common-task-1-electron-photon-classification

March 26, 2024

```
[]: pip install --upgrade --quiet tensorflow
```

Note: you may need to restart the kernel to use updated packages.

```
[]: import h5py
     import requests
     import numpy as np
     import matplotlib.pyplot as plt
     import os
     import PIL
     import tensorflow as tf
     from tensorflow import keras
     from tensorflow.keras import layers
     from tensorflow.keras.layers import Dense, Flatten, Dropout
     from tensorflow.keras.models import Sequential
     from tensorflow.keras.optimizers import Adam
     from sklearn.model_selection import train_test_split
     from tensorflow.keras.callbacks import ModelCheckpoint
     from sklearn.metrics import roc_auc_score
     from tensorflow.keras.optimizers import Adam
```

```
[]: with h5py.File(photon_filename, 'r') as f_photon:
        X_photon, y_photon = f_photon['X'][:], f_photon['y'][:]
     with h5py.File(electron_filename, 'r') as f_electron:
        X_electron, y_electron = f_electron['X'][:], f_electron['y'][:]
     X_combined = np.concatenate((X_photon, X_electron), axis=0)
     y_combined = np.concatenate((y_photon, y_electron), axis=0)
     print("Shape of combined data array:", X_combined.shape)
     print("Shape of combined target array:", y_combined.shape)
    Shape of combined data array: (498000, 32, 32, 2)
    Shape of combined target array: (498000,)
[]: energy_channel = X_combined[:, :, :, 0]
     time_channel = X_combined[:, :, :, 1]
     # Create a new array by stacking the two channels along with a zeros array for
     ⇔the third channel
     X_combined_3channel = np.stack([energy_channel, time_channel, np.
      ⇒zeros_like(time_channel)], axis=-1)
[]: x_train, x_test, y_train, y_test = train_test_split(X_combined, y_combined, u_
      →test_size=0.2, shuffle=True, random_state=42, stratify=y_combined)
[]: y_train[0]
[]: 0.0
[]: x_train.shape
[]: (398400, 32, 32, 2)
[]: del X_combined_3channel
     del y_combined
     del energy_channel
     del time_channel
     del X_photon
     del y_photon
     del X_electron
     del y_electron
[]: model = Sequential()
     resnet50_model = keras.applications.ResNet50(
         include_top=True,
```

```
weights=None,
         input_shape=(32, 32, 2),
         pooling='max',
         classes=1,
         classifier_activation="sigmoid",
     # for layer in resnet50_model.layers:
         layer.trainable = False
     model.add(resnet50_model)
     # model.add(Flatten())
     # model.add(Dense(512, activation='relu'))
     # model.add(Dropout(0.5))
     # model.add(Dense(1, activation='sigmoid'))
     model.summary()
    /opt/conda/lib/python3.10/site-packages/keras/src/applications/resnet.py:125:
    UserWarning: This model usually expects 1 or 3 input channels. However, it was
    passed an input_shape with 2 input channels.
      input_shape = imagenet_utils.obtain_input_shape(
    Model: "sequential"
     Layer (type)
                                                                      Param #
                                       Output Shape
     resnet50 (Functional)
                                                                   23,586,625
     Total params: 23,586,625 (89.98 MB)
     Trainable params: 23,533,505 (89.77 MB)
     Non-trainable params: 53,120 (207.50 KB)
[]: model.compile(loss="binary_crossentropy", optimizer=Adam(learning_rate=0.001),__
      →metrics=["accuracy"])
[]: | # rm -rf /kaggle/working/model_output
[]: output_dir = 'model_output/resnet50_maxpool_adam001'
     modelcheckpoint = ModelCheckpoint(filepath=output_dir+"/weights.{epoch:02d}.
      ⇔keras")
```

```
if not os.path.exists(output_dir):
         os.makedirs(output_dir)
     callback = keras.callbacks.EarlyStopping(
         monitor="val_loss",
         patience=3,
         verbose=1,
         restore_best_weights=True,
     )
[]: epochs = 50
     batch_size = 32
     history = model.fit(
         x_train,
         y_train,
         batch_size=batch_size,
         epochs=epochs,
         verbose=1,
         validation_data=(x_test, y_test),
         callbacks=[callback, modelcheckpoint]
     )
    Epoch 1/50
    WARNING: All log messages before absl::InitializeLog() is called are written to
    STDERR
    I0000 00:00:1711103069.426778
                                      881 service.cc:145] XLA service 0x7f8414002570
    initialized for platform CUDA (this does not guarantee that XLA will be used).
    Devices:
    I0000 00:00:1711103069.426822
                                      881 service.cc:153]
                                                             StreamExecutor device
    (0): Tesla P100-PCIE-16GB, Compute Capability 6.0
        3/12450
                            6:48 33ms/step -
    accuracy: 0.4740 - loss: 1.8538
                                     881 device_compiler.h:188] Compiled cluster
    I0000 00:00:1711103089.079931
    using XLA! This line is logged at most once for the lifetime of the process.
    12450/12450
                            416s
    29ms/step - accuracy: 0.5651 - loss: 0.7258 - val_accuracy: 0.6027 - val_loss:
    0.6663
    Epoch 2/50
    12450/12450
                            343s
    28ms/step - accuracy: 0.6050 - loss: 0.6702 - val_accuracy: 0.6134 - val_loss:
    0.6592
```

343s

Epoch 3/50 12450/12450

```
28ms/step - accuracy: 0.6331 - loss: 0.6456 - val_accuracy: 0.6649 - val_loss:
0.6316
Epoch 4/50
12450/12450
                        345s
28ms/step - accuracy: 0.6610 - loss: 0.6226 - val_accuracy: 0.6728 - val_loss:
0.6117
Epoch 5/50
12450/12450
                        341s
27ms/step - accuracy: 0.6822 - loss: 0.6028 - val_accuracy: 0.6820 - val_loss:
0.6078
Epoch 6/50
12450/12450
                        336s
27ms/step - accuracy: 0.6966 - loss: 0.5877 - val_accuracy: 0.6975 - val_loss:
0.5835
Epoch 7/50
12450/12450
                        338s
27ms/step - accuracy: 0.7056 - loss: 0.5783 - val_accuracy: 0.7092 - val_loss:
0.5720
Epoch 8/50
12450/12450
                        340s
27ms/step - accuracy: 0.7087 - loss: 0.5733 - val_accuracy: 0.7072 - val_loss:
0.5751
Epoch 9/50
12450/12450
                        340s
27ms/step - accuracy: 0.7122 - loss: 0.5686 - val_accuracy: 0.6910 - val_loss:
0.5904
Epoch 10/50
12450/12450
                        341s
27ms/step - accuracy: 0.7142 - loss: 0.5668 - val_accuracy: 0.7141 - val_loss:
0.5670
Epoch 11/50
12450/12450
                        343s
28ms/step - accuracy: 0.7171 - loss: 0.5629 - val_accuracy: 0.7129 - val_loss:
0.5670
Epoch 12/50
12450/12450
                        339s
27ms/step - accuracy: 0.7204 - loss: 0.5592 - val_accuracy: 0.7163 - val_loss:
0.5651
Epoch 13/50
12450/12450
                        359s
29ms/step - accuracy: 0.7223 - loss: 0.5560 - val_accuracy: 0.7184 - val_loss:
0.5612
Epoch 14/50
                        362s
12450/12450
29ms/step - accuracy: 0.7212 - loss: 0.5564 - val_accuracy: 0.7160 - val_loss:
0.5608
Epoch 15/50
12450/12450
                        363s
```

```
0.6014
    Epoch 16/50
    12450/12450
                            368s
    30ms/step - accuracy: 0.7269 - loss: 0.5506 - val_accuracy: 0.7175 - val_loss:
    0.5643
    Epoch 17/50
    12450/12450
                            352s
    28ms/step - accuracy: 0.7268 - loss: 0.5496 - val_accuracy: 0.7212 - val_loss:
    0.5569
    Epoch 18/50
    12450/12450
                            342s
    27ms/step - accuracy: 0.7277 - loss: 0.5488 - val_accuracy: 0.7227 - val_loss:
    0.5552
    Epoch 19/50
    12450/12450
                            342s
    27ms/step - accuracy: 0.7293 - loss: 0.5466 - val_accuracy: 0.5319 - val_loss:
    0.8686
    Epoch 20/50
    12450/12450
                            341s
    27ms/step - accuracy: 0.7310 - loss: 0.5447 - val_accuracy: 0.7200 - val_loss:
    0.5588
    Epoch 21/50
    12450/12450
                            342s
    27ms/step - accuracy: 0.7324 - loss: 0.5422 - val_accuracy: 0.6662 - val_loss:
    0.6207
    Epoch 21: early stopping
    Restoring model weights from the end of the best epoch: 18.
[]: model.save("trainedResNet50.keras")
[]: x = keras.models.load model('trainedResNet50.keras')
    /opt/conda/lib/python3.10/site-packages/keras/src/saving/saving lib.py:396:
    UserWarning: Skipping variable loading for optimizer 'rmsprop', because it has
    216 variables whereas the saved optimizer has 431 variables.
      trackable.load_own_variables(weights_store.get(inner_path))
[]: x.summary()
    Model: "sequential"
     Layer (type)
                                        Output Shape
                                                                       Param #
     resnet50 (Functional)
                                        (32, 1)
                                                                    23,586,625
```

29ms/step - accuracy: 0.7236 - loss: 0.5534 - val_accuracy: 0.6699 - val_loss:

```
Trainable params: 23,533,505 (89.77 MB)
     Non-trainable params: 53,120 (207.50 KB)
     Optimizer params: 23,533,507 (89.77 MB)
[]: import os
     import subprocess
     from IPython.display import FileLink, display
     def download_file(path, download_file_name):
         os.chdir('/kaggle/working/')
         zip_name = f"/kaggle/working/{download_file_name}.zip"
         command = f"zip {zip_name} {path} -r"
         result = subprocess.run(command, shell=True, capture_output=True, text=True)
         if result.returncode != 0:
             print("Unable to run zip command!")
             print(result.stderr)
             return
         display(FileLink(f'{download_file_name}.zip'))
[]: download_file('/kaggle/working/trainedResNet50.keras', 'trainedResNet50.keras')
    /kaggle/working/trainedResNet50.keras.zip
[]: model = Sequential()
     resnet50_model = keras.applications.ResNet50(
         include top=True,
         weights=None,
         input_shape=(32, 32, 2),
         pooling='avg',
         classes=1,
         classifier_activation="sigmoid",
     # for layer in resnet50_model.layers:
           layer.trainable = False
     model.add(resnet50_model)
     # model.add(Flatten())
     # model.add(Dense(512, activation='relu'))
```

Total params: 47,120,132 (179.75 MB)

```
# model.add(Dropout(0.5))
# model.add(Dense(1, activation='sigmoid'))
model.summary()
epochs = 50
batch_size = 32
history = model.fit(
    x_train,
    y_train,
    batch_size=batch_size,
    epochs=epochs,
    verbose=1,
    validation_data=(x_test, y_test),
    callbacks=[callback, modelcheckpoint]
)
Epoch 1/50
WARNING: All log messages before absl::InitializeLog() is called are written to
STDERR
I0000 00:00:1711099400.521850
                                  362 service.cc:145] XLA service 0x7e0c9c005fc0
initialized for platform CUDA (this does not guarantee that XLA will be used).
Devices:
I0000 00:00:1711099400.521893
                                  362 service.cc:153]
                                                        StreamExecutor device
(0): Tesla P100-PCIE-16GB, Compute Capability 6.0
    3/12450
                       6:37 32ms/step -
accuracy: 0.4549 - loss: 1.7112
I0000 00:00:1711099419.214896 362 device_compiler.h:188] Compiled cluster
using XLA! This line is logged at most once for the lifetime of the process.
12450/12450
                        400s
28ms/step - accuracy: 0.5691 - loss: 0.7245 - val_accuracy: 0.5965 - val_loss:
0.6680
Epoch 2/50
12450/12450
                       335s
27ms/step - accuracy: 0.6134 - loss: 0.6711 - val_accuracy: 0.6056 - val_loss:
0.6691
Epoch 3/50
12450/12450
                        333s
27ms/step - accuracy: 0.6672 - loss: 0.6235 - val_accuracy: 0.6711 - val_loss:
0.6168
Epoch 4/50
                       333s
12450/12450
27ms/step - accuracy: 0.6861 - loss: 0.6033 - val_accuracy: 0.6932 - val_loss:
0.5926
Epoch 5/50
```

```
12450/12450
                            336s
    27ms/step - accuracy: 0.6966 - loss: 0.5894 - val_accuracy: 0.6884 - val_loss:
    0.6562
    Epoch 6/50
    12450/12450
                            337s
    27ms/step - accuracy: 0.7010 - loss: 0.5837 - val_accuracy: 0.7055 - val_loss:
    0.5942
    Epoch 7/50
    12450/12450
                            337s
    27ms/step - accuracy: 0.7049 - loss: 0.5800 - val_accuracy: 0.6794 - val_loss:
    1.0661
    Epoch 7: early stopping
    Restoring model weights from the end of the best epoch: 4.
[]: y_train.shape
[]: (398400,)
[]: x_train.shape
[]: (398400, 32, 32, 2)
[]:
```