

1. Import Libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report, confusion_matrix, roc_auc_score
import streamlit as st
import os
```

2. Load Data

```
DATA_PATH = "collegiate_athlete_injury_dataset.csv"
```

```
if not os.path.exists(DATA_PATH):
```

```
    st.error(f" Dataset not found: {DATA_PATH}")
```

```
else:
```

```
    df = pd.read_csv(DATA_PATH)
```

```
    df.columns = [col.strip() for col in df.columns] # Remove accidental spaces
```

#3. Data Cleaning

```
st.write(" Data Cleaning Summary")
```

```
st.write(df.isnull().sum()) # Show missing values
```

```
#Convert to categorical
```

```
for col in ['Gender', 'Position']:
```

```
    df[col] = df[col].astype('category')
```

```
# Encode categorical variables
```

```
le_gender = LabelEncoder()
```

```
le_position = LabelEncoder()
```

```
df['Gender_num'] = le_gender.fit_transform(df['Gender'])
```

```
df['Position_num'] = le_position.fit_transform(df['Position'])
```

4. Exploratory Data Analysis (EDA)

```
st.write("Injury Case Distribution")
```

```
fig1, ax1 = plt.subplots(figsize=(6,4))
```

```
sns.countplot(x='Injury_Indicator', data=df, ax=ax1)
```

```

ax1.set_title('Injury Case Distribution')
st.pyplot(fig1)
st.write("Fatigue by Injury Status")
fig2, ax2 = plt.subplots(figsize=(6,4))
sns.boxplot(x='Injury_Indicator', y='Fatigue_Score', data=df, ax=ax2)
ax2.set_title('Fatigue by Injury Status')
st.pyplot(fig2)

# 5. Feature Engineering

features = [
    'Age','Height_cm','Weight_kg','Training_Intensity','Training_Hours_Per_Week',
    'Recovery_Days_Per_Week','Match_Count_Per_Week','Rest_Between_Events_Days',
    'Fatigue_Score','Performance_Score','Team_Contribution_Score',
    'Load_Balance_Score','ACL_Risk_Score','Gender_num','Position_num'
]
X = df[features]
y = df['Injury_Indicator']
# Normalize features
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)

# 6. Train/Test Split
X_train, X_test, y_train, y_test = train_test_split(
    X_scaled, y, test_size=0.2, random_state=42, stratify=y
)

# 7. Model Building
model = RandomForestClassifier(n_estimators=100, random_state=42)
model.fit(X_train, y_train)
y_pred = model.predict(X_test)

# 8. Model Evaluation
st.write("Model Evaluation Results")
report = classification_report(y_test, y_pred, output_dict=True)
report_df = pd.DataFrame(report).transpose()
st.dataframe(report_df)
cm = confusion_matrix(y_test, y_pred)

```

```

st.write("**Confusion Matrix:**")

st.write(cm)

roc_score = roc_auc_score(y_test, model.predict_proba(X_test)[:, 1])

st.write(f"**ROC AUC Score:** {roc_score:.3f}")

# 9. Streamlit App - Prediction Section

st.write("Athlete Injury Risk Prediction")

uploaded_file = st.file_uploader(" Upload Athlete Data CSV", type="csv")

if uploaded_file is not None:

    input_df = pd.read_csv(uploaded_file)

    for col in ['Gender', 'Position']:

        input_df[col] = input_df[col].astype('category')

    # Apply same encoders

    input_df['Gender_num'] = le_gender.transform(input_df['Gender'])

    input_df['Position_num'] = le_position.transform(input_df['Position'])

    X_input = input_df[features]

    X_input_scaled = scaler.transform(X_input)

    risk_pred = model.predict(X_input_scaled)

    risk_prob = model.predict_proba(X_input_scaled)[:, 1]

    input_df['Injury_Risk_Prediction'] = risk_pred

    input_df['Injury_Risk_Probability'] = risk_prob

    st.write("Prediction Results")

    st.dataframe(input_df)

    st.success(" Predictions generated successfully!")

st.info("This tool uses a Random Forest model trained on athlete workload and health metrics.")

```