```
In [1]:
         | import os
            import numpy as np
            from scipy.io import loadmat
            from scipy.signal import spectrogram
            from sklearn.model_selection import train_test_split
            import tensorflow as tf
            from tensorflow.keras.models import Sequential
            from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dens
            import matplotlib.pyplot as plt
        # Function to load ECG data from the folders and convert to spectrogram
In [2]:
            def load ecg data from folders(root folder, nperseg=256, noverlap=128):
                all data = []
                all labels = []
                # Traverse through the root folder and subfolders
                for subdir, dirs, files in os.walk(root_folder):
                    print(f"Checking folder: {subdir}")
                    for file in files:
                        if file.endswith('.mat'):
                            file_path = os.path.join(subdir, file)
                            print(f"Loading MATLAB file: {file_path}")
                            try:
                                # Load the MATLAB file
                                mat_data = loadmat(file_path)
                                print(f"Keys in MATLAB file: {mat_data.keys()}")
                                # Assuming 'val' contains ECG data
                                if 'val' in mat_data:
                                    ecg_data = mat_data['val'].squeeze() # Squeeze
                                    # Convert ECG data to spectrogram
                                    f, t, Sxx = spectrogram(ecg_data, nperseg=npers
                                    Sxx = np.log(Sxx + 1e-8) # Apply Log transform
                                    # Resize spectrogram to a fixed size
                                    desired_size = (128, 128) # Example size, adju
                                    Sxx_resized = np.resize(Sxx, desired_size)
                                    all_data.append(Sxx_resized)
                                    all labels.append(np.random.randint(2)) # Plac
                                else:
                                    print(f"'val' key not found in {file_path}")
                            except Exception as e:
                                print(f"Error loading MATLAB file {file path}: {e}"
                return np.array(all data), np.array(all labels)
                                                                                  \triangleright
In [3]:
            # Root folder path where your .mat data is stored
            root_folder = r'C:\Users\diyu2\OneDrive - AUT University\AUT YEAR 4\IND
```

```
In [4]:
         # Load the ECG data from folders
            data, labels = load_ecg_data_from_folders(root_folder)
            or-arrhythmia\WFDBRecords\34\348\JS34254.mat
            Keys in MATLAB file: dict keys(['val'])
            Loading MATLAB file: C:\Users\diyu2\OneDrive - AUT University\AUT Y
            EAR 4\INDUSTRIAL PROJECT (Mechanical)\Reports Part B\ECG database-f
            or-arrhythmia\WFDBRecords\34\348\JS34255.mat
            Keys in MATLAB file: dict keys(['val'])
            Loading MATLAB file: C:\Users\diyu2\OneDrive - AUT University\AUT Y
            EAR 4\INDUSTRIAL PROJECT (Mechanical)\Reports Part B\ECG database-f
            or-arrhythmia\WFDBRecords\34\348\JS34256.mat
            Keys in MATLAB file: dict_keys(['val'])
            Loading MATLAB file: C:\Users\diyu2\OneDrive - AUT University\AUT Y
            EAR 4\INDUSTRIAL PROJECT (Mechanical)\Reports Part B\ECG database-f
            or-arrhythmia\WFDBRecords\34\348\JS34257.mat
            Keys in MATLAB file: dict_keys(['val'])
            Loading MATLAB file: C:\Users\diyu2\OneDrive - AUT University\AUT Y
            EAR 4\INDUSTRIAL PROJECT (Mechanical)\Reports Part B\ECG database-f
            or-arrhythmia\WFDBRecords\34\348\JS34258.mat
            Keys in MATLAB file: dict_keys(['val'])
            Loading MATLAB file: C:\Users\diyu2\OneDrive - AUT University\AUT Y
            FAR 4\TNDUSTRIAL PROJECT (Mechanical)\Renorts Part B\FCG database-f
In [5]:
         # Check the shape of the data
            print(f"Original shape of data: {data.shape}")
```

Original shape of data: (45152, 128, 128)

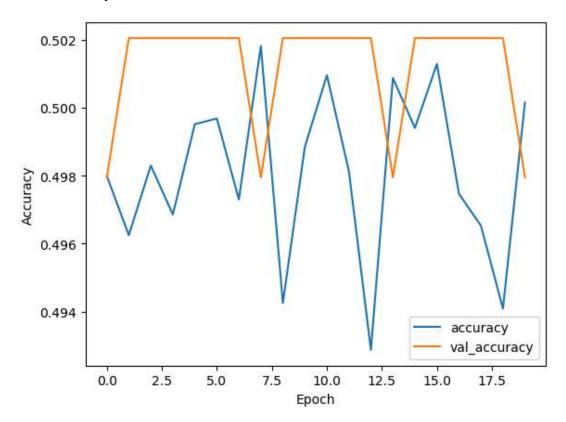
```
In [6]: ▶ # Ensure data is Loaded before proceeding
            if data.shape[0] > 0:
                # Normalize data to [0, 1]
                data = (data - np.min(data)) / (np.max(data) - np.min(data))
                # Expand dimensions for CNN input (add channel dimension)
                data = np.expand_dims(data, axis=-1)
                # Split the data into training and testing sets
                X train, X test, y train, y test = train test split(data, labels, t
                # Build the CNN model
                model = Sequential()
                # 2D CNN layers for processing spectrogram images
                model.add(Conv2D(filters=32, kernel_size=(3, 3), activation='relu',
                model.add(MaxPooling2D(pool size=(2, 2)))
                model.add(Conv2D(filters=64, kernel_size=(3, 3), activation='relu')
                model.add(MaxPooling2D(pool size=(2, 2)))
                model.add(Conv2D(filters=128, kernel_size=(3, 3), activation='relu'
                model.add(MaxPooling2D(pool_size=(2, 2)))
                model.add(Flatten())
                # Fully connected layers
                model.add(Dense(128, activation='relu'))
                model.add(Dense(64, activation='relu'))
                model.add(Dense(1, activation='sigmoid')) # Assuming binary classi
                # Compile the model
                model.compile(optimizer='adam', loss='binary_crossentropy', metrics
                # Train the model
                history = model.fit(X_train, y_train, epochs=20, validation_data=(X
                # Evaluate the model on the test data
                test_loss, test_acc = model.evaluate(X_test, y_test)
                print(f'Test Accuracy: {test_acc}')
                # Plotting the training and validation accuracy
                plt.plot(history.history['accuracy'], label='accuracy')
                plt.plot(history.history['val_accuracy'], label='val_accuracy')
                plt.xlabel('Epoch')
                plt.ylabel('Accuracy')
                plt.legend(loc='lower right')
                plt.show()
                # Save the trained model
                model.save('ecg cnn model.h5')
            else:
                print("No data was loaded, exiting.")
```

C:\Users\diyu2\anaconda3\Lib\site-packages\keras\src\layers\convolutio
nal\base_conv.py:107: UserWarning: Do not pass an `input_shape`/`input
_dim` argument to a layer. When using Sequential models, prefer using
an `Input(shape)` object as the first layer in the model instead.
 super().__init__(activity_regularizer=activity_regularizer, **kwarg
s)

```
Epoch 1/20
           230s 200ms/step - accuracy: 0.4977 - lo
1129/1129 -
ss: 0.6935 - val accuracy: 0.4980 - val loss: 0.6932
Epoch 2/20
1129/1129 -
                        253s 224ms/step - accuracy: 0.4989 - lo
ss: 0.6932 - val accuracy: 0.5020 - val loss: 0.6931
Epoch 3/20
1129/1129 -
                        264s 233ms/step - accuracy: 0.4952 - lo
ss: 0.6932 - val_accuracy: 0.5020 - val_loss: 0.6931
Epoch 4/20
1129/1129 -
                   320s 283ms/step - accuracy: 0.4915 - lo
ss: 0.6932 - val accuracy: 0.5020 - val loss: 0.6931
Epoch 5/20
1129/1129 -
            271s 240ms/step - accuracy: 0.5013 - lo
ss: 0.6932 - val_accuracy: 0.5020 - val_loss: 0.6931
Epoch 6/20
1129/1129 -
                         329s 292ms/step - accuracy: 0.4981 - lo
ss: 0.6932 - val_accuracy: 0.5020 - val_loss: 0.6931
Epoch 7/20
                     321s 284ms/step - accuracy: 0.4966 - lo
1129/1129 -
ss: 0.6932 - val_accuracy: 0.5020 - val_loss: 0.6931
Epoch 8/20
           264s 234ms/step - accuracy: 0.5061 - lo
1129/1129 -
ss: 0.6931 - val_accuracy: 0.4980 - val_loss: 0.6932
Epoch 9/20
1129/1129 -
                    290s 257ms/step - accuracy: 0.4915 - lo
ss: 0.6932 - val_accuracy: 0.5020 - val_loss: 0.6931
Epoch 10/20
                      286s 253ms/step - accuracy: 0.5004 - lo
1129/1129 -
ss: 0.6932 - val_accuracy: 0.5020 - val_loss: 0.6931
Epoch 11/20
                285s 252ms/step - accuracy: 0.5038 - lo
1129/1129 -
ss: 0.6932 - val_accuracy: 0.5020 - val_loss: 0.6931
Epoch 12/20
1129/1129 —
              288s 255ms/step - accuracy: 0.4960 - lo
ss: 0.6932 - val_accuracy: 0.5020 - val_loss: 0.6931
Epoch 13/20
                    284s 252ms/step - accuracy: 0.4933 - lo
1129/1129 -
ss: 0.6932 - val_accuracy: 0.5020 - val_loss: 0.6931
Epoch 14/20
1129/1129 -
                          - 274s 242ms/step - accuracy: 0.5054 - lo
ss: 0.6931 - val_accuracy: 0.4980 - val_loss: 0.6932
Epoch 15/20
1129/1129 277s 245ms/step - accuracy: 0.4970 - lo
ss: 0.6932 - val_accuracy: 0.5020 - val_loss: 0.6931
Epoch 16/20
                    244s 216ms/step - accuracy: 0.5039 - lo
1129/1129 -
ss: 0.6931 - val accuracy: 0.5020 - val loss: 0.6931
Epoch 17/20
                    254s 225ms/step - accuracy: 0.4956 - lo
1129/1129 ————
ss: 0.6932 - val accuracy: 0.5020 - val loss: 0.6932
Epoch 18/20
                    258s 228ms/step - accuracy: 0.5007 - lo
1129/1129 —
ss: 0.6932 - val accuracy: 0.5020 - val loss: 0.6931
Epoch 19/20
            245s 217ms/step - accuracy: 0.4951 - lo
1129/1129 -
ss: 0.6932 - val_accuracy: 0.5020 - val_loss: 0.6931
Epoch 20/20
1129/1129 -
                     249s 220ms/step - accuracy: 0.5040 - lo
ss: 0.6931 - val accuracy: 0.4980 - val loss: 0.6932
283/283 —
           18s 64ms/step - accuracy: 0.4968 - loss:
```

0.6932

Test Accuracy: 0.4979515075683594



WARNING:absl:You are saving your model as an HDF5 file via `model.save ()` or `keras.saving.save_model(model)`. This file format is considere d legacy. We recommend using instead the native Keras format, e.g. `mo del.save('my_model.keras')` or `keras.saving.save_model(model, 'my_mod el.keras')`.