

## ARRHYTHMIA DETECTION USING ECG SIGNALS

```
import os
import wfdb
import numpy as np
import matplotlib.pyplot as plt
import pywt
from scipy.signal import butter, filtfilt, resample
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout

# Configuration
DATA_DIRS = ["mitdb"]
FS_TARGET = 360
WINDOW = 216
SCALES = np.arange(1, 64)
WAVELET = 'morl'

# Bandpass filter
def bandpass_filter(signal, lowcut=0.5, highcut=40.0, fs=360, order=4):
    nyq = 0.5 * fs
    b, a = butter(order, [lowcut / nyq, highcut / nyq], btype='band')
    return filtfilt(b, a, signal)

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```

```

# Label mapping
label_map = {
    'N': 'Normal', 'L': 'Normal', 'R': 'Normal',
    'V': 'Ventricular', 'A': 'Atrial', 'F': 'Fusion', 'E': 'Escape'
}

X_scalograms = []
y_labels = []
plotted = False # Only plot one raw vs filtered and one scalogram

# Load and preprocess
for data_dir in DATA_DIRS:
    if not os.path.isdir(data_dir):
        print(f"Directory not found: {data_dir}")
        continue

    for file in os.listdir(data_dir):
        if file.endswith('.dat'):
            record_name = file.replace('.dat', '')
            print(f"Processing: {record_name}")
            try:
                record = wfdb.rdrecord(os.path.join(data_dir, record_name))
                annotation = wfdb.rdann(os.path.join(data_dir, record_name), 'atr')

                signal = record.p_signal[:, 0]
                fs = record.fs
                if fs != FS_TARGET:
                    signal = resample(signal, int(len(signal) * FS_TARGET / fs))
                    r_peaks = (annotation.sample * FS_TARGET / fs).astype(int)
                else:
                    r_peaks = annotation.sample

```

```
symbols = annotation.symbol
```

```
filtered = bandpass_filter(signal, fs=FS_TARGET)
```

```
for r, sym in zip(r_peaks, symbols):
```

```
    if r > 108 and r + 108 < len(signal) and sym in label_map:
```

```
        raw = signal[r - 108:r + 108]
```

```
        filt = filtered[r - 108:r + 108]
```

```
    if not plotted:
```

```
        # Plot raw vs filtered once
```

```
        plt.figure(figsize=(10, 4))
```

```
        plt.plot(raw, label='Raw')
```

```
        plt.plot(filt, label='Filtered')
```

```
        plt.title('Raw vs Filtered ECG Beat')
```

```
        plt.legend()
```

```
        plt.tight_layout()
```

```
        plt.show()
```

```
coeffs, _ = pywt.cwt(filt, SCALES, WAVELET, 1 / FS_TARGET)
```

```
scalogram = np.abs(coeffs)
```

```
X_scalograms.append(scalogram)
```

```
y_labels.append(label_map[sym])
```

```
if not plotted:
```

```
    # Plot scalogram once
```

```
    plt.figure(figsize=(8, 4))
```

```
    plt.imshow(scalogram, extent=[0, WINDOW / FS_TARGET, 1, len(SCALES)],
```

```
               aspect='auto', cmap='jet', origin='lower')
```

```
    plt.title('Scalogram of ECG Beat')
```

```
    plt.xlabel('Time (s)')
```

```
    plt.ylabel('Scale')
```

```

plt.colorbar(label='Magnitude')

plt.tight_layout()

plt.show()

plotted = True

except Exception as e:
    print(f"Skipping {record_name}: {e}")

# Ensure data is collected
if not X_scalograms:
    print("No ECG beats were processed. Check your dataset.")
    exit()

# Convert and encode
X = np.array(X_scalograms)[..., np.newaxis]
y = np.array(y_labels)
le = LabelEncoder()
y_encoded = le.fit_transform(y)
y_cat = to_categorical(y_encoded)

# Split
X_train, X_test, y_train, y_test = train_test_split(X, y_cat, test_size=0.2, random_state=42)

# CNN model
model = Sequential([
    Conv2D(32, (3, 3), activation='relu', input_shape=X.shape[1:]),
    MaxPooling2D((2, 2)),
    Dropout(0.25),

    Conv2D(64, (3, 3), activation='relu'),
    MaxPooling2D((2, 2)),
    Dropout(0.25),

```

```
    Flatten(),
    Dense(128, activation='relu'),
    Dropout(0.5),
    Dense(y_cat.shape[1], activation='softmax')
])
```

```
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()
```

```
# Train
```

```
history = model.fit(X_train, y_train, epochs=10, batch_size=64, validation_data=(X_test, y_test))
```

```
# Evaluate
```

```
loss, acc = model.evaluate(X_test, y_test)
print(f"\nTest Accuracy: {acc*100:.2f}%")
```

```
# Plot training curves
```

```
plt.figure(figsize=(10, 4))
```

```
plt.subplot(1, 2, 1)
```

```
plt.plot(history.history['accuracy'], label='Train')
```

```
plt.plot(history.history['val_accuracy'], label='Val')
```

```
plt.title('Model Accuracy')
```

```
plt.xlabel('Epoch')
```

```
plt.ylabel('Accuracy')
```

```
plt.legend()
```

```
plt.subplot(1, 2, 2)
```

```
plt.plot(history.history['loss'], label='Train')
```

```
plt.plot(history.history['val_loss'], label='Val')
```

```
plt.title('Model Loss')
```

```
plt.xlabel('Epoch')
```

```
plt.ylabel('Loss')
```

```
plt.legend()
```

```
plt.tight_layout()
```

```
plt.show()
```