full-puzzle-model

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```
[1]: import numpy as np
  import pandas as pd
  import tensorflow as tf
  from tensorflow.keras import models, layers
  from tensorflow.keras.callbacks import EarlyStopping
  from data_generation import ChessPositionGen

import datetime
%load_ext tensorboard
```

We'll set early stopping and tensorboard call backs.

```
[4]: # Memory management, likely not necessary, but used as a safety as per the

documentation recommendations on using GPUS

gpus = tf.config.list_physical_devices('GPU')

if gpus:

try:

# Currently, memory growth needs to be the same across GPUs

for gpu in gpus:

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

Train test split and data cleaning was performed in the dataorg notebook, so we can load our standard cleaned data. As a reminder, this is a list of FEN positions with the target as being the correct move. The data generator will turn this into 5 different potential positions, with the correct move labeled 1 and all other labeled 0.

```
[5]: train = pd.read_csv('fens/train.csv')
val = pd.read_csv('fens/val.csv')
```

```
[6]: train_gen = ChessPositionGen(train, batch_size=512)
val_gen = ChessPositionGen(val, batch_size=512)
```

The CNN is pretty much the same as was used in misc-notebooks/baseline-models.ipynb, but with the improved data generator, it can be trained on our full training set.

```
[7]: full_puzzle_model = models.Sequential()
     full_puzzle_model.add(layers.Conv2D(64, 4, padding='same',_
     →input_shape=(8,8,13), activation='relu'))
     full_puzzle_model.add(layers.MaxPooling2D(2))
     full_puzzle_model.add(layers.Conv2D(32, 2, padding='same', activation='relu'))
     full_puzzle_model.add(layers.Flatten())
     full_puzzle_model.add(layers.Dense(64, activation='relu'))
     full_puzzle_model.add(layers.Dense(1, activation='sigmoid'))
     full_puzzle_model.compile(optimizer="adam", loss="binary_crossentropy", __
     →metrics=['acc'])
     full_puzzle_model.summary()
     # Fitting the model
     full_puzzle_history = full_puzzle_model.fit(x=train_gen,
                         validation_data=val_gen,
                         epochs=30,
                         callbacks=[earlystop, tensorboard_callback]
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 8, 8, 64)	13376
max_pooling2d (MaxPooling2D)	(None, 4, 4, 64)	0
conv2d_1 (Conv2D)	(None, 4, 4, 32)	8224
flatten (Flatten)	(None, 512)	0

```
dense (Dense)
            (None, 64)
                        32832
        (None, 1)
dense_1 (Dense)
                       65
______
Total params: 54,497
Trainable params: 54,497
Non-trainable params: 0
    ______
Epoch 1/30
acc: 0.8451 - val_loss: 0.3744 - val_acc: 0.8628
Epoch 2/30
acc: 0.8653 - val_loss: 0.3647 - val_acc: 0.8667
Epoch 3/30
acc: 0.8691 - val_loss: 0.3559 - val_acc: 0.8703
Epoch 4/30
acc: 0.8716 - val_loss: 0.3518 - val_acc: 0.8712
Epoch 5/30
acc: 0.8731 - val_loss: 0.3440 - val_acc: 0.8737
Epoch 6/30
acc: 0.8744 - val_loss: 0.3381 - val_acc: 0.8745
Epoch 7/30
acc: 0.8755 - val_loss: 0.3356 - val_acc: 0.8748
Epoch 8/30
acc: 0.8762 - val_loss: 0.3298 - val_acc: 0.8769
Epoch 9/30
acc: 0.8772 - val_loss: 0.3319 - val_acc: 0.8762
Epoch 10/30
acc: 0.8778 - val_loss: 0.3253 - val_acc: 0.8781
Epoch 11/30
acc: 0.8785 - val_loss: 0.3230 - val_acc: 0.8786
acc: 0.8791 - val_loss: 0.3235 - val_acc: 0.8787
Epoch 13/30
acc: 0.8796 - val_loss: 0.3200 - val_acc: 0.8794
```

```
Epoch 14/30
acc: 0.8801 - val_loss: 0.3189 - val_acc: 0.8796
Epoch 15/30
acc: 0.8807 - val_loss: 0.3173 - val_acc: 0.8808
Epoch 16/30
acc: 0.8810 - val_loss: 0.3183 - val_acc: 0.8804
Epoch 17/30
acc: 0.8813 - val_loss: 0.3172 - val_acc: 0.8806
Epoch 18/30
acc: 0.8817 - val_loss: 0.3155 - val_acc: 0.8818
Epoch 19/30
acc: 0.8819 - val_loss: 0.3153 - val_acc: 0.8815
Epoch 20/30
acc: 0.8822 - val_loss: 0.3128 - val_acc: 0.8822
Epoch 21/30
acc: 0.8826 - val_loss: 0.3150 - val_acc: 0.8815
Epoch 22/30
acc: 0.8829 - val_loss: 0.3121 - val_acc: 0.8819
Epoch 23/30
acc: 0.8831 - val_loss: 0.3099 - val_acc: 0.8831
Epoch 24/30
acc: 0.8834 - val_loss: 0.3114 - val_acc: 0.8827
Epoch 25/30
acc: 0.8836 - val_loss: 0.3091 - val_acc: 0.8831
Epoch 26/30
acc: 0.8837 - val_loss: 0.3080 - val_acc: 0.8839
Epoch 27/30
acc: 0.8840 - val_loss: 0.3106 - val_acc: 0.8828
acc: 0.8842 - val_loss: 0.3084 - val_acc: 0.8836
Epoch 29/30
acc: 0.8844 - val_loss: 0.3071 - val_acc: 0.8839
```