baseline-models

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[1]: import numpy as np
    import pandas as pd
    import tensorflow as tf
    from tensorflow.keras import models, layers
    from fenpreprocessing import fen_to_array
    from tensorflow.keras.callbacks import EarlyStopping
    from data_generation import position_generator, fix_positions, PosGen
    import datetime
    %load_ext tensorboard
[2]: tf.config.list_physical_devices()
[2]: [PhysicalDevice(name='/physical_device:CPU:0', device_type='CPU'),
     PhysicalDevice(name='/physical_device:GPU:0', device_type='GPU')]
[3]: # Setting paramaters on early stopping
    earlystop = EarlyStopping(monitor='val_loss',
                              min delta=0,
                              patience=20,
                               verbose=1.
                              mode='min',
                              restore_best_weights=True)
    log_dir = "logs/fit/baseline" + datetime.datetime.now().
     tensorboard_callback = tf.keras.callbacks.TensorBoard(log_dir=log_dir,_u
      →histogram_freq=1)
[4]: # Memory management, likely not necessary, but used as a safety as per the
     \rightarrow documentation recommendations on using GPUS
    gpus = tf.config.list_physical_devices('GPU')
    if gpus:
         # Currently, memory growth needs to be the same across GPUs
        for gpu in gpus:
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tf.config.experimental.set_memory_growth(gpu, True)
         logical_gpus = tf.config.list_logical_devices('GPU')
         print(len(gpus), "Physical GPUs,", len(logical_gpus), "Logical GPUs")
        except RuntimeError as e:
          # Memory growth must be set before GPUs have been initialized
         print(e)
     1 Physical GPUs, 1 Logical GPUs
[5]: small train = pd.read csv('fens/converted train partial.csv')
     small_val = pd.read_csv('fens/converted_val_partial.csv')
[6]: print(f"Train: {small_train.shape}, Validation: {small_val.shape}")
     Train: (338855, 2), Validation: (19123, 2)
[7]: small_val.shape[0] / 32
[7]: 597.59375
[8]: # train_gen = position_generator(small_train)
      # val_gen = position_generator(small_val)
     train_gen = PosGen(small_train, 'Position', 'Target')
     val_gen = PosGen(small_val, 'Position', 'Target')
[10]: baseline_model = models.Sequential()
     baseline_model.add(layers.Conv2D(64, 4, padding='same', input_shape=(8,8,13),__
      →activation='relu'))
     baseline_model.add(layers.MaxPooling2D(2))
     baseline_model.add(layers.Conv2D(32, 2, padding='same', activation='relu'))
     baseline_model.add(layers.Flatten())
     baseline_model.add(layers.Dense(64, activation='relu'))
     baseline_model.add(layers.Dense(1, activation='sigmoid'))
     baseline_model.compile(optimizer="adam", loss="binary_crossentropy", __
      →metrics=['acc'])
     baseline_model.summary()
      # Fitting the model
     baseline_history = baseline_model.fit(x=train_gen,
                          validation_data=val_gen,
                          # steps_per_epoch=100,
                          epochs=30,
                          callbacks=[earlystop, tensorboard_callback]
     Model: "sequential_1"
     Layer (type)
                                  Output Shape
                                                           Param #
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______
conv2d_2 (Conv2D)
                 (None, 8, 8, 64)
                                  13376
max_pooling2d_1 (MaxPooling2 (None, 4, 4, 64)
      ._____
                 (None, 4, 4, 32)
conv2d 3 (Conv2D)
_____
flatten 1 (Flatten)
                  (None, 512)
______
dense_2 (Dense)
                 (None, 64)
                                  32832
dense_3 (Dense)
           (None, 1)
                                  65
_____
Total params: 54,497
Trainable params: 54,497
Non-trainable params: 0
           ______
Epoch 1/30
10590/10590 [============== ] - 148s 14ms/step - loss: 0.1348 -
acc: 0.9681 - val_loss: 0.1263 - val_acc: 0.9693
acc: 0.9684 - val_loss: 0.1221 - val_acc: 0.9691
Epoch 3/30
acc: 0.9689 - val_loss: 0.1180 - val_acc: 0.9695
Epoch 4/30
10590/10590 [============== ] - 149s 14ms/step - loss: 0.1157 -
acc: 0.9695 - val_loss: 0.1189 - val_acc: 0.9696
Epoch 5/30
acc: 0.9701 - val_loss: 0.1176 - val_acc: 0.9695
Epoch 6/30
acc: 0.9706 - val loss: 0.1176 - val acc: 0.9697
Epoch 7/30
10590/10590 [============== ] - 155s 15ms/step - loss: 0.1079 -
acc: 0.9711 - val_loss: 0.1209 - val_acc: 0.9680
Epoch 8/30
10590/10590 [============== ] - 149s 14ms/step - loss: 0.1056 -
acc: 0.9717 - val_loss: 0.1239 - val_acc: 0.9672
Epoch 9/30
10590/10590 [============== ] - 147s 14ms/step - loss: 0.1033 -
acc: 0.9724 - val_loss: 0.1288 - val_acc: 0.9671
Epoch 10/30
10590/10590 [=============== ] - 146s 14ms/step - loss: 0.1012 -
acc: 0.9726 - val_loss: 0.1245 - val_acc: 0.9679
Epoch 11/30
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acc: 0.9732 - val_loss: 0.1262 - val_acc: 0.9686
Epoch 12/30
acc: 0.9737 - val_loss: 0.1322 - val_acc: 0.9655
Epoch 13/30
10590/10590 [============== ] - 149s 14ms/step - loss: 0.0953 -
acc: 0.9743 - val_loss: 0.1306 - val_acc: 0.9678
Epoch 14/30
acc: 0.9746 - val_loss: 0.1326 - val_acc: 0.9652
10590/10590 [============== ] - 147s 14ms/step - loss: 0.0918 -
acc: 0.9750 - val_loss: 0.1365 - val_acc: 0.9647
acc: 0.9754 - val_loss: 0.1345 - val_acc: 0.9657
Epoch 17/30
acc: 0.9758 - val_loss: 0.1414 - val_acc: 0.9646
Epoch 18/30
acc: 0.9763 - val_loss: 0.1427 - val_acc: 0.9645
Epoch 19/30
acc: 0.9765 - val_loss: 0.1461 - val_acc: 0.9635
Epoch 20/30
acc: 0.9771 - val_loss: 0.1484 - val_acc: 0.9647
Epoch 21/30
acc: 0.9773 - val_loss: 0.1479 - val_acc: 0.9646
Epoch 22/30
acc: 0.9775 - val loss: 0.1626 - val acc: 0.9614
Epoch 23/30
10590/10590 [============== ] - 145s 14ms/step - loss: 0.0799 -
acc: 0.9778 - val_loss: 0.1533 - val_acc: 0.9626
Epoch 24/30
10590/10590 [============= ] - 145s 14ms/step - loss: 0.0784 -
acc: 0.9783 - val_loss: 0.1617 - val_acc: 0.9607
Epoch 25/30
10590/10590 [============== ] - 143s 14ms/step - loss: 0.0771 -
acc: 0.9785 - val_loss: 0.1632 - val_acc: 0.9610
Restoring model weights from the end of the best epoch.
Epoch 00025: early stopping
```