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In [ ]:
         import numpy as np
         import tensorflow as tf
         from tensorflow import keras
         from tensorflow.keras import layers
         import h5py
         import warnings
         warnings.filterwarnings('ignore')
In [ ]:
         file_electron = "SingleElectronPt50_IMGCROPS_n249k_RHv1.hdf5"
         file_photon = "SinglePhotonPt50_IMGCROPS_n249k_RHv1.hdf5"
         with h5py.File(file_electron, "r") as f1:
             X_elec = np.array(f1['X'][:])
             y_elec = np.array(f1['y'][:])
         with h5py.File(file_photon, "r") as f2:
             X_phot = np.array(f2['X'][:])
             y_phot = np.array(f2['y'][:])
         print(X_elec.shape)
         print(X_phot.shape)
        (249000, 32, 32, 2)
        (249000, 32, 32, 2)
In [ ]:
         num_classes = 2
         input\_shape = (32, 32, 2)
         X = np.append(X_elec, X_phot, axis=0)
         y = np.append(y_elec, y_phot)
         X.shape
        (498000, 32, 32, 2)
Out[]:
In [ ]:
         from sklearn.model_selection import train_test_split
         x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=.2, random_state=4, stratify=y)
In [ ]:
         y_train = keras.utils.to_categorical(y_train, num_classes=2)
         y_test = keras.utils.to_categorical(y_test, num_classes=2)
In [ ]:
         learning_rate = 0.0001
         batch_size = 256
         num_epochs = 50
         patch_size = 2
         num_patches = (input_shape[0]//patch_size)**2
         projection_dim = 64
         num_heads = 2
         transformer_units = [
             projection_dim * 2,
             projection_dim,
         transformer_layers = 2
         mlp_head_units = [512, 256]
In [ ]:
         def mlp(x, hidden_units):
             for units in hidden_units:
                 x = layers.Dense(units, activation=tf.nn.gelu)(x)
             return x
In [ ]:
         class Patches(layers.Layer):
             def __init__(self, patch_size):
                 super(Patches, self).__init__()
                 self.patch_size = patch_size
             def call(self, images):
                 batch_size = tf.shape(images)[0]
                 patches = tf.image.extract_patches(
                     images=images,
                     sizes=[1, self.patch size, self.patch size, 1],
                     strides=[1, self.patch_size, self.patch_size, 1],
                     rates=[1, 1, 1, 1],
                     padding="VALID",
                 patch_dims = patches.shape[-1]
                 patches = tf.reshape(patches, [batch_size, -1, patch_dims])
                 return patches
In [ ]:
         class PatchEncoder(layers.Layer):
             def init (self, num patches, projection dim):
                 super(PatchEncoder, self). init ()
                 self.num_patches = num_patches
                 self.projection = layers.Dense(units=projection_dim)
                 self.position_embedding = layers.Embedding(
                     input_dim=num_patches, output_dim=projection_dim
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def call(self, patch):
               positions = tf.range(start=0, limit=self.num_patches, delta=1)
               encoded = self.projection(patch) + self.position_embedding(positions)
               return encoded
In [ ]:
       def create_vit_classifier():
           inputs = layers.Input(shape=input_shape)
           patches = Patches(patch_size)(inputs)
           encoded_patches = PatchEncoder(num_patches, projection_dim)(patches)
           for _ in range(transformer_layers):
               x1 = layers.LayerNormalization(epsilon=1e-6)(encoded_patches)
              attention_output = layers.MultiHeadAttention(
                  num_heads=num_heads, key_dim=projection_dim, dropout=0
              (x1, x1)
              x2 = layers.Add()([attention_output, encoded_patches])
              x3 = layers.LayerNormalization(epsilon=1e-6)(x2)
              x3 = mlp(x3, hidden_units=transformer_units)
               encoded_patches = layers.Add()([x3, x2])
           representation = layers.LayerNormalization(epsilon=1e-6)(encoded_patches)
           representation = layers.Flatten()(representation)
           features = mlp(representation, hidden_units=mlp_head_units)
           outputs = layers.Dense(num_classes, activation='softmax')(features)
           model = keras.Model(inputs=inputs, outputs=outputs)
           return model
In [ ]:
       def run_experiment(model):
           model.compile(optimizer=tf.optimizers.Adam(learning_rate=learning_rate), loss='categorical_crossentropy', metrics=[tf.keras.me
           reduce_lr = tf.keras.callbacks.ReduceLROnPlateau(monitor='val_auc', factor=0.2,
                                 patience=10, min_lr=1e-10 , verbose=1)
           history = model.fit(
              x=x_train,
              y=y_train,
              batch size=batch size,
               epochs=num_epochs,
              validation_split=0.1,
               callbacks=[reduce_lr]
           return model, history
In [ ]:
       vit_classifier = create_vit_classifier()
       with tf.device('/gpu:0'):
           model, history = run_experiment(vit_classifier)
       2022-03-23 01:37:27.243773: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:936] successful NUMA node read from SysFS had n
       egative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
       2022-03-23 01:37:27.261132: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:936] successful NUMA node read from SysFS had n
       egative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
       2022-03-23 01:37:27.263336: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:936] successful NUMA node read from SysFS had n
       egative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
       2022-03-23 01:37:27.266516: I tensorflow/core/platform/cpu_feature_guard.cc:151] This TensorFlow binary is optimized with oneAPI D
       eep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operations: AVX2 FMA
       To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.
       2022-03-23 01:37:27.268711: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:936] successful NUMA node read from SysFS had n
       egative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
       2022-03-23 01:37:27.270947: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:936] successful NUMA node read from SysFS had n
       egative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
       2022-03-23 01:37:27.273037: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:936] successful NUMA node read from SysFS had n
       egative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
       2022-03-23 01:37:28.160767: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:936] successful NUMA node read from SysFS had n
       egative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
       2022-03-23 01:37:28.163021: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:936] successful NUMA node read from SysFS had n
       egative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
       2022-03-23 01:37:28.165119: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:936] successful NUMA node read from SysFS had n
       egative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
       2022-03-23 01:37:28.167292: I tensorflow/core/common_runtime/gpu/gpu_device.cc:1525] Created device /job:localhost/replica:0/task:
       0/device:GPU:0 with 47216 MB memory: -> device: 0, name: Quadro RTX 8000, pci bus id: 0000:04:00.0, compute capability: 7.5
       Epoch 1/50
       1.0000e-04
       Epoch 2/50
       1.0000e-04
       Epoch 3/50
       1.0000e-04
       Epoch 4/50
       1.0000e-04
       Epoch 5/50
       1.0000e-04
       Epoch 6/50
       1.0000e-04
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Epoch 7/50
1.0000e-04
Epoch 8/50
1.0000e-04
Epoch 9/50
1401/1401 [======
     1.0000e-04
Epoch 10/50
1401/1401 [=
     1.0000e-04
Epoch 11/50
Epoch 11: ReduceLROnPlateau reducing learning rate to 1.9999999494757503e-05.
1.0000e-04
Epoch 12/50
2.0000e-05
Epoch 13/50
2.0000e-05
Epoch 14/50
2.0000e-05
Epoch 15/50
     :============] - 87s 62ms/step - loss: 0.5561 - auc: 0.7879 - val_loss: 0.5643 - val_auc: 0.7799 - lr:
1401/1401 [=
2.0000e-05
Epoch 16/50
2.0000e-05
Epoch 17/50
2.0000e-05
Epoch 18/50
2.0000e-05
Epoch 19/50
2.0000e-05
Epoch 20/50
2.0000e-05
Epoch 21/50
Epoch 21: ReduceLROnPlateau reducing learning rate to 3.999999898951501e-06.
2.0000e-05
Epoch 22/50
4.0000e-06
Epoch 23/50
4.0000e-06
Epoch 24/50
4.0000e-06
Epoch 25/50
1401/1401 [=
     ==============] - 87s 62ms/step - loss: 0.5418 - auc: 0.8014 - val_loss: 0.5624 - val_auc: 0.7820 - lr:
4.0000e-06
Epoch 26/50
     1401/1401 [=
4.0000e-06
Epoch 27/50
4.0000e-06
Epoch 28/50
4.0000e-06
Epoch 29/50
1401/1401
   4.0000e-06
Epoch 30/50
4.0000e-06
Epoch 31/50
Epoch 31: ReduceLROnPlateau reducing learning rate to 7.999999979801942e-07.
Epoch 32/50
8.0000e-07
Epoch 33/50
8.0000e-07
Epoch 34/50
8.0000e-07
Epoch 35/50
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8.0000e-07
 Epoch 36/50
 8.0000e-07
 Epoch 37/50
 8.0000e-07
 Epoch 38/50
 8.0000e-07
 Epoch 39/50
 8.0000e-07
 Epoch 40/50
 8.0000e-07
 Epoch 41/50
 Epoch 41: ReduceLROnPlateau reducing learning rate to 1.600000018697756e-07.
 8.0000e-07
 Epoch 42/50
 1.6000e-07
 Epoch 43/50
 1.6000e-07
 Epoch 44/50
 1.6000e-07
 Epoch 45/50
 1.6000e-07
 Epoch 46/50
 1.6000e-07
 Epoch 47/50
 1.6000e-07
 Fnoch 48/50
 1.6000e-07
 Epoch 49/50
 1.6000e-07
 Epoch 50/50
 1.6000e-07
In [ ]:
 model.evaluate(x_test,y_test)
 [0.5674085021018982, 0.7774555683135986]
Out[ ]:
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Results

Train AUC Score: 0.806

Validation AUC Score: 0.782

Test AUC Score: 0.777