

openmic

December 19, 2019

```
[1]: import librosa as lb
import librosa.display
import pandas as pd
import scipy
import json
import numpy as np
import sklearn
from sklearn.metrics import classification_report
from sklearn.model_selection import train_test_split
import os
from pylab import plot, show, figure, imshow, xlim, ylim, title
import matplotlib.pyplot as plt
import keras
from keras.utils import np_utils
from keras import layers
from keras import models
```

Using TensorFlow backend.

```
[2]: #CONSTANTS

DATA_DIR = "openmic-2018/"
CATEGORY_COUNT = 8

[3]: df = pd.read_csv('openmic-2018/openmic-2018-aggregated-labels.csv')
del df['relevance']
del df['num_responses']

[4]: labels = df.values
labels

[4]: array([[ '000046_3840', 'clarinet'],
            [ '000046_3840', 'flute'],
            [ '000046_3840', 'trumpet'],
            ...,
            [ '155311_453120', 'saxophone'],
            [ '155311_453120', 'trumpet'],
            [ '155311_453120', 'trombone']], dtype=object)
```

```
[5]: y, sr = lb.load(DATA_DIR + 'audio/000/000963_88320.ogg')
S = lb.feature.melspectrogram(y=y, sr=sr)

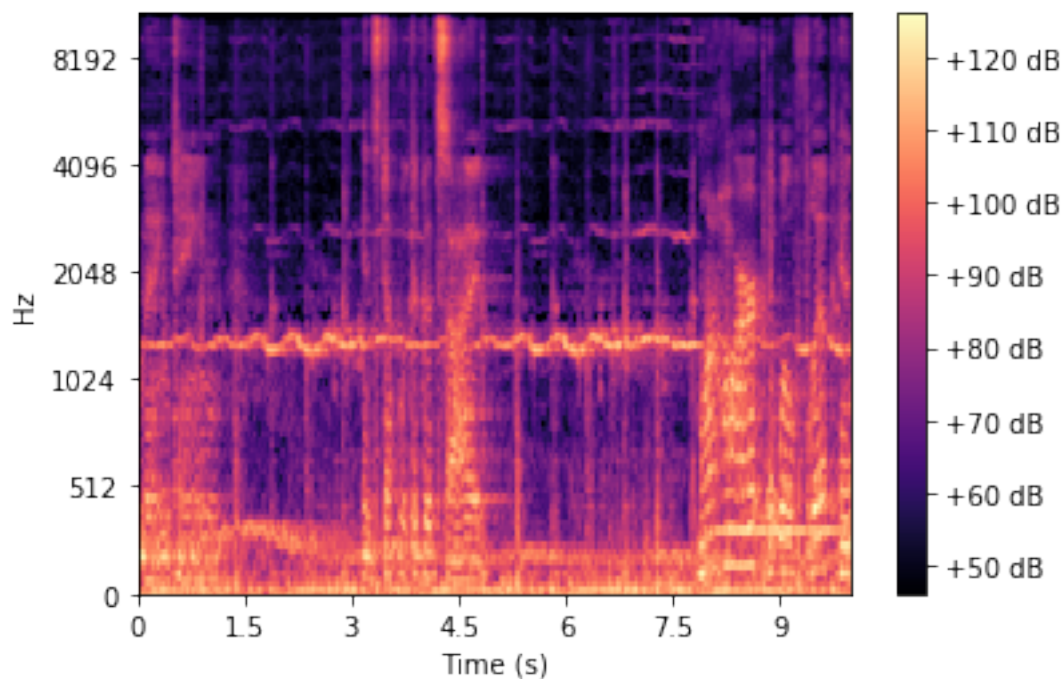
S_dB = lb.power_to_db(S, ref=0) # 10 * log10(S / ref)

print(y.shape)
print(sr)
print(S.shape)
print(S_dB.shape)
```

```
(220512,)
22050
(128, 431)
(128, 431)
```

```
[6]: librosa.display.specshow(S_dB, x_axis='s', y_axis='mel')
plt.colorbar(format='%+2.0f dB')
```

```
[6]: <matplotlib.colorbar.Colorbar at 0x197a1f621d0>
```



```
[7]: OPENMIC = np.load(os.path.join(DATA_DIR, 'openmic-2018.npz'),
↳ allow_pickle=True)
print(list(OPENMIC.keys()))
```

```
['X', 'Y_true', 'Y_mask', 'sample_key']
```

```
[8]: X, Y_true, Y_mask, sample_key = OPENMIC['X'], OPENMIC['Y_true'],  
      ↪OPENMIC['Y_mask'], OPENMIC['sample_key']  
      #print(X.shape)  
      #X = []  
      #print(len(sample_key))  
      #for key in sample_key:  
      #    key_dir = key[:3]  
      #    y, sr = lb.load(DATA_DIR + 'audio/' + key_dir + '/' + key + '.ogg')  
      #    X.append(lb.feature.melspectrogram(y=y, sr=sr))  
      #    print(len(X))
```

```
[9]: with open(os.path.join(DATA_DIR, 'class-map.json'), 'r') as f:  
      class_map = json.load(f)
```

```
[10]: split_train, split_test, X_train, X_test, Y_true_train, Y_true_test,  
      ↪Y_mask_train, Y_mask_test = train_test_split(sample_key, X, Y_true, Y_mask)  
      split_val, split_test, X_val, X_test, Y_true_val, Y_true_test, Y_mask_val,  
      ↪Y_mask_test = train_test_split(split_test, X_test, Y_true_test, Y_mask_test,  
      ↪test_size=0.5)  
  
      train_set = np.asarray(set(split_train))  
      test_set = np.asarray(set(split_test))  
      print('# Train: {}, # Val: {}, # Test: {}'.format(len(split_train),  
      ↪len(split_test), len(split_val)))
```

```
# Train: 15000, # Val: 2500, # Test: 2500
```

```
[11]: print(X_train.shape)  
      print(X_val.shape)  
      print(X_test.shape)
```

```
(15000, 10, 128)
```

```
(2500, 10, 128)
```

```
(2500, 10, 128)
```

```
[12]: THRESHOLD = 0.5  
  
      # This dictionary will include the classifiers for each model  
      mymodels = dict()  
  
      # We'll iterate over all instrument classes, and fit a model for each one  
      # After training, we'll print a classification report for each instrument  
      for instrument in class_map:  
  
          # Map the instrument name to its column number  
          inst_num = class_map[instrument]
```

```

# Step 1: sub-sample the data

# First, we need to select down to the data for which we have annotations
# This is what the mask arrays are for
train_inst = Y_mask_train[:, inst_num]
val_inst = Y_mask_val[:, inst_num]
test_inst = Y_mask_test[:, inst_num]

# Here, we're using the Y_mask_train array to slice out only the training
→examples
# for which we have annotations for the given class
X_train_inst = X_train[train_inst]
X_val_inst = X_val[val_inst]

# Step 3: simplify the data by averaging over time

# Let's arrange the data for a sklearn Random Forest model
# Instead of having time-varying features, we'll summarize each track by
→its mean feature vector over time
X_train_inst_sklearn = np.mean(X_train_inst, axis=1)

# Again, we slice the labels to the annotated examples
# We threshold the label likelihoods at 0.5 to get binary labels
Y_true_train_inst = Y_true_train[train_inst, inst_num] >= THRESHOLD
Y_true_val_inst = Y_true_val[val_inst, inst_num] >= THRESHOLD

# Repeat the above slicing and dicing but for the test set
X_test_inst = X_test[test_inst]
X_test_inst_sklearn = np.mean(X_test_inst, axis=1)
Y_true_test_inst = Y_true_test[test_inst, inst_num] >= THRESHOLD

X_train_inst = X_train_inst.astype('float32')
X_val_inst = X_val_inst.astype('float32')
X_train_inst_sklearn = X_train_inst_sklearn.astype('float32')
X_train_inst_sklearn = lb.util.normalize(X_train_inst_sklearn)
# X_train_inst = S_dB
print(X_train_inst.shape)
shape = X_train_inst.shape
X_train_inst = X_train_inst.reshape(shape[0],1, shape[1], shape[2])
shape = X_val_inst.shape
X_val_inst = X_val_inst.reshape(shape[0],1, shape[1], shape[2])
shape = X_test_inst.shape
X_test_inst = X_test_inst.reshape(shape[0],1, shape[1], shape[2])
#X_train_inst = X_train_inst.reshape(1,1,431,128)
print(X_train_inst.shape)

```

```

print(Y_true_train_inst[0])
# Step 3.
# Initialize a new classifier
import keras, os
from keras.models import Sequential
from keras.layers import Dense, Conv2D, MaxPool2D, Flatten, Dropout
from keras.preprocessing.image import ImageDataGenerator
import numpy as np
model = models.Sequential()

# model.add(layers.Conv2D(filters=8, kernel_size=(3,3), activation='relu',
→input_shape=(10,128,1,)))
model.
→add(Conv2D(input_shape=(1,10,128), data_format="channels_first", filters=64, kernel_size=(3,3)
→activation="relu"))
model.add(Conv2D(filters=32, kernel_size=(3,3), padding="same",
→activation="relu"))
model.add(MaxPool2D(pool_size=(3,3), strides=(2,2)))
model.add(Conv2D(filters=128, kernel_size=(3,3), padding="same",
→activation="relu"))
model.add(Conv2D(filters=128, kernel_size=(3,3), padding="same",
→activation="relu"))
model.add(MaxPool2D(pool_size=(2,2), strides=(2,2)))
model.add(Dropout(0.2))
model.add(Conv2D(filters=256, kernel_size=(3,3), padding="same",
→activation="relu"))
model.add(Conv2D(filters=256, kernel_size=(3,3), padding="same",
→activation="relu"))
model.add(Conv2D(filters=256, kernel_size=(3,3), padding="same",
→activation="relu"))
model.add(MaxPool2D(pool_size=(2,2), strides=(2,2)))
model.add(Dropout(0.2))
model.add(Conv2D(filters=512, kernel_size=(3,3), padding="same",
→activation="relu"))
model.add(Conv2D(filters=512, kernel_size=(3,3), padding="same",
→activation="relu"))
model.add(Conv2D(filters=512, kernel_size=(3,3), padding="same",
→activation="relu"))
model.add(layers.Flatten())
model.add(layers.Dense(units=4096, activation='relu'))
model.add(layers.Dense(units=4096, activation='relu'))
model.add(layers.Dense(units=1, activation='sigmoid'))

model.compile(loss='binary_crossentropy',
              optimizer=keras.optimizers.Adam(lr=0.0001),
              metrics = ['accuracy'])

```

```

# model.summary()
# Step 4.
history = model.fit(X_train_inst,Y_true_train_inst , epochs=10,
→batch_size=32, validation_data=(X_val_inst,Y_true_val_inst))

plt.figure(figsize=(9,4))

plt.subplot(1,2,1)
plt.plot(history.history['acc'])
plt.plot(history.history['val_acc'])
plt.title('Model accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(['Train accuracy', 'Validation accuracy'], loc='upper left')

plt.subplot(1,2,2)
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('Model loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['Train loss', 'Validation loss'], loc='upper left')
plt.show()

loss, acc = model.evaluate(X_test_inst, Y_true_test_inst)
print('Test loss: {}'.format(loss))
print('Test accuracy: {:.2%}'.format(acc))

# Step 5.
# Finally, we'll evaluate the model on both train and test
Y_pred_train = model.predict(X_train_inst)
Y_pred_test = model.predict(X_test_inst)
Y_pred_train_bool = Y_pred_train > THRESHOLD - 0.3 #THRESHOLD (should be
→lower than 0.5)
Y_pred_test_bool = Y_pred_test > THRESHOLD - 0.3 #THRESHOLD (should be
→lower than 0.5)
print(Y_pred_train[0])
print('-' * 52)
print(instrument)
print('\tTRAIN')
print(classification_report(Y_true_train_inst, Y_pred_train_bool))

print(Y_true_train_inst[3])
print(Y_pred_train[3])
print('\tTEST')
print(classification_report(Y_true_test_inst, Y_pred_test_bool))

```

```

sum = 0
# for i, prob in enumerate(Y_pred_train):
#     print (i)
#     print (prob)
#     sum += prob
# print(sum)
# Store the classifier in our dictionary
mymodels[instrument] = model

```

(1594, 10, 128)

(1594, 1, 10, 128)

True

WARNING:tensorflow:From C:\Users\hjani\Documents\Conda\lib\site-packages\tensorflow\python\framework\op_def_library.py:263: colocate_with (from tensorflow.python.framework.ops) is deprecated and will be removed in a future version.

Instructions for updating:

Colocations handled automatically by placer.

WARNING:tensorflow:From C:\Users\hjani\Documents\Conda\lib\site-packages\keras\backend\tensorflow_backend.py:3445: calling dropout (from tensorflow.python.ops.nn_ops) with keep_prob is deprecated and will be removed in a future version.

Instructions for updating:

Please use `rate` instead of `keep_prob`. Rate should be set to `rate = 1 - keep_prob`.

WARNING:tensorflow:From C:\Users\hjani\Documents\Conda\lib\site-packages\tensorflow\python\ops\math_ops.py:3066: to_int32 (from tensorflow.python.ops.math_ops) is deprecated and will be removed in a future version.

Instructions for updating:

Use tf.cast instead.

Train on 1594 samples, validate on 256 samples

Epoch 1/10

1594/1594 [=====] - 116s 73ms/step - loss: 0.5646 - acc: 0.7641 - val_loss: 0.5210 - val_acc: 0.7695

Epoch 2/10

1594/1594 [=====] - 117s 74ms/step - loss: 0.5260 - acc: 0.7647 - val_loss: 0.4802 - val_acc: 0.7695

Epoch 3/10

1594/1594 [=====] - 118s 74ms/step - loss: 0.4534 - acc: 0.7823 - val_loss: 0.3944 - val_acc: 0.8164

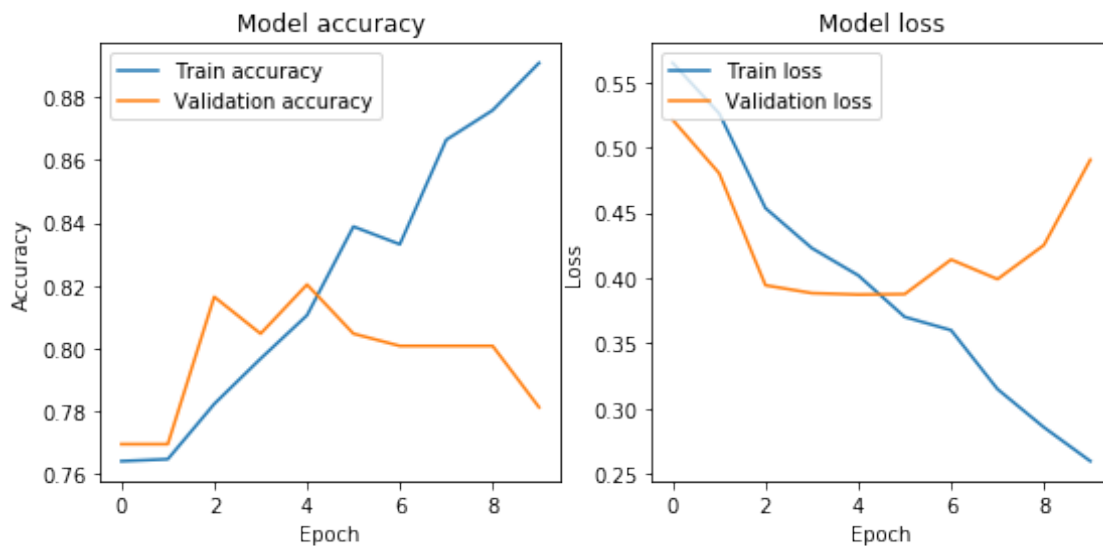
Epoch 4/10

1594/1594 [=====] - 121s 76ms/step - loss: 0.4228 - acc: 0.7967 - val_loss: 0.3884 - val_acc: 0.8047

Epoch 5/10

1594/1594 [=====] - 116s 73ms/step - loss: 0.4019 - acc: 0.8105 - val_loss: 0.3873 - val_acc: 0.8203

Epoch 6/10
 1594/1594 [=====] - 114s 72ms/step - loss: 0.3702 -
 acc: 0.8388 - val_loss: 0.3876 - val_acc: 0.8047
 Epoch 7/10
 1594/1594 [=====] - 112s 70ms/step - loss: 0.3600 -
 acc: 0.8331 - val_loss: 0.4140 - val_acc: 0.8008
 Epoch 8/10
 1594/1594 [=====] - 111s 70ms/step - loss: 0.3149 -
 acc: 0.8664 - val_loss: 0.3991 - val_acc: 0.8008
 Epoch 9/10
 1594/1594 [=====] - 112s 70ms/step - loss: 0.2857 -
 acc: 0.8758 - val_loss: 0.4251 - val_acc: 0.8008
 Epoch 10/10
 1594/1594 [=====] - 110s 69ms/step - loss: 0.2598 -
 acc: 0.8908 - val_loss: 0.4904 - val_acc: 0.7812



221/221 [=====] - 2s 9ms/step
 Test loss: 0.437995489636158
 Test accuracy: 81.45%
 [0.6548375]

 accordion

	TRAIN			
	precision	recall	f1-score	support
False	0.99	0.91	0.94	1219
True	0.76	0.96	0.85	375
accuracy				0.92 1594

macro avg	0.87	0.93	0.90	1594
weighted avg	0.93	0.92	0.92	1594

False

[0.10981479]

TEST

	precision	recall	f1-score	support
False	0.90	0.83	0.87	166
True	0.59	0.73	0.65	55
accuracy			0.81	221
macro avg	0.75	0.78	0.76	221
weighted avg	0.82	0.81	0.81	221

(1664, 10, 128)

(1664, 1, 10, 128)

False

Train on 1664 samples, validate on 265 samples

Epoch 1/10

1664/1664 [=====] - 121s 73ms/step - loss: 0.6420 -
acc: 0.6635 - val_loss: 0.5529 - val_acc: 0.6981

Epoch 2/10

1664/1664 [=====] - 119s 71ms/step - loss: 0.5296 -
acc: 0.7194 - val_loss: 0.4689 - val_acc: 0.7660

Epoch 3/10

1664/1664 [=====] - 116s 70ms/step - loss: 0.4610 -
acc: 0.7686 - val_loss: 0.4292 - val_acc: 0.7849

Epoch 4/10

1664/1664 [=====] - 135s 81ms/step - loss: 0.4598 -
acc: 0.7915 - val_loss: 0.5193 - val_acc: 0.7774

Epoch 5/10

1664/1664 [=====] - 123s 74ms/step - loss: 0.4405 -
acc: 0.7837 - val_loss: 0.4020 - val_acc: 0.8075

Epoch 6/10

1664/1664 [=====] - 122s 73ms/step - loss: 0.3896 -
acc: 0.7999 - val_loss: 0.4002 - val_acc: 0.7925

Epoch 7/10

1664/1664 [=====] - 120s 72ms/step - loss: 0.3789 -
acc: 0.8155 - val_loss: 0.4013 - val_acc: 0.7887

Epoch 8/10

1664/1664 [=====] - 124s 75ms/step - loss: 0.3661 -
acc: 0.8263 - val_loss: 0.4295 - val_acc: 0.7962

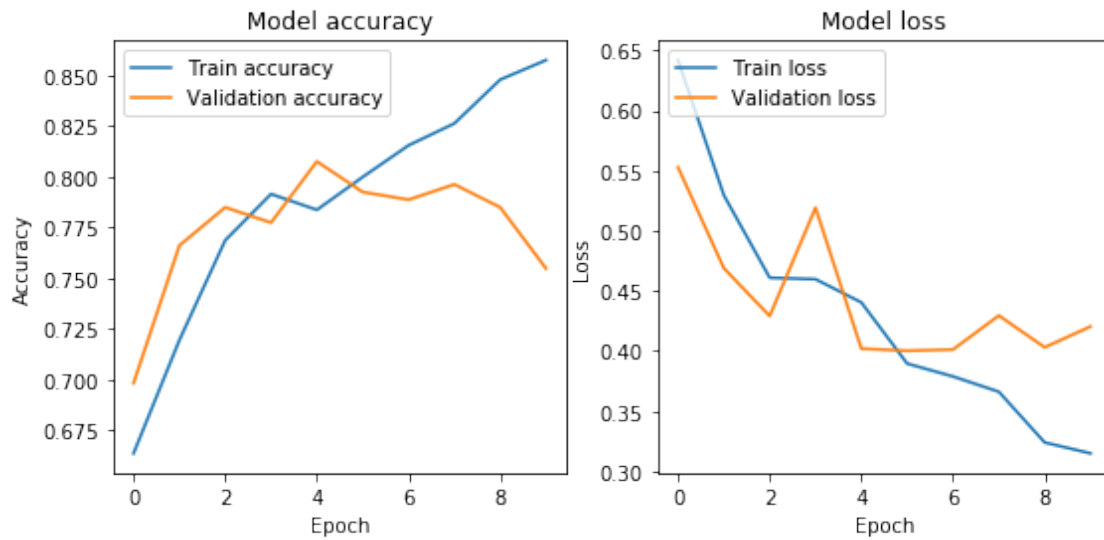
Epoch 9/10

1664/1664 [=====] - 128s 77ms/step - loss: 0.3239 -
acc: 0.8480 - val_loss: 0.4031 - val_acc: 0.7849

Epoch 10/10

1664/1664 [=====] - 130s 78ms/step - loss: 0.3149 -

acc: 0.8576 - val_loss: 0.4204 - val_acc: 0.7547



289/289 [=====] - 3s 10ms/step

Test loss: 0.47478529078737675

Test accuracy: 75.78%

[0.13815153]

banjo

TRAIN					
	precision	recall	f1-score	support	
False	0.99	0.78	0.87	1114	
True	0.69	0.98	0.81	550	
accuracy			0.85	1664	
macro avg	0.84	0.88	0.84	1664	
weighted avg	0.89	0.85	0.85	1664	

False

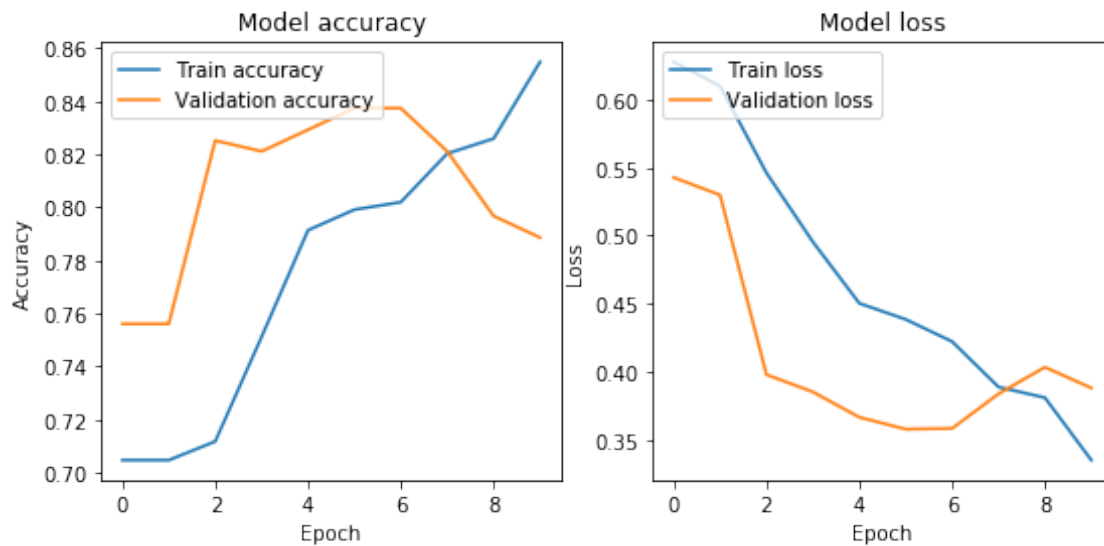
[0.00191763]

TEST					
	precision	recall	f1-score	support	
False	0.89	0.71	0.79	187	
True	0.61	0.83	0.70	102	
accuracy			0.75	289	
macro avg	0.75	0.77	0.74	289	
weighted avg	0.79	0.75	0.76	289	

```

(1419, 10, 128)
(1419, 1, 10, 128)
False
Train on 1419 samples, validate on 246 samples
Epoch 1/10
1419/1419 [=====] - 112s 79ms/step - loss: 0.6278 -
acc: 0.7047 - val_loss: 0.5428 - val_acc: 0.7561
Epoch 2/10
1419/1419 [=====] - 110s 78ms/step - loss: 0.6102 -
acc: 0.7047 - val_loss: 0.5299 - val_acc: 0.7561
Epoch 3/10
1419/1419 [=====] - 109s 77ms/step - loss: 0.5461 -
acc: 0.7118 - val_loss: 0.3977 - val_acc: 0.8252
Epoch 4/10
1419/1419 [=====] - 112s 79ms/step - loss: 0.4954 -
acc: 0.7512 - val_loss: 0.3850 - val_acc: 0.8211
Epoch 5/10
1419/1419 [=====] - 109s 77ms/step - loss: 0.4502 -
acc: 0.7914 - val_loss: 0.3663 - val_acc: 0.8293
Epoch 6/10
1419/1419 [=====] - 104s 74ms/step - loss: 0.4384 -
acc: 0.7992 - val_loss: 0.3576 - val_acc: 0.8374
Epoch 7/10
1419/1419 [=====] - 105s 74ms/step - loss: 0.4221 -
acc: 0.8020 - val_loss: 0.3581 - val_acc: 0.8374
Epoch 8/10
1419/1419 [=====] - 107s 75ms/step - loss: 0.3886 -
acc: 0.8203 - val_loss: 0.3836 - val_acc: 0.8211
Epoch 9/10
1419/1419 [=====] - 114s 80ms/step - loss: 0.3808 -
acc: 0.8259 - val_loss: 0.4032 - val_acc: 0.7967
Epoch 10/10
1419/1419 [=====] - 120s 85ms/step - loss: 0.3348 -
acc: 0.8548 - val_loss: 0.3879 - val_acc: 0.7886

```



```
223/223 [=====] - 3s 11ms/step
Test loss: 0.45042256349405363
Test accuracy: 76.23%
[0.01511016]
```

base

TRAIN				
	precision	recall	f1-score	support
False	1.00	0.57	0.72	1000
True	0.49	1.00	0.66	419
accuracy			0.70	1419
macro avg	0.75	0.78	0.69	1419
weighted avg	0.85	0.70	0.71	1419

False

```
[0.3186671]
```

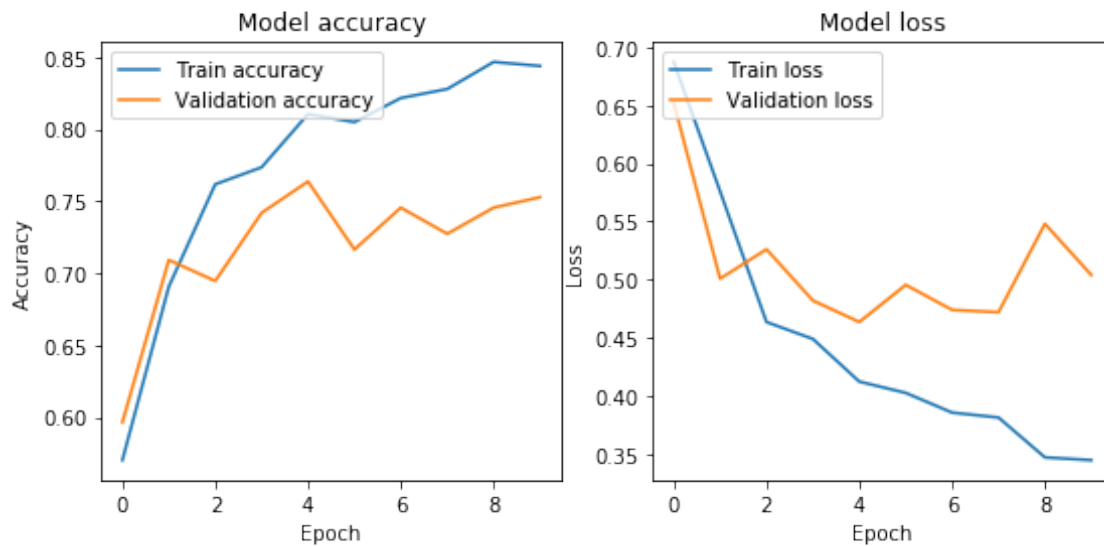
TEST				
	precision	recall	f1-score	support
False	0.94	0.54	0.69	153
True	0.48	0.93	0.63	70
accuracy			0.66	223
macro avg	0.71	0.74	0.66	223
weighted avg	0.80	0.66	0.67	223

(1435, 10, 128)

```

(1435, 1, 10, 128)
False
Train on 1435 samples, validate on 275 samples
Epoch 1/10
1435/1435 [=====] - 120s 84ms/step - loss: 0.6873 -
acc: 0.5700 - val_loss: 0.6523 - val_acc: 0.5964
Epoch 2/10
1435/1435 [=====] - 103s 72ms/step - loss: 0.5760 -
acc: 0.6906 - val_loss: 0.5007 - val_acc: 0.7091
Epoch 3/10
1435/1435 [=====] - 102s 71ms/step - loss: 0.4634 -
acc: 0.7617 - val_loss: 0.5260 - val_acc: 0.6945
Epoch 4/10
1435/1435 [=====] - 111s 78ms/step - loss: 0.4489 -
acc: 0.7735 - val_loss: 0.4819 - val_acc: 0.7418
Epoch 5/10
1435/1435 [=====] - 101s 71ms/step - loss: 0.4121 -
acc: 0.8105 - val_loss: 0.4635 - val_acc: 0.7636
Epoch 6/10
1435/1435 [=====] - 101s 70ms/step - loss: 0.4024 -
acc: 0.8049 - val_loss: 0.4954 - val_acc: 0.7164
Epoch 7/10
1435/1435 [=====] - 115s 80ms/step - loss: 0.3855 -
acc: 0.8216 - val_loss: 0.4739 - val_acc: 0.7455
Epoch 8/10
1435/1435 [=====] - 104s 72ms/step - loss: 0.3812 -
acc: 0.8279 - val_loss: 0.4719 - val_acc: 0.7273
Epoch 9/10
1435/1435 [=====] - 103s 72ms/step - loss: 0.3469 -
acc: 0.8467 - val_loss: 0.5480 - val_acc: 0.7455
Epoch 10/10
1435/1435 [=====] - 99s 69ms/step - loss: 0.3445 - acc:
0.8439 - val_loss: 0.5039 - val_acc: 0.7527

```



239/239 [=====] - 2s 9ms/step

Test loss: 0.5433883659510433

Test accuracy: 74.90%

[0.00171691]

cello

TRAIN					
	precision	recall	f1-score	support	
False	0.98	0.79	0.87	833	
True	0.77	0.98	0.86	602	
accuracy			0.87	1435	
macro avg	0.88	0.88	0.87	1435	
weighted avg	0.89	0.87	0.87	1435	

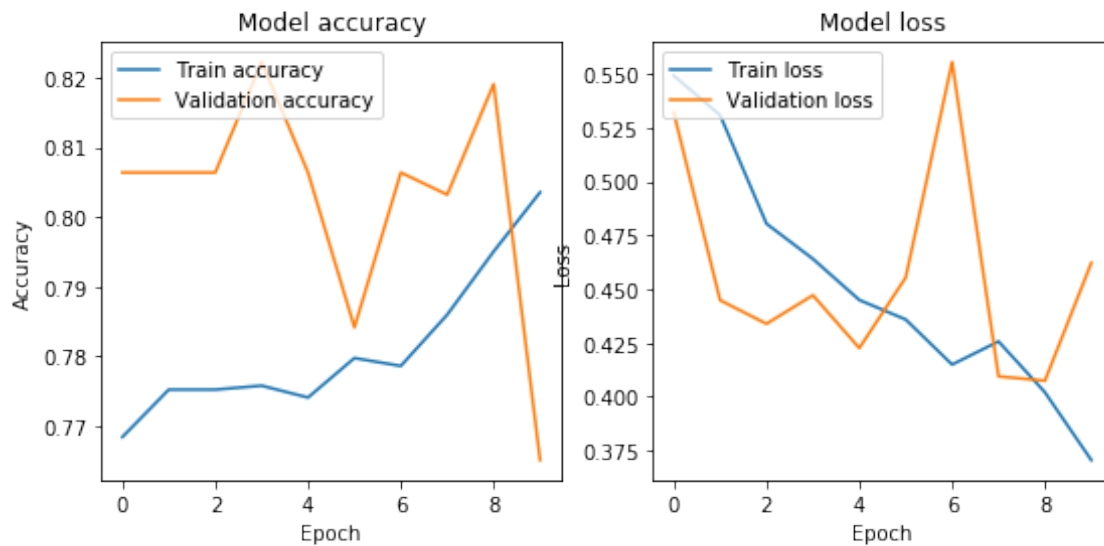
False

[0.00046462]

TEST					
	precision	recall	f1-score	support	
False	0.83	0.63	0.72	128	
True	0.67	0.85	0.75	111	
accuracy			0.73	239	
macro avg	0.75	0.74	0.73	239	
weighted avg	0.75	0.73	0.73	239	

(1766, 10, 128)

```
(1766, 1, 10, 128)
True
Train on 1766 samples, validate on 315 samples
Epoch 1/10
1766/1766 [=====] - 135s 77ms/step - loss: 0.5493 -
acc: 0.7684 - val_loss: 0.5322 - val_acc: 0.8063
Epoch 2/10
1766/1766 [=====] - 127s 72ms/step - loss: 0.5309 -
acc: 0.7752 - val_loss: 0.4448 - val_acc: 0.8063
Epoch 3/10
1766/1766 [=====] - 129s 73ms/step - loss: 0.4804 -
acc: 0.7752 - val_loss: 0.4339 - val_acc: 0.8063
Epoch 4/10
1766/1766 [=====] - 127s 72ms/step - loss: 0.4641 -
acc: 0.7758 - val_loss: 0.4471 - val_acc: 0.8222
Epoch 5/10
1766/1766 [=====] - 127s 72ms/step - loss: 0.4449 -
acc: 0.7741 - val_loss: 0.4225 - val_acc: 0.8063
Epoch 6/10
1766/1766 [=====] - 125s 71ms/step - loss: 0.4358 -
acc: 0.7797 - val_loss: 0.4554 - val_acc: 0.7841
Epoch 7/10
1766/1766 [=====] - 129s 73ms/step - loss: 0.4149 -
acc: 0.7786 - val_loss: 0.5557 - val_acc: 0.8063
Epoch 8/10
1766/1766 [=====] - 129s 73ms/step - loss: 0.4257 -
acc: 0.7860 - val_loss: 0.4094 - val_acc: 0.8032
Epoch 9/10
1766/1766 [=====] - 129s 73ms/step - loss: 0.4020 -
acc: 0.7950 - val_loss: 0.4074 - val_acc: 0.8190
Epoch 10/10
1766/1766 [=====] - 124s 70ms/step - loss: 0.3704 -
acc: 0.8035 - val_loss: 0.4622 - val_acc: 0.7651
```



304/304 [=====] - 3s 9ms/step

Test loss: 0.5573157260292455

Test accuracy: 66.45%

[0.5319122]

clarinet

TRAIN				
	precision	recall	f1-score	support
False	1.00	0.46	0.63	1369
True	0.35	1.00	0.52	397
accuracy			0.58	1766
macro avg	0.67	0.73	0.58	1766
weighted avg	0.85	0.58	0.61	1766

False

[0.0020178]

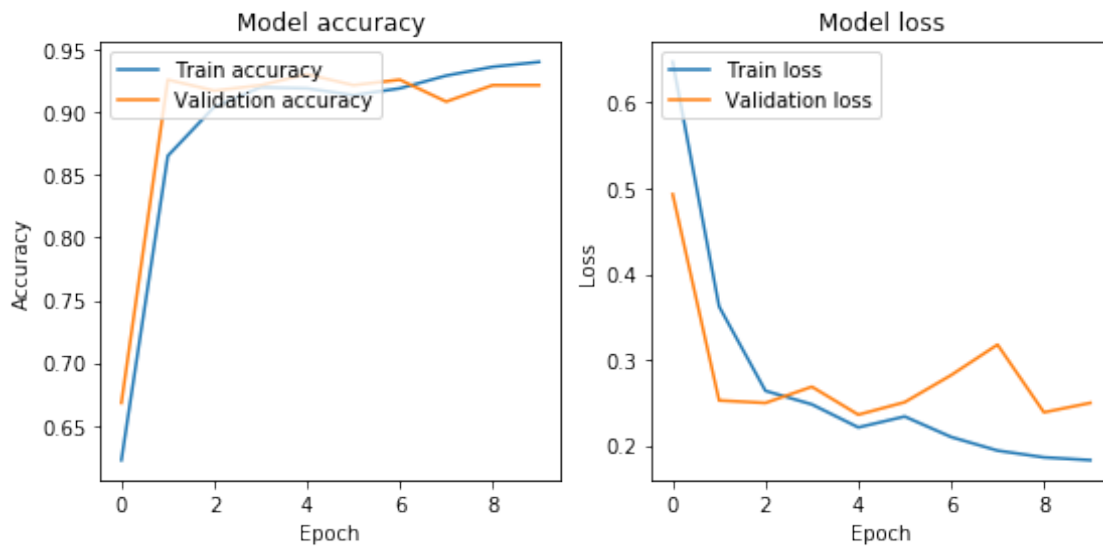
TEST				
	precision	recall	f1-score	support
False	0.94	0.39	0.55	229
True	0.33	0.92	0.49	75
accuracy			0.52	304
macro avg	0.63	0.66	0.52	304
weighted avg	0.79	0.52	0.54	304

(1282, 10, 128)


```

(1282, 1, 10, 128)
False
Train on 1282 samples, validate on 229 samples
Epoch 1/10
1282/1282 [=====] - 94s 74ms/step - loss: 0.6474 - acc:
0.6225 - val_loss: 0.4931 - val_acc: 0.6681
Epoch 2/10
1282/1282 [=====] - 93s 72ms/step - loss: 0.3621 - acc:
0.8651 - val_loss: 0.2525 - val_acc: 0.9258
Epoch 3/10
1282/1282 [=====] - 94s 73ms/step - loss: 0.2636 - acc:
0.9041 - val_loss: 0.2496 - val_acc: 0.9170
Epoch 4/10
1282/1282 [=====] - 93s 73ms/step - loss: 0.2479 - acc:
0.9197 - val_loss: 0.2684 - val_acc: 0.9214
Epoch 5/10
1282/1282 [=====] - 91s 71ms/step - loss: 0.2210 - acc:
0.9189 - val_loss: 0.2358 - val_acc: 0.9301
Epoch 6/10
1282/1282 [=====] - 90s 70ms/step - loss: 0.2337 - acc:
0.9134 - val_loss: 0.2504 - val_acc: 0.9214
Epoch 7/10
1282/1282 [=====] - 90s 70ms/step - loss: 0.2098 - acc:
0.9189 - val_loss: 0.2820 - val_acc: 0.9258
Epoch 8/10
1282/1282 [=====] - 90s 70ms/step - loss: 0.1940 - acc:
0.9290 - val_loss: 0.3176 - val_acc: 0.9083
Epoch 9/10
1282/1282 [=====] - 94s 73ms/step - loss: 0.1861 - acc:
0.9360 - val_loss: 0.2386 - val_acc: 0.9214
Epoch 10/10
1282/1282 [=====] - 93s 72ms/step - loss: 0.1828 - acc:
0.9399 - val_loss: 0.2496 - val_acc: 0.9214

```



224/224 [=====] - 2s 10ms/step

Test loss: 0.2819361814430782

Test accuracy: 91.52%

[0.00207818]

cymbals

TRAIN					
	precision	recall	f1-score	support	
False	1.00	0.84	0.91	469	
True	0.92	1.00	0.96	813	
accuracy			0.94	1282	
macro avg	0.96	0.92	0.93	1282	
weighted avg	0.95	0.94	0.94	1282	

True

[0.95123166]

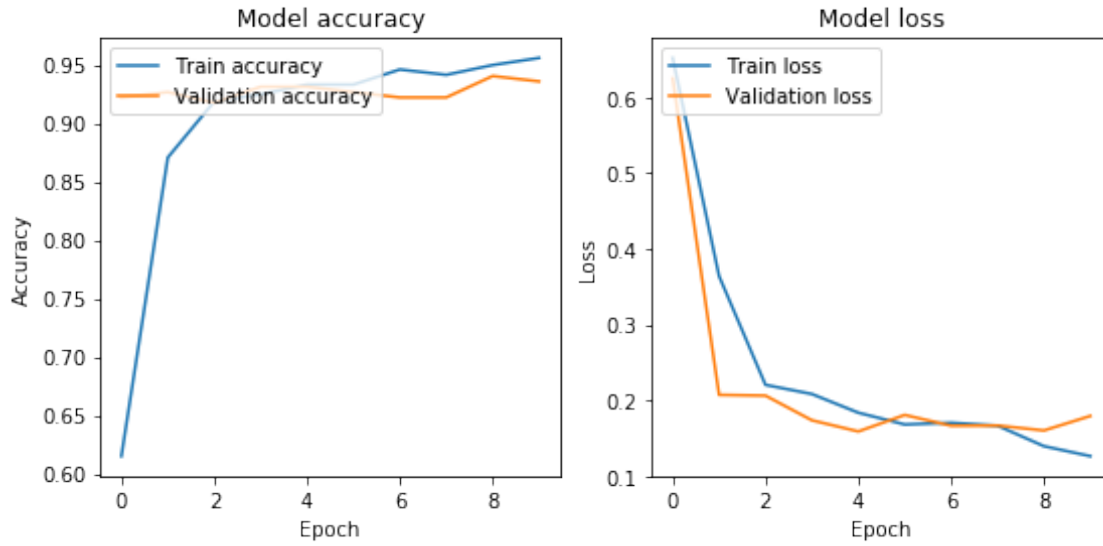
TEST					
	precision	recall	f1-score	support	
False	0.96	0.72	0.82	75	
True	0.88	0.99	0.93	149	
accuracy			0.90	224	
macro avg	0.92	0.85	0.88	224	
weighted avg	0.90	0.90	0.89	224	

(1314, 10, 128)

```

(1314, 1, 10, 128)
False
Train on 1314 samples, validate on 218 samples
Epoch 1/10
1314/1314 [=====] - 98s 74ms/step - loss: 0.6521 - acc:
0.6149 - val_loss: 0.6250 - val_acc: 0.9220
Epoch 2/10
1314/1314 [=====] - 94s 71ms/step - loss: 0.3639 - acc:
0.8706 - val_loss: 0.2071 - val_acc: 0.9266
Epoch 3/10
1314/1314 [=====] - 93s 71ms/step - loss: 0.2203 - acc:
0.9186 - val_loss: 0.2061 - val_acc: 0.9174
Epoch 4/10
1314/1314 [=====] - 93s 71ms/step - loss: 0.2082 - acc:
0.9247 - val_loss: 0.1735 - val_acc: 0.9312
Epoch 5/10
1314/1314 [=====] - 95s 72ms/step - loss: 0.1835 - acc:
0.9330 - val_loss: 0.1588 - val_acc: 0.9312
Epoch 6/10
1314/1314 [=====] - 97s 74ms/step - loss: 0.1680 - acc:
0.9330 - val_loss: 0.1805 - val_acc: 0.9266
Epoch 7/10
1314/1314 [=====] - 96s 73ms/step - loss: 0.1702 - acc:
0.9460 - val_loss: 0.1663 - val_acc: 0.9220
Epoch 8/10
1314/1314 [=====] - 96s 73ms/step - loss: 0.1664 - acc:
0.9414 - val_loss: 0.1665 - val_acc: 0.9220
Epoch 9/10
1314/1314 [=====] - 93s 71ms/step - loss: 0.1393 - acc:
0.9498 - val_loss: 0.1601 - val_acc: 0.9404
Epoch 10/10
1314/1314 [=====] - 92s 70ms/step - loss: 0.1262 - acc:
0.9559 - val_loss: 0.1792 - val_acc: 0.9358

```



215/215 [=====] - 2s 9ms/step
 Test loss: 0.20257972887782164
 Test accuracy: 93.95%
 [0.0016264]

drums

TRAIN				
	precision	recall	f1-score	support
False	1.00	0.92	0.96	485
True	0.96	1.00	0.98	829
accuracy			0.97	1314
macro avg	0.98	0.96	0.97	1314
weighted avg	0.97	0.97	0.97	1314

False

[0.47499835]

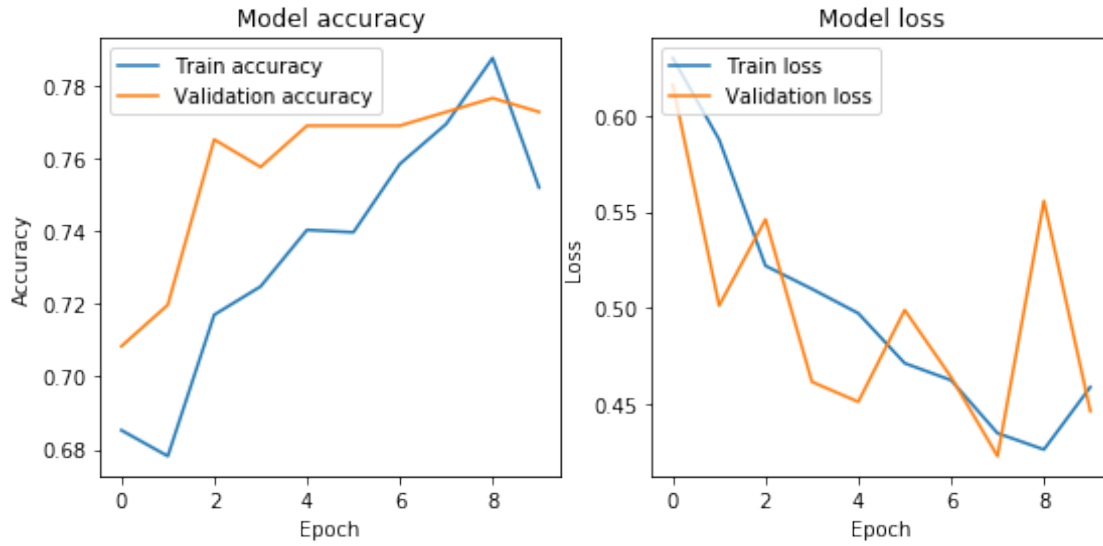
TEST				
	precision	recall	f1-score	support
False	0.96	0.86	0.91	81
True	0.92	0.98	0.95	134
accuracy			0.93	215
macro avg	0.94	0.92	0.93	215
weighted avg	0.94	0.93	0.93	215

(1544, 10, 128)

```

(1544, 1, 10, 128)
True
Train on 1544 samples, validate on 264 samples
Epoch 1/10
1544/1544 [=====] - 118s 77ms/step - loss: 0.6297 -
acc: 0.6852 - val_loss: 0.6158 - val_acc: 0.7083
Epoch 2/10
1544/1544 [=====] - 109s 71ms/step - loss: 0.5874 -
acc: 0.6781 - val_loss: 0.5013 - val_acc: 0.7197
Epoch 3/10
1544/1544 [=====] - 113s 73ms/step - loss: 0.5219 -
acc: 0.7170 - val_loss: 0.5460 - val_acc: 0.7652
Epoch 4/10
1544/1544 [=====] - 110s 71ms/step - loss: 0.5099 -
acc: 0.7247 - val_loss: 0.4618 - val_acc: 0.7576
Epoch 5/10
1544/1544 [=====] - 109s 71ms/step - loss: 0.4973 -
acc: 0.7403 - val_loss: 0.4514 - val_acc: 0.7689
Epoch 6/10
1544/1544 [=====] - 110s 71ms/step - loss: 0.4714 -
acc: 0.7396 - val_loss: 0.4990 - val_acc: 0.7689
Epoch 7/10
1544/1544 [=====] - 110s 71ms/step - loss: 0.4626 -
acc: 0.7584 - val_loss: 0.4642 - val_acc: 0.7689
Epoch 8/10
1544/1544 [=====] - 109s 70ms/step - loss: 0.4351 -
acc: 0.7694 - val_loss: 0.4232 - val_acc: 0.7727
Epoch 9/10
1544/1544 [=====] - 113s 73ms/step - loss: 0.4266 -
acc: 0.7876 - val_loss: 0.5556 - val_acc: 0.7765
Epoch 10/10
1544/1544 [=====] - 110s 71ms/step - loss: 0.4590 -
acc: 0.7519 - val_loss: 0.4465 - val_acc: 0.7727

```



276/276 [=====] - 3s 9ms/step
 Test loss: 0.6236255212106566
 Test accuracy: 70.29%
 [0.65752745]

 flute

TRAIN					
	precision	recall	f1-score	support	
False	0.97	0.69	0.80	1073	
True	0.57	0.95	0.71	471	
accuracy			0.77	1544	
macro avg	0.77	0.82	0.76	1544	
weighted avg	0.85	0.77	0.78	1544	

True

[0.74695337]

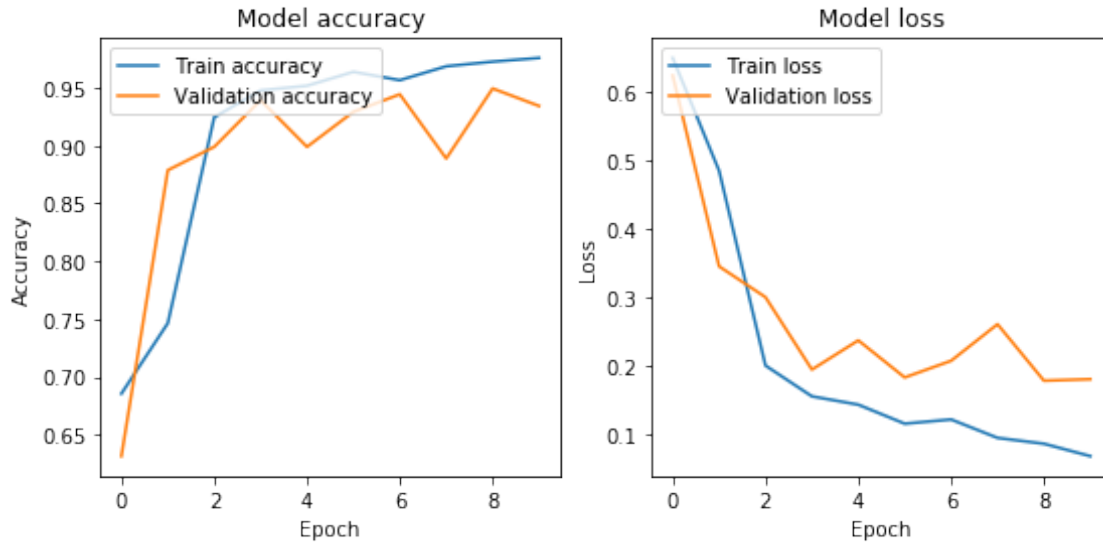
TEST					
	precision	recall	f1-score	support	
False	0.84	0.58	0.69	177	
True	0.52	0.81	0.63	99	
accuracy			0.66	276	
macro avg	0.68	0.70	0.66	276	
weighted avg	0.73	0.66	0.67	276	

(1246, 10, 128)

```

(1246, 1, 10, 128)
True
Train on 1246 samples, validate on 198 samples
Epoch 1/10
1246/1246 [=====] - 96s 77ms/step - loss: 0.6507 - acc:
0.6854 - val_loss: 0.6250 - val_acc: 0.6313
Epoch 2/10
1246/1246 [=====] - 91s 73ms/step - loss: 0.4856 - acc:
0.7464 - val_loss: 0.3456 - val_acc: 0.8788
Epoch 3/10
1246/1246 [=====] - 88s 71ms/step - loss: 0.2004 - acc:
0.9246 - val_loss: 0.3005 - val_acc: 0.8990
Epoch 4/10
1246/1246 [=====] - 90s 73ms/step - loss: 0.1554 - acc:
0.9478 - val_loss: 0.1945 - val_acc: 0.9394
Epoch 5/10
1246/1246 [=====] - 89s 72ms/step - loss: 0.1432 - acc:
0.9518 - val_loss: 0.2373 - val_acc: 0.8990
Epoch 6/10
1246/1246 [=====] - 88s 70ms/step - loss: 0.1154 - acc:
0.9639 - val_loss: 0.1830 - val_acc: 0.9293
Epoch 7/10
1246/1246 [=====] - 87s 70ms/step - loss: 0.1215 - acc:
0.9567 - val_loss: 0.2074 - val_acc: 0.9444
Epoch 8/10
1246/1246 [=====] - 88s 71ms/step - loss: 0.0946 - acc:
0.9687 - val_loss: 0.2610 - val_acc: 0.8889
Epoch 9/10
1246/1246 [=====] - 87s 70ms/step - loss: 0.0860 - acc:
0.9727 - val_loss: 0.1783 - val_acc: 0.9495
Epoch 10/10
1246/1246 [=====] - 87s 70ms/step - loss: 0.0679 - acc:
0.9759 - val_loss: 0.1805 - val_acc: 0.9343

```



206/206 [=====] - 2s 9ms/step

Test loss: 0.1476010004673334

Test accuracy: 96.60%

[0.9957912]

guitar

TRAIN				
	precision	recall	f1-score	support
False	1.00	0.95	0.97	376
True	0.98	1.00	0.99	870
accuracy			0.98	1246
macro avg	0.99	0.97	0.98	1246
weighted avg	0.99	0.98	0.98	1246

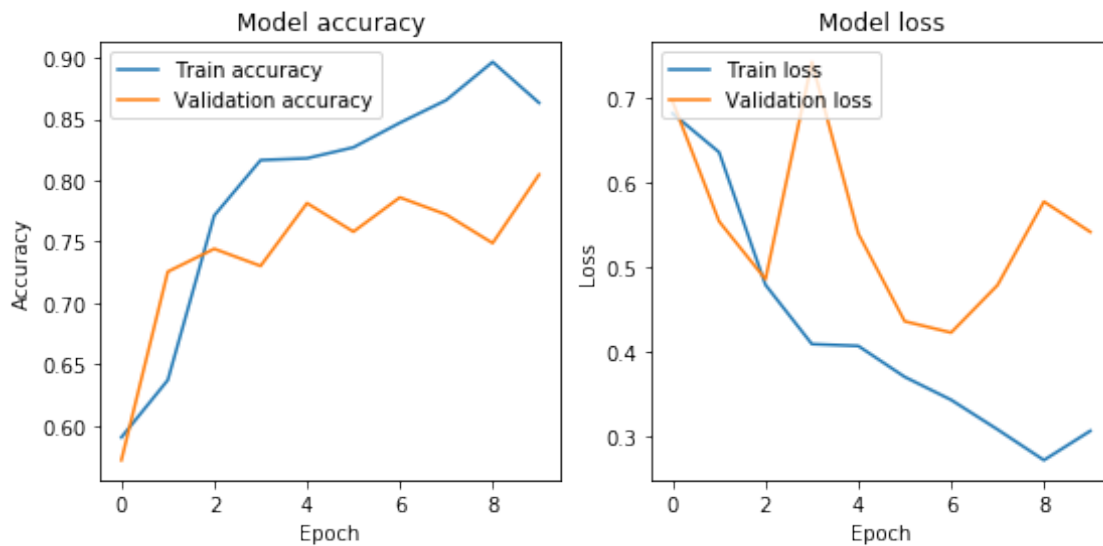
True

[0.9896692]

TEST				
	precision	recall	f1-score	support
False	0.91	0.95	0.93	63
True	0.98	0.96	0.97	143
accuracy			0.96	206
macro avg	0.94	0.96	0.95	206
weighted avg	0.96	0.96	0.96	206

(1351, 10, 128)


```
(1351, 1, 10, 128)
True
Train on 1351 samples, validate on 215 samples
Epoch 1/10
1351/1351 [=====] - 106s 79ms/step - loss: 0.6811 -
acc: 0.5907 - val_loss: 0.6950 - val_acc: 0.5721
Epoch 2/10
1351/1351 [=====] - 97s 71ms/step - loss: 0.6354 - acc:
0.6373 - val_loss: 0.5536 - val_acc: 0.7256
Epoch 3/10
1351/1351 [=====] - 96s 71ms/step - loss: 0.4786 - acc:
0.7713 - val_loss: 0.4852 - val_acc: 0.7442
Epoch 4/10
1351/1351 [=====] - 96s 71ms/step - loss: 0.4085 - acc:
0.8164 - val_loss: 0.7421 - val_acc: 0.7302
Epoch 5/10
1351/1351 [=====] - 95s 70ms/step - loss: 0.4063 - acc:
0.8179 - val_loss: 0.5391 - val_acc: 0.7814
Epoch 6/10
1351/1351 [=====] - 96s 71ms/step - loss: 0.3697 - acc:
0.8268 - val_loss: 0.4354 - val_acc: 0.7581
Epoch 7/10
1351/1351 [=====] - 95s 70ms/step - loss: 0.3426 - acc:
0.8468 - val_loss: 0.4221 - val_acc: 0.7860
Epoch 8/10
1351/1351 [=====] - 95s 70ms/step - loss: 0.3076 - acc:
0.8653 - val_loss: 0.4782 - val_acc: 0.7721
Epoch 9/10
1351/1351 [=====] - 99s 74ms/step - loss: 0.2713 - acc:
0.8964 - val_loss: 0.5770 - val_acc: 0.7488
Epoch 10/10
1351/1351 [=====] - 97s 72ms/step - loss: 0.3059 - acc:
0.8631 - val_loss: 0.5411 - val_acc: 0.8047
```



236/236 [=====] - 2s 10ms/step

Test loss: 0.5606073989706525

Test accuracy: 78.39%

[0.9939307]

mallet_percussion

TRAIN					
	precision	recall	f1-score	support	
False	0.99	0.90	0.94	798	
True	0.87	0.99	0.93	553	
accuracy			0.94	1351	
macro avg	0.93	0.94	0.94	1351	
weighted avg	0.94	0.94	0.94	1351	

False

[0.00234887]

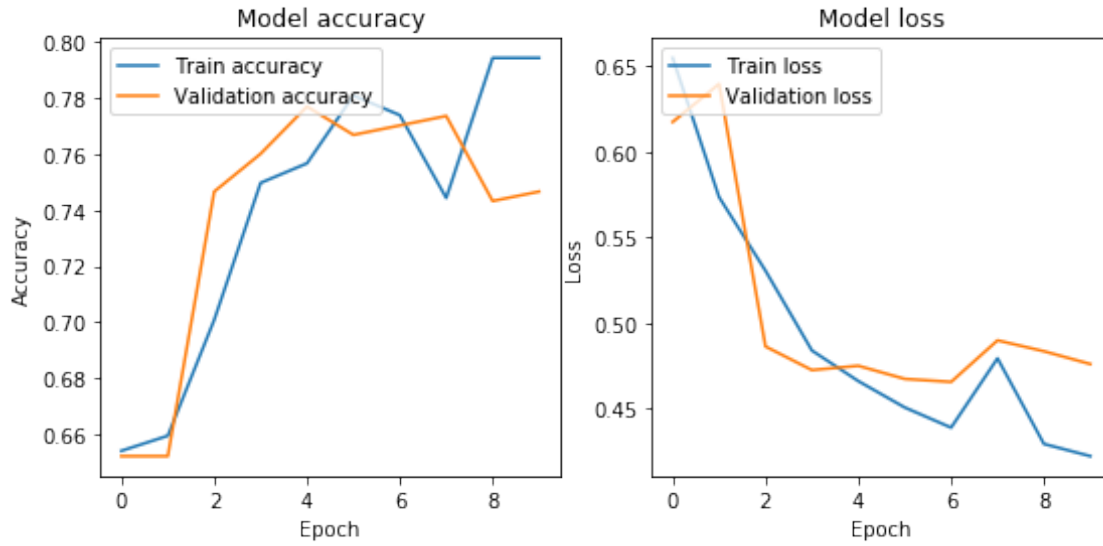
TEST					
	precision	recall	f1-score	support	
False	0.88	0.78	0.83	148	
True	0.69	0.82	0.75	88	
accuracy			0.80	236	
macro avg	0.79	0.80	0.79	236	
weighted avg	0.81	0.80	0.80	236	

(1858, 10, 128)

```

(1858, 1, 10, 128)
True
Train on 1858 samples, validate on 296 samples
Epoch 1/10
1858/1858 [=====] - 137s 74ms/step - loss: 0.6545 -
acc: 0.6539 - val_loss: 0.6172 - val_acc: 0.6520
Epoch 2/10
1858/1858 [=====] - 131s 71ms/step - loss: 0.5734 -
acc: 0.6593 - val_loss: 0.6394 - val_acc: 0.6520
Epoch 3/10
1858/1858 [=====] - 133s 72ms/step - loss: 0.5305 -
acc: 0.7008 - val_loss: 0.4863 - val_acc: 0.7466
Epoch 4/10
1858/1858 [=====] - 134s 72ms/step - loss: 0.4840 -
acc: 0.7497 - val_loss: 0.4726 - val_acc: 0.7601
Epoch 5/10
1858/1858 [=====] - 130s 70ms/step - loss: 0.4662 -
acc: 0.7567 - val_loss: 0.4751 - val_acc: 0.7770
Epoch 6/10
1858/1858 [=====] - 131s 70ms/step - loss: 0.4508 -
acc: 0.7809 - val_loss: 0.4674 - val_acc: 0.7669
Epoch 7/10
1858/1858 [=====] - 131s 71ms/step - loss: 0.4390 -
acc: 0.7740 - val_loss: 0.4657 - val_acc: 0.7703
Epoch 8/10
1858/1858 [=====] - 134s 72ms/step - loss: 0.4794 -
acc: 0.7443 - val_loss: 0.4899 - val_acc: 0.7736
Epoch 9/10
1858/1858 [=====] - 132s 71ms/step - loss: 0.4295 -
acc: 0.7944 - val_loss: 0.4836 - val_acc: 0.7432
Epoch 10/10
1858/1858 [=====] - 130s 70ms/step - loss: 0.4224 -
acc: 0.7944 - val_loss: 0.4761 - val_acc: 0.7466

```



310/310 [=====] - 3s 9ms/step
 Test loss: 0.47123505319318465
 Test accuracy: 77.10%
 [0.56316614]

 mandolin

TRAIN				
	precision	recall	f1-score	support
False	0.99	0.57	0.72	1215
True	0.55	0.99	0.71	643
accuracy			0.71	1858
macro avg	0.77	0.78	0.71	1858
weighted avg	0.84	0.71	0.72	1858

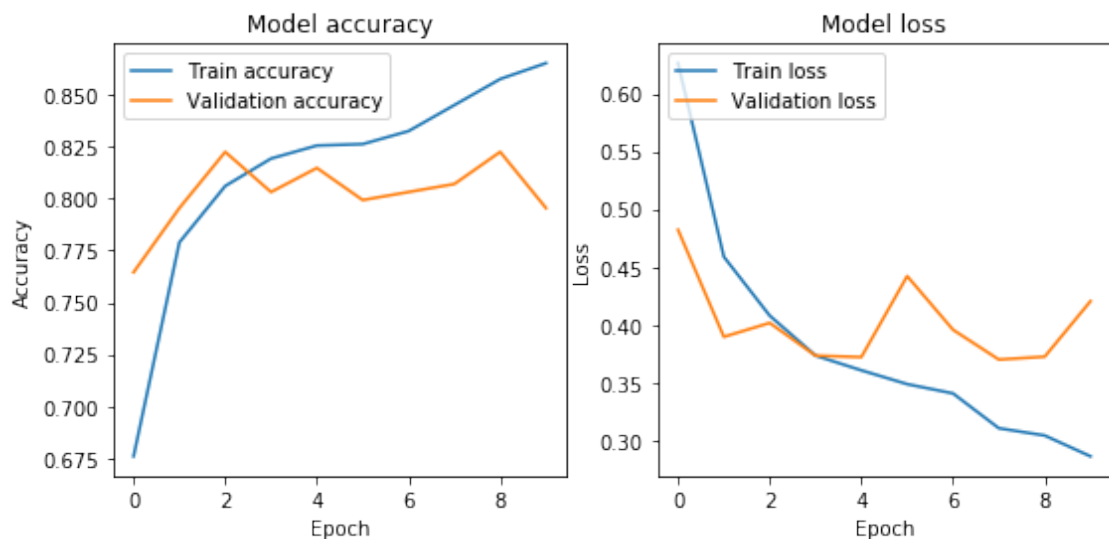
True

[0.5705857]

TEST				
	precision	recall	f1-score	support
False	0.96	0.49	0.65	211
True	0.47	0.96	0.63	99
accuracy			0.64	310
macro avg	0.72	0.73	0.64	310
weighted avg	0.81	0.64	0.65	310

(1438, 10, 128)

```
(1438, 1, 10, 128)
True
Train on 1438 samples, validate on 259 samples
Epoch 1/10
1438/1438 [=====] - 108s 75ms/step - loss: 0.6263 -
acc: 0.6759 - val_loss: 0.4821 - val_acc: 0.7645
Epoch 2/10
1438/1438 [=====] - 103s 72ms/step - loss: 0.4587 -
acc: 0.7789 - val_loss: 0.3895 - val_acc: 0.7954
Epoch 3/10
1438/1438 [=====] - 104s 73ms/step - loss: 0.4078 -
acc: 0.8060 - val_loss: 0.4015 - val_acc: 0.8224
Epoch 4/10
1438/1438 [=====] - 102s 71ms/step - loss: 0.3733 -
acc: 0.8192 - val_loss: 0.3732 - val_acc: 0.8031
Epoch 5/10
1438/1438 [=====] - 101s 70ms/step - loss: 0.3605 -
acc: 0.8255 - val_loss: 0.3720 - val_acc: 0.8147
Epoch 6/10
1438/1438 [=====] - 103s 72ms/step - loss: 0.3486 -
acc: 0.8261 - val_loss: 0.4420 - val_acc: 0.7992
Epoch 7/10
1438/1438 [=====] - 103s 71ms/step - loss: 0.3406 -
acc: 0.8324 - val_loss: 0.3956 - val_acc: 0.8031
Epoch 8/10
1438/1438 [=====] - 102s 71ms/step - loss: 0.3104 -
acc: 0.8449 - val_loss: 0.3698 - val_acc: 0.8069
Epoch 9/10
1438/1438 [=====] - 104s 72ms/step - loss: 0.3042 -
acc: 0.8574 - val_loss: 0.3723 - val_acc: 0.8224
Epoch 10/10
1438/1438 [=====] - 103s 72ms/step - loss: 0.2861 -
acc: 0.8651 - val_loss: 0.4204 - val_acc: 0.7954
```



193/193 [=====] - 2s 10ms/step
 Test loss: 0.4051659104892009
 Test accuracy: 80.83%
 [0.7450688]

 organ

TRAIN					
	precision	recall	f1-score	support	
False	0.99	0.86	0.92	972	
True	0.77	0.98	0.86	466	
accuracy			0.90	1438	
macro avg	0.88	0.92	0.89	1438	
weighted avg	0.92	0.90	0.90	1438	

True

[0.9714943]

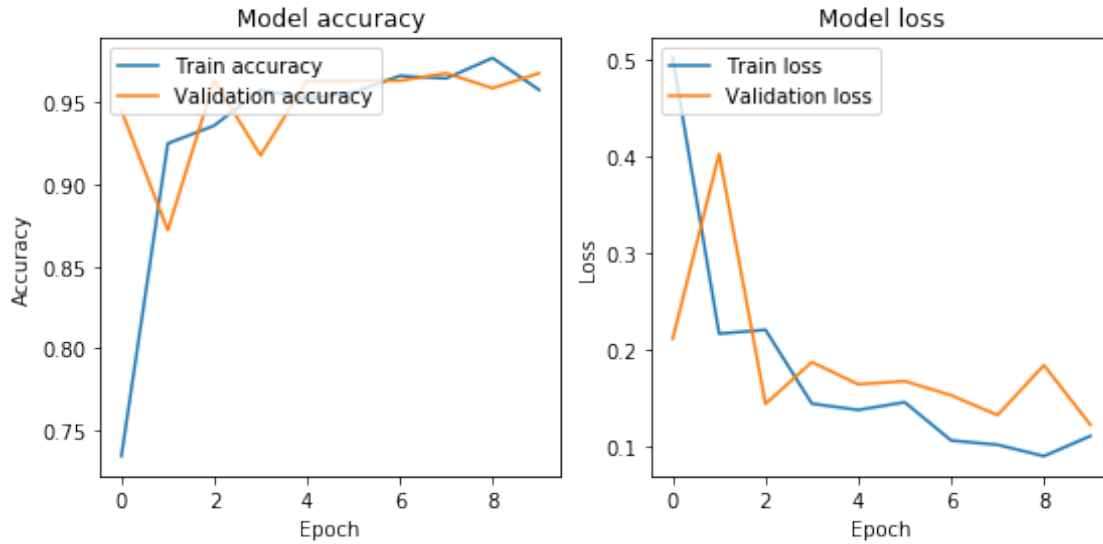
TEST					
	precision	recall	f1-score	support	
False	0.89	0.83	0.86	138	
True	0.63	0.75	0.68	55	
accuracy			0.80	193	
macro avg	0.76	0.79	0.77	193	
weighted avg	0.82	0.80	0.81	193	

(1282, 10, 128)

```

(1282, 1, 10, 128)
False
Train on 1282 samples, validate on 219 samples
Epoch 1/10
1282/1282 [=====] - 100s 78ms/step - loss: 0.5013 -
acc: 0.7340 - val_loss: 0.2113 - val_acc: 0.9452
Epoch 2/10
1282/1282 [=====] - 93s 73ms/step - loss: 0.2166 - acc:
0.9251 - val_loss: 0.4023 - val_acc: 0.8721
Epoch 3/10
1282/1282 [=====] - 93s 73ms/step - loss: 0.2206 - acc:
0.9360 - val_loss: 0.1442 - val_acc: 0.9635
Epoch 4/10
1282/1282 [=====] - 94s 73ms/step - loss: 0.1443 - acc:
0.9579 - val_loss: 0.1873 - val_acc: 0.9178
Epoch 5/10
1282/1282 [=====] - 95s 74ms/step - loss: 0.1379 - acc:
0.9524 - val_loss: 0.1644 - val_acc: 0.9635
Epoch 6/10
1282/1282 [=====] - 93s 73ms/step - loss: 0.1459 - acc:
0.9563 - val_loss: 0.1676 - val_acc: 0.9635
Epoch 7/10
1282/1282 [=====] - 92s 72ms/step - loss: 0.1064 - acc:
0.9665 - val_loss: 0.1530 - val_acc: 0.9635
Epoch 8/10
1282/1282 [=====] - 93s 72ms/step - loss: 0.1019 - acc:
0.9649 - val_loss: 0.1326 - val_acc: 0.9680
Epoch 9/10
1282/1282 [=====] - 92s 72ms/step - loss: 0.0902 - acc:
0.9774 - val_loss: 0.1841 - val_acc: 0.9589
Epoch 10/10
1282/1282 [=====] - 93s 73ms/step - loss: 0.1110 - acc:
0.9579 - val_loss: 0.1227 - val_acc: 0.9680

```



219/219 [=====] - 2s 9ms/step
 Test loss: 0.16597979757339443
 Test accuracy: 94.52%
 [0.0208216]

 piano

TRAIN					
	precision	recall	f1-score	support	
False	1.00	0.87	0.93	402	
True	0.94	1.00	0.97	880	
accuracy			0.96	1282	
macro avg	0.97	0.94	0.95	1282	
weighted avg	0.96	0.96	0.96	1282	

True

[0.9986913]

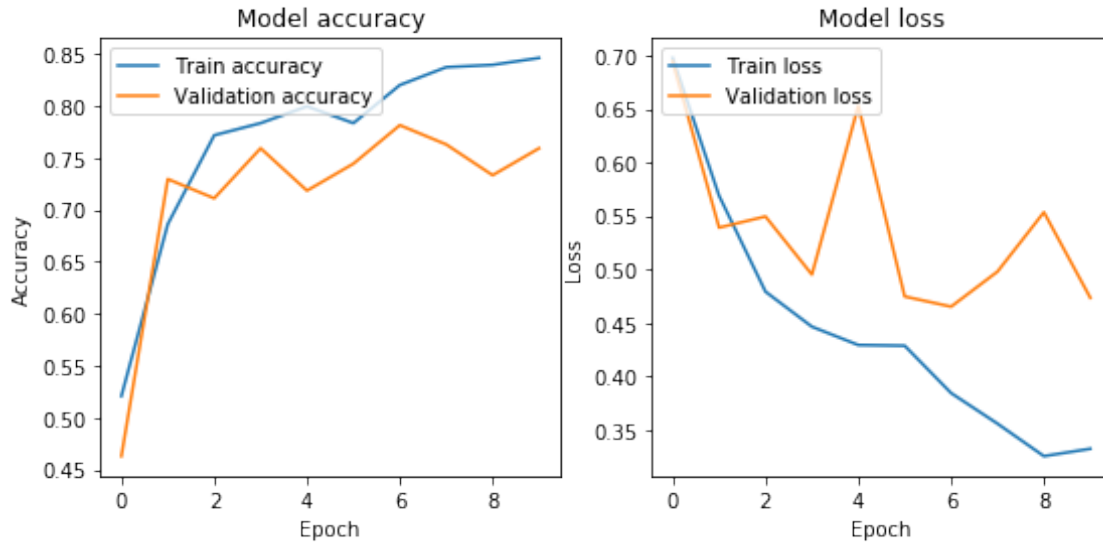
TEST					
	precision	recall	f1-score	support	
False	0.98	0.79	0.88	73	
True	0.91	0.99	0.95	146	
accuracy			0.93	219	
macro avg	0.94	0.89	0.91	219	
weighted avg	0.93	0.93	0.92	219	

(1801, 10, 128)


```

(1801, 1, 10, 128)
False
Train on 1801 samples, validate on 270 samples
Epoch 1/10
1801/1801 [=====] - 137s 76ms/step - loss: 0.6976 -
acc: 0.5208 - val_loss: 0.6941 - val_acc: 0.4630
Epoch 2/10
1801/1801 [=====] - 129s 71ms/step - loss: 0.5688 -
acc: 0.6863 - val_loss: 0.5390 - val_acc: 0.7296
Epoch 3/10
1801/1801 [=====] - 127s 71ms/step - loss: 0.4792 -
acc: 0.7718 - val_loss: 0.5494 - val_acc: 0.7111
Epoch 4/10
1801/1801 [=====] - 127s 70ms/step - loss: 0.4464 -
acc: 0.7835 - val_loss: 0.4951 - val_acc: 0.7593
Epoch 5/10
1801/1801 [=====] - 129s 72ms/step - loss: 0.4293 -
acc: 0.7996 - val_loss: 0.6528 - val_acc: 0.7185
Epoch 6/10
1801/1801 [=====] - 131s 73ms/step - loss: 0.4286 -
acc: 0.7835 - val_loss: 0.4746 - val_acc: 0.7444
Epoch 7/10
1801/1801 [=====] - 126s 70ms/step - loss: 0.3844 -
acc: 0.8201 - val_loss: 0.4651 - val_acc: 0.7815
Epoch 8/10
1801/1801 [=====] - 129s 71ms/step - loss: 0.3558 -
acc: 0.8373 - val_loss: 0.4979 - val_acc: 0.7630
Epoch 9/10
1801/1801 [=====] - 127s 71ms/step - loss: 0.3254 -
acc: 0.8395 - val_loss: 0.5534 - val_acc: 0.7333
Epoch 10/10
1801/1801 [=====] - 128s 71ms/step - loss: 0.3323 -
acc: 0.8462 - val_loss: 0.4734 - val_acc: 0.7593

```



294/294 [=====] - 3s 10ms/step
 Test loss: 0.4525670425421527
 Test accuracy: 76.87%
 [0.00114265]

 saxophone

TRAIN				
	precision	recall	f1-score	support
False	1.00	0.54	0.70	927
True	0.67	1.00	0.80	874
accuracy			0.76	1801
macro avg	0.83	0.77	0.75	1801
weighted avg	0.84	0.76	0.75	1801

True

[0.54964817]

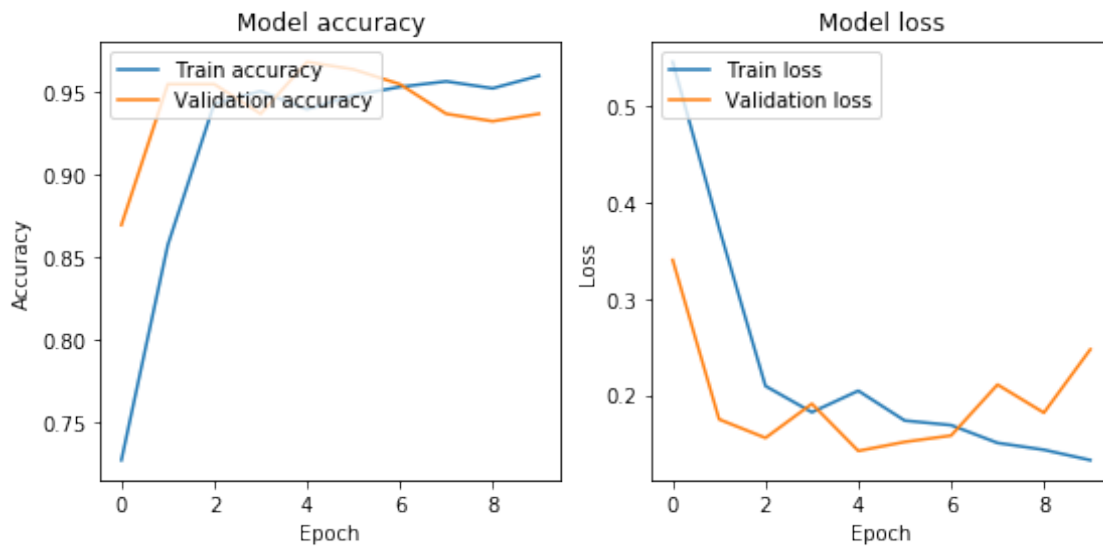
TEST				
	precision	recall	f1-score	support
False	0.97	0.44	0.60	158
True	0.60	0.99	0.75	136
accuracy			0.69	294
macro avg	0.79	0.71	0.67	294
weighted avg	0.80	0.69	0.67	294

(1177, 10, 128)

```

(1177, 1, 10, 128)
True
Train on 1177 samples, validate on 222 samples
Epoch 1/10
1177/1177 [=====] - 90s 77ms/step - loss: 0.5461 - acc:
0.7264 - val_loss: 0.3404 - val_acc: 0.8694
Epoch 2/10
1177/1177 [=====] - 84s 71ms/step - loss: 0.3736 - acc:
0.8573 - val_loss: 0.1749 - val_acc: 0.9550
Epoch 3/10
1177/1177 [=====] - 83s 71ms/step - loss: 0.2094 - acc:
0.9431 - val_loss: 0.1556 - val_acc: 0.9550
Epoch 4/10
1177/1177 [=====] - 83s 71ms/step - loss: 0.1822 - acc:
0.9507 - val_loss: 0.1910 - val_acc: 0.9369
Epoch 5/10
1177/1177 [=====] - 84s 71ms/step - loss: 0.2045 - acc:
0.9397 - val_loss: 0.1420 - val_acc: 0.9685
Epoch 6/10
1177/1177 [=====] - 85s 72ms/step - loss: 0.1736 - acc:
0.9482 - val_loss: 0.1514 - val_acc: 0.9640
Epoch 7/10
1177/1177 [=====] - 85s 72ms/step - loss: 0.1688 - acc:
0.9533 - val_loss: 0.1580 - val_acc: 0.9550
Epoch 8/10
1177/1177 [=====] - 86s 73ms/step - loss: 0.1504 - acc:
0.9567 - val_loss: 0.2109 - val_acc: 0.9369
Epoch 9/10
1177/1177 [=====] - 84s 71ms/step - loss: 0.1432 - acc:
0.9524 - val_loss: 0.1817 - val_acc: 0.9324
Epoch 10/10
1177/1177 [=====] - 86s 73ms/step - loss: 0.1325 - acc:
0.9601 - val_loss: 0.2475 - val_acc: 0.9369

```



203/203 [=====] - 2s 10ms/step
 Test loss: 0.24100707397965962
 Test accuracy: 92.61%
 [0.9997945]

 synthesizer

TRAIN					
	precision	recall	f1-score	support	
False	1.00	0.80	0.89	370	
True	0.91	1.00	0.95	807	
accuracy			0.94	1177	
macro avg	0.96	0.90	0.92	1177	
weighted avg	0.94	0.94	0.93	1177	

True

[0.9982294]

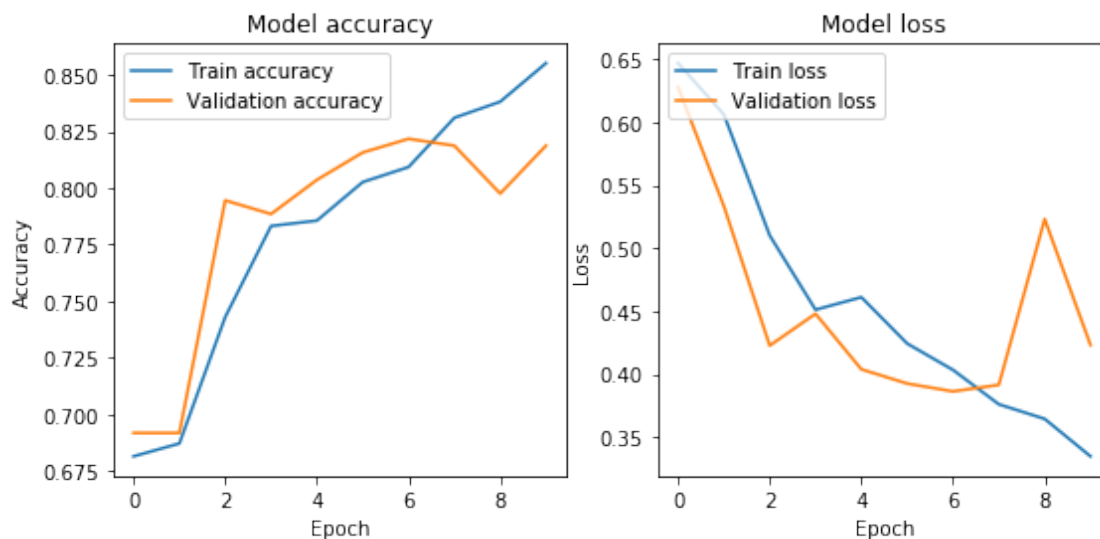
TEST					
	precision	recall	f1-score	support	
False	0.97	0.77	0.86	75	
True	0.88	0.98	0.93	128	
accuracy			0.91	203	
macro avg	0.92	0.88	0.89	203	
weighted avg	0.91	0.91	0.90	203	

(2113, 10, 128)

```

(2113, 1, 10, 128)
False
Train on 2113 samples, validate on 331 samples
Epoch 1/10
2113/2113 [=====] - 161s 76ms/step - loss: 0.6467 -
acc: 0.6815 - val_loss: 0.6277 - val_acc: 0.6918
Epoch 2/10
2113/2113 [=====] - 152s 72ms/step - loss: 0.6059 -
acc: 0.6872 - val_loss: 0.5333 - val_acc: 0.6918
Epoch 3/10
2113/2113 [=====] - 151s 72ms/step - loss: 0.5101 -
acc: 0.7430 - val_loss: 0.4226 - val_acc: 0.7946
Epoch 4/10
2113/2113 [=====] - 149s 71ms/step - loss: 0.4511 -
acc: 0.7832 - val_loss: 0.4479 - val_acc: 0.7885
Epoch 5/10
2113/2113 [=====] - 150s 71ms/step - loss: 0.4610 -
acc: 0.7856 - val_loss: 0.4039 - val_acc: 0.8036
Epoch 6/10
2113/2113 [=====] - 151s 72ms/step - loss: 0.4243 -
acc: 0.8027 - val_loss: 0.3925 - val_acc: 0.8157
Epoch 7/10
2113/2113 [=====] - 152s 72ms/step - loss: 0.4033 -
acc: 0.8093 - val_loss: 0.3863 - val_acc: 0.8218
Epoch 8/10
2113/2113 [=====] - 149s 71ms/step - loss: 0.3761 -
acc: 0.8310 - val_loss: 0.3915 - val_acc: 0.8187
Epoch 9/10
2113/2113 [=====] - 151s 71ms/step - loss: 0.3645 -
acc: 0.8381 - val_loss: 0.5230 - val_acc: 0.7976
Epoch 10/10
2113/2113 [=====] - 152s 72ms/step - loss: 0.3348 -
acc: 0.8552 - val_loss: 0.4231 - val_acc: 0.8187

```



316/316 [=====] - 3s 10ms/step
 Test loss: 0.4415015563180175
 Test accuracy: 82.59%
 [0.00079224]

 trombone

TRAIN					
	precision	recall	f1-score	support	
False	0.99	0.82	0.89	1452	
True	0.71	0.98	0.82	661	
accuracy			0.87	2113	
macro avg	0.85	0.90	0.86	2113	
weighted avg	0.90	0.87	0.87	2113	

False

[0.03959638]

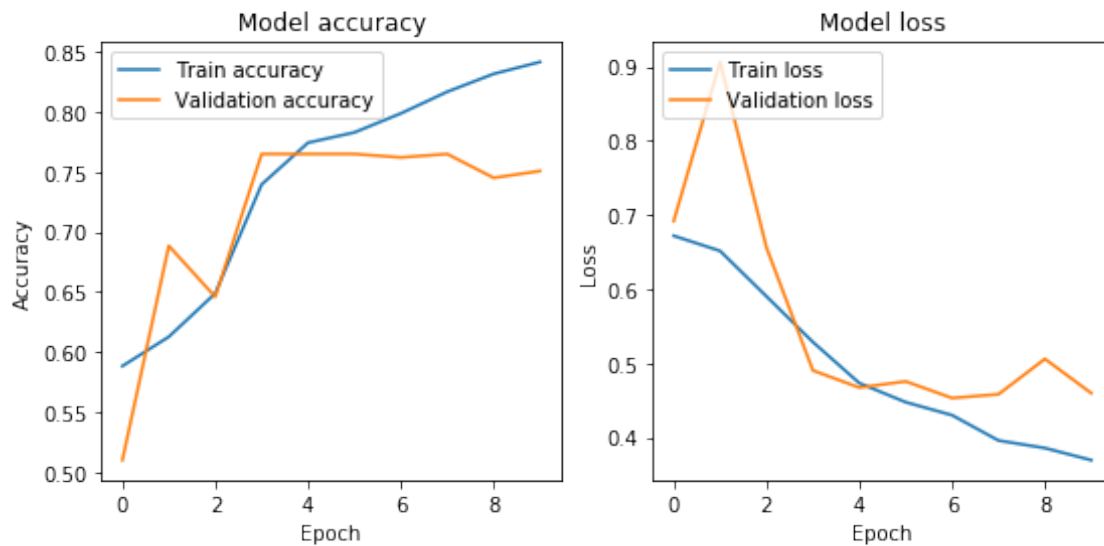
TEST					
	precision	recall	f1-score	support	
False	0.94	0.72	0.82	216	
True	0.60	0.90	0.72	100	
accuracy			0.78	316	
macro avg	0.77	0.81	0.77	316	
weighted avg	0.83	0.78	0.79	316	

(2215, 10, 128)

```

(2215, 1, 10, 128)
False
Train on 2215 samples, validate on 353 samples
Epoch 1/10
2215/2215 [=====] - 166s 75ms/step - loss: 0.6718 -
acc: 0.5883 - val_loss: 0.6917 - val_acc: 0.5099
Epoch 2/10
2215/2215 [=====] - 158s 71ms/step - loss: 0.6514 -
acc: 0.6126 - val_loss: 0.9064 - val_acc: 0.6884
Epoch 3/10
2215/2215 [=====] - 160s 72ms/step - loss: 0.5900 -
acc: 0.6488 - val_loss: 0.6558 - val_acc: 0.6459
Epoch 4/10
2215/2215 [=====] - 158s 71ms/step - loss: 0.5284 -
acc: 0.7395 - val_loss: 0.4900 - val_acc: 0.7649
Epoch 5/10
2215/2215 [=====] - 160s 72ms/step - loss: 0.4729 -
acc: 0.7743 - val_loss: 0.4669 - val_acc: 0.7649
Epoch 6/10
2215/2215 [=====] - 156s 71ms/step - loss: 0.4472 -
acc: 0.7828 - val_loss: 0.4753 - val_acc: 0.7649
Epoch 7/10
2215/2215 [=====] - 158s 72ms/step - loss: 0.4296 -
acc: 0.7986 - val_loss: 0.4528 - val_acc: 0.7620
Epoch 8/10
2215/2215 [=====] - 161s 73ms/step - loss: 0.3955 -
acc: 0.8167 - val_loss: 0.4579 - val_acc: 0.7649
Epoch 9/10
2215/2215 [=====] - 158s 71ms/step - loss: 0.3853 -
acc: 0.8316 - val_loss: 0.5056 - val_acc: 0.7450
Epoch 10/10
2215/2215 [=====] - 158s 71ms/step - loss: 0.3690 -
acc: 0.8415 - val_loss: 0.4593 - val_acc: 0.7507

```



348/348 [=====] - 3s 9ms/step

Test loss: 0.4435253341992696

Test accuracy: 81.03%

[0.10742149]

trumpet

TRAIN					
	precision	recall	f1-score	support	
False	0.98	0.65	0.78	1347	
True	0.64	0.98	0.78	868	
accuracy			0.78	2215	
macro avg	0.81	0.82	0.78	2215	
weighted avg	0.85	0.78	0.78	2215	

False

[0.30723488]

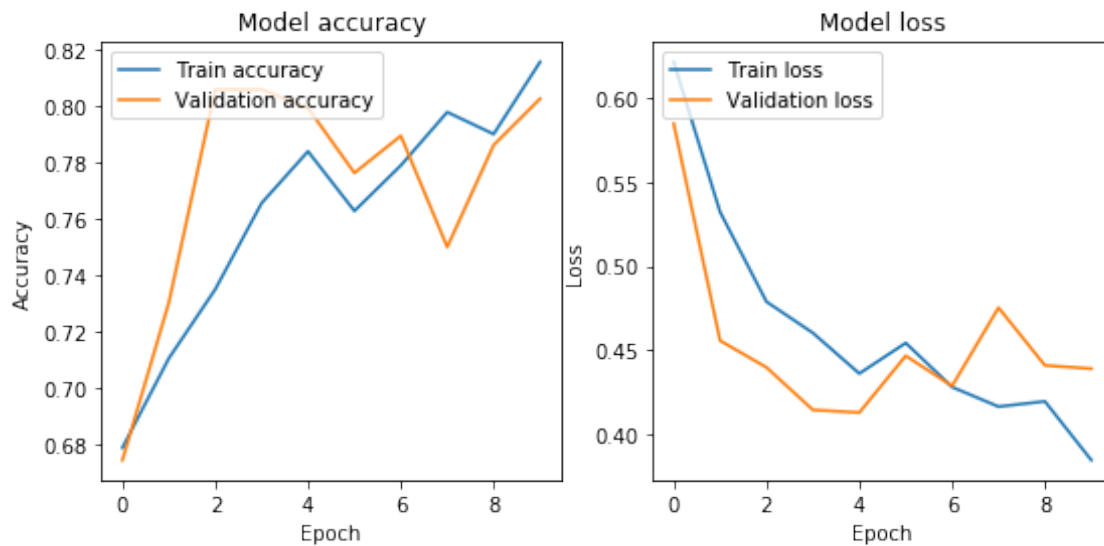
TEST					
	precision	recall	f1-score	support	
False	0.93	0.56	0.69	209	
True	0.58	0.94	0.72	139	
accuracy			0.71	348	
macro avg	0.76	0.75	0.71	348	
weighted avg	0.79	0.71	0.70	348	

(1796, 10, 128)


```

(1796, 1, 10, 128)
False
Train on 1796 samples, validate on 304 samples
Epoch 1/10
1796/1796 [=====] - 140s 78ms/step - loss: 0.6214 -
acc: 0.6787 - val_loss: 0.5847 - val_acc: 0.6743
Epoch 2/10
1796/1796 [=====] - 132s 73ms/step - loss: 0.5322 -
acc: 0.7105 - val_loss: 0.4558 - val_acc: 0.7303
Epoch 3/10
1796/1796 [=====] - 134s 75ms/step - loss: 0.4789 -
acc: 0.7350 - val_loss: 0.4399 - val_acc: 0.8059
Epoch 4/10
1796/1796 [=====] - 131s 73ms/step - loss: 0.4604 -
acc: 0.7656 - val_loss: 0.4146 - val_acc: 0.8059
Epoch 5/10
1796/1796 [=====] - 132s 74ms/step - loss: 0.4362 -
acc: 0.7840 - val_loss: 0.4131 - val_acc: 0.7993
Epoch 6/10
1796/1796 [=====] - 130s 72ms/step - loss: 0.4544 -
acc: 0.7628 - val_loss: 0.4468 - val_acc: 0.7763
Epoch 7/10
1796/1796 [=====] - 130s 72ms/step - loss: 0.4282 -
acc: 0.7790 - val_loss: 0.4289 - val_acc: 0.7895
Epoch 8/10
1796/1796 [=====] - 130s 72ms/step - loss: 0.4166 -
acc: 0.7979 - val_loss: 0.4754 - val_acc: 0.7500
Epoch 9/10
1796/1796 [=====] - 131s 73ms/step - loss: 0.4197 -
acc: 0.7901 - val_loss: 0.4411 - val_acc: 0.7862
Epoch 10/10
1796/1796 [=====] - 129s 72ms/step - loss: 0.3848 -
acc: 0.8157 - val_loss: 0.4392 - val_acc: 0.8026

```



```
325/325 [=====] - 3s 10ms/step
Test loss: 0.4593540314527658
Test accuracy: 76.00%
[0.3134371]
```

ukulele

TRAIN				
	precision	recall	f1-score	support
False	0.99	0.58	0.73	1255
True	0.50	0.99	0.67	541
accuracy			0.70	1796
macro avg	0.75	0.78	0.70	1796
weighted avg	0.85	0.70	0.71	1796

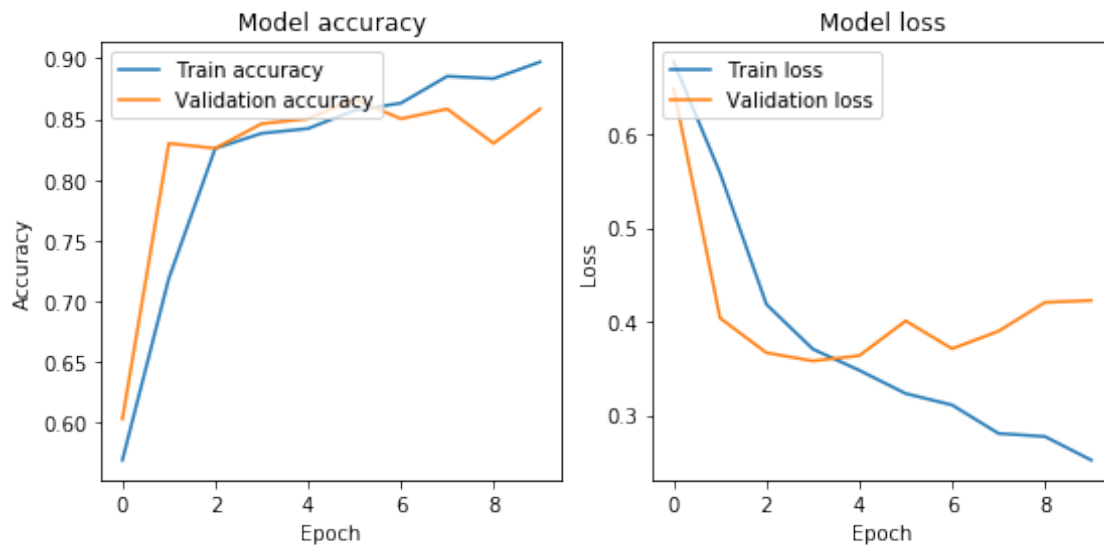
True

[0.5016497]

TEST				
	precision	recall	f1-score	support
False	0.95	0.52	0.67	227
True	0.46	0.94	0.62	98
accuracy			0.65	325
macro avg	0.70	0.73	0.64	325
weighted avg	0.80	0.65	0.66	325

(1533, 10, 128)

```
(1533, 1, 10, 128)
False
Train on 1533 samples, validate on 247 samples
Epoch 1/10
1533/1533 [=====] - 117s 76ms/step - loss: 0.6766 -
acc: 0.5695 - val_loss: 0.6478 - val_acc: 0.6032
Epoch 2/10
1533/1533 [=====] - 111s 73ms/step - loss: 0.5582 -
acc: 0.7189 - val_loss: 0.4036 - val_acc: 0.8300
Epoch 3/10
1533/1533 [=====] - 110s 72ms/step - loss: 0.4181 -
acc: 0.8258 - val_loss: 0.3666 - val_acc: 0.8259
Epoch 4/10
1533/1533 [=====] - 111s 73ms/step - loss: 0.3705 -
acc: 0.8382 - val_loss: 0.3579 - val_acc: 0.8462
Epoch 5/10
1533/1533 [=====] - 109s 71ms/step - loss: 0.3479 -
acc: 0.8421 - val_loss: 0.3636 - val_acc: 0.8502
Epoch 6/10
1533/1533 [=====] - 112s 73ms/step - loss: 0.3231 -
acc: 0.8571 - val_loss: 0.4007 - val_acc: 0.8664
Epoch 7/10
1533/1533 [=====] - 111s 72ms/step - loss: 0.3110 -
acc: 0.8630 - val_loss: 0.3709 - val_acc: 0.8502
Epoch 8/10
1533/1533 [=====] - 110s 71ms/step - loss: 0.2806 -
acc: 0.8852 - val_loss: 0.3896 - val_acc: 0.8583
Epoch 9/10
1533/1533 [=====] - 109s 71ms/step - loss: 0.2774 -
acc: 0.8832 - val_loss: 0.4204 - val_acc: 0.8300
Epoch 10/10
1533/1533 [=====] - 111s 72ms/step - loss: 0.2522 -
acc: 0.8969 - val_loss: 0.4224 - val_acc: 0.8583
```



253/253 [=====] - 2s 9ms/step

Test loss: 0.43692774358003034

Test accuracy: 79.84%

[0.8562643]

violin

TRAIN					
	precision	recall	f1-score	support	
False	1.00	0.73	0.84	653	
True	0.83	1.00	0.91	880	
accuracy			0.88	1533	
macro avg	0.92	0.86	0.88	1533	
weighted avg	0.90	0.88	0.88	1533	

True

[0.7658575]

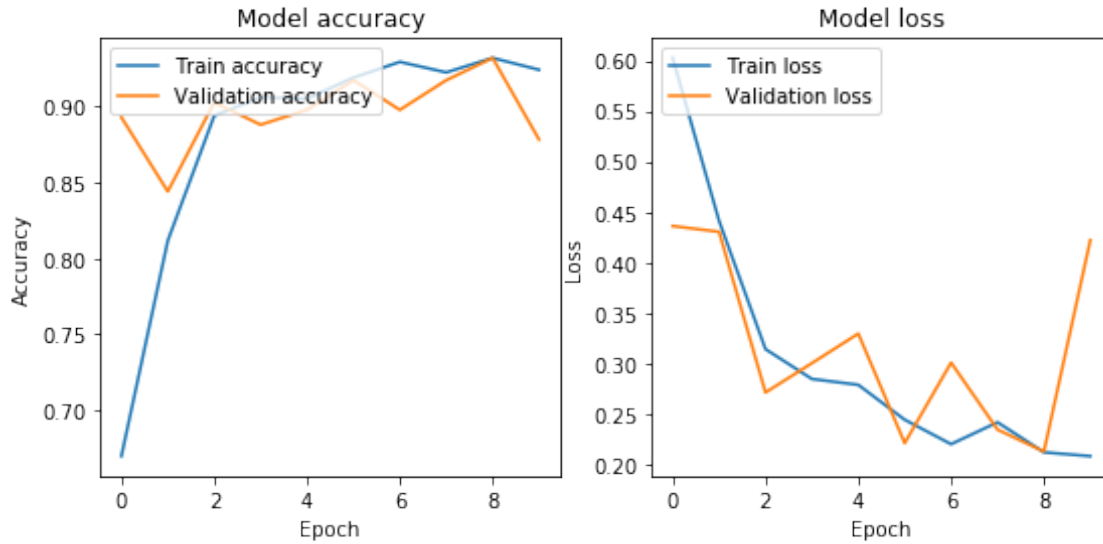
TEST					
	precision	recall	f1-score	support	
False	0.94	0.54	0.69	109	
True	0.74	0.97	0.84	144	
accuracy			0.79	253	
macro avg	0.84	0.76	0.76	253	
weighted avg	0.82	0.79	0.77	253	

(1157, 10, 128)

```

(1157, 1, 10, 128)
True
Train on 1157 samples, validate on 205 samples
Epoch 1/10
1157/1157 [=====] - 92s 79ms/step - loss: 0.6026 - acc:
0.6698 - val_loss: 0.4360 - val_acc: 0.8927
Epoch 2/10
1157/1157 [=====] - 84s 72ms/step - loss: 0.4411 - acc:
0.8116 - val_loss: 0.4303 - val_acc: 0.8439
Epoch 3/10
1157/1157 [=====] - 86s 75ms/step - loss: 0.3140 - acc:
0.8937 - val_loss: 0.2711 - val_acc: 0.9024
Epoch 4/10
1157/1157 [=====] - 84s 73ms/step - loss: 0.2845 - acc:
0.9058 - val_loss: 0.3001 - val_acc: 0.8878
Epoch 5/10
1157/1157 [=====] - 84s 72ms/step - loss: 0.2787 - acc:
0.9049 - val_loss: 0.3294 - val_acc: 0.8976
Epoch 6/10
1157/1157 [=====] - 86s 74ms/step - loss: 0.2441 - acc:
0.9188 - val_loss: 0.2209 - val_acc: 0.9171
Epoch 7/10
1157/1157 [=====] - 85s 73ms/step - loss: 0.2199 - acc:
0.9291 - val_loss: 0.3007 - val_acc: 0.8976
Epoch 8/10
1157/1157 [=====] - 85s 73ms/step - loss: 0.2416 - acc:
0.9222 - val_loss: 0.2343 - val_acc: 0.9171
Epoch 9/10
1157/1157 [=====] - 84s 73ms/step - loss: 0.2118 - acc:
0.9317 - val_loss: 0.2126 - val_acc: 0.9317
Epoch 10/10
1157/1157 [=====] - 85s 73ms/step - loss: 0.2080 - acc:
0.9239 - val_loss: 0.4221 - val_acc: 0.8780

```



202/202 [=====] - 2s 10ms/step
 Test loss: 0.3783179719259243
 Test accuracy: 89.60%
 [0.9990358]

voice

TRAIN				
	precision	recall	f1-score	support
False	0.98	0.62	0.76	434
True	0.81	0.99	0.89	723
accuracy			0.85	1157
macro avg	0.90	0.80	0.82	1157
weighted avg	0.87	0.85	0.84	1157

True

[0.99848545]

TEST				
	precision	recall	f1-score	support
False	1.00	0.54	0.70	59
True	0.84	1.00	0.91	143
accuracy			0.87	202
macro avg	0.92	0.77	0.81	202
weighted avg	0.89	0.87	0.85	202

```
[13]: print(X_train_inst_sklearn)
      print(Y_pred_train)
```

```
[[0.921332  0.40947548 0.6402511 ... 0.4017433 0.4174872 1.      ]
 [0.87017375 0.7986464 0.8297372 ... 0.5835975 0.17802285 1.      ]
 [0.8291506  0.02876481 0.5637505 ... 0.08914421 0.35053173 1.      ]
 ...
 [0.9300193  0.76480544 0.7944292 ... 0.16798733 0.1346987 1.      ]
 [0.9329151  0.48223352 0.8422911 ... 0.11568939 0.6384404 1.      ]
 [0.9092664  0.42808798 0.64848965 ... 0.36648178 0.48956284 1.      ]]
[[0.9990358 ]
 [0.99189687]
 [0.0078997 ]
 ...
 [0.9992817 ]
 [0.9992617 ]
 [0.99744934]]
```

```
[ ]:
```