



# **Model Building**

#### 1.Train-Test Split

Split the dataset to evaluate model performance.
from sklearn.model\_selection import train\_test\_split

X = df.drop('Cirrhosis', axis=1)
y = df['Cirrhosis']

X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y, test\_size=0.2, random\_state=42)

#### 2. Model Selection

Try multiple algorithms and compare:

- Logistic Regression
- Random Forest
- **XGBoost**
- Support Vector Machine (SVM)
- K-Nearest Neighbors (KNN)

#### 3. Training Models

### **Example: Random Forest**

from sklearn.ensemble import RandomForestClassifier rf = RandomForestClassifier(n\_estimators=100, random\_state=42) rf.fit(X\_train, y\_train)

# **Example: XGBoost**

import xgboost as xgb

xgb\_model = xgb.XGBClassifier(use\_label\_encoder=False, eval\_metric='logloss')
xgb\_model.fit(X\_train, y\_train)

#### 4. Model Evaluation

Use metrics like:

- Accuracy
- Precision, Recall, F1-Score
- Confusion Matrix
- ROC-AUC Score

from sklearn.metrics import classification\_report, confusion\_matrix, roc\_auc\_score y\_pred = rf.predict(X\_test)
print(classification\_report(y\_test, y\_pred))
print(confusion\_matrix(y\_test, y\_pred))
print("ROC AUC Score:", roc\_auc\_score(y\_test, rf.predict\_proba(X\_test)[:,1]))

## **5.** Hyperparameter Tuning (Optional but Recommended)

```
from sklearn.model_selection import GridSearchCV
param_grid = {
    'n_estimators': [50, 100, 200],
    'max_depth': [4, 6, 8]
}
grid = GridSearchCV(RandomForestClassifier(), param_grid, cv=5)
grid.fit(X_train, y_train)
```

## **6. Model Comparison Table (Example Output)**

Model	Accuracy	ROC AUC	F1-Score
LogisticRegression	82%	0.85	0.79
RandomForest	89%	0.91	0.86
XGBoost	90%	0.93	0.88
SVM	85%	0.87	0.82

# Final Step: Model Selection & Save Best Model

import joblib

joblib.dump(xgb\_model, 'best\_model\_liver.pkl')