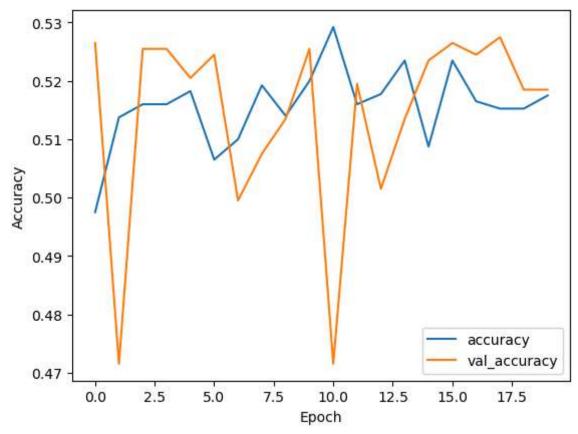
```
In [3]: import numpy as np
        import cv2
        from scipy import signal
        import tensorflow as tf
        from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense
        from sklearn.model selection import train test split
        import matplotlib.pyplot as plt
        def convert_to_spectrogram(data, fs=1000, nperseg=256):
            Convert ECG time-series data to spectrogram images.
            :param data: Raw ECG data as a list of numpy arrays.
            :param fs: Sampling frequency of the ECG data.
            :param nperseg: Length of each segment for the spectrogram.
            :return: Spectrogram images as a numpy array.
            spectrogram images = []
            for signal_data in data:
                if len(signal_data) < nperseg:</pre>
                    # Pad the signal if it's shorter than nperseg
                    signal data = np.pad(signal data, (0, nperseg - len(signal data)),
                f, t, Sxx = signal.spectrogram(signal_data, fs=fs, nperseg=nperseg)
                # Normalize the spectrogram
                Sxx = np.log(Sxx + 1e-8)
                # Resize spectrogram to fixed size
                img = cv2.resize(Sxx, (128, 128))
                spectrogram_images.append(img)
            return np.array(spectrogram_images)
        # Load the ECG data from the text file
        file_path = r'C:\Users\diyu2\OneDrive - AUT University\AUT YEAR 4\INDUSTRIAL PRO
        # Reading the file and converting the data into an array
        with open(file_path, 'r') as file:
            lines = file.readlines()
        # Process only numeric lines
        data = []
        for line in lines:
            trv:
                # Split the line into numbers and convert to integers
                numbers = list(map(int, line.strip().split()))
                data.append(numbers)
            except ValueError:
                # Ignore lines that can't be converted to integers
                continue
        # Convert the ECG data to spectrogram images
        spectrogram_images = convert_to_spectrogram(data)
        # Check the shape of the images and normalize pixel values
        spectrogram images = spectrogram images.reshape((spectrogram images.shape[0], 12
        spectrogram_images = spectrogram_images.astype('float32') / np.max(spectrogram_i
        # Generate Labels (for simplicity, using random Labels for now)
        labels = np.random.randint(2, size=spectrogram_images.shape[0])
```

```
# Split the data into training and testing sets
 X_train, X_test, y_train, y_test = train_test_split(spectrogram_images, labels,
 # Build the CNN model for image data
 model = Sequential()
 model.add(Conv2D(filters=64, kernel_size=(3, 3), activation='relu', input_shape=
 model.add(MaxPooling2D(pool_size=(2, 2)))
 model.add(Flatten())
 model.add(Dense(128, activation='relu'))
 model.add(Dense(64, activation='relu'))
 model.add(Dense(1, activation='sigmoid')) # Assuming binary classification
 # Compile the model
 model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy']
 # Train the model
 history = model.fit(X_train, y_train, epochs=20, validation_data=(X_test, y_test
 # 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')
C:\Users\diyu2\anaconda3\Lib\site-packages\keras\src\layers\convolutional\base_co
nv.py:107: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a la
yer. When using Sequential models, prefer using an `Input(shape)` object as the f
irst layer in the model instead.
  super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

```
Epoch 1/20
            50s 348ms/step - accuracy: 0.4871 - loss: 3.5732 - v
126/126 ——
al_accuracy: 0.5265 - val_loss: 0.7074
Epoch 2/20
126/126 ---
                    44s 350ms/step - accuracy: 0.5315 - loss: 0.6950 - v
al_accuracy: 0.4715 - val_loss: 0.6945
Epoch 3/20
                         - 41s 327ms/step - accuracy: 0.5191 - loss: 0.6915 - v
126/126 -
al_accuracy: 0.5255 - val_loss: 0.6927
Epoch 4/20
                     40s 319ms/step - accuracy: 0.5189 - loss: 0.6928 - v
126/126 -
al accuracy: 0.5255 - val loss: 0.6931
Epoch 5/20
              40s 318ms/step - accuracy: 0.5262 - loss: 0.6918 - v
126/126 —
al accuracy: 0.5205 - val loss: 0.6917
Epoch 6/20
                        - 43s 340ms/step - accuracy: 0.5157 - loss: 0.6906 - v
126/126 -
al accuracy: 0.5245 - val loss: 0.6926
Epoch 7/20
                        — 43s 341ms/step - accuracy: 0.5036 - loss: 0.6952 - v
126/126 -
al_accuracy: 0.4995 - val_loss: 0.6934
Epoch 8/20
126/126 -
              al_accuracy: 0.5075 - val_loss: 0.6935
Epoch 9/20
                        — 43s 341ms/step - accuracy: 0.5276 - loss: 0.6911 - v
al_accuracy: 0.5135 - val_loss: 0.6926
Epoch 10/20
                 41s 325ms/step - accuracy: 0.5358 - loss: 0.6922 - v
126/126 -
al accuracy: 0.5255 - val loss: 0.6931
Epoch 11/20
126/126 -
                    40s 320ms/step - accuracy: 0.5349 - loss: 0.6903 - v
al accuracy: 0.4715 - val loss: 0.6988
Epoch 12/20
126/126 40s 314ms/step - accuracy: 0.5241 - loss: 0.6954 - v
al_accuracy: 0.5195 - val_loss: 0.6929
Epoch 13/20
                        — 40s 317ms/step - accuracy: 0.5212 - loss: 0.6923 - v
126/126 -
al_accuracy: 0.5015 - val_loss: 0.6933
Epoch 14/20
                         - 49s 392ms/step - accuracy: 0.5267 - loss: 0.6941 - v
126/126 -
al accuracy: 0.5135 - val loss: 0.6948
Epoch 15/20
                52s 411ms/step - accuracy: 0.5163 - loss: 0.6912 - v
126/126 —
al_accuracy: 0.5235 - val loss: 0.6920
Epoch 16/20
126/126 —
                        — 47s 373ms/step - accuracy: 0.5196 - loss: 0.6921 - v
al accuracy: 0.5265 - val loss: 0.6934
Epoch 17/20
                     42s 334ms/step - accuracy: 0.5149 - loss: 0.6927 - v
126/126 -
al_accuracy: 0.5245 - val_loss: 0.6926
Epoch 18/20
                        — 41s 323ms/step - accuracy: 0.5198 - loss: 0.6899 - v
126/126 -
al accuracy: 0.5275 - val loss: 0.6930
Epoch 19/20
                   40s 315ms/step - accuracy: 0.5270 - loss: 0.6891 - v
126/126 ——
al_accuracy: 0.5185 - val_loss: 0.6920
Epoch 20/20
                40s 317ms/step - accuracy: 0.5248 - loss: 0.6892 - v
126/126 —
al accuracy: 0.5185 - val loss: 0.6925
```

32/32 2s 49ms/step - accuracy: 0.5315 - loss: 0.6928

Test Accuracy: 0.5184814929962158



WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `ker as.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `ke ras.saving.save_model(model, 'my_model.keras')`.

In []: