```
1 import numpy as np
2 import pandas as pd
3 import tensorflow as tf
4 import datetime
```

Loading and running our previous MNIST handwritten digits dataset

Using 20 Epochs

```
1 from keras.datasets import mnist
 2 from keras.utils import to_categorical
 3
 4 (train_images, train_labels), (test_images, test_labels) = mnist.load_data()
 6 train images = train images.reshape((60000, 28, 28, 1))
 7 train_images = train_images.astype('float32') / 255
 9 test_images = test_images.reshape((10000, 28, 28, 1))
10 test_images = test_images.astype('float32') / 255
12 train_labels = to_categorical(train_labels)
13 test labels = to categorical(test labels)
15 from keras import layers
16 from keras import models
17 from keras.layers import Dropout
18
19 model = models.Sequential()
20 model.add(layers.Conv2D(32, (3, 3), activation='relu', input shape=(28, 28, 1)))
21 model.add(Dropout(0.5))
22 model.add(layers.MaxPooling2D((2, 2)))
23 model.add(layers.Conv2D(16, (3, 3), activation='relu'))
24 model.add(layers.Flatten())
25 model.add(layers.Dense(16, activation='relu'))
26 model.add(Dropout(0.5))
27 model.add(layers.Dense(10, activation='sigmoid'))
28
29 model.summary()
30
31 model.compile(optimizer='rmsprop',
32
                 loss='categorical crossentropy',
33
                 metrics=['accuracy'])
34
35 model.fit(train images, train labels,
             epochs=20,
36
37
             batch size=64,
38
             validation split = 0.2)
```

	7 1331511	mem_5 (2).ipyi	10 00.	iaboratory			
dropout (Dropout)	(None,	26,	26,	32))	0		
<pre>max_pooling2d (MaxPooling2D)</pre>	(None,	13,	13,	32)) (0	_	
conv2d_1 (Conv2D)	(None,	11,	11,	16))	4624	_	
flatten (Flatten)	(None,	1936)		(0	_	
dense (Dense)	(None,	16)				30992	_	
dropout_1 (Dropout)	(None,	16)			()	_	
dense_1 (Dense)	(None,					170	=	
Total params: 36,106 Trainable params: 36,106 Non-trainable params: 0								
Epoch 1/20 750/750 [====================================	=====	====]	_	44s	57ms/ste	o - loss:		- accura
Epoch 2/20 750/750 [====================================	======	====]	_	42s	57ms/ste	o - loss:	0.5766	- accura
Epoch 3/20 750/750 [====================================	======	====]	_	43s	57ms/ste	o - loss:	0.4781	- accura
Epoch 4/20 750/750 [====================================	======	====]	_	43s	57ms/ste	o - loss:	0.4366	- accura
Epoch 5/20 750/750 [====================================	======	====]	_	43s	58ms/ste	o - loss:	0.4164	- accura
Epoch 6/20 750/750 [====================================	======	====]	_	44s	59ms/ste	o - loss:	0.4172	- accura
Epoch 7/20 750/750 [====================================	======	====]	_	44s	59ms/step	p - loss:	0.4150	- accura
Epoch 8/20 750/750 [====================================	=====	====]	_	44s	59ms/step	o - loss:	0.4064	- accura
Epoch 9/20 750/750 [====================================	======	====]	_	45s	60ms/ste	o - loss:	0.4108	- accura
Epoch 10/20 750/750 [====================================	======	====]	-	45s	60ms/ste	o - loss:	0.4087	- accura
Epoch 11/20 750/750 [====================================	======	====]	-	44s	59ms/ste	o - loss:	0.4174	- accura
Epoch 12/20 750/750 [====================================	======	====]	-	45s	60ms/step	o - loss:	0.4185	- accura
Epoch 13/20 750/750 [====================================	=====	====]	_	46s	61ms/step	o - loss:	0.4139	- accura
Epoch 14/20 750/750 [====================================	=====	====]	_	42s	57ms/ste	o - loss:	0.4172	- accura
Epoch 15/20 750/750 [====================================	=====	====]	-	43s	58ms/ste	o - loss:	0.4173	- accura
750/750 [====================================	=====	====]	-	43s	58ms/ste	o - loss:	0.4192	- accura
Epoch 17/20 750/750 [====================================	=====	====]	-	44s	59ms/ste	o - loss:	0.4132	- accura
Epoch 18/20 750/750 [====================================	=====	====]	_	46s	61ms/step	o - loss:	0.4186	- accura
Epoch 19/20 750/750 [====================================	=====	====]	_	43s	57ms/step	o - loss:	0.4178	- accura

Evaluating the model on the test set

Automatically reducing overall runtime by usning earlystopping

- we can use callback_early_stopping to interrupt training once a target metric being monitored
 has stopped improving for a fixed number of epochs.
- this callback allows us to interrupt training as soon as we start overfitting, thus avoiding having to retrain our model for a smaller number of epochs.

MODEL CHECKPOINT AND EARLY STOPPING CALLBACKS

```
1 import keras
 3 callbacks list = [
       keras.callbacks.EarlyStopping(
 5
           monitor='acc',
 6
           patience=1,
 7
 8
       keras.callbacks.ModelCheckpoint(
 9
           filepath='my model.h5',
10
           monitor='val loss',
           save best only=True,
11
12
13 1
14
15 model.compile(optimizer='rmsprop',
                 loss='binary crossentropy',
16
                 metrics=['acc'])
17
18
19 model.fit(train images, train labels,
20
             epochs=20,
             batch size=64,
21
22
             callbacks=callbacks list,
23
             validation data=(test images, test labels))
```

REDUCING LEARNING RATE ON PLATEAU CALLBACK

```
1 callbacks_list = [
 2
       keras.callbacks.ReduceLROnPlateau(
           monitor='val loss',
 3
 4
           factor=0.1,
 5
           patience=10,
 6
       )
 7 ]
 8
 9 model.fit(train_images, train_labels,
10
             epochs=20,
             batch size=32,
11
             callbacks=callbacks list,
12
13
             validation_data=(test_images, test_labels))
```

```
Epoch 1/20
Epoch 2/20
Epoch 3/20
Epoch 4/20
Epoch 5/20
Epoch 6/20
Epoch 7/20
Epoch 8/20
Epoch 9/20
Epoch 10/20
Epoch 11/20
Epoch 12/20
Epoch 13/20
Epoch 14/20
Epoch 15/20
Epoch 16/20
Epoch 17/20
Epoch 18/20
```

```
1 import keras
 2 import numpy as np
 4 class ActivationLogger(keras.callbacks.Callback):
 5
 6
       def set model(self, model):
 7
           self.model = model
 8
           layer_outputs = [layer.output for layer in model.layers]
           self.activations model = keras.models.Model(model.input,
 9
                                                         layer_outputs)
10
11
12
       def on epoch end(self, epoch, logs=None):
13
           if self.validation_data is None:
               raise RuntimeError('Requires validation_data.')
14
15
           validation sample = self.validation data[0][0:1]
16
           activations = self.activations model.predict(validation sample)
17
18
           f = open('activations at epoch ' + str(epoch) + '.npz', 'w')
           np.savez(f, activations)
19
           f.close()
20
```

Training the model with a TensorBoard callback

```
1 from keras.callbacks import TensorBoard
 2 tbCallBack = TensorBoard(log_dir='./log', histogram_freq=1,
 3
                             write graph=True,
 4
                             write grads=True,
 5
                             batch size=128,
 6
                             write images=True)
 7
 8 model.fit(train images, train labels,
 9
             batch size=128,
10
             epochs=20,
11
             verbose=1,
             validation data=(test_images, test_labels),
12
13
             callbacks=[tbCallBack])
```

Epoch 3/20

```
Epoch 4/20
Epoch 5/20
Epoch 6/20
Epoch 7/20
Epoch 8/20
Epoch 9/20
Epoch 10/20
Epoch 11/20
Epoch 12/20
Epoch 13/20
Epoch 14/20
Epoch 15/20
Epoch 16/20
Epoch 17/20
Epoch 18/20
Epoch 19/20
Epoch 20/20
<tensorflow.python.keras.callbacks.History at 0x7f159fa4a890>
```

```
1 model.save_weights('model.h2')
2 model.load_weights('model.h2')
3 model.evaluate(test_images, test_labels)
```

Visualizing the Tensorboard outcome

```
1 %load_ext tensorboard
2 %tensorboard --logdir log
```

INACTIVE

IMAGES

TensorBoard

SCALARS

Show data download links epoch_acc Ignore outliers in chart scaling **Tooltip sorting** epoch_acc default method: 0.86 0.82 Smoothing 0.78 0.6 0 0.74 0.7 0.66 Horizontal Axis 0.62 **RELATIVE** STEP 6 8 10 12 14 16 run to download WALL epoch_loss Runs Write a regex to filter runs epoch_loss train 0.2) validation 0.16 **TOGGLE ALL RUNS** 0.12 log 0.08 0.04 0 10 12 14 16 run to download

INACTIVE

TensorBoard

The tensorboard extension is already loaded. To reload it, use: ${\tt \$reload_ext\ tensorboard}$

IMAGES

SCALARS

Reusing TensorBoard on port 6006 (pid 969), started 0:17:22 ago. (Use '!kill 969

Main Graph Search nodes. Regexes suppor... Fit to Screen AssignAdd... AssignAddVariab.. AssignAdd... AssignAdd... Download PNG Run Identity_1 train (1) Tag Default div_no_nan_1 (3) Upload Choose File AssignAddVariab. Graph AssignAddVariab. AssignAddVaria Conceptual Graph Profile Trace inputs Color O Structure **RMSprop** Device ✓ Close legend. IteratorGet... Graph (* = expandable) Namespace*? OpNode ?

The tensorboard extension is already loaded. To reload it, use:
 %reload_ext tensorboard
Reusing TensorBoard on port 6006 (pid 969), started 0:17:26 ago. (Use '!kill 969)

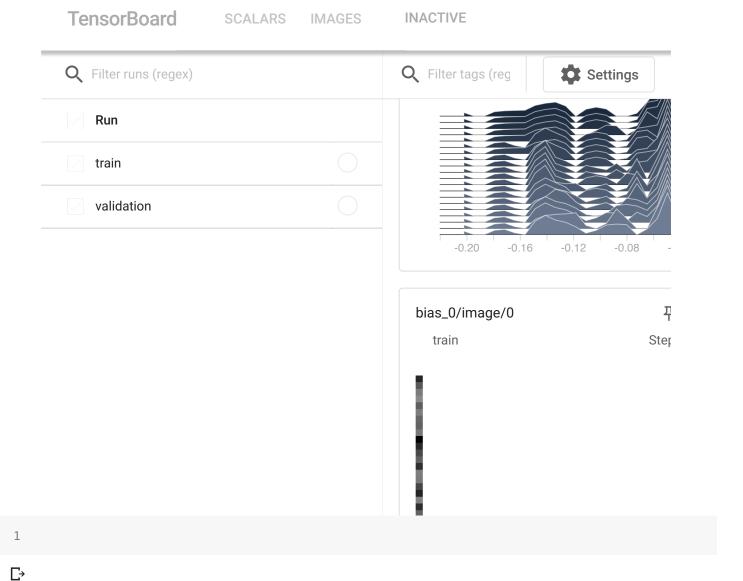
TensorBoard SCALARS IMAGES INACTIVE

Q Filter tags (regular expressions supported) Histogram mode **OVERLAY OFFSET** conv2d_1 2 ^ Offset time axis train conv2d_1/bias_0 **STEP RELATIVE** WALL Runs Write a regex to filter runs -0.35 -0.25 -0.15 -0.05 train validation train conv2d_1/kernel_0 **TOGGLE ALL RUNS** log

1

The tensorboard extension is already loaded. To reload it, use: %reload_ext tensorboard

Reusing TensorBoard on port 6006 (pid 969), started 0:23:00 ago. (Use '!kill 969



 $https://colab.research.google.com/drive/1eC-Nku_kQiI6bBTq6OJ1dnY-9C5xP6rC\#scrollTo=uHzGbQZh5NUt\&printMode=true$

The tensorboard extension is already loaded. To reload it, use: %reload_ext tensorboard

Reusing TensorBoard on port 6006 (pid 969), started 0:23:09 ago. (Use '!kill 969



✓ 0s completed at 8:05 PM