Spectrogram Project - Train from Scratch

The data used for this project was obtained from the University of Iowa Electronic Music Studios.

The random seed is so the splitting between training, validation, and testing is consistent. If you are training on a GPU, your networks may not be the same if you train the same network multiple times.

```
rng(123)
```

Load and split data

The images are stored in the folder `Spectrograms`. The data for each instrument is separated in subfolders, labelled by the folder name.

Split the data so that 80% of the images are used for training, 10% are used for validation, and the rest are used for testing. If changing the image size, make augmented image datastores.

```
ids=imageDatastore("Projects\MusicProject\Spectrograms","IncludeSubfolders",true,"LabelSource",
[trig,vlig,tsig]=splitEachLabel(ids,0.8,0.1,0.1,"randomized");

trds=augmentedImageDatastore([100 100],trig);
vlds=augmentedImageDatastore([100 100], vlig);
tsds= augmentedImageDatastore([100 100], tsig);
```

Series architecture from the documentation

You can find the example that uses this architecture in the documentation here: Speech Recognition

The images are sized `[200 200 3]`.

```
lbls=14;
drp = 0.2;
layers = [
    imageInputLayer([100 100 3])

    convolution2dLayer(3,16,"Padding","same")
    batchNormalizationLayer
    reluLayer

    maxPooling2dLayer(2,"Stride",2)

    convolution2dLayer(3,32,"Padding","same")
    batchNormalizationLayer
    reluLayer

    maxPooling2dLayer(2,"Stride",2,"Padding",[0,1])

    dropoutLayer(drp)
    convolution2dLayer(3,64,"Padding","same")
    batchNormalizationLayer
    reluLayer
```

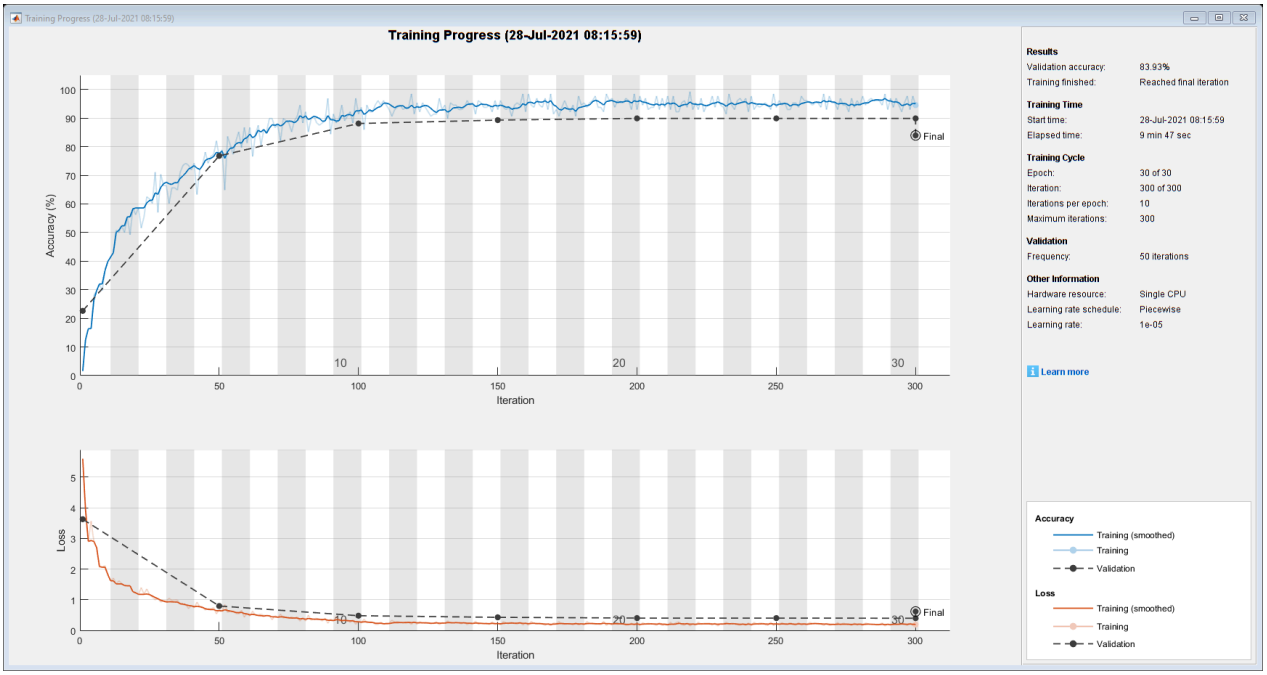
```
    dropoutLayer(drp)
    convolution2dLayer(3,64,"Padding","same")
    batchNormalizationLayer
    reluLayer

    maxPooling2dLayer(2,"Stride",2,"Padding",[0,1])

    dropoutLayer(drp)
    convolution2dLayer(3,64,"Padding","same")
    batchNormalizationLayer
    reluLayer

    dropoutLayer(drp)
    convolution2dLayer(3,64,"Padding","same")
    batchNormalizationLayer
    reluLayer

    maxPooling2dLayer([1 13])

    fullyConnectedLayer(lbls)
    softmaxLayer
    classificationLayer];
```

Train the network

Set the training options and train the network.

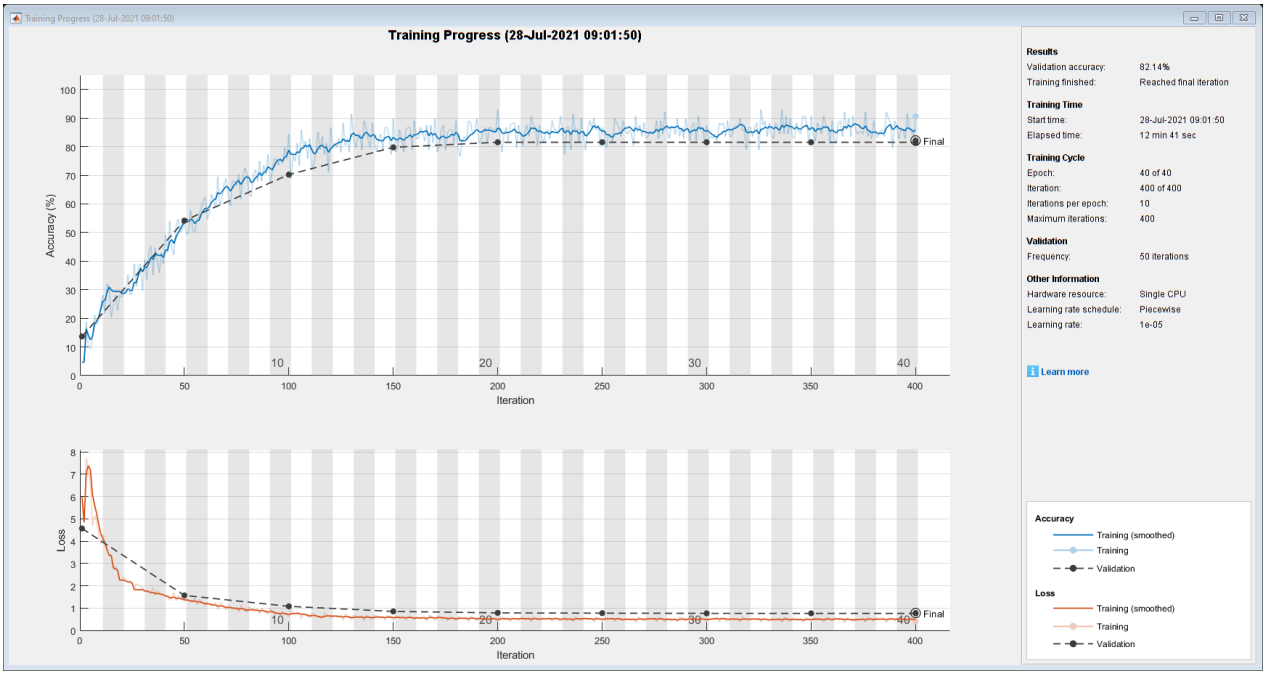
```
opt=trainingOptions("adam","Plots","training-progress","Verbose",true,"LearnRateSchedule","piec
scrnet= trainNetwork(trds,layers,opt);
```

```
Training on single CPU.
Initializing input data normalization.
|=========================================================================================================================
|  Epoch  |  Iteration  |  Time Elapsed  |  Mini-batch  |  Validation  |  Mini-batch  |  Validation  |  Base Learnir
|         |             |   (hh:mm:ss)   |   Accuracy   |   Accuracy   |     Loss     |     Loss     |      Rate
|=========================================================================================================================
|       1 |           1 |       00:00:05 |        1.56% |       22.62% |       5.6014 |       3.6272 |          0.00
|       5 |          50 |       00:01:50 |       76.56% |       76.79% |       0.6757 |       0.7976 |          0.00
|      10 |         100 |       00:03:33 |       91.41% |       88.10% |       0.2912 |       0.4813 |          0.00
|      15 |         150 |       00:05:13 |       92.97% |       89.29% |       0.2334 |       0.4283 |          0.00
|      20 |         200 |       00:06:46 |       92.97% |       89.88% |       0.2513 |       0.4040 |          0.00
|      25 |         250 |       00:08:17 |       93.75% |       89.88% |       0.2181 |       0.4033 |      1.0000e-
|      30 |         300 |       00:09:46 |       94.53% |       89.88% |       0.1914 |       0.4004 |      1.0000e-
|=========================================================================================================================
```



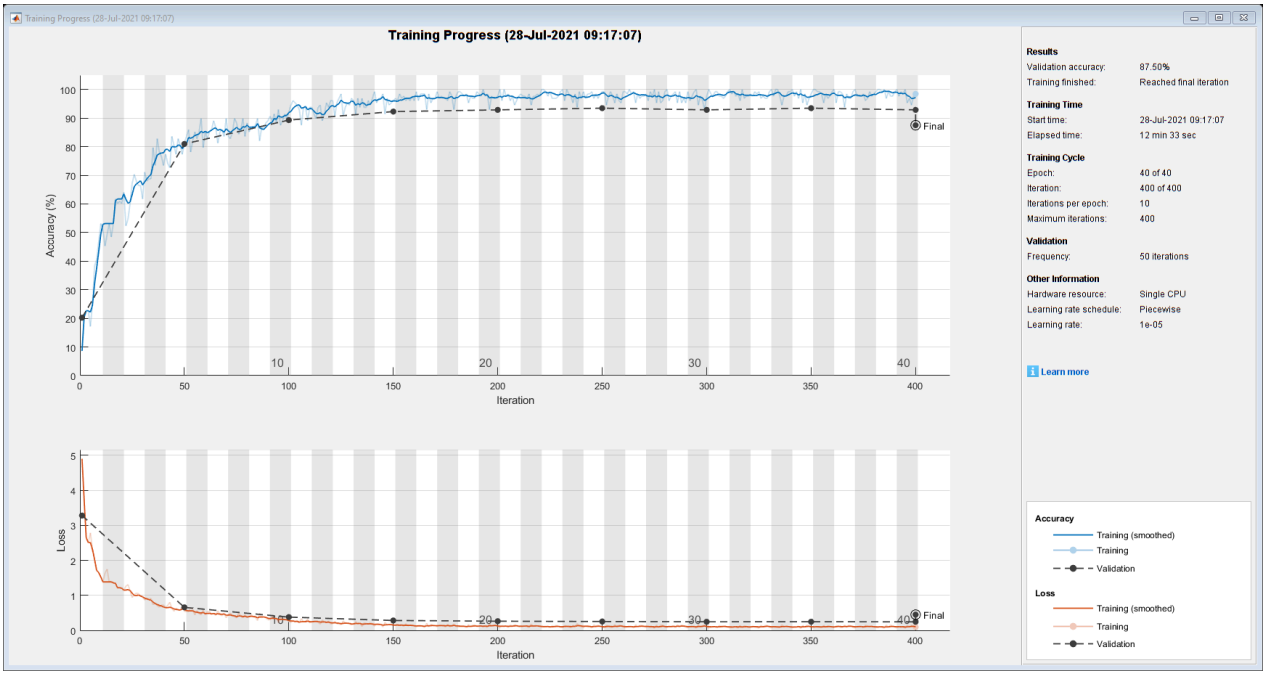

```
opt2=trainingOptions("adam","Plots","training-progress","Verbose",true,"LearnRateSchedule","pie
scrnet2= trainNetwork(trds,layers,opt2);
```

```
Training on single CPU.
Initializing input data normalization.
|=========================================================================================================================
|  Epoch  |  Iteration  |  Time Elapsed  |  Mini-batch  |  Validation  |  Mini-batch  |  Validation  |  Base Learnir
|         |             |   (hh:mm:ss)   |   Accuracy   |   Accuracy   |     Loss     |     Loss     |      Rate
|=========================================================================================================================
|       1 |           1 |       00:00:04 |        4.69% |       13.69% |       5.9819 |       4.5680 |          0.01
|       5 |          50 |       00:01:36 |       54.69% |       54.17% |       1.2881 |       1.5699 |          0.01
|      10 |         100 |       00:03:09 |       84.38% |       70.24% |       0.6160 |       1.0809 |          0.01
|      15 |         150 |       00:04:42 |       89.84% |       79.76% |       0.4717 |       0.8539 |          0.00
|      20 |         200 |       00:06:14 |       92.97% |       81.55% |       0.4105 |       0.7867 |          0.00
|      25 |         250 |       00:07:54 |       86.72% |       81.55% |       0.4382 |       0.7763 |       1.0000e-
|      30 |         300 |       00:09:36 |       87.50% |       81.55% |       0.4319 |       0.7670 |       1.0000e-
|      35 |         350 |       00:11:07 |       90.63% |       81.55% |       0.4041 |       0.7662 |       1.0000e-
|      40 |         400 |       00:12:40 |       90.63% |       81.55% |       0.4123 |       0.7659 |       1.0000e-
|=========================================================================================================================
```



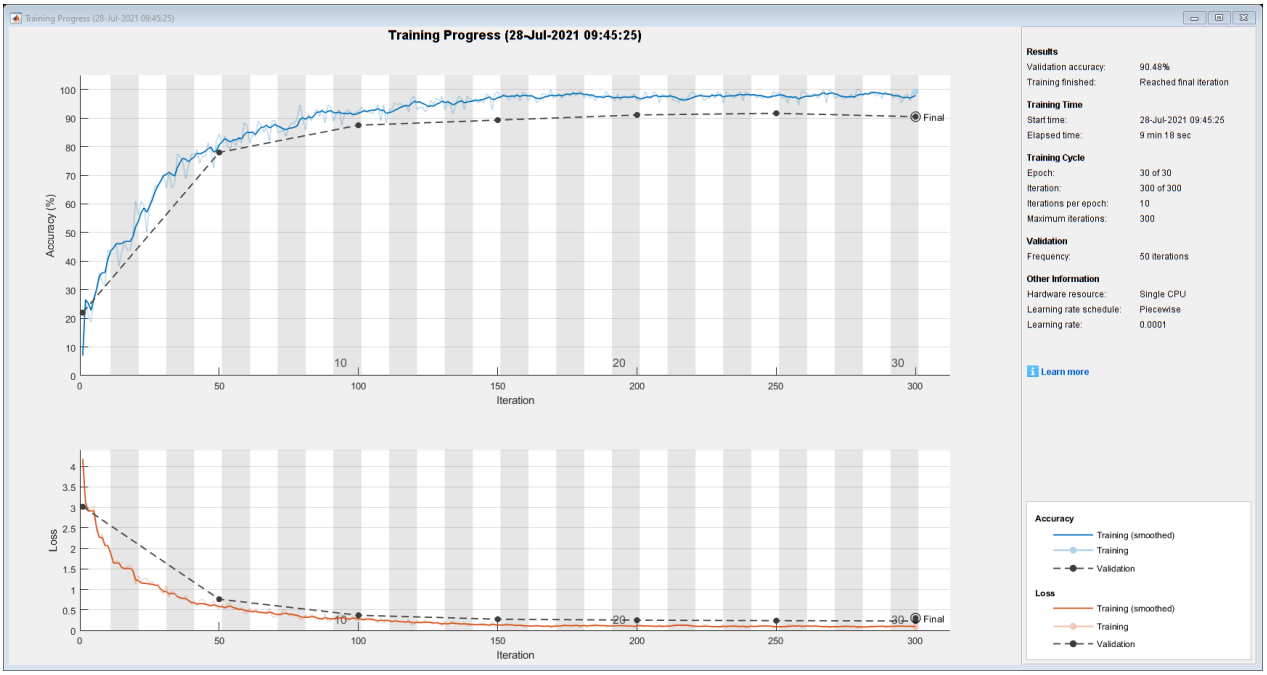

```
opt3=trainingOptions("adam","Plots","training-progress","Verbose",true,"LearnRateSchedule","pie
scrnet3= trainNetwork(trds,layers,opt3);
```

```
Training on single CPU.
Initializing input data normalization.
|=========================================================================================================================
|  Epoch  |  Iteration  |  Time Elapsed  |  Mini-batch  |  Validation  |  Mini-batch  |  Validation  |  Base Learnir
|         |             |   (hh:mm:ss)   |   Accuracy   |   Accuracy   |     Loss     |     Loss     |      Rate
|=========================================================================================================================
|       1 |           1 |       00:00:03 |        8.59% |       20.24% |       4.8994 |       3.2791 |          0.00
|       5 |          50 |       00:01:35 |       81.25% |       80.95% |       0.5430 |       0.6594 |          0.00
|      10 |         100 |       00:03:09 |       92.19% |       89.29% |       0.2668 |       0.3809 |          0.00
|      15 |         150 |       00:04:43 |       96.09% |       92.26% |       0.1566 |       0.2846 |          0.00
|      20 |         200 |       00:06:17 |       97.66% |       92.86% |       0.1069 |       0.2649 |          0.00
|      25 |         250 |       00:07:50 |       99.22% |       93.45% |       0.1063 |       0.2558 |          0.00
|      30 |         300 |       00:09:22 |       97.66% |       92.86% |       0.0923 |       0.2478 |          0.00
|      35 |         350 |       00:10:56 |       98.44% |       93.45% |       0.0854 |       0.2486 |       1.0000e-
|      40 |         400 |       00:12:33 |       98.44% |       92.86% |       0.0997 |       0.2471 |       1.0000e-
|=========================================================================================================================
```



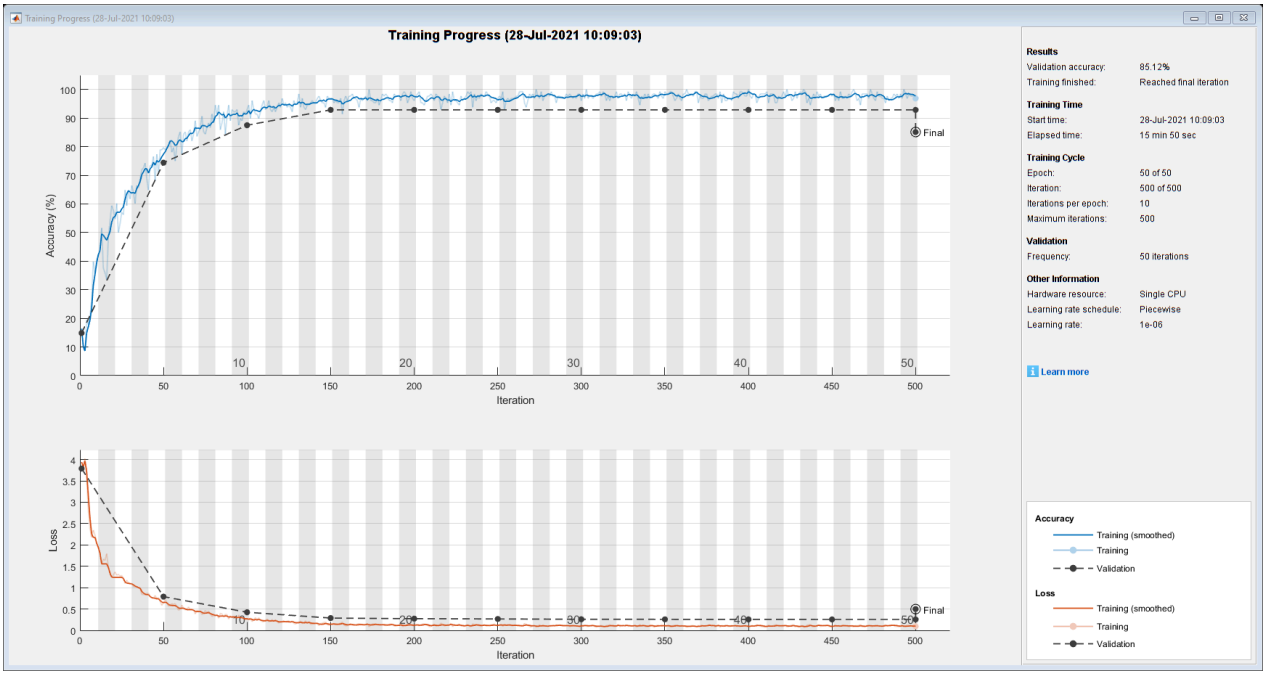

```
opt4=trainingOptions("adam","Plots","training-progress","Verbose",true,"LearnRateSchedule","pie
scrnet4= trainNetwork(trds,layers,opt4);
```

```
Training on single CPU.
Initializing input data normalization.
|=========================================================================================================================================|
|  Epoch  |  Iteration  |  Time Elapsed  |  Mini-batch  |  Validation  |  Mini-batch  |  Validation  |  Base Learnin |
|         |             |   (hh:mm:ss)   |   Accuracy   |   Accuracy   |     Loss     |     Loss     |      Rate     |
|=========================================================================================================================================|
|       1 |           1 |       00:00:03 |        7.03% |       22.02% |       4.1875 |       3.0182 |          0.00 |
|       5 |          50 |       00:01:34 |       84.38% |       77.98% |       0.5283 |       0.7625 |          0.00 |
|      10 |         100 |       00:03:07 |       92.19% |       87.50% |       0.2535 |       0.3705 |          0.00 |
|      15 |         150 |       00:04:40 |       99.22% |       89.29% |       0.0968 |       0.2751 |          0.00 |
|      20 |         200 |       00:06:12 |       96.09% |       91.07% |       0.1210 |       0.2506 |          0.00 |
|      25 |         250 |       00:07:46 |       98.44% |       91.67% |       0.0779 |       0.2387 |          0.00 |
|      30 |         300 |       00:09:18 |       99.22% |       90.48% |       0.0861 |       0.2309 |          0.00 |
|=========================================================================================================================================|
```




```
opt5=trainingOptions("adam","Plots","training-progress","Verbose",true,"LearnRateSchedule","pie
scrnet5= trainNetwork(trds,layers,opt5);
```

```
Training on single CPU.
Initializing input data normalization.
|=========================================================================================================================
|  Epoch  |  Iteration  |  Time Elapsed  |  Mini-batch  |  Validation  |  Mini-batch  |  Validation  |  Base Learnir
|         |             |   (hh:mm:ss)   |   Accuracy   |   Accuracy   |     Loss     |     Loss     |      Rate
|=========================================================================================================================
|       1 |           1 |       00:00:03 |       16.41% |       14.88% |       3.9389 |       3.7883 |          0.00
|       5 |          50 |       00:01:35 |       79.69% |       74.40% |       0.6547 |       0.7913 |          0.00
|      10 |         100 |       00:03:08 |       92.97% |       87.50% |       0.2711 |       0.4260 |          0.00
|      15 |         150 |       00:04:43 |       96.88% |       92.86% |       0.1349 |       0.2900 |          0.00
|      20 |         200 |       00:06:16 |       96.09% |       92.86% |       0.1065 |       0.2761 |          0.00
|      25 |         250 |       00:07:49 |       96.88% |       92.86% |       0.1080 |       0.2693 |          0.00
|      30 |         300 |       00:09:24 |       96.09% |       92.86% |       0.1266 |       0.2615 |          0.00
|      35 |         350 |       00:11:01 |       99.22% |       92.86% |       0.0967 |       0.2605 |       1.0000e-
|      40 |         400 |       00:12:37 |      100.00% |       92.86% |       0.0770 |       0.2587 |       1.0000e-
|      45 |         450 |       00:14:13 |       97.66% |       92.86% |       0.0847 |       0.2584 |       1.0000e-
|      50 |         500 |       00:15:49 |       96.88% |       92.86% |       0.0983 |       0.2584 |       1.0000e-
|=========================================================================================================================
```



Evaluate network

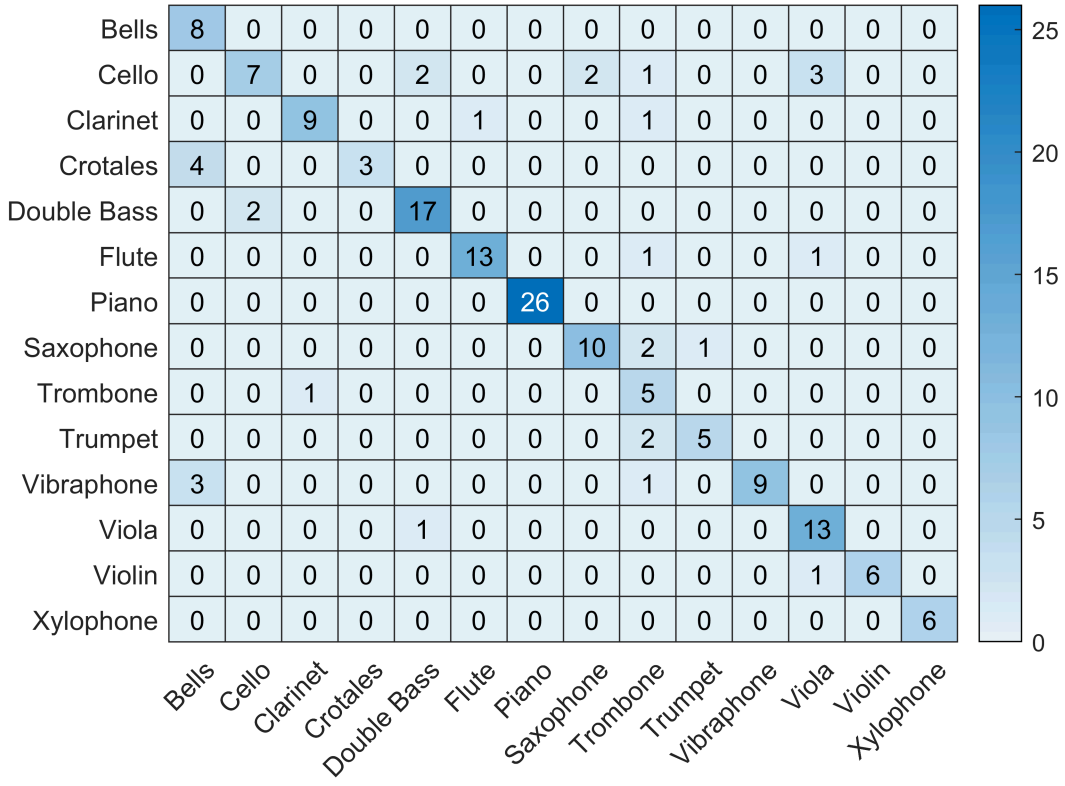
Calculate the accuracy on the test dataset.

```
pred1 = classify(scrnet,tsds);
nnz(pred1 == tsig.Labels)/numel(tsig.Labels)
```

```
ans = 0.8204
```

Plot confusion matrix

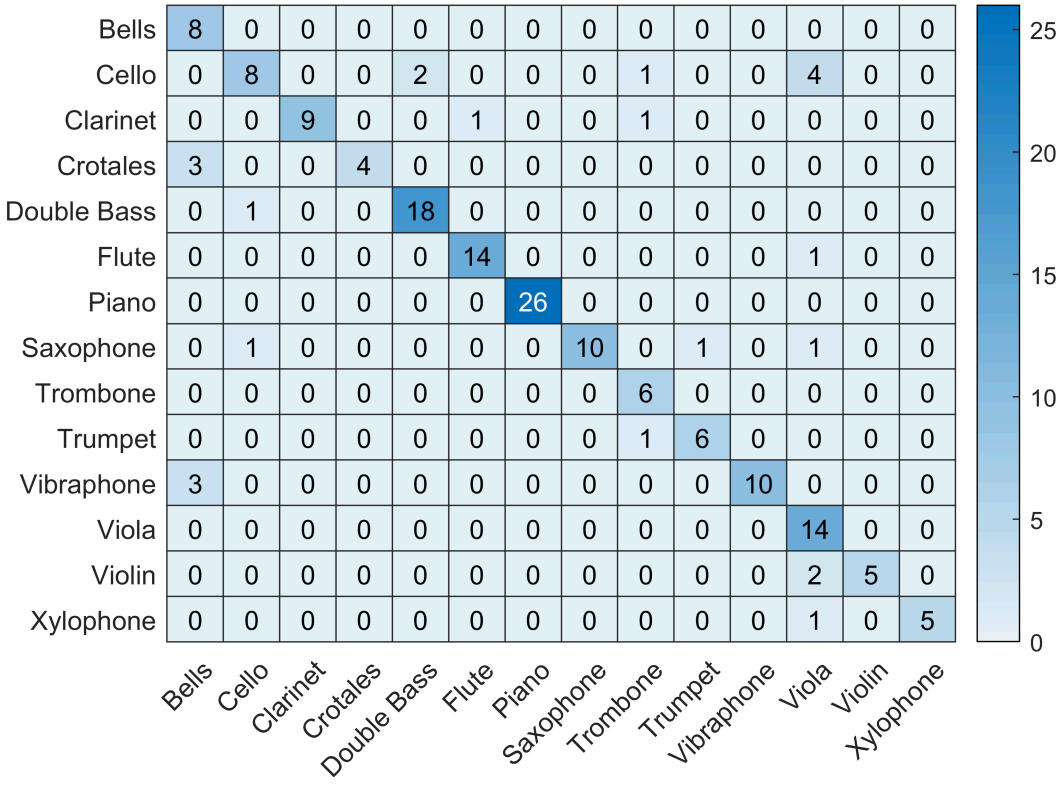
```
[cmap,clabel] = confusionmat(tsig.Labels,pred1);
heatmap(clabel,clabel,cmap)
```




```
pred5 = classify(scrnet5,tsds);
nnz(pred5 == tsig.Labels)/numel(tsig.Labels)
```

```
ans = 0.8563
```

```
[cm,cl] = confusionmat(tsig.Labels,pred5);
heatmap(cl,cl,cm)
```


Bells 8 0 0 0 0 0 0 0 0 0 0 0 0 0
Cello 0 8 0 0 2 0 0 0 1 0 0 4 0 0
Clarinet 0 0 9 0 0 1 0 0 1 0 0 0 0 0
Crotales 3 0 0 4 0 0 0 0 0 0 0 0 0 0
Double Bass 0 1 0 0 18 0 0 0 0 0 0 0 0 0
Flute 0 0 0 0 0 14 0 0 0 0 0 1 0 0
Piano 0 0 0 0 0 0 26 0 0 0 0 0 0 0
Saxophone 0 1 0 0 0 0 0 10 0 1 0 1 0 0
Trombone 0 0 0 0 0 0 0 0 6 0 0 0 0 0
Trumpet 0 0 0 0 0 0 0 0 1 6 0 0 0 0
Vibraphone 3 0 0 0 0 0 0 0 0 0 10 0 0 0
Viola 0 0 0 0 0 0 0 0 0 0 0 14 0 0
Violin 0 0 0 0 0 0 0 0 0 0 0 2 5 0
Xylophone 0 0 0 0 0 0 0 0 0 0 0 1 0 5
Bells
Cello
Clarinet
Crotales
Double Bass
Flute
Piano
Saxophone
Trombone
Trumpet
Vibraphone
Viola
Violin
Xylophone
25
20
15
10
5
0