Name : Devdeep Shetranjiwala Email ID : devdeep0702@gmail.com

Specific Task: Vision Transformers

Description:

Train a Transformer model of your choice on the dataset below to achieve the performance closest to your CNN model's performance in Task 1. </br>
by Discuss the resulting performance of the 2 chosen architectures.

(The discussion is done at the end of the code)

Datasets (Same as in Task 1): </br>
https://cernbox.cern.ch/index.php/s/AtBT8y4MiQYFcgc
(Photons) </br>
https://cernbox.cern.ch/index.php/s/FbXw3V4XNyYB3oA
(Electrons)

```
In [ ]:
import numpy as np
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras import layers
import h5py
from sklearn.model selection import train test split
In [ ]:
print("Num of GPUs Available: ", len(tf.config.list physical devices('GPU')))
Num of GPUs Available: 1
In [ ]:
import requests
url = 'https://cernbox.cern.ch/remote.php/dav/public-files/AtBT8y4MiQYFcqc/SinqlePhotonPt
50 IMGCROPS n249k RHv1.hdf5'
r = requests.get(url, allow_redirects=True)
open('photons.hdf5', 'wb').write(r.content)
url = 'https://cernbox.cern.ch/remote.php/dav/public-files/FbXw3V4XNyYB3oA/SingleElectron
Pt50 IMGCROPS n249k RHv1.hdf5'
r = requests.get(url, allow_redirects=True)
open('electrons.hdf5', 'wb').write(r.content)
Out[]:
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In [ ]:
file electron = "electrons.hdf5"
file photon = "photons.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 \text{ phot} = \text{np.array}(f2['X'][:])
    y phot = np.array(f2['y'][:])
print(X elec.shape)
print(X phot.shape)
del(file electron, file photon)
(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
del(X elec, X phot, y elec, y phot)
In [ ]:
x train, x test, y train, y test = train test split(X, y, test size=0.2, random state=42
, stratify=y)
del(X, 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

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 []:

In []:

```
vit classifier = create vit classifier()
model, history = run experiment(vit classifier)
Epoch 1/50
- val loss: 0.6478 - val auc: 0.6729 - lr: 1.0000e-04
Epoch 2/50
- val loss: 0.6121 - val auc: 0.7287 - lr: 1.0000e-04
Epoch 3/50
- val loss: 0.6269 - val auc: 0.7157 - lr: 1.0000e-04
Epoch 4/50
- val loss: 0.6086 - val auc: 0.7398 - lr: 1.0000e-04
Epoch 5/50
- val loss: 0.5999 - val auc: 0.7424 - lr: 1.0000e-04
Enoch 6/50
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120011 0,00
- val loss: 0.5970 - val auc: 0.7497 - lr: 1.0000e-04
Epoch 7/50
- val loss: 0.5846 - val auc: 0.7589 - lr: 1.0000e-04
Epoch 8/50
- val loss: 0.5804 - val auc: 0.7631 - lr: 1.0000e-04
Epoch 9/50
- val loss: 0.5825 - val auc: 0.7609 - lr: 1.0000e-04
Epoch 10/50
- val_loss: 0.5732 - val_auc: 0.7715 - lr: 1.0000e-04
Epoch 11/50
Epoch 11: ReduceLROnPlateau reducing learning rate to 1.9999999494757503e-05.
- val_loss: 0.5851 - val_auc: 0.7607 - lr: 1.0000e-04
Epoch 12/50
- val loss: 0.5681 - val auc: 0.7759 - lr: 2.0000e-05
Epoch 13/50
- val loss: 0.5667 - val auc: 0.7781 - lr: 2.0000e-05
Epoch 14/50
- val loss: 0.5656 - val auc: 0.7787 - lr: 2.0000e-05
Epoch 15/50
- val loss: 0.5662 - val auc: 0.7780 - lr: 2.0000e-05
Epoch 16/50
- val loss: 0.5657 - val auc: 0.7786 - 1r: 2.0000e-05
Epoch 17/50
- val loss: 0.5663 - val auc: 0.7782 - 1r: 2.0000e-05
Epoch 18/50
- val loss: 0.5658 - val auc: 0.7792 - lr: 2.0000e-05
Epoch 19/50
- val loss: 0.5680 - val auc: 0.7776 - lr: 2.0000e-05
Epoch 20/50
- val loss: 0.5671 - val auc: 0.7773 - lr: 2.0000e-05
Epoch 21/50
Epoch 21: ReduceLROnPlateau reducing learning rate to 3.999999898951501e-06.
- val_loss: 0.5667 - val_auc: 0.7783 - 1r: 2.0000e-05
Epoch 22/50
- val loss: 0.5653 - val auc: 0.7794 - lr: 4.0000e-06
Epoch 23/50
- val loss: 0.5659 - val auc: 0.7794 - lr: 4.0000e-06
Epoch 24/50
- val loss: 0.5666 - val auc: 0.7782 - lr: 4.0000e-06
Epoch 25/50
- val loss: 0.5658 - val auc: 0.7795 - lr: 4.0000e-06
Epoch 26/50
- val loss: 0.5662 - val auc: 0.7793 - lr: 4.0000e-06
Epoch 27/50
- val loss: 0.5666 - val auc: 0.7788 - lr: 4.0000e-06
Epoch 28/50
```

- val loss. 0 5665 - val auc. 0 7784 - 1r. 4 0000e-06

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vai 1000. 0.000
        Epoch 29/50
- val loss: 0.5667 - val auc: 0.7780 - lr: 4.0000e-06
Epoch 30/50
- val loss: 0.5668 - val auc: 0.7789 - lr: 4.0000e-06
Epoch 31/50
Epoch 31: ReduceLROnPlateau reducing learning rate to 7.999999979801942e-07.
- val loss: 0.5670 - val auc: 0.7786 - lr: 4.0000e-06
Epoch 32/50
- val_loss: 0.5667 - val_auc: 0.7792 - 1r: 8.0000e-07
Epoch 33/50
- val loss: 0.5666 - val auc: 0.7788 - lr: 8.0000e-07
Epoch 34/50
- val loss: 0.5669 - val auc: 0.7789 - lr: 8.0000e-07
Epoch 35/50
- val loss: 0.5667 - val auc: 0.7786 - lr: 8.0000e-07
Epoch 36/50
- val loss: 0.5671 - val auc: 0.7786 - lr: 8.0000e-07
Epoch 37/50
- val loss: 0.5670 - val auc: 0.7788 - lr: 8.0000e-07
- val loss: 0.5670 - val auc: 0.7790 - lr: 8.0000e-07
Epoch 39/50
- val loss: 0.5669 - val auc: 0.7786 - lr: 8.0000e-07
Epoch 40/50
- val loss: 0.5669 - val auc: 0.7787 - lr: 8.0000e-07
Epoch 41/50
Epoch 41: ReduceLROnPlateau reducing learning rate to 1.600000018697756e-07.
- val loss: 0.5668 - val auc: 0.7785 - lr: 8.0000e-07
Epoch 42/50
- val loss: 0.5670 - val auc: 0.7787 - lr: 1.6000e-07
Epoch 43/50
- val_loss: 0.5669 - val_auc: 0.7788 - lr: 1.6000e-07
Epoch 44/50
- val loss: 0.5670 - val auc: 0.7787 - lr: 1.6000e-07
Epoch 45/50
- val loss: 0.5670 - val auc: 0.7788 - lr: 1.6000e-07
Epoch 46/50
- val loss: 0.5670 - val auc: 0.7787 - lr: 1.6000e-07
Epoch 47/50
- val loss: 0.5670 - val auc: 0.7788 - lr: 1.6000e-07
Epoch 48/50
- val loss: 0.5670 - val auc: 0.7788 - lr: 1.6000e-07
Epoch 49/50
- val loss: 0.5670 - val auc: 0.7788 - lr: 1.6000e-07
Epoch 50/50
- val_loss: 0.5670 - val_auc: 0.7787 - lr: 1.6000e-07
```

Discuss the resulting performance of the 2 chosen architectures.

The AUC score is low because of lower no. of epochs still it is comperable to keras and pytorch score of 0.8.