

Perfect! Since you've downloaded the **ECG5000 dataset**, I'll walk you through the **entire end-to-end process** to build a **machine learning model** that classifies ECG signals.

We'll go step-by-step:

Project Goal

Classify ECG heartbeats as **normal or abnormal** using the ECG5000 dataset and **machine learning**.

Step-by-Step Workflow:

1. Import Libraries
 2. Load Dataset
 3. Explore & Preprocess Data
 4. Train-Test Split
 5. Train ML Model (e.g., RandomForest)
 6. Evaluate the Model
 7. Visualize Results (optional)
-

Step 1: Import Libraries

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

Step 2: Load Dataset

If it's a .txt file like ECG5000_TRAIN.txt and ECG5000_TEST.txt:

```
# Load both train and test sets  
train_data = pd.read_csv("ECG5000_TRAIN.txt", delim_whitespace=True, header=None)
```

```
test_data = pd.read_csv("ECG5000_TEST.txt", delim_whitespace=True, header=None)
```

```
# Combine them
```

```
df = pd.concat([train_data, test_data], ignore_index=True)
```

```
If you already have a merged ECG5000.csv:
```

```
df = pd.read_csv("ECG5000.csv", header=None)
```

Step 3: Explore & Preprocess

```
# Check shape and preview
```

```
print(df.shape)
```

```
print(df.head())
```

```
# First column is the label
```

```
X = df.iloc[:, 1:] # features
```

```
y = df.iloc[:, 0] # labels
```

```
# For binary classification (normal vs abnormal), convert labels
```

```
# Label 1: Normal | Label 2-5: Abnormal
```

```
y_binary = y.apply(lambda x: 0 if x == 1 else 1) # 0: normal, 1: abnormal
```

Step 4: Train-Test Split

```
X_train, X_test, y_train, y_test = train_test_split(
```

```
    X, y_binary, test_size=0.2, random_state=42, stratify=y_binary
```

```
)
```

```
# Normalize the features
```

```
scaler = StandardScaler()
```

```
X_train_scaled = scaler.fit_transform(X_train)
```

```
X_test_scaled = scaler.transform(X_test)
```

Step 5: Train ML Model

```
model = RandomForestClassifier(n_estimators=100, random_state=42)  
model.fit(X_train_scaled, y_train)
```

Step 6: Evaluate the Model

```
y_pred = model.predict(X_test_scaled)
```

```
print("Classification Report:")  
print(classification_report(y_test, y_pred))
```

```
# Confusion Matrix
```

```
cm = confusion_matrix(y_test, y_pred)  
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues')  
plt.title('Confusion Matrix')  
plt.xlabel('Predicted')  
plt.ylabel('True')  
plt.show()
```

Optional: Deep Learning Version (LSTM or CNN)

Let me know if you want the same project done using **Keras (TensorFlow)** with an **LSTM** or **1D CNN**, which works better for time series like ECG.

Output

You will get:

- Classification report with precision, recall, F1-score
 - Confusion matrix heatmap
 - A trained model that predicts normal vs abnormal heartbeats
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Would you like:

- This project in a **.ipynb Jupyter notebook?**
- Or deployed as a **Flask web app?**
Let me know — I can generate that for you too.

