic Blood Pressure (mmHg)', 40, 120, 80)
pain_median = st.sidebar.slider('Pain Score (0-10)', 0, 10, 2)

if st.sidebar.button('**Predict Severity**', use_container_width=True):
    # Prepare data for the model
    new_patient_data = {
        'temperature': temperature, 'heartrate': heartrate, 'resprate': resprate,
        'o2sat': o2sat, 'sbp': sbp, 'dbp': dbp, 'pain_median': pain_median,
        'heartrate_median': heartrate, 'heartrate_min': heartrate - 5,
        'heartrate_max': heartrate + 5, 'resprate_median': resprate,
        'o2sat_median': o2sat, 'o2sat_min': o2sat - 1,
        'temperature_median': temperature, 'sbp_median': sbp, 'dbp_median': dbp,
        'shock_index': np.nan, 'pulse_pressure': np.nan,
        'hypoxia_flag': np.nan, 'fever_flag': np.nan
    }
    input_df = pd.DataFrame([new_patient_data])
    input_df_engineered = add_engineered_features(input_df)
    original_features = [
        'temperature', 'heartrate', 'resprate', 'o2sat', 'sbp', 'dbp', 'heartrate_median',
        'heartrate_min', 'heartrate_max', 'resprate_median', 'o2sat_median', 'o2sat_min',
        'temperature_median', 'sbp_median', 'dbp_median', 'pain_median', 'shock_index',
        'pulse_pressure', 'hypoxia_flag', 'fever_flag'
    ]
    input_df_final = input_df_engineered[original_features]

    # Make and display prediction
    prediction_index = model.predict(input_df_final)[0]
    prediction_probabilities = model.predict_proba(input_df_final)[0]
    severity_map = {0: 'Critical', 1: 'Moderate', 2: 'Low Urgency'}
    predicted_severity = severity_map[prediction_index]

```

```
st.subheader('Prediction Result')
if predicted_severity == 'Critical':
    st.error(f'### Predicted Severity: **{predicted_severity}**')
elif predicted_severity == 'Moderate':
    st.warning(f'### Predicted Severity: **{predicted_severity}**')
else:
    st.success(f'### Predicted Severity: **{predicted_severity}**')

st.write('**Confidence Scores:**')
prob_df = pd.DataFrame({
    'Severity': ['Critical', 'Moderate', 'Low Urgency'],
    'Confidence': prediction_probabilities
})
st.bar_chart(prob_df.set_index('Severity'))
```

Step 6: Run Your Web App

1. **Save** the app.py file.
2. Go to your Anaconda Prompt and run this command:
Bash
streamlit run app.py
3. This will open a new tab in your web browser with the interactive application running.