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In [27]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
import keras
import keras.backend as K
from keras.optimizers import Adam
from keras.models import Sequential
from keras.utils import Sequence
from keras.layers import *

# open csv file, please change this if you're opening a file from your end
df = pd.read_csv(r"C:\Users\GLUTEUSMAXIMUS&OPPAI\Desktop\352_project\sudoku.csv")
rating_df = pd.read_csv(r"C:\Users\GLUTEUSMAXIMUS&OPPAI\Desktop\352_project\sudoku-3m")
# taking a quick glance at the data
df.head()
```

Out[27]:

		quizzes	solutions
0	0043002090050090010700600430060020871900074000...	8643712593258497619712658434361925871986574322...	
1	0401000501070039605200080000000000170009068008...	3461792581875239645296483719658324174729168358...	
2	6001203840084590720000060050002640300700800069...	6951273841384596727248369158512647392739815469...	
3	4972000001004000050000160986203000403009000000...	4972583161864397252537164986293815473759641828...	
4	0059103080094030600275001000300002010008200070...	4659123781894735623275681497386452919548216372...	

```
In [28]: # initializing the data generator class of keras
class DataGenerator(Sequence):
    def __init__(self, df, batch_size = 16, subset = "train", shuffle = False):
        super().__init__()
        self.df = df
        self.batch_size = batch_size
        self.shuffle = shuffle
        self.subset = subset
        self.data_path = "\sudoku.csv"
        self.on_epoch_end()

    def __len__(self):
        return int(np.floor(len(self.df)/self.batch_size))

    def on_epoch_end(self):
        self.indexes = np.arange(len(self.df))
        if self.shuffle==True:
            np.random.shuffle(self.indexes)

    def __getitem__(self, index):
        X = np.empty((self.batch_size, 9,9,1))
        y = np.empty((self.batch_size,81,1))
        indexes = self.indexes[index*self.batch_size:(index+1)*self.batch_size]
        for i,f in enumerate(self.df['quizzes'].iloc[indexes]):
            self.info[index*self.batch_size+i]=f
            X[i,:] = (np.array(list(map(int,list(f)))).reshape((9,9,1))/9)-0.5
        if self.subset == 'train':
            for i,f in enumerate(self.df['solutions'].iloc[indexes]):
                self.info[index*self.batch_size+i]=f
                y[i,:] = np.array(list(map(int,list(f)))).reshape((81,1)) - 1
            if self.subset == 'train': return X, y
        else: return X
```

```
In [29]: #building the sequential model
model = Sequential()
model.add(Conv2D(64, kernel_size=(3,3), activation='relu', padding='same', input_shape=(9,9,1)))
model.add(BatchNormalization())
model.add(Conv2D(64, kernel_size=(3,3), activation='relu', padding='same'))
model.add(BatchNormalization())
model.add(Conv2D(128, kernel_size=(1,1), activation='relu', padding='same'))
model.add(Flatten())
model.add(Dense(81*9))
model.add(Reshape((-1, 9)))
model.add(Activation('softmax'))
adam = keras.optimizers.Adam(lr=.001)
model.compile(loss='sparse_categorical_crossentropy', optimizer=adam, metrics=['accuracy'])
```

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In [30]: model.summary()

Model: "sequential_2"

Layer (type)                                     Output Shape                                     Param #
-----
conv2d_6 (Conv2D)                                (None, 9, 9, 64)                                640
batch_normalization_4 (Batch Normalization)      (None, 9, 9, 64)                                256
conv2d_7 (Conv2D)                                (None, 9, 9, 64)                                36928
batch_normalization_5 (Batch Normalization)      (None, 9, 9, 64)                                256
conv2d_8 (Conv2D)                                (None, 9, 9, 128)                               8320
flatten_2 (Flatten)                              (None, 10368)                                    0
dense_2 (Dense)                                   (None, 729)                                       7559001
reshape_2 (Reshape)                              (None, 81, 9)                                    0
activation_2 (Activation)                        (None, 81, 9)                                    0
Total params: 7,605,401
Trainable params: 7,605,145
Non-trainable params: 256
```

```
In [31]: train_idx = int(len(df)*0.95)
df = df.sample(frac=1).reset_index(drop=True)
training_generator = DataGenerator(df.iloc[train_idx:], subset = "train", batch_size=batch_size)
validation_generator = DataGenerator(df.iloc[train_idx:], subset = "train", batch_size=batch_size)
```

```
In [32]: from keras.callbacks import Callback, ModelCheckpoint, ReduceLROnPlateau
filepath1="weights-improvement-{epoch:02d}-{val_accuracy:.2f}.hdf5"
filepath2 = "best_weights.hdf5"
checkpoint1 = ModelCheckpoint(filepath1, monitor='val_accuracy', verbose=1, save_best_only=True)
checkpoint2 = ModelCheckpoint(filepath2, monitor='val_accuracy', verbose=1, save_best_only=True)

reduce_lr = ReduceLROnPlateau(
    monitor='val_loss',
    patience=3,
    verbose=1,
    min_lr=1e-6
)
callbacks_list = [checkpoint1,checkpoint2,reduce_lr]
```

```
In [21]: history = model.fit(training_generator, validation_data = validation_generator, epochs=100, callbacks=callbacks_list)

Epoch 1/5
1484/1484 [=====] - 426s 287ms/step - loss: 0.3713 - accuracy: 0.1110 - val_loss: 0.3688 - val_accuracy: 0.1132

Epoch 00001: val accuracy improved from 0.10992 to 0.11315, saving model to weights-improvement-01-0.11.hdf5

Epoch 00001: val_accuracy improved from 0.10992 to 0.11315, saving model to best_weights.hdf5
Epoch 2/5
1484/1484 [=====] - 429s 289ms/step - loss: 0.3570 - accuracy: 0.1111 - val_loss: 0.3624 - val_accuracy: 0.1066

Epoch 00002: val_accuracy did not improve from 0.11315

Epoch 00002: val_accuracy did not improve from 0.11315
Epoch 3/5
1484/1484 [=====] - 434s 292ms/step - loss: 0.3472 - accuracy: 0.1112 - val_loss: 0.3595 - val_accuracy: 0.1109

Epoch 00003: val_accuracy did not improve from 0.11315

Epoch 00003: val_accuracy did not improve from 0.11315
Epoch 4/5
1484/1484 [=====] - 423s 285ms/step - loss: 0.3394 - accuracy: 0.1112 - val_loss: 0.3588 - val_accuracy: 0.1090

Epoch 00004: val_accuracy did not improve from 0.11315

Epoch 00004: val_accuracy did not improve from 0.11315
Epoch 5/5
1484/1484 [=====] - 432s 291ms/step - loss: 0.3327 - accuracy: 0.1112 - val_loss: 0.3588 - val_accuracy: 0.1142

Epoch 00005: val accuracy improved from 0.11315 to 0.11417, saving model to weights-improvement-05-0.11.hdf5

Epoch 00005: val_accuracy improved from 0.11315 to 0.11417, saving model to best_weights.hdf5
```

```
In [33]: # saving the best weights
model.load_weights('best_weights.hdf5')
```

```
In [49]: # using the model to solve, reference the CNN guide for this section of the reformatted code
def solve_sudoku(game):
    game = np.array(game).reshape((9,9,1))
    game = (game/9)-0.5
    while(1):
        out = model.predict(game.reshape((1,9,9,1)))
        out = out.squeeze()
        pred = np.argmax(out, axis=1).reshape((9,9))+1
        prob = np.around(np.max(out, axis=1).reshape((9,9)), 2)
        game = ((game+.5)*9).reshape((9,9))
        mask = (game==0)
        if(mask.sum()==0):
            break
        prob_new = prob*mask
        ind = np.argmax(prob_new)
        x, y = (ind//9),(ind%9)
        val = pred[x][y]
        game[x][y] = val
        game = (game/9)-0.5
    return pred
    return game
```

```
In [50]: # sample board input
game = [[0, 0, 0, 2, 4, 8, 7, 1, 9],
        [7, 0, 0, 0, 0, 0, 5, 0, 0],
        [0, 2, 0, 6, 0, 5, 0, 0, 4],
        [0, 0, 0, 4, 6, 0, 3, 7, 8],
        [0, 1, 3, 0, 0, 7, 9, 0, 6],
        [6, 0, 0, 0, 0, 2, 0, 5, 1],
        [0, 7, 2, 1, 0, 0, 0, 0, 3],
        [8, 0, 0, 0, 0, 6, 0, 0, 7],
        [9, 0, 0, 0, 2, 0, 1, 8, 0]]

game = solve_sudoku(game)
print(game)

[[3 5 6 2 4 8 7 1 9]
 [7 4 8 3 1 9 5 6 2]
 [1 2 9 6 7 5 8 3 4]
 [2 9 5 4 6 1 3 7 8]
 [4 1 3 8 5 7 9 2 6]
 [6 8 7 9 3 2 4 5 1]
 [5 7 2 1 8 4 6 9 3]
 [8 3 1 5 9 6 2 4 7]
 [9 6 4 7 2 3 1 8 5]]
```

```
In [51]: #check if the array is correct
np.sum(game, axis=1)
```

Out[51]: array([45, 45, 45, 45, 45, 45, 45, 45, 45], dtype=int64)

In [ ]: