Mit_Aug

November 17, 2024

```
[1]: import numpy as np
     import pandas as pd
     import tensorflow as tf
     import os
     import os
     import itertools
     import numpy as np
     import pandas as pd
     import seaborn as sns
     import tensorflow as tf
     from itertools import cycle
     from sklearn import metrics
     import matplotlib.pyplot as plt
     from sklearn.metrics import confusion_matrix, classification_report
     from sklearn.model_selection import cross_val_score
     import keras
     from keras.utils import np_utils
     from keras.models import Sequential,load_model
     from keras.utils.vis_utils import plot_model
     from keras.layers import Conv1D, MaxPooling1D, Dense, Dropout, Flatten, U
      →Reshape, GlobalAveragePooling1D
     import random
     from scipy.interpolate import CubicSpline
     %matplotlib inline
     tf.__version__
     #from keras.models import load_model
```

[1]: '2.10.0'

```
print(test_y)
            return test_x,test_y
[3]: def load_train():
            #Test ori data
            train_x = np.load(r"D:\Research Work\Thesis\New Data\Data\MIT_
        Processed\Train Dataset\Original\ecg_ori_all_reshaped_tr_segments.npy")
            train_y = np.load(r"D:\Research Work\Thesis\New Data\Data\MIT_
        →Processed\Train Dataset\Original\ecg_ori_all_tr_labels.npy")
            print(train_x.shape)
            print(train_y)
            return train_x, train_y
[4]: train x, train y = load train()
      test_x,test_y = load_test()
      print(train x.shape,train y.shape,test x.shape,test y.shape)
      (460, 1800, 1)
      [1\ 3\ 0\ 0\ 3\ 2\ 2\ 1\ 2\ 0\ 3\ 0\ 2\ 1\ 3\ 3\ 2\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 2\ 0\ 0\ 0\ 2\ 3\ 0\ 0\ 1\ 1\ 0\ 0
       1\ 1\ 0\ 1\ 3\ 3\ 2\ 2\ 3\ 3\ 2\ 0\ 0\ 0\ 1\ 0\ 3\ 2\ 1\ 2\ 2\ 1\ 2\ 1\ 3\ 0\ 1\ 2\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 1\ 3
       \begin{smallmatrix} 0 & 0 & 0 & 1 & 0 & 0 & 0 & 3 & 1 & 1 & 1 & 2 & 1 & 1 & 2 & 0 & 0 & 0 & 1 & 0 & 3 & 0 & 0 & 3 & 2 & 0 & 1 & 3 & 3 & 2 & 1 & 0 & 0 & 0 & 2 \\ \end{smallmatrix}
       \begin{smallmatrix} 2 & 3 & 1 & 2 & 2 & 2 & 0 & 2 & 1 & 0 & 1 & 0 & 2 & 1 & 1 & 0 & 3 & 3 & 3 & 0 & 2 & 0 & 2 & 2 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 2 & 0 & 2 & 0 & 0 & 3 \end{smallmatrix}
       1 \; 1 \; 3 \; 0 \; 1 \; 1 \; 0 \; 0 \; 0 \; 3 \; 0 \; 0 \; 1 \; 0 \; 0 \; 0 \; 2 \; 3 \; 0 \; 2 \; 0 \; 2 \; 0 \; 3 \; 2 \; 1 \; 0 \; 1 \; 0 \; 3 \; 3 \; 0 \; 0 \; 0 \; 2 \; 2 \; 0
       \begin{smallmatrix} 0 & 3 & 1 & 0 & 1 & 1 & 0 & 0 & 0 & 2 & 0 & 2 & 2 & 2 & 0 & 3 & 3 & 0 & 1 & 0 & 3 & 0 & 0 & 2 & 0 & 1 & 3 & 2 & 1 & 0 & 1 & 3 & 3 & 2 & 0 & 2 & 1 & 2 \\ \end{smallmatrix}
       2\;0\;0\;0\;2\;0\;2\;1\;2\;0\;0\;1\;2\;3\;0\;2\;0\;2\;3\;1\;0\;3\;2\;3\;1\;0\;3\;1\;0\;1\;0\;3\;0\;0\;1\;2\;1
       3\ 1\ 0\ 0\ 2\ 3\ 0\ 1\ 2\ 2\ 3\ 0\ 3\ 0\ 2\ 0\ 0\ 0\ 0\ 1\ 0\ 1\ 1\ 0\ 2\ 0\ 2\ 0\ 3\ 2\ 0\ 1\ 1\ 0\ 1\ 0
       3\ 2\ 1\ 3\ 0\ 1\ 0\ 0\ 0\ 0\ 3\ 3\ 0\ 0\ 2\ 0\ 2\ 1\ 0\ 0\ 2\ 3\ 3\ 2\ 2\ 2\ 2\ 0\ 2\ 0\ 1\ 2\ 0\ 3\ 2\ 0\ 0
       \begin{smallmatrix} 0 & 0 & 2 & 0 & 1 & 0 & 3 & 3 & 0 & 1 & 2 & 0 & 0 & 0 & 3 & 0 & 0 & 2 & 0 & 1 & 2 & 3 & 2 & 2 & 0 & 2 & 0 & 1 & 0 & 3 & 3 & 0 & 3 & 0 & 2 & 2 \\ \end{smallmatrix}
       1 2 0 0 3 1 3 0 3 1 3 0 3 0 3 2]
      (192, 1800, 1)
      [1 2 3 0 1 0 2 1 0 0 2 2 3 1 1 0 0 0 0 1 1 0 2 0 0 0 2 1 0 0 0 1 0 2 3 0 0
       \begin{smallmatrix} 0 & 3 & 0 & 2 & 3 & 1 & 0 & 2 & 2 & 3 & 0 & 1 & 0 & 0 & 0 & 1 & 2 & 0 & 3 & 2 & 1 & 3 & 0 & 0 & 0 & 2 & 2 & 3 & 2 & 1 & 0 & 0 & 3 & 3 & 0 & 0 & 1 \end{smallmatrix}
       \begin{smallmatrix} 0 & 0 & 3 & 0 & 2 & 1 & 3 & 0 & 0 & 3 & 1 & 3 & 2 & 1 & 1 & 0 & 0 & 0 & 2 & 2 & 2 & 3 & 0 & 0 & 1 & 0 & 2 & 0 & 0 & 3 & 0 & 0 & 2 & 2 & 1 & 2 \\ \end{smallmatrix}
       2\ 1\ 3\ 1\ 0\ 3\ 0\ 0\ 0\ 0\ 2\ 3\ 0\ 3\ 1\ 2\ 3\ 1\ 1\ 0\ 0\ 2\ 0\ 0\ 2\ 0\ 1\ 3\ 3\ 0\ 0\ 3\ 2\ 3\ 0\ 1
       3 0 1 1 0 2 2]
      (460, 1800, 1) (460,) (192, 1800, 1) (192,)
[5]: from collections import Counter
      c = Counter(train_y)
      print(c)
      Counter({0: 198, 1: 95, 2: 94, 3: 73})
```

```
[6]: #Different Augmentation Functions
     def no_augmentation(x):
         return x
     # Jittering
     def jitter(x, sigma=0.05):
         return x + np.random.normal(loc=0, scale=sigma, size=x.shape)
     # Scaling Has issue with loading into memory.
     def scale(x, sigma=0.5):
         x \text{ scaled} = np.copy(x)
         for i in range(x_scaled.shape[0]):
             factor = np.random.normal(loc=1.0, scale=sigma)
             x_scaled[i] *= factor
         return x_scaled
     # Flipping
     def flip(x):
         # Check if data is 2D with a single feature
         if x.ndim == 3 and x.shape[2] == 1:
             return np.flip(x, axis=1)
         else:
             # Fallback for 1D data
            return np.flip(x, axis=0)
     # Permutation
     def permute(x, num_segments=10):
         split = np.array_split(x, num_segments, axis=0)
         np.random.shuffle(split)
         return np.concatenate(split, axis=0)
     # Magnitude Warping
     def magnitude_warp(x, sigma=0.001, num_knots=4):
         time_steps = np.linspace(0, 1, num=x.shape[0])
         knot_positions = np.linspace(0, 1, num=num_knots)
         knot_values = np.random.normal(loc=1.0, scale=sigma, size=num_knots)
         spline = CubicSpline(knot_positions, knot_values)
         warp_values = spline(time_steps)
         return list(x * warp_values)
     # Rotate
     def rotate(x):
```

```
if x.ndim != 2:
        raise ValueError("Input data must be a 2D array.")
    num_features = x.shape[1]
    random_rotation_matrix = np.random.normal(size=(num_features, num_features))
    q, r = np.linalg.qr(random_rotation_matrix)
    x_rotated = np.dot(x, q)
    return x_rotated
# Time Warping
def time_warp(x, sigma=0.2, num_knots=4):
    if x.ndim == 1:
        # For 1D data, we work directly with x
        original_steps = np.arange(len(x))
        new_steps = np.linspace(0, len(x) - 1, num=len(x))
    elif x.ndim == 2 and x.shape[1] == 1:
        # For 2D data with a single feature
        original_steps = np.arange(x.shape[0])
        new steps = np.linspace(0, x.shape[0] - 1, num=x.shape[0])
    else:
        raise ValueError("Unsupported data shape for time_warp")
    knot_positions = np.linspace(0, len(new_steps) - 1, num=num_knots)
    knot_values = np.random.normal(loc=0.0, scale=sigma, size=num_knots) +__
 ⇔knot_positions
    spline = CubicSpline(knot_positions, knot_values)
    warped steps = spline(new steps)
    warped_steps[warped_steps < 0] = 0</pre>
    warped_steps[warped_steps > len(new_steps) - 1] = len(new_steps) - 1
    return np.array([x[int(warped_step)] if x.ndim == 1 else_

¬x[int(warped_step), 0] for warped_step in warped_steps])

# Window Slicing
def window_slice(x, window_size=50):
    # Ensure the window size is smaller than the length of the data
    if window_size >= len(x):
        raise ValueError("Window size must be smaller than the length of the \sqcup

data")
```

```
start = np.random.randint(0, len(x) - window_size)
         windowed_data = x[start:start + window_size]
         padding_length = len(x) - window_size
         padded_data = np.pad(windowed_data, (0, padding_length), mode='constant')
         return padded_data
     # Window Warping
     def window_warp(x, window_size=50, scale=2):
         if len(x) <= window size:</pre>
             raise ValueError("The window size must be less than the size of the \sqcup
      ⇔data.")
         start = random.randint(0, len(x) - window_size)
         windowed_data = x[start:start + window_size]
         warped_window = np.repeat(windowed_data, scale)
         remove_length = len(warped_window) - window_size
         if remove_length > 0:
             warped_window = warped_window[:-remove_length]
         return np.concatenate([x[:start], warped_window, x[start + window_size:]])
[]: aug_function = jitter
     def balance_and_augment(train_x, train_y, aug_function):
         #aug_function = globals()[function_name]
         c = Counter(train_y)
         m = max(c.values())
         aug_train, aug_y = [], []
         for k, v in c.items():
             val = m - v
             while val > 0:
                 gen_idx = random.randrange(0, train_x.shape[0])
                 if k != 3:
                     augmented_data = aug_function(train_x[gen_idx])
                 else:
                     aug_train.append(train_x[gen_idx])
                 if len(augmented_data) == 1800:
                     aug_train.append(augmented_data)
                     aug_y.append(k)
                 val -= 1
         print(np.array(aug_train).shape,train_x.shape)
         for i in range(len(train_x)):
             aug_train.append(train_x[i])
```

aug_y.append(train_y[i])

```
return np.array(aug_train), np.array(aug_y)
     aug_train,aug_y = balance_and augment(np.squeeze(train_x), train_y,_
      ⇒aug_function)
     aug train = np.expand dims(aug train,axis=2)
     c_ = Counter(aug_y)
     print(c_,aug_train.shape)
    (457, 1800) (460, 1800)
    Counter({1: 198, 3: 198, 2: 198, 0: 198}) (917, 1800, 1)
[8]: #Hyper-parameters
     BATCH SIZE = 16
     EPOCHS = 100
     TIME_PERIODS = 1800 # The number of steps within one time segment
     STEP_DISTANCE = 1800 # The steps to take from one segment to the next
     #if step value is equal to TIME_PERIODS, then there is no overlap between the
      \hookrightarrow segments
     LABELS = [0,1,2,3] #nsr, afib, pvc, lb
     LABEL_CONDITION=['NSR','PVC','AFIB','LB']
     num_classes = 4
[9]: print("\n--- Reshape data to be accepted by Keras ---\n")
     train samples, train time periods, train sensors = train x.shape[0], train x.
      ⇒shape[1], train_x.shape[2]
     test_samples,test_time_periods, test_sensors = test_x.shape[0], test_x.
      ⇒shape[1], test_x.shape[2]
     train_input_shape = (train_time_periods*train_sensors)
     train_x = (train_x.reshape(train_samples, train_time_periods)).astype("float64")
     print('train_x shape: ', train_x.shape)
     print('Train input shape:', train_input_shape)
     test input shape = (test time periods*test sensors)
     test_x = (test_x.reshape(test_samples, test_time_periods)).astype("float64")
     print('test_x shape: ', test_x.shape)
     print('Test input shape:', test_input_shape)
     # Convert type for Keras otherwise Keras cannot process the data
     train_y = train_y.astype("float64")
     print('train_y shape: ', train_y.shape)
     test_y = test_y.astype("float64")
```

```
print('test_y shape: ', test_y.shape)
     --- Reshape data to be accepted by Keras ---
     train x shape: (460, 1800)
     Train input shape: 1800
     test x shape: (192, 1800)
     Test input shape: 1800
     train_y shape: (460,)
     test_y shape: (192,)
[10]: def plot_graphs(history):
          plt.plot(history.epoch, history.history["loss"], 'b', label='Training loss')
          plt.plot(history.epoch, history.history["val_loss"], 'g', label='Validation_
       ⇔loss')
          plt.legend()
          plt.title("Loss")
          plt.show()
          plt.plot(history.epoch, history.history["accuracy"], 'b', label='Training_
       →accuracy')
          plt.plot(history.epoch, history.history["val_accuracy"], 'g',__
       ⇔label='Validation accuracy')
          plt.legend()
          plt.title("Accuracy")
          plt.show()
          plt.plot(history.epoch, history.history["AUC"], 'b', label='Training AUC')
          plt.plot(history.epoch, history.history["val_AUC"], 'g', label='Validation_
       →AUC')
          plt.legend()
          plt.title("AUC")
          plt.show()
[11]: def classifier():
          model = Sequential()
          #model.add(Reshape((time, features), input_shape=(input_shape,)))
          model.add(Conv1D(32, 18, name='conv0',activation='relu', u
       →input_shape=(1800,1)))
          model.add(Conv1D(32, 18, name='conv1',activation='relu'))
          model.add(Conv1D(64, 18, name='conv2',activation='relu'))
          model.add(Conv1D(64, 18, name='conv3',activation='relu'))
          model.add(Conv1D(128, 18, name='conv4',activation='relu'))
          model.add(Conv1D(128, 18, name='conv5',activation='relu'))
          model.add(Conv1D(256, 18, name='conv6',activation='relu'))
          model.add(Conv1D(256, 18, name='conv7',activation='relu'))
```

```
model.add(MaxPooling1D(3,name='max1'))
model.add(Conv1D(32, 18, name='conv8',activation='relu'))
model.add(Conv1D(32, 18, name='conv9',activation='relu'))
model.add(Conv1D(64, 18, name='conv10',activation='relu'))
model.add(Conv1D(64, 18, name='conv11',activation='relu'))
model.add(Conv1D(128, 18, name='conv12',activation='relu'))
model.add(Conv1D(256, 18, name='conv13',activation='relu'))
model.add(GlobalAveragePooling1D(name='gap1'))
model.add(Dropout(.5,name='drop1'))
model.add(Dense(4, name='dense1',activation='softmax'))
return model
```

```
[12]: from keras.regularizers import 12
      from keras.layers import Conv1D, MaxPooling1D, Dense, Dropout, Flatten, L
       -Reshape, GlobalAveragePooling1D,BatchNormalization,Input,ELU
      def build_improved_model():
          model = Sequential()
          # Convolutional block 1
          model.add(Conv1D(32, kernel_size=3, padding='same', input_shape=(1800, 1), u
       ⇔kernel_regularizer=12(0.001), name='conv0'))
          model.add(BatchNormalization(name='bn0'))
          model.add(ELU(name='elu0'))
          model.add(Dropout(0.2, name='dropout0'))
          # Convolutional block 2
          model.add(Conv1D(32, kernel_size=3, padding='same', kernel_regularizer=12(0.
       \hookrightarrow 001), name='conv1'))
          model.add(BatchNormalization(name='bn1'))
          model.add(ELU(name='elu1'))
          model.add(MaxPooling1D(pool_size=3, name='max0'))
          model.add(Dropout(0.2, name='dropout1'))
          # Convolutional block 3
          model.add(Conv1D(64, kernel_size=3, padding='same', kernel_regularizer=12(0.
       →001), name='conv2'))
          model.add(BatchNormalization(name='bn2'))
          model.add(ELU(name='elu2'))
          model.add(Dropout(0.3, name='dropout2'))
          # Convolutional block 4
          model.add(Conv1D(64, kernel_size=3, padding='same', kernel_regularizer=12(0.
       →001), name='conv3'))
          model.add(BatchNormalization(name='bn3'))
          model.add(ELU(name='elu3'))
```

```
model.add(MaxPooling1D(pool_size=3, name='max1'))
    model.add(Dropout(0.3, name='dropout3'))
    # Convolutional block 5
    model.add(Conv1D(128, kernel_size=3, padding='same',_
 ⇔kernel_regularizer=12(0.001), name='conv4'))
    model.add(BatchNormalization(name='bn4'))
    model.add(ELU(name='elu4'))
    model.add(Dropout(0.4, name='dropout4'))
    # Convolutional block 6
    model.add(Conv1D(128, kernel_size=3, padding='same',__
 ⇔kernel_regularizer=12(0.001), name='conv5'))
    model.add(BatchNormalization(name='bn5'))
    model.add(ELU(name='elu5'))
    model.add(MaxPooling1D(pool_size=3, name='max2'))
    model.add(Dropout(0.4, name='dropout5'))
    # Flatten and Dense Layers
    model.add(Flatten())
    model.add(Dense(units=4, activation='softmax', kernel_regularizer=12(0.
 ⇔001), name='dense_final'))
    return model
model = build_improved_model()
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv0 (Conv1D)	(None, 1800, 32)	128
bn0 (BatchNormalization)	(None, 1800, 32)	128
eluO (ELU)	(None, 1800, 32)	0
dropout0 (Dropout)	(None, 1800, 32)	0
conv1 (Conv1D)	(None, 1800, 32)	3104
bn1 (BatchNormalization)	(None, 1800, 32)	128
elu1 (ELU)	(None, 1800, 32)	0

max0 (MaxPooling1D)	(None, 600, 32)	0
dropout1 (Dropout)	(None, 600, 32)	0
conv2 (Conv1D)	(None, 600, 64)	6208
bn2 (BatchNormalization)	(None, 600, 64)	256
elu2 (ELU)	(None, 600, 64)	0
dropout2 (Dropout)	(None, 600, 64)	0
conv3 (Conv1D)	(None, 600, 64)	12352
bn3 (BatchNormalization)	(None, 600, 64)	256
elu3 (ELU)	(None, 600, 64)	0
max1 (MaxPooling1D)	(None, 200, 64)	0
dropout3 (Dropout)	(None, 200, 64)	0
conv4 (Conv1D)	(None, 200, 128)	24704
bn4 (BatchNormalization)	(None, 200, 128)	512
elu4 (ELU)	(None, 200, 128)	0
dropout4 (Dropout)	(None, 200, 128)	0
conv5 (Conv1D)	(None, 200, 128)	49280
bn5 (BatchNormalization)	(None, 200, 128)	512
elu5 (ELU)	(None, 200, 128)	0
max2 (MaxPooling1D)	(None, 66, 128)	0
dropout5 (Dropout)	(None, 66, 128)	0
flatten (Flatten)	(None, 8448)	0
dense_final (Dense)	(None, 4)	33796

Total params: 131,364 Trainable params: 130,468 Non-trainable params: 896 ______

```
[13]: batchsize = 32
    ckpt_folder = os.path.join(os.getcwd(), 'Aug_No_Aug_T2')
    ckpt_path = os.path.join(ckpt_folder, 'No_Aug_Model_{epoch}')
    model_checkpoint_callback = tf.keras.callbacks.ModelCheckpoint(
        filepath=ckpt_path,
        save weights only=True,
        monitor='val_accuracy',
        mode='max',
        save_best_only=True, verbose=1)
    model = build_improved_model()
    model.compile(loss=tf.keras.losses.CategoricalCrossentropy(), optimizer=tf.
      -keras.optimizers.Adam(decay=.01,learning_rate=0.001,beta_1=.009,beta_2=.
      →8,epsilon=1e-08), metrics=[tf.keras.metrics.CategoricalAccuracy(
           name='accuracy'),tf.keras.metrics.Recall(name='Recall'),tf.keras.
     history = model.fit(train_x,np_utils.to_categorical(train_y, num_classes),_u
     ⇔epochs=150, batch_size = batchsize,
                    validation_data=(test_x,np_utils.to_categorical(test_y,__
     →num_classes)), callbacks=[model_checkpoint_callback])
    plot_graphs(history)
    Epoch 1/150
    0.3239 - Recall: 0.3022 - Precision: 0.3286 - AUC: 0.5528
    Epoch 1: val accuracy improved from -inf to 0.33333, saving model to d:\Research
    Work\Thesis\New Data\Code-20240119T232854Z-001\Code\Aug No_Aug_T2\No_Aug_Model_1
    15/15 [============ ] - 6s 132ms/step - loss: 3.8298 -
    accuracy: 0.3239 - Recall: 0.3022 - Precision: 0.3286 - AUC: 0.5528 - val_loss:
    1.7503 - val_accuracy: 0.3333 - val_Recall: 0.0000e+00 - val_Precision:
    0.0000e+00 - val_AUC: 0.5623
    Epoch 2/150
    0.3783 - Recall: 0.3196 - Precision: 0.3731 - AUC: 0.6232
    Epoch 2: val_accuracy did not improve from 0.33333
```

```
0.3783 - Recall: 0.3196 - Precision: 0.3731 - AUC: 0.6232 - val_loss: 1.9146 -
val_accuracy: 0.2396 - val_Recall: 0.0833 - val_Precision: 0.2857 - val_AUC:
0.5460
Epoch 3/150
0.4391 - Recall: 0.4000 - Precision: 0.4792 - AUC: 0.6847
Epoch 3: val_accuracy did not improve from 0.33333
0.4391 - Recall: 0.4000 - Precision: 0.4792 - AUC: 0.6847 - val_loss: 1.9177 -
val_accuracy: 0.2656 - val_Recall: 0.1198 - val_Precision: 0.2091 - val_AUC:
0.5685
Epoch 4/150
0.4174 - Recall: 0.3661 - Precision: 0.4444 - AUC: 0.6900
Epoch 4: val accuracy improved from 0.33333 to 0.35938, saving model to
d:\Research Work\Thesis\New
Data\Code-20240119T232854Z-001\Code\Aug_No_Aug_T2\No_Aug_Model_4
0.4174 - Recall: 0.3674 - Precision: 0.4459 - AUC: 0.6896 - val_loss: 2.6863 -
val_accuracy: 0.3594 - val_Recall: 0.3333 - val_Precision: 0.3699 - val_AUC:
0.5497
Epoch 5/150
0.4688 - Recall: 0.3996 - Precision: 0.5129 - AUC: 0.7033
Epoch 5: val_accuracy did not improve from 0.35938
0.4630 - Recall: 0.3935 - Precision: 0.5056 - AUC: 0.7024 - val loss: 2.9544 -
val_accuracy: 0.2552 - val_Recall: 0.2448 - val_Precision: 0.2513 - val_AUC:
0.5297
Epoch 6/150
0.4509 - Recall: 0.4018 - Precision: 0.4878 - AUC: 0.7289
Epoch 6: val_accuracy did not improve from 0.35938
0.4522 - Recall: 0.4022 - Precision: 0.4881 - AUC: 0.7293 - val loss: 2.2653 -
val_accuracy: 0.2500 - val_Recall: 0.2135 - val_Precision: 0.2455 - val_AUC:
0.5360
Epoch 7/150
0.4808 - Recall: 0.4303 - Precision: 0.5457 - AUC: 0.7427
Epoch 7: val_accuracy did not improve from 0.35938
0.4804 - Recall: 0.4283 - Precision: 0.5427 - AUC: 0.7449 - val_loss: 3.1197 -
val_accuracy: 0.2344 - val_Recall: 0.2240 - val_Precision: 0.2287 - val_AUC:
0.5345
Epoch 8/150
0.4913 - Recall: 0.4217 - Precision: 0.5374 - AUC: 0.7535
```

```
Epoch 8: val_accuracy did not improve from 0.35938
accuracy: 0.4913 - Recall: 0.4217 - Precision: 0.5374 - AUC: 0.7535 - val_loss:
2.5408 - val_accuracy: 0.2396 - val_Recall: 0.2344 - val_Precision: 0.2432 -
val AUC: 0.5368
Epoch 9/150
0.5087 - Recall: 0.4522 - Precision: 0.5547 - AUC: 0.7644
Epoch 9: val_accuracy did not improve from 0.35938
0.5087 - Recall: 0.4522 - Precision: 0.5547 - AUC: 0.7644 - val_loss: 2.1164 -
val_accuracy: 0.3385 - val_Recall: 0.1458 - val_Precision: 0.3333 - val_AUC:
0.6154
Epoch 10/150
0.5370 - Recall: 0.4891 - Precision: 0.6032 - AUC: 0.7933
Epoch 10: val_accuracy did not improve from 0.35938
0.5370 - Recall: 0.4891 - Precision: 0.6032 - AUC: 0.7933 - val_loss: 1.9580 -
val_accuracy: 0.2812 - val_Recall: 0.1771 - val_Precision: 0.2556 - val_AUC:
0.6045
Epoch 11/150
0.5196 - Recall: 0.4565 - Precision: 0.5585 - AUC: 0.7705
Epoch 11: val_accuracy did not improve from 0.35938
accuracy: 0.5196 - Recall: 0.4565 - Precision: 0.5585 - AUC: 0.7705 - val_loss:
2.4001 - val_accuracy: 0.2083 - val_Recall: 0.1927 - val_Precision: 0.2022 -
val_AUC: 0.5621
Epoch 12/150
0.5283 - Recall: 0.4630 - Precision: 0.5741 - AUC: 0.7709
Epoch 12: val_accuracy did not improve from 0.35938
accuracy: 0.5283 - Recall: 0.4630 - Precision: 0.5741 - AUC: 0.7709 - val loss:
2.8364 - val_accuracy: 0.2240 - val_Recall: 0.2031 - val_Precision: 0.2167 -
val AUC: 0.5101
Epoch 13/150
0.5500 - Recall: 0.4674 - Precision: 0.5907 - AUC: 0.7985
Epoch 13: val_accuracy did not improve from 0.35938
accuracy: 0.5500 - Recall: 0.4674 - Precision: 0.5907 - AUC: 0.7985 - val_loss:
2.3968 - val_accuracy: 0.2188 - val_Recall: 0.2031 - val_Precision: 0.2179 -
val_AUC: 0.5359
Epoch 14/150
0.5761 - Recall: 0.4826 - Precision: 0.6271 - AUC: 0.8056
```

```
Epoch 14: val_accuracy did not improve from 0.35938
accuracy: 0.5761 - Recall: 0.4826 - Precision: 0.6271 - AUC: 0.8056 - val_loss:
2.8049 - val_accuracy: 0.2083 - val_Recall: 0.2031 - val_Precision: 0.2120 -
val AUC: 0.5036
Epoch 15/150
0.5826 - Recall: 0.4848 - Precision: 0.6126 - AUC: 0.8192
Epoch 15: val_accuracy did not improve from 0.35938
accuracy: 0.5826 - Recall: 0.4848 - Precision: 0.6126 - AUC: 0.8192 - val_loss:
2.8770 - val accuracy: 0.2396 - val Recall: 0.2240 - val Precision: 0.2324 -
val_AUC: 0.5835
Epoch 16/150
0.5891 - Recall: 0.5174 - Precision: 0.6347 - AUC: 0.8249
Epoch 16: val_accuracy did not improve from 0.35938
accuracy: 0.5891 - Recall: 0.5174 - Precision: 0.6347 - AUC: 0.8249 - val_loss:
2.5025 - val_accuracy: 0.2135 - val_Recall: 0.2031 - val_Precision: 0.2179 -
val AUC: 0.5010
Epoch 17/150
0.5891 - Recall: 0.5217 - Precision: 0.6486 - AUC: 0.8259
Epoch 17: val_accuracy did not improve from 0.35938
accuracy: 0.5891 - Recall: 0.5217 - Precision: 0.6486 - AUC: 0.8259 - val_loss:
2.1836 - val_accuracy: 0.2292 - val_Recall: 0.1927 - val_Precision: 0.2229 -
val_AUC: 0.5654
Epoch 18/150
0.5913 - Recall: 0.4978 - Precision: 0.6343 - AUC: 0.8333
Epoch 18: val_accuracy did not improve from 0.35938
accuracy: 0.5913 - Recall: 0.4978 - Precision: 0.6343 - AUC: 0.8333 - val loss:
2.3851 - val_accuracy: 0.2292 - val_Recall: 0.1719 - val_Precision: 0.2037 -
val AUC: 0.5276
Epoch 19/150
0.5804 - Recall: 0.5174 - Precision: 0.6398 - AUC: 0.8197
Epoch 19: val_accuracy did not improve from 0.35938
accuracy: 0.5804 - Recall: 0.5174 - Precision: 0.6398 - AUC: 0.8197 - val_loss:
2.2931 - val_accuracy: 0.2188 - val_Recall: 0.1458 - val_Precision: 0.1892 -
val_AUC: 0.5196
Epoch 20/150
0.5781 - Recall: 0.5000 - Precision: 0.6275 - AUC: 0.8169
```

```
Epoch 20: val_accuracy did not improve from 0.35938
accuracy: 0.5804 - Recall: 0.5022 - Precision: 0.6311 - AUC: 0.8182 - val_loss:
2.3469 - val_accuracy: 0.1979 - val_Recall: 0.1823 - val_Precision: 0.2096 -
val AUC: 0.5646
Epoch 21/150
0.6304 - Recall: 0.5565 - Precision: 0.6684 - AUC: 0.8349
Epoch 21: val_accuracy did not improve from 0.35938
0.6304 - Recall: 0.5565 - Precision: 0.6684 - AUC: 0.8349 - val_loss: 2.2525 -
val_accuracy: 0.2500 - val_Recall: 0.1979 - val_Precision: 0.2452 - val_AUC:
0.5533
Epoch 22/150
0.6391 - Recall: 0.5696 - Precision: 0.7198 - AUC: 0.8596
Epoch 22: val_accuracy did not improve from 0.35938
0.6391 - Recall: 0.5696 - Precision: 0.7198 - AUC: 0.8596 - val_loss: 2.7998 -
val_accuracy: 0.1979 - val_Recall: 0.1927 - val_Precision: 0.2033 - val_AUC:
0.5649
Epoch 23/150
0.6043 - Recall: 0.5587 - Precision: 0.6693 - AUC: 0.8394
Epoch 23: val_accuracy did not improve from 0.35938
0.6043 - Recall: 0.5587 - Precision: 0.6693 - AUC: 0.8394 - val loss: 2.6960 -
val_accuracy: 0.1927 - val_Recall: 0.1823 - val_Precision: 0.2011 - val_AUC:
0.5477
Epoch 24/150
0.6205 - Recall: 0.5670 - Precision: 0.6632 - AUC: 0.8498
Epoch 24: val_accuracy did not improve from 0.35938
0.6239 - Recall: 0.5674 - Precision: 0.6658 - AUC: 0.8508 - val loss: 2.4260 -
val_accuracy: 0.2656 - val_Recall: 0.2031 - val_Precision: 0.2335 - val_AUC:
0.5813
Epoch 25/150
0.6761 - Recall: 0.5891 - Precision: 0.7021 - AUC: 0.8757
Epoch 25: val_accuracy did not improve from 0.35938
0.6761 - Recall: 0.5891 - Precision: 0.7021 - AUC: 0.8757 - val_loss: 2.1683 -
val_accuracy: 0.2292 - val_Recall: 0.1406 - val_Precision: 0.1915 - val_AUC:
0.5546
Epoch 26/150
0.6987 - Recall: 0.6228 - Precision: 0.7440 - AUC: 0.8910
```

```
Epoch 26: val_accuracy did not improve from 0.35938
0.7000 - Recall: 0.6261 - Precision: 0.7442 - AUC: 0.8919 - val_loss: 2.2536 -
val_accuracy: 0.2344 - val_Recall: 0.1875 - val_Precision: 0.2264 - val_AUC:
0.5407
Epoch 27/150
0.6283 - Recall: 0.5500 - Precision: 0.6951 - AUC: 0.8526
Epoch 27: val_accuracy did not improve from 0.35938
0.6283 - Recall: 0.5500 - Precision: 0.6951 - AUC: 0.8526 - val_loss: 2.3849 -
val_accuracy: 0.2344 - val_Recall: 0.1823 - val_Precision: 0.2174 - val_AUC:
0.5431
Epoch 28/150
0.6540 - Recall: 0.5781 - Precision: 0.6907 - AUC: 0.8685
Epoch 28: val_accuracy did not improve from 0.35938
0.6565 - Recall: 0.5804 - Precision: 0.6935 - AUC: 0.8686 - val_loss: 2.8759 -
val_accuracy: 0.2083 - val_Recall: 0.1927 - val_Precision: 0.1989 - val_AUC:
0.5629
Epoch 29/150
0.6652 - Recall: 0.6071 - Precision: 0.7273 - AUC: 0.8930
Epoch 29: val_accuracy did not improve from 0.35938
0.6652 - Recall: 0.6065 - Precision: 0.7285 - AUC: 0.8927 - val loss: 2.8262 -
val_accuracy: 0.2188 - val_Recall: 0.2083 - val_Precision: 0.2198 - val_AUC:
0.5625
Epoch 30/150
0.6783 - Recall: 0.5891 - Precision: 0.7207 - AUC: 0.8844
Epoch 30: val_accuracy did not improve from 0.35938
0.6783 - Recall: 0.5891 - Precision: 0.7207 - AUC: 0.8844 - val loss: 2.9301 -
val_accuracy: 0.2344 - val_Recall: 0.2240 - val_Precision: 0.2429 - val_AUC:
0.5742
Epoch 31/150
15/15 [============== ] - ETA: Os - loss: 1.1476 - accuracy:
0.6870 - Recall: 0.6217 - Precision: 0.7352 - AUC: 0.8943
Epoch 31: val_accuracy did not improve from 0.35938
0.6870 - Recall: 0.6217 - Precision: 0.7352 - AUC: 0.8943 - val_loss: 2.8948 -
val_accuracy: 0.2396 - val_Recall: 0.2292 - val_Precision: 0.2472 - val_AUC:
0.5438
Epoch 32/150
0.7022 - Recall: 0.6391 - Precision: 0.7350 - AUC: 0.9062
```

```
Epoch 32: val_accuracy did not improve from 0.35938
0.7022 - Recall: 0.6391 - Precision: 0.7350 - AUC: 0.9062 - val_loss: 2.6266 -
val_accuracy: 0.2188 - val_Recall: 0.2031 - val_Precision: 0.2254 - val_AUC:
0.5285
Epoch 33/150
0.6783 - Recall: 0.6261 - Precision: 0.7347 - AUC: 0.8824
Epoch 33: val_accuracy did not improve from 0.35938
0.6783 - Recall: 0.6261 - Precision: 0.7347 - AUC: 0.8824 - val_loss: 2.6693 -
val_accuracy: 0.2448 - val_Recall: 0.2344 - val_Precision: 0.2473 - val_AUC:
0.5465
Epoch 34/150
0.7009 - Recall: 0.6473 - Precision: 0.7572 - AUC: 0.8983
Epoch 34: val_accuracy did not improve from 0.35938
0.7000 - Recall: 0.6435 - Precision: 0.7590 - AUC: 0.8981 - val_loss: 2.8690 -
val_accuracy: 0.2135 - val_Recall: 0.1927 - val_Precision: 0.2044 - val_AUC:
0.5660
Epoch 35/150
0.6739 - Recall: 0.6130 - Precision: 0.7231 - AUC: 0.8824
Epoch 35: val_accuracy did not improve from 0.35938
0.6739 - Recall: 0.6130 - Precision: 0.7231 - AUC: 0.8824 - val_loss: 3.0737 -
val_accuracy: 0.2448 - val_Recall: 0.2188 - val_Precision: 0.2320 - val_AUC:
0.5432
Epoch 36/150
0.7087 - Recall: 0.6370 - Precision: 0.7571 - AUC: 0.9022
Epoch 36: val_accuracy did not improve from 0.35938
0.7087 - Recall: 0.6370 - Precision: 0.7571 - AUC: 0.9022 - val loss: 2.6076 -
val_accuracy: 0.2500 - val_Recall: 0.2448 - val_Precision: 0.2640 - val_AUC:
0.5522
Epoch 37/150
0.6942 - Recall: 0.6451 - Precision: 0.7468 - AUC: 0.8967
Epoch 37: val_accuracy did not improve from 0.35938
0.6935 - Recall: 0.6435 - Precision: 0.7437 - AUC: 0.8944 - val_loss: 2.5457 -
val_accuracy: 0.2604 - val_Recall: 0.2188 - val_Precision: 0.2428 - val_AUC:
0.5691
Epoch 38/150
0.7076 - Recall: 0.6473 - Precision: 0.7572 - AUC: 0.9019
```

```
Epoch 38: val_accuracy did not improve from 0.35938
0.7000 - Recall: 0.6413 - Precision: 0.7526 - AUC: 0.9003 - val_loss: 2.6216 -
val_accuracy: 0.2552 - val_Recall: 0.2135 - val_Precision: 0.2384 - val_AUC:
0.5734
Epoch 39/150
0.7308 - Recall: 0.6779 - Precision: 0.7726 - AUC: 0.9169
Epoch 39: val_accuracy did not improve from 0.35938
0.7348 - Recall: 0.6804 - Precision: 0.7786 - AUC: 0.9193 - val_loss: 2.4113 -
val_accuracy: 0.2812 - val_Recall: 0.2656 - val_Precision: 0.2849 - val_AUC:
0.5904
Epoch 40/150
0.7411 - Recall: 0.6830 - Precision: 0.7927 - AUC: 0.9221
Epoch 40: val_accuracy did not improve from 0.35938
0.7435 - Recall: 0.6848 - Precision: 0.7955 - AUC: 0.9226 - val_loss: 2.6513 -
val_accuracy: 0.2656 - val_Recall: 0.2552 - val_Precision: 0.2663 - val_AUC:
0.5592
Epoch 41/150
0.7326 - Recall: 0.6761 - Precision: 0.7814 - AUC: 0.9140
Epoch 41: val_accuracy did not improve from 0.35938
0.7326 - Recall: 0.6761 - Precision: 0.7814 - AUC: 0.9140 - val loss: 2.4944 -
val_accuracy: 0.2865 - val_Recall: 0.2604 - val_Precision: 0.2809 - val_AUC:
0.5734
Epoch 42/150
0.7174 - Recall: 0.6587 - Precision: 0.7671 - AUC: 0.9051
Epoch 42: val_accuracy did not improve from 0.35938
0.7174 - Recall: 0.6587 - Precision: 0.7671 - AUC: 0.9051 - val loss: 2.3875 -
val_accuracy: 0.3125 - val_Recall: 0.2865 - val_Precision: 0.3179 - val_AUC:
0.6074
Epoch 43/150
0.7283 - Recall: 0.6717 - Precision: 0.7863 - AUC: 0.9182
Epoch 43: val_accuracy did not improve from 0.35938
0.7283 - Recall: 0.6717 - Precision: 0.7863 - AUC: 0.9182 - val_loss: 2.9629 -
val_accuracy: 0.2708 - val_Recall: 0.2396 - val_Precision: 0.2599 - val_AUC:
0.5650
Epoch 44/150
0.7239 - Recall: 0.6652 - Precision: 0.7747 - AUC: 0.9204
```

```
Epoch 44: val_accuracy did not improve from 0.35938
0.7239 - Recall: 0.6652 - Precision: 0.7747 - AUC: 0.9204 - val_loss: 2.2801 -
val_accuracy: 0.2917 - val_Recall: 0.2760 - val_Precision: 0.3272 - val_AUC:
0.5986
Epoch 45/150
0.7435 - Recall: 0.6913 - Precision: 0.7930 - AUC: 0.9214
Epoch 45: val_accuracy did not improve from 0.35938
0.7435 - Recall: 0.6913 - Precision: 0.7930 - AUC: 0.9214 - val_loss: 2.4677 -
val_accuracy: 0.3021 - val_Recall: 0.2865 - val_Precision: 0.3022 - val_AUC:
0.6021
Epoch 46/150
0.7500 - Recall: 0.6870 - Precision: 0.7980 - AUC: 0.9261
Epoch 46: val_accuracy did not improve from 0.35938
0.7500 - Recall: 0.6870 - Precision: 0.7980 - AUC: 0.9261 - val_loss: 2.7109 -
val_accuracy: 0.3021 - val_Recall: 0.2812 - val_Precision: 0.2967 - val_AUC:
0.5970
Epoch 47/150
0.7630 - Recall: 0.6957 - Precision: 0.8081 - AUC: 0.9249
Epoch 47: val_accuracy did not improve from 0.35938
0.7630 - Recall: 0.6957 - Precision: 0.8081 - AUC: 0.9249 - val loss: 2.3455 -
val_accuracy: 0.3125 - val_Recall: 0.2969 - val_Precision: 0.3132 - val_AUC:
0.6212
Epoch 48/150
0.7500 - Recall: 0.6891 - Precision: 0.8005 - AUC: 0.9163
Epoch 48: val_accuracy did not improve from 0.35938
0.7500 - Recall: 0.6891 - Precision: 0.8005 - AUC: 0.9163 - val loss: 2.5876 -
val_accuracy: 0.3385 - val_Recall: 0.3125 - val_Precision: 0.3315 - val_AUC:
0.6242
Epoch 49/150
0.7522 - Recall: 0.7130 - Precision: 0.8000 - AUC: 0.9351
Epoch 49: val_accuracy did not improve from 0.35938
0.7522 - Recall: 0.7130 - Precision: 0.8000 - AUC: 0.9351 - val_loss: 2.5062 -
val_accuracy: 0.3229 - val_Recall: 0.3021 - val_Precision: 0.3169 - val_AUC:
0.6103
Epoch 50/150
0.7478 - Recall: 0.6978 - Precision: 0.7887 - AUC: 0.9298
```

```
Epoch 50: val_accuracy did not improve from 0.35938
0.7478 - Recall: 0.6978 - Precision: 0.7887 - AUC: 0.9298 - val_loss: 2.5011 -
val_accuracy: 0.3333 - val_Recall: 0.3021 - val_Precision: 0.3258 - val_AUC:
0.6108
Epoch 51/150
0.7804 - Recall: 0.7239 - Precision: 0.8024 - AUC: 0.9347
Epoch 51: val_accuracy did not improve from 0.35938
0.7804 - Recall: 0.7239 - Precision: 0.8024 - AUC: 0.9347 - val_loss: 2.5827 -
val_accuracy: 0.3177 - val_Recall: 0.2969 - val_Precision: 0.3098 - val_AUC:
0.6079
Epoch 52/150
0.7217 - Recall: 0.6935 - Precision: 0.7687 - AUC: 0.9317
Epoch 52: val_accuracy did not improve from 0.35938
0.7217 - Recall: 0.6935 - Precision: 0.7687 - AUC: 0.9317 - val_loss: 2.6402 -
val_accuracy: 0.3281 - val_Recall: 0.3021 - val_Precision: 0.3187 - val_AUC:
0.6045
Epoch 53/150
0.7522 - Recall: 0.7130 - Precision: 0.7942 - AUC: 0.9295
Epoch 53: val_accuracy improved from 0.35938 to 0.37500, saving model to
d:\Research Work\Thesis\New
Data\Code-20240119T232854Z-001\Code\Aug No_Aug_T2\No_Aug_Model_53
0.7522 - Recall: 0.7130 - Precision: 0.7942 - AUC: 0.9295 - val_loss: 2.3715 -
val_accuracy: 0.3750 - val_Recall: 0.3229 - val_Precision: 0.3543 - val_AUC:
0.6417
Epoch 54/150
0.7768 - Recall: 0.7277 - Precision: 0.8191 - AUC: 0.9473
Epoch 54: val accuracy did not improve from 0.37500
0.7804 - Recall: 0.7304 - Precision: 0.8215 - AUC: 0.9480 - val_loss: 2.4946 -
val_accuracy: 0.3333 - val_Recall: 0.3177 - val_Precision: 0.3333 - val_AUC:
0.6286
Epoch 55/150
0.7304 - Recall: 0.6957 - Precision: 0.7786 - AUC: 0.9240
Epoch 55: val_accuracy did not improve from 0.37500
0.7304 - Recall: 0.6957 - Precision: 0.7786 - AUC: 0.9240 - val_loss: 2.5281 -
val_accuracy: 0.3281 - val_Recall: 0.3073 - val_Precision: 0.3172 - val_AUC:
0.6320
Epoch 56/150
```

```
0.7826 - Recall: 0.7217 - Precision: 0.8117 - AUC: 0.9431
Epoch 56: val_accuracy did not improve from 0.37500
0.7826 - Recall: 0.7217 - Precision: 0.8117 - AUC: 0.9431 - val loss: 2.5887 -
val_accuracy: 0.3177 - val_Recall: 0.3177 - val_Precision: 0.3262 - val_AUC:
Epoch 57/150
15/15 [============== ] - ETA: Os - loss: 0.9450 - accuracy:
0.7674 - Recall: 0.7065 - Precision: 0.8207 - AUC: 0.9385
Epoch 57: val accuracy improved from 0.37500 to 0.40625, saving model to
d:\Research Work\Thesis\New
Data\Code-20240119T232854Z-001\Code\Aug_No_Aug_T2\No_Aug_Model_57
0.7674 - Recall: 0.7065 - Precision: 0.8207 - AUC: 0.9385 - val_loss: 2.3708 -
val_accuracy: 0.4062 - val_Recall: 0.3646 - val_Precision: 0.3955 - val_AUC:
0.6558
Epoch 58/150
0.7522 - Recall: 0.7009 - Precision: 0.8072 - AUC: 0.9328
Epoch 58: val_accuracy did not improve from 0.40625
0.7522 - Recall: 0.7022 - Precision: 0.8055 - AUC: 0.9323 - val_loss: 2.4145 -
val_accuracy: 0.3594 - val_Recall: 0.3229 - val_Precision: 0.3464 - val_AUC:
0.6429
Epoch 59/150
0.7761 - Recall: 0.7217 - Precision: 0.8177 - AUC: 0.9430
Epoch 59: val_accuracy did not improve from 0.40625
0.7761 - Recall: 0.7217 - Precision: 0.8177 - AUC: 0.9430 - val_loss: 2.5701 -
val_accuracy: 0.3385 - val_Recall: 0.3125 - val_Precision: 0.3315 - val_AUC:
0.6316
Epoch 60/150
0.7717 - Recall: 0.7304 - Precision: 0.8058 - AUC: 0.9450
Epoch 60: val accuracy did not improve from 0.40625
0.7717 - Recall: 0.7304 - Precision: 0.8058 - AUC: 0.9450 - val_loss: 2.4733 -
val_accuracy: 0.3594 - val_Recall: 0.3333 - val_Precision: 0.3596 - val_AUC:
0.6452
Epoch 61/150
15/15 [============== ] - ETA: Os - loss: 0.8766 - accuracy:
0.7870 - Recall: 0.7348 - Precision: 0.8325 - AUC: 0.9515
Epoch 61: val_accuracy did not improve from 0.40625
0.7870 - Recall: 0.7348 - Precision: 0.8325 - AUC: 0.9515 - val_loss: 2.4790 -
val_accuracy: 0.3542 - val_Recall: 0.3385 - val_Precision: 0.3533 - val_AUC:
```

```
0.6548
Epoch 62/150
15/15 [============== ] - ETA: Os - loss: 0.9691 - accuracy:
0.7739 - Recall: 0.7196 - Precision: 0.8054 - AUC: 0.9338
Epoch 62: val accuracy did not improve from 0.40625
0.7739 - Recall: 0.7196 - Precision: 0.8054 - AUC: 0.9338 - val loss: 2.6926 -
val_accuracy: 0.3281 - val_Recall: 0.3073 - val_Precision: 0.3242 - val_AUC:
0.6179
Epoch 63/150
15/15 [============== ] - ETA: Os - loss: 0.8381 - accuracy:
0.8087 - Recall: 0.7674 - Precision: 0.8445 - AUC: 0.9587
Epoch 63: val_accuracy did not improve from 0.40625
0.8087 - Recall: 0.7674 - Precision: 0.8445 - AUC: 0.9587 - val_loss: 2.5731 -
val_accuracy: 0.3646 - val_Recall: 0.3333 - val_Precision: 0.3596 - val_AUC:
0.6479
Epoch 64/150
0.8022 - Recall: 0.7652 - Precision: 0.8482 - AUC: 0.9498
Epoch 64: val_accuracy did not improve from 0.40625
0.8022 - Recall: 0.7652 - Precision: 0.8482 - AUC: 0.9498 - val_loss: 2.4504 -
val_accuracy: 0.4010 - val_Recall: 0.3750 - val_Precision: 0.4068 - val_AUC:
0.6558
Epoch 65/150
0.7723 - Recall: 0.7299 - Precision: 0.8175 - AUC: 0.9447
Epoch 65: val_accuracy did not improve from 0.40625
0.7739 - Recall: 0.7283 - Precision: 0.8211 - AUC: 0.9454 - val_loss: 2.5314 -
val_accuracy: 0.3490 - val_Recall: 0.3281 - val_Precision: 0.3443 - val_AUC:
0.6440
Epoch 66/150
0.8087 - Recall: 0.7630 - Precision: 0.8499 - AUC: 0.9562
Epoch 66: val accuracy did not improve from 0.40625
0.8087 - Recall: 0.7630 - Precision: 0.8499 - AUC: 0.9562 - val_loss: 2.8011 -
val_accuracy: 0.3333 - val_Recall: 0.3229 - val_Precision: 0.3407 - val_AUC:
0.6179
Epoch 67/150
15/15 [============== ] - ETA: Os - loss: 0.8591 - accuracy:
0.8065 - Recall: 0.7652 - Precision: 0.8482 - AUC: 0.9540
Epoch 67: val_accuracy did not improve from 0.40625
0.8065 - Recall: 0.7652 - Precision: 0.8482 - AUC: 0.9540 - val_loss: 2.7521 -
val_accuracy: 0.3385 - val_Recall: 0.3125 - val_Precision: 0.3297 - val_AUC:
```

```
0.6245
Epoch 68/150
15/15 [============== ] - ETA: Os - loss: 0.9282 - accuracy:
0.7696 - Recall: 0.7304 - Precision: 0.8038 - AUC: 0.9414
Epoch 68: val accuracy did not improve from 0.40625
0.7696 - Recall: 0.7304 - Precision: 0.8038 - AUC: 0.9414 - val loss: 2.4055 -
val_accuracy: 0.3802 - val_Recall: 0.3646 - val_Precision: 0.3911 - val_AUC:
0.6668
Epoch 69/150
15/15 [============== ] - ETA: Os - loss: 0.9163 - accuracy:
0.7913 - Recall: 0.7478 - Precision: 0.8171 - AUC: 0.9450
Epoch 69: val_accuracy did not improve from 0.40625
0.7913 - Recall: 0.7478 - Precision: 0.8171 - AUC: 0.9450 - val_loss: 2.5493 -
val_accuracy: 0.4062 - val_Recall: 0.3958 - val_Precision: 0.4153 - val_AUC:
0.6655
Epoch 70/150
0.8087 - Recall: 0.7522 - Precision: 0.8398 - AUC: 0.9546
Epoch 70: val_accuracy did not improve from 0.40625
0.8087 - Recall: 0.7522 - Precision: 0.8398 - AUC: 0.9546 - val_loss: 2.4747 -
val_accuracy: 0.3750 - val_Recall: 0.3490 - val_Precision: 0.3722 - val_AUC:
0.6632
Epoch 71/150
0.8043 - Recall: 0.7500 - Precision: 0.8354 - AUC: 0.9553
Epoch 71: val_accuracy did not improve from 0.40625
0.8043 - Recall: 0.7500 - Precision: 0.8354 - AUC: 0.9553 - val_loss: 2.5830 -
val_accuracy: 0.3490 - val_Recall: 0.3333 - val_Precision: 0.3556 - val_AUC:
0.6470
Epoch 72/150
0.7978 - Recall: 0.7739 - Precision: 0.8416 - AUC: 0.9591
Epoch 72: val accuracy did not improve from 0.40625
0.7978 - Recall: 0.7739 - Precision: 0.8416 - AUC: 0.9591 - val_loss: 2.5065 -
val_accuracy: 0.4010 - val_Recall: 0.3594 - val_Precision: 0.3920 - val_AUC:
0.6613
Epoch 73/150
15/15 [============== ] - ETA: Os - loss: 0.8208 - accuracy:
0.7978 - Recall: 0.7696 - Precision: 0.8329 - AUC: 0.9606
Epoch 73: val_accuracy did not improve from 0.40625
0.7978 - Recall: 0.7696 - Precision: 0.8329 - AUC: 0.9606 - val_loss: 2.5491 -
val_accuracy: 0.3646 - val_Recall: 0.3438 - val_Precision: 0.3646 - val_AUC:
```

```
0.6604
Epoch 74/150
0.7870 - Recall: 0.7565 - Precision: 0.8386 - AUC: 0.9506
Epoch 74: val accuracy did not improve from 0.40625
0.7870 - Recall: 0.7565 - Precision: 0.8386 - AUC: 0.9506 - val loss: 2.8800 -
val_accuracy: 0.3281 - val_Recall: 0.3125 - val_Precision: 0.3333 - val_AUC:
0.6092
Epoch 75/150
15/15 [============== ] - ETA: Os - loss: 0.7875 - accuracy:
0.8196 - Recall: 0.7761 - Precision: 0.8520 - AUC: 0.9663
Epoch 75: val_accuracy did not improve from 0.40625
0.8196 - Recall: 0.7761 - Precision: 0.8520 - AUC: 0.9663 - val_loss: 2.5392 -
val_accuracy: 0.3698 - val_Recall: 0.3490 - val_Precision: 0.3702 - val_AUC:
0.6559
Epoch 76/150
0.8149 - Recall: 0.7861 - Precision: 0.8538 - AUC: 0.9555
Epoch 76: val_accuracy did not improve from 0.40625
0.8174 - Recall: 0.7848 - Precision: 0.8555 - AUC: 0.9569 - val_loss: 2.4737 -
val_accuracy: 0.3854 - val_Recall: 0.3698 - val_Precision: 0.3966 - val_AUC:
0.6672
Epoch 77/150
0.8217 - Recall: 0.7957 - Precision: 0.8612 - AUC: 0.9619
Epoch 77: val_accuracy did not improve from 0.40625
0.8217 - Recall: 0.7957 - Precision: 0.8612 - AUC: 0.9619 - val_loss: 2.8932 -
val_accuracy: 0.3385 - val_Recall: 0.3333 - val_Precision: 0.3478 - val_AUC:
0.6226
Epoch 78/150
0.8149 - Recall: 0.7909 - Precision: 0.8393 - AUC: 0.9581
Epoch 78: val accuracy did not improve from 0.40625
0.8217 - Recall: 0.7978 - Precision: 0.8476 - AUC: 0.9605 - val_loss: 2.5130 -
val_accuracy: 0.3698 - val_Recall: 0.3542 - val_Precision: 0.3799 - val_AUC:
0.6551
Epoch 79/150
15/15 [============== ] - ETA: Os - loss: 0.7996 - accuracy:
0.8174 - Recall: 0.7957 - Precision: 0.8453 - AUC: 0.9636
Epoch 79: val_accuracy did not improve from 0.40625
0.8174 - Recall: 0.7957 - Precision: 0.8453 - AUC: 0.9636 - val_loss: 2.5585 -
val_accuracy: 0.3542 - val_Recall: 0.3281 - val_Precision: 0.3559 - val_AUC:
```

```
0.6365
Epoch 80/150
0.8304 - Recall: 0.7991 - Precision: 0.8689 - AUC: 0.9623
Epoch 80: val accuracy did not improve from 0.40625
0.8217 - Recall: 0.7891 - Precision: 0.8582 - AUC: 0.9563 - val loss: 2.7580 -
val_accuracy: 0.3542 - val_Recall: 0.3229 - val_Precision: 0.3483 - val_AUC:
0.6251
Epoch 81/150
15/15 [============== ] - ETA: Os - loss: 0.8058 - accuracy:
0.8065 - Recall: 0.7913 - Precision: 0.8387 - AUC: 0.9628
Epoch 81: val_accuracy did not improve from 0.40625
0.8065 - Recall: 0.7913 - Precision: 0.8387 - AUC: 0.9628 - val_loss: 2.7811 -
val_accuracy: 0.3490 - val_Recall: 0.3281 - val_Precision: 0.3520 - val_AUC:
0.6305
Epoch 82/150
0.8239 - Recall: 0.7783 - Precision: 0.8524 - AUC: 0.9525
Epoch 82: val_accuracy did not improve from 0.40625
0.8239 - Recall: 0.7783 - Precision: 0.8524 - AUC: 0.9525 - val_loss: 2.8991 -
val_accuracy: 0.3490 - val_Recall: 0.3229 - val_Precision: 0.3407 - val_AUC:
0.6275
Epoch 83/150
0.8510 - Recall: 0.8197 - Precision: 0.8789 - AUC: 0.9732
Epoch 83: val_accuracy did not improve from 0.40625
0.8435 - Recall: 0.8130 - Precision: 0.8759 - AUC: 0.9720 - val_loss: 2.7162 -
val_accuracy: 0.3542 - val_Recall: 0.3385 - val_Precision: 0.3611 - val_AUC:
0.6357
Epoch 84/150
0.8196 - Recall: 0.7652 - Precision: 0.8523 - AUC: 0.9581
Epoch 84: val accuracy did not improve from 0.40625
0.8196 - Recall: 0.7652 - Precision: 0.8523 - AUC: 0.9581 - val_loss: 2.5759 -
val_accuracy: 0.3698 - val_Recall: 0.3385 - val_Precision: 0.3672 - val_AUC:
0.6444
Epoch 85/150
0.8221 - Recall: 0.7981 - Precision: 0.8579 - AUC: 0.9610
Epoch 85: val_accuracy did not improve from 0.40625
0.8130 - Recall: 0.7891 - Precision: 0.8501 - AUC: 0.9588 - val_loss: 2.9574 -
val_accuracy: 0.3542 - val_Recall: 0.3229 - val_Precision: 0.3407 - val_AUC:
```

```
0.6210
Epoch 86/150
15/15 [============== ] - ETA: Os - loss: 0.7890 - accuracy:
0.8174 - Recall: 0.7804 - Precision: 0.8589 - AUC: 0.9653
Epoch 86: val accuracy did not improve from 0.40625
0.8174 - Recall: 0.7804 - Precision: 0.8589 - AUC: 0.9653 - val loss: 2.5426 -
val_accuracy: 0.3802 - val_Recall: 0.3698 - val_Precision: 0.3944 - val_AUC:
0.6691
Epoch 87/150
15/15 [============== ] - ETA: Os - loss: 0.7651 - accuracy:
0.8565 - Recall: 0.8261 - Precision: 0.8920 - AUC: 0.9690
Epoch 87: val_accuracy did not improve from 0.40625
0.8565 - Recall: 0.8261 - Precision: 0.8920 - AUC: 0.9690 - val_loss: 2.9345 -
val_accuracy: 0.3542 - val_Recall: 0.3229 - val_Precision: 0.3444 - val_AUC:
0.6154
Epoch 88/150
0.8261 - Recall: 0.7957 - Precision: 0.8632 - AUC: 0.9704
Epoch 88: val_accuracy did not improve from 0.40625
0.8261 - Recall: 0.7957 - Precision: 0.8632 - AUC: 0.9704 - val_loss: 2.6203 -
val_accuracy: 0.3854 - val_Recall: 0.3698 - val_Precision: 0.3944 - val_AUC:
0.6737
Epoch 89/150
0.8413 - Recall: 0.8043 - Precision: 0.8706 - AUC: 0.9645
Epoch 89: val_accuracy did not improve from 0.40625
0.8413 - Recall: 0.8043 - Precision: 0.8706 - AUC: 0.9645 - val_loss: 2.8939 -
val_accuracy: 0.3490 - val_Recall: 0.3281 - val_Precision: 0.3520 - val_AUC:
0.6142
Epoch 90/150
0.8435 - Recall: 0.8022 - Precision: 0.8765 - AUC: 0.9687
Epoch 90: val accuracy did not improve from 0.40625
0.8435 - Recall: 0.8022 - Precision: 0.8765 - AUC: 0.9687 - val_loss: 2.6254 -
val_accuracy: 0.3750 - val_Recall: 0.3594 - val_Precision: 0.3791 - val_AUC:
0.6492
Epoch 91/150
0.8582 - Recall: 0.8389 - Precision: 0.8972 - AUC: 0.9760
Epoch 91: val_accuracy did not improve from 0.40625
0.8522 - Recall: 0.8326 - Precision: 0.8886 - AUC: 0.9762 - val_loss: 2.5682 -
val_accuracy: 0.3802 - val_Recall: 0.3646 - val_Precision: 0.3867 - val_AUC:
```

```
0.6673
Epoch 92/150
0.8728 - Recall: 0.8326 - Precision: 0.8923 - AUC: 0.9758
Epoch 92: val accuracy did not improve from 0.40625
0.8739 - Recall: 0.8348 - Precision: 0.8930 - AUC: 0.9756 - val loss: 2.6248 -
val_accuracy: 0.3646 - val_Recall: 0.3438 - val_Precision: 0.3708 - val_AUC:
0.6584
Epoch 93/150
0.8371 - Recall: 0.8080 - Precision: 0.8599 - AUC: 0.9675
Epoch 93: val_accuracy improved from 0.40625 to 0.42188, saving model to
d:\Research Work\Thesis\New
Data\Code-20240119T232854Z-001\Code\Aug_No_Aug_T2\No_Aug_Model_93
0.8348 - Recall: 0.8065 - Precision: 0.8588 - AUC: 0.9668 - val_loss: 2.4279 -
val_accuracy: 0.4219 - val_Recall: 0.4010 - val_Precision: 0.4425 - val_AUC:
0.6893
Epoch 94/150
0.8435 - Recall: 0.8087 - Precision: 0.8671 - AUC: 0.9683
Epoch 94: val_accuracy did not improve from 0.42188
0.8435 - Recall: 0.8087 - Precision: 0.8671 - AUC: 0.9683 - val_loss: 2.6438 -
val_accuracy: 0.3698 - val_Recall: 0.3385 - val_Precision: 0.3652 - val_AUC:
0.6550
Epoch 95/150
0.8435 - Recall: 0.8130 - Precision: 0.8657 - AUC: 0.9679
Epoch 95: val_accuracy did not improve from 0.42188
0.8435 - Recall: 0.8130 - Precision: 0.8657 - AUC: 0.9679 - val_loss: 2.6867 -
val_accuracy: 0.3646 - val_Recall: 0.3594 - val_Precision: 0.3855 - val_AUC:
0.6615
Epoch 96/150
0.8728 - Recall: 0.8460 - Precision: 0.8981 - AUC: 0.9748
Epoch 96: val_accuracy did not improve from 0.42188
0.8674 - Recall: 0.8391 - Precision: 0.8956 - AUC: 0.9715 - val_loss: 2.9108 -
val_accuracy: 0.3542 - val_Recall: 0.3333 - val_Precision: 0.3556 - val_AUC:
0.6266
Epoch 97/150
0.8237 - Recall: 0.7902 - Precision: 0.8530 - AUC: 0.9645
Epoch 97: val_accuracy did not improve from 0.42188
```

```
0.8239 - Recall: 0.7891 - Precision: 0.8541 - AUC: 0.9648 - val_loss: 3.0239 -
val_accuracy: 0.3490 - val_Recall: 0.3333 - val_Precision: 0.3478 - val_AUC:
0.6227
Epoch 98/150
0.8326 - Recall: 0.7991 - Precision: 0.8627 - AUC: 0.9621
Epoch 98: val accuracy did not improve from 0.42188
0.8239 - Recall: 0.7913 - Precision: 0.8565 - AUC: 0.9610 - val_loss: 2.7882 -
val_accuracy: 0.3802 - val_Recall: 0.3542 - val_Precision: 0.3757 - val_AUC:
0.6577
Epoch 99/150
0.8549 - Recall: 0.8214 - Precision: 0.8659 - AUC: 0.9710
Epoch 99: val_accuracy did not improve from 0.42188
0.8543 - Recall: 0.8217 - Precision: 0.8650 - AUC: 0.9715 - val_loss: 2.7055 -
val_accuracy: 0.3594 - val_Recall: 0.3385 - val_Precision: 0.3693 - val_AUC:
0.6468
Epoch 100/150
0.8504 - Recall: 0.8192 - Precision: 0.8759 - AUC: 0.9736
Epoch 100: val_accuracy did not improve from 0.42188
0.8478 - Recall: 0.8174 - Precision: 0.8744 - AUC: 0.9736 - val_loss: 3.2655 -
val_accuracy: 0.3385 - val_Recall: 0.3229 - val_Precision: 0.3388 - val_AUC:
0.6105
Epoch 101/150
0.8478 - Recall: 0.8087 - Precision: 0.8732 - AUC: 0.9663
Epoch 101: val_accuracy did not improve from 0.42188
0.8478 - Recall: 0.8087 - Precision: 0.8732 - AUC: 0.9663 - val_loss: 2.8191 -
val_accuracy: 0.3594 - val_Recall: 0.3385 - val_Precision: 0.3611 - val_AUC:
0.6448
Epoch 102/150
0.8527 - Recall: 0.8170 - Precision: 0.8862 - AUC: 0.9757
Epoch 102: val_accuracy did not improve from 0.42188
0.8522 - Recall: 0.8174 - Precision: 0.8868 - AUC: 0.9739 - val_loss: 2.8389 -
val_accuracy: 0.3646 - val_Recall: 0.3333 - val_Precision: 0.3616 - val_AUC:
0.6498
Epoch 103/150
0.8826 - Recall: 0.8630 - Precision: 0.9064 - AUC: 0.9748
Epoch 103: val_accuracy did not improve from 0.42188
```

```
0.8826 - Recall: 0.8630 - Precision: 0.9064 - AUC: 0.9748 - val_loss: 2.8609 -
val_accuracy: 0.3594 - val_Recall: 0.3385 - val_Precision: 0.3611 - val_AUC:
0.6462
Epoch 104/150
0.8370 - Recall: 0.8065 - Precision: 0.8729 - AUC: 0.9662
Epoch 104: val accuracy did not improve from 0.42188
0.8370 - Recall: 0.8065 - Precision: 0.8729 - AUC: 0.9662 - val_loss: 2.7004 -
val_accuracy: 0.3698 - val_Recall: 0.3542 - val_Precision: 0.3757 - val_AUC:
0.6488
Epoch 105/150
0.8393 - Recall: 0.8147 - Precision: 0.8732 - AUC: 0.9697
Epoch 105: val_accuracy did not improve from 0.42188
0.8370 - Recall: 0.8130 - Precision: 0.8698 - AUC: 0.9691 - val_loss: 2.8101 -
val_accuracy: 0.3698 - val_Recall: 0.3490 - val_Precision: 0.3681 - val_AUC:
0.6536
Epoch 106/150
0.8457 - Recall: 0.8239 - Precision: 0.8814 - AUC: 0.9693
Epoch 106: val_accuracy did not improve from 0.42188
0.8457 - Recall: 0.8239 - Precision: 0.8814 - AUC: 0.9693 - val_loss: 2.8538 -
val_accuracy: 0.3646 - val_Recall: 0.3385 - val_Precision: 0.3714 - val_AUC:
0.6463
Epoch 107/150
0.8504 - Recall: 0.8080 - Precision: 0.8808 - AUC: 0.9682
Epoch 107: val_accuracy did not improve from 0.42188
0.8522 - Recall: 0.8109 - Precision: 0.8839 - AUC: 0.9687 - val_loss: 2.9015 -
val_accuracy: 0.3750 - val_Recall: 0.3490 - val_Precision: 0.3764 - val_AUC:
0.6365
Epoch 108/150
0.8587 - Recall: 0.8174 - Precision: 0.8826 - AUC: 0.9771
Epoch 108: val_accuracy did not improve from 0.42188
0.8587 - Recall: 0.8174 - Precision: 0.8826 - AUC: 0.9771 - val_loss: 2.8764 -
val_accuracy: 0.3646 - val_Recall: 0.3594 - val_Precision: 0.3750 - val_AUC:
0.6556
Epoch 109/150
0.8616 - Recall: 0.8259 - Precision: 0.8831 - AUC: 0.9766
Epoch 109: val_accuracy did not improve from 0.42188
```

```
0.8609 - Recall: 0.8239 - Precision: 0.8834 - AUC: 0.9763 - val_loss: 3.0106 -
val_accuracy: 0.3490 - val_Recall: 0.3333 - val_Precision: 0.3575 - val_AUC:
0.6390
Epoch 110/150
0.8435 - Recall: 0.8065 - Precision: 0.8668 - AUC: 0.9747
Epoch 110: val accuracy did not improve from 0.42188
0.8435 - Recall: 0.8065 - Precision: 0.8668 - AUC: 0.9747 - val_loss: 2.6747 -
val_accuracy: 0.4010 - val_Recall: 0.3802 - val_Precision: 0.4033 - val_AUC:
0.6783
Epoch 111/150
0.8460 - Recall: 0.8281 - Precision: 0.8668 - AUC: 0.9746
Epoch 111: val_accuracy did not improve from 0.42188
0.8457 - Recall: 0.8261 - Precision: 0.8676 - AUC: 0.9737 - val_loss: 2.7979 -
val_accuracy: 0.3750 - val_Recall: 0.3594 - val_Precision: 0.3833 - val_AUC:
0.6649
Epoch 112/150
0.8413 - Recall: 0.8152 - Precision: 0.8661 - AUC: 0.9730
Epoch 112: val_accuracy did not improve from 0.42188
0.8413 - Recall: 0.8152 - Precision: 0.8661 - AUC: 0.9730 - val_loss: 2.8239 -
val_accuracy: 0.3646 - val_Recall: 0.3594 - val_Precision: 0.3730 - val_AUC:
0.6648
Epoch 113/150
0.8772 - Recall: 0.8504 - Precision: 0.9050 - AUC: 0.9813
Epoch 113: val_accuracy did not improve from 0.42188
0.8717 - Recall: 0.8435 - Precision: 0.8981 - AUC: 0.9794 - val_loss: 2.9473 -
val_accuracy: 0.3698 - val_Recall: 0.3438 - val_Precision: 0.3646 - val_AUC:
0.6386
Epoch 114/150
0.8587 - Recall: 0.8239 - Precision: 0.8814 - AUC: 0.9758
Epoch 114: val_accuracy did not improve from 0.42188
0.8587 - Recall: 0.8239 - Precision: 0.8814 - AUC: 0.9758 - val_loss: 2.9639 -
val_accuracy: 0.3698 - val_Recall: 0.3385 - val_Precision: 0.3631 - val_AUC:
0.6439
Epoch 115/150
0.8549 - Recall: 0.8170 - Precision: 0.8819 - AUC: 0.9743
Epoch 115: val_accuracy did not improve from 0.42188
```

```
0.8478 - Recall: 0.8109 - Precision: 0.8735 - AUC: 0.9730 - val_loss: 2.7018 -
val_accuracy: 0.3906 - val_Recall: 0.3490 - val_Precision: 0.3807 - val_AUC:
0.6687
Epoch 116/150
0.8549 - Recall: 0.8304 - Precision: 0.8753 - AUC: 0.9743
Epoch 116: val accuracy did not improve from 0.42188
0.8587 - Recall: 0.8348 - Precision: 0.8787 - AUC: 0.9752 - val_loss: 2.7229 -
val_accuracy: 0.3906 - val_Recall: 0.3646 - val_Precision: 0.3933 - val_AUC:
0.6719
Epoch 117/150
0.8705 - Recall: 0.8460 - Precision: 0.8960 - AUC: 0.9781
Epoch 117: val_accuracy did not improve from 0.42188
0.8717 - Recall: 0.8478 - Precision: 0.8966 - AUC: 0.9782 - val_loss: 2.6919 -
val_accuracy: 0.3906 - val_Recall: 0.3646 - val_Precision: 0.3911 - val_AUC:
0.6758
Epoch 118/150
0.8761 - Recall: 0.8565 - Precision: 0.9037 - AUC: 0.9830
Epoch 118: val_accuracy did not improve from 0.42188
0.8761 - Recall: 0.8565 - Precision: 0.9037 - AUC: 0.9830 - val_loss: 2.7631 -
val_accuracy: 0.3906 - val_Recall: 0.3646 - val_Precision: 0.3867 - val_AUC:
0.6677
Epoch 119/150
0.8638 - Recall: 0.8304 - Precision: 0.8815 - AUC: 0.9737
Epoch 119: val_accuracy did not improve from 0.42188
0.8630 - Recall: 0.8283 - Precision: 0.8799 - AUC: 0.9733 - val_loss: 2.9408 -
val_accuracy: 0.3698 - val_Recall: 0.3490 - val_Precision: 0.3764 - val_AUC:
0.6422
Epoch 120/150
0.8891 - Recall: 0.8674 - Precision: 0.9110 - AUC: 0.9827
Epoch 120: val_accuracy did not improve from 0.42188
0.8891 - Recall: 0.8674 - Precision: 0.9110 - AUC: 0.9827 - val_loss: 2.8885 -
val_accuracy: 0.3698 - val_Recall: 0.3490 - val_Precision: 0.3722 - val_AUC:
0.6569
Epoch 121/150
0.8683 - Recall: 0.8371 - Precision: 0.8886 - AUC: 0.9774
Epoch 121: val_accuracy did not improve from 0.42188
```

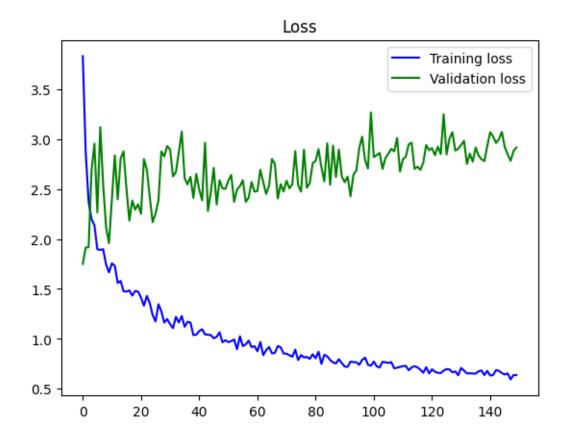
```
0.8674 - Recall: 0.8370 - Precision: 0.8871 - AUC: 0.9775 - val_loss: 2.9095 -
val_accuracy: 0.3594 - val_Recall: 0.3438 - val_Precision: 0.3687 - val_AUC:
0.6552
Epoch 122/150
0.8795 - Recall: 0.8460 - Precision: 0.9002 - AUC: 0.9812
Epoch 122: val accuracy did not improve from 0.42188
0.8804 - Recall: 0.8478 - Precision: 0.9007 - AUC: 0.9806 - val_loss: 2.8381 -
val_accuracy: 0.3802 - val_Recall: 0.3698 - val_Precision: 0.3880 - val_AUC:
0.6629
Epoch 123/150
0.8728 - Recall: 0.8527 - Precision: 0.8884 - AUC: 0.9827
Epoch 123: val_accuracy did not improve from 0.42188
0.8696 - Recall: 0.8500 - Precision: 0.8846 - AUC: 0.9818 - val_loss: 2.9241 -
val_accuracy: 0.3750 - val_Recall: 0.3646 - val_Precision: 0.3763 - val_AUC:
0.6674
Epoch 124/150
0.8705 - Recall: 0.8438 - Precision: 0.8936 - AUC: 0.9824
Epoch 124: val_accuracy did not improve from 0.42188
0.8717 - Recall: 0.8457 - Precision: 0.8943 - AUC: 0.9825 - val_loss: 2.8494 -
val_accuracy: 0.3698 - val_Recall: 0.3542 - val_Precision: 0.3757 - val_AUC:
0.6664
Epoch 125/150
0.8783 - Recall: 0.8457 - Precision: 0.8922 - AUC: 0.9779
Epoch 125: val_accuracy did not improve from 0.42188
0.8783 - Recall: 0.8457 - Precision: 0.8922 - AUC: 0.9779 - val_loss: 3.2480 -
val_accuracy: 0.3646 - val_Recall: 0.3490 - val_Precision: 0.3661 - val_AUC:
0.6271
Epoch 126/150
0.8839 - Recall: 0.8594 - Precision: 0.9102 - AUC: 0.9771
Epoch 126: val_accuracy did not improve from 0.42188
0.8848 - Recall: 0.8587 - Precision: 0.9101 - AUC: 0.9762 - val_loss: 2.8458 -
val_accuracy: 0.3958 - val_Recall: 0.3698 - val_Precision: 0.3859 - val_AUC:
0.6735
Epoch 127/150
0.8594 - Recall: 0.8348 - Precision: 0.8842 - AUC: 0.9794
Epoch 127: val_accuracy did not improve from 0.42188
```

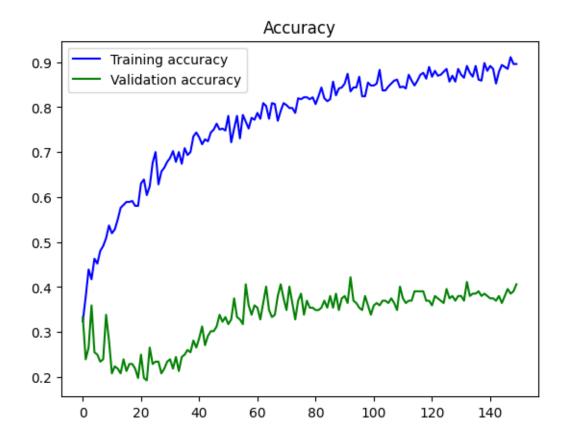
```
0.8565 - Recall: 0.8326 - Precision: 0.8845 - AUC: 0.9770 - val_loss: 3.0031 -
val_accuracy: 0.3750 - val_Recall: 0.3542 - val_Precision: 0.3736 - val_AUC:
0.6382
Epoch 128/150
0.8705 - Recall: 0.8438 - Precision: 0.9087 - AUC: 0.9812
Epoch 128: val accuracy did not improve from 0.42188
0.8696 - Recall: 0.8435 - Precision: 0.9065 - AUC: 0.9813 - val_loss: 3.0686 -
val_accuracy: 0.3802 - val_Recall: 0.3490 - val_Precision: 0.3661 - val_AUC:
0.6336
Epoch 129/150
0.8565 - Recall: 0.8304 - Precision: 0.8843 - AUC: 0.9809
Epoch 129: val_accuracy did not improve from 0.42188
0.8565 - Recall: 0.8304 - Precision: 0.8843 - AUC: 0.9809 - val_loss: 2.8847 -
val_accuracy: 0.3698 - val_Recall: 0.3542 - val_Precision: 0.3757 - val AUC:
0.6648
Epoch 130/150
0.8848 - Recall: 0.8587 - Precision: 0.9101 - AUC: 0.9854
Epoch 130: val_accuracy did not improve from 0.42188
0.8848 - Recall: 0.8587 - Precision: 0.9101 - AUC: 0.9854 - val_loss: 2.9020 -
val_accuracy: 0.3802 - val_Recall: 0.3490 - val_Precision: 0.3743 - val_AUC:
0.6584
Epoch 131/150
0.8717 - Recall: 0.8500 - Precision: 0.8907 - AUC: 0.9753
Epoch 131: val_accuracy did not improve from 0.42188
0.8717 - Recall: 0.8500 - Precision: 0.8907 - AUC: 0.9753 - val_loss: 2.9408 -
val_accuracy: 0.3802 - val_Recall: 0.3542 - val_Precision: 0.3778 - val_AUC:
0.6451
Epoch 132/150
0.8638 - Recall: 0.8415 - Precision: 0.8934 - AUC: 0.9776
Epoch 132: val_accuracy did not improve from 0.42188
0.8652 - Recall: 0.8435 - Precision: 0.8961 - AUC: 0.9783 - val_loss: 2.9823 -
val_accuracy: 0.3698 - val_Recall: 0.3490 - val_Precision: 0.3702 - val_AUC:
0.6493
Epoch 133/150
0.8973 - Recall: 0.8772 - Precision: 0.9161 - AUC: 0.9831
Epoch 133: val_accuracy did not improve from 0.42188
```

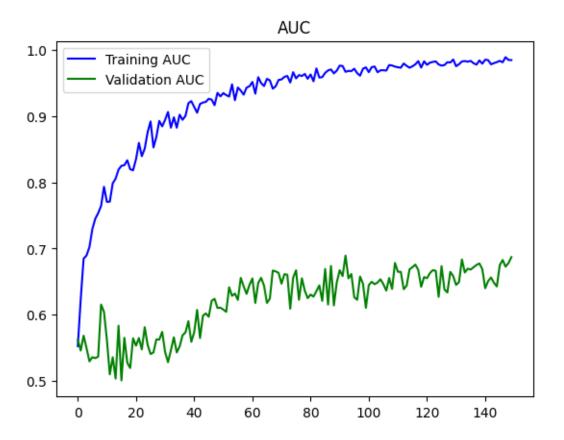
```
0.8913 - Recall: 0.8717 - Precision: 0.9114 - AUC: 0.9825 - val_loss: 2.7508 -
val_accuracy: 0.4115 - val_Recall: 0.3802 - val_Precision: 0.4011 - val_AUC:
0.6836
Epoch 134/150
0.8750 - Recall: 0.8594 - Precision: 0.8995 - AUC: 0.9827
Epoch 134: val accuracy did not improve from 0.42188
0.8761 - Recall: 0.8609 - Precision: 0.9021 - AUC: 0.9831 - val_loss: 2.8538 -
val_accuracy: 0.3802 - val_Recall: 0.3490 - val_Precision: 0.3743 - val_AUC:
0.6638
Epoch 135/150
0.8661 - Recall: 0.8504 - Precision: 0.8986 - AUC: 0.9821
Epoch 135: val_accuracy did not improve from 0.42188
0.8674 - Recall: 0.8500 - Precision: 0.8989 - AUC: 0.9821 - val_loss: 2.7728 -
val_accuracy: 0.3854 - val_Recall: 0.3490 - val_Precision: 0.3743 - val_AUC:
0.6697
Epoch 136/150
0.8884 - Recall: 0.8571 - Precision: 0.9035 - AUC: 0.9828
Epoch 136: val_accuracy did not improve from 0.42188
0.8913 - Recall: 0.8587 - Precision: 0.9060 - AUC: 0.9833 - val_loss: 2.9149 -
val_accuracy: 0.3854 - val_Recall: 0.3698 - val_Precision: 0.3923 - val_AUC:
0.6683
Epoch 137/150
0.8609 - Recall: 0.8326 - Precision: 0.8784 - AUC: 0.9798
Epoch 137: val_accuracy did not improve from 0.42188
accuracy: 0.8609 - Recall: 0.8326 - Precision: 0.8784 - AUC: 0.9798 - val_loss:
2.8352 - val_accuracy: 0.3906 - val_Recall: 0.3750 - val_Precision: 0.3913 -
val AUC: 0.6719
Epoch 138/150
0.8587 - Recall: 0.8348 - Precision: 0.8787 - AUC: 0.9777
Epoch 138: val_accuracy did not improve from 0.42188
15/15 [============= ] - 1s 101ms/step - loss: 0.6834 -
accuracy: 0.8587 - Recall: 0.8348 - Precision: 0.8787 - AUC: 0.9777 - val_loss:
2.7978 - val accuracy: 0.3802 - val Recall: 0.3646 - val Precision: 0.3911 -
val_AUC: 0.6752
Epoch 139/150
0.8996 - Recall: 0.8661 - Precision: 0.9216 - AUC: 0.9843
Epoch 139: val_accuracy did not improve from 0.42188
```

```
accuracy: 0.8978 - Recall: 0.8652 - Precision: 0.9213 - AUC: 0.9840 - val_loss:
2.7793 - val_accuracy: 0.3854 - val_Recall: 0.3646 - val_Precision: 0.3911 -
val_AUC: 0.6775
Epoch 140/150
0.8795 - Recall: 0.8482 - Precision: 0.8879 - AUC: 0.9785
Epoch 140: val accuracy did not improve from 0.42188
0.8804 - Recall: 0.8500 - Precision: 0.8886 - AUC: 0.9791 - val_loss: 2.9315 -
val_accuracy: 0.3802 - val_Recall: 0.3698 - val_Precision: 0.3859 - val_AUC:
0.6690
Epoch 141/150
0.8913 - Recall: 0.8674 - Precision: 0.9089 - AUC: 0.9851
Epoch 141: val_accuracy did not improve from 0.42188
accuracy: 0.8913 - Recall: 0.8674 - Precision: 0.9089 - AUC: 0.9851 - val_loss:
3.0686 - val_accuracy: 0.3750 - val_Recall: 0.3594 - val_Precision: 0.3770 -
val AUC: 0.6401
Epoch 142/150
0.8848 - Recall: 0.8674 - Precision: 0.9110 - AUC: 0.9845
Epoch 142: val_accuracy did not improve from 0.42188
accuracy: 0.8848 - Recall: 0.8674 - Precision: 0.9110 - AUC: 0.9845 - val_loss:
3.0259 - val_accuracy: 0.3750 - val_Recall: 0.3542 - val_Precision: 0.3778 -
val_AUC: 0.6508
Epoch 143/150
0.8527 - Recall: 0.8304 - Precision: 0.8836 - AUC: 0.9782
Epoch 143: val_accuracy did not improve from 0.42188
0.8522 - Recall: 0.8283 - Precision: 0.8819 - AUC: 0.9781 - val_loss: 2.9609 -
val_accuracy: 0.3698 - val_Recall: 0.3490 - val_Precision: 0.3722 - val_AUC:
0.6563
Epoch 144/150
15/15 [=================== ] - ETA: Os - loss: 0.6800 - accuracy:
0.8783 - Recall: 0.8457 - Precision: 0.8984 - AUC: 0.9799
Epoch 144: val_accuracy did not improve from 0.42188
0.8783 - Recall: 0.8457 - Precision: 0.8984 - AUC: 0.9799 - val_loss: 2.9990 -
val_accuracy: 0.3802 - val_Recall: 0.3594 - val_Precision: 0.3730 - val_AUC:
0.6484
Epoch 145/150
0.8935 - Recall: 0.8674 - Precision: 0.9068 - AUC: 0.9815
Epoch 145: val_accuracy did not improve from 0.42188
```

```
0.8935 - Recall: 0.8674 - Precision: 0.9068 - AUC: 0.9815 - val loss: 3.0718 -
val_accuracy: 0.3646 - val_Recall: 0.3490 - val_Precision: 0.3722 - val_AUC:
0.6428
Epoch 146/150
0.8862 - Recall: 0.8661 - Precision: 0.9087 - AUC: 0.9828
Epoch 146: val accuracy did not improve from 0.42188
0.8891 - Recall: 0.8696 - Precision: 0.9112 - AUC: 0.9834 - val_loss: 2.9190 -
val_accuracy: 0.3802 - val_Recall: 0.3698 - val_Precision: 0.3817 - val_AUC:
0.6750
Epoch 147/150
0.8839 - Recall: 0.8638 - Precision: 0.8938 - AUC: 0.9808
Epoch 147: val_accuracy did not improve from 0.42188
0.8848 - Recall: 0.8652 - Precision: 0.8944 - AUC: 0.9812 - val_loss: 2.8479 -
val_accuracy: 0.3958 - val_Recall: 0.3802 - val_Precision: 0.3989 - val_AUC:
0.6828
Epoch 148/150
0.9109 - Recall: 0.8826 - Precision: 0.9312 - AUC: 0.9890
Epoch 148: val_accuracy did not improve from 0.42188
0.9109 - Recall: 0.8826 - Precision: 0.9312 - AUC: 0.9890 - val_loss: 2.7819 -
val_accuracy: 0.3854 - val_Recall: 0.3698 - val_Precision: 0.3989 - val_AUC:
0.6727
Epoch 149/150
0.8957 - Recall: 0.8696 - Precision: 0.9112 - AUC: 0.9847
Epoch 149: val_accuracy did not improve from 0.42188
0.8957 - Recall: 0.8696 - Precision: 0.9112 - AUC: 0.9847 - val_loss: 2.8796 -
val_accuracy: 0.3906 - val_Recall: 0.3750 - val_Precision: 0.3892 - val_AUC:
0.6781
Epoch 150/150
0.8957 - Recall: 0.8783 - Precision: 0.9120 - AUC: 0.9845
Epoch 150: val_accuracy did not improve from 0.42188
0.8957 - Recall: 0.8783 - Precision: 0.9120 - AUC: 0.9845 - val_loss: 2.9162 -
val_accuracy: 0.4062 - val_Recall: 0.3906 - val_Precision: 0.4098 - val_AUC:
0.6871
```







```
[14]: train_y
[14]: array([1., 3., 0., 0., 3., 2., 2., 1., 2., 0., 3., 0., 2., 1., 3., 3., 2.,
            0., 0., 0., 0., 0., 0., 1., 2., 0., 0., 0., 2., 3., 0., 0., 1.,
             1., 0., 0., 3., 0., 0., 3., 2., 0., 0., 0., 1., 0., 1., 1., 0., 0.,
            0., 0., 0., 0., 1., 3., 1., 0., 0., 0., 0., 0., 0., 2., 0., 1.,
             1., 3., 1., 0., 1., 0., 0., 0., 1., 3., 1., 0., 2., 0., 2., 1., 1.,
            2., 2., 1., 0., 1., 1., 2., 0., 2., 0., 0., 0., 2., 2., 3., 2., 0.,
             1., 0., 1., 0., 0., 0., 0., 0., 1., 1., 0., 1., 3., 3., 2., 2.,
             3., 3., 2., 0., 0., 0., 1., 0., 3., 2., 1., 2., 2., 1., 2., 1., 3.,
            0., 1., 2., 0., 0., 0., 1., 0., 0., 1., 3., 0., 0., 0., 1., 0.,
            0., 0., 3., 1., 1., 1., 2., 1., 1., 2., 0., 0., 0., 0., 1., 0., 3.,
            0., 0., 3., 2., 0., 1., 3., 3., 2., 1., 0., 0., 0., 0., 2., 2., 3.,
             1., 2., 2., 2., 0., 2., 1., 0., 1., 0., 2., 1., 1., 0., 3., 3., 3.,
            0., 2., 0., 2., 2., 1., 1., 1., 1., 1., 0., 0., 2., 0., 2., 0., 0.,
            3., 1., 1., 3., 0., 1., 1., 0., 0., 0., 3., 0., 0., 1., 0., 0., 0.,
            2., 3., 0., 2., 0., 2., 0., 3., 2., 1., 0., 1., 0., 3., 3., 0., 0.,
            0., 2., 2., 0., 0., 3., 1., 0., 1., 1., 0., 0., 0., 2., 0., 2., 2.
            0., 3., 3., 0., 1., 0., 3., 0., 0., 2., 0., 1., 3., 2., 1., 0., 1.,
            3., 3., 2., 0., 2., 1., 2., 2., 0., 0., 0., 2., 0., 2., 1., 2., 0.,
            0., 1., 2., 3., 0., 2., 0., 2., 3., 1., 0., 3., 2., 3., 1., 0., 3.,
```

```
1., 0., 1., 0., 3., 0., 0., 1., 2., 1., 3., 1., 0., 0., 2., 3., 0., 1., 2., 2., 3., 0., 3., 0., 2., 0., 0., 0., 0., 0., 1., 0., 1., 1., 0., 2., 0., 2., 0., 3., 2., 0., 1., 1., 0., 1., 0., 3., 2., 1., 3., 0., 1., 0., 0., 0., 0., 3., 3., 0., 0., 2., 0., 2., 1., 0., 0., 2., 3., 3., 2., 2., 2., 2., 2., 0., 2., 0., 1., 2., 0., 3., 2., 0., 0., 0., 0., 0., 2., 0., 1., 2., 0., 1., 0., 3., 3., 0., 1., 2., 0., 0., 0., 3., 3., 0., 3., 0., 2., 0., 1., 2., 3., 2., 2., 0., 2., 0., 0., 1., 0., 3., 3., 0., 3., 0., 3., 0., 2., 2., 1., 2., 0., 0., 3., 1., 3., 0., 3., 1., 3., 0., 3., 0., 3., 2.])
```

[]: