

```
clc; clear
cd 'G:\Jiaxu Flashdrive Backup\code';
addpath 'G:\Jiaxu Flashdrive Backup\code\functions'
```

```
net = load("efficientnet_6class_untrained.mat");
% mobilenetV2_6va = net.mobilenetV2_6variable;
efficientnet_6var = net.efficientnet_6class;
fracTrainFiles = 0.8;
fracValFiles = 0.2;
training_imds = imageDatastore("G:\Machine Learning\NW_pitch_30_Nws_6_intervals\train", ...
    "IncludeSubfolders",true,...
    "LabelSource","foldernames");
shuffle_training_imds = shuffle(training_imds);
[trainImgs,validImgs] = splitEachLabel(shuffle_training_imds,fracTrainFiles,fracValFiles,"random");
numClasses = numel(categories(training_imds.Labels));
testing_imds = imageDatastore("G:\Machine Learning\NW_pitch_30_Nws_6_intervals\test" ...
    ,"IncludeSubfolders",true,...
    "LabelSource","foldernames");
```

```
options = trainingOptions('adam', ...
    'InitialLearnRate',0.001, ...
    'MaxEpochs',50, ...
    'Shuffle','every-epoch', ...
    'ValidationData',validImgs,...
    'MiniBatchSize',64,...
    'LearnRateSchedule','piecewise',...
    'LearnRateDropFactor',0.9,...
    'LearnRateDropPeriod',10,...
    'ValidationPatience',6,...
    'ExecutionEnvironment','multi-gpu',...
    'Plots','training-progress');

aug = imageDataAugmenter("RandScale",[0.8 1.3], ...
    "RandYTranslation",[-40 40], ...
    "RandXReflection",true, ...
    "RandYReflection",true);
auimds = augmentedImageDatastore([150 150 3],trainImgs,'DataAugmentation',aug);

% inputlayer = imageInputLayer([150,150,3],'Name','input','Normalization','none');
```

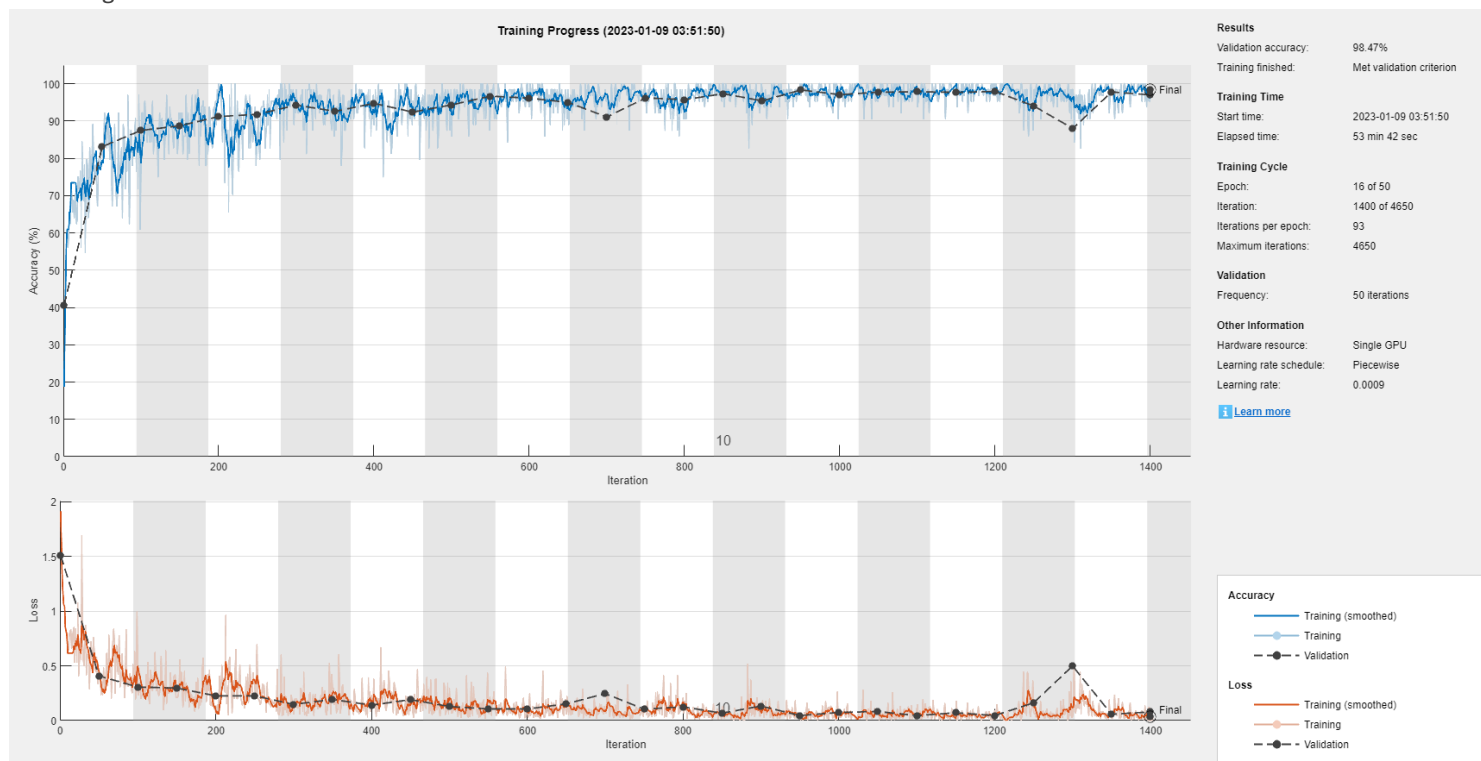
```
[efficientnet_6inv_pitchangle,info] = trainNetwork(auimds,efficientnet_6var,options);
```

Initializing input data normalization.

Epoch	Iteration	Time Elapsed (hh:mm:ss)	Mini-batch Accuracy	Validation Accuracy	Mini-batch Loss	Validation Loss	Base Learning Rate
1	1	00:00:09	18.75%	40.67%	1.9150	1.5072	0.001
1	50	00:02:02	81.25%	83.13%	0.4727	0.4088	0.001
2	100	00:03:56	79.69%	87.47%	0.3894	0.3068	0.001
2	150	00:05:51	90.62%	88.73%	0.2132	0.2973	0.001
3	200	00:07:45	100.00%	91.20%	0.0757	0.2263	0.001
3	250	00:09:39	85.94%	91.80%	0.3462	0.2261	0.001

4	300	00:11:33	98.44%	94.27%	0.0618	0.1479	0.00
4	350	00:13:27	92.19%	92.60%	0.2880	0.1933	0.00
5	400	00:15:22	89.06%	94.73%	0.2278	0.1411	0.00
5	450	00:17:17	95.31%	92.40%	0.0891	0.1914	0.00
6	500	00:19:12	85.94%	94.33%	0.2939	0.1292	0.00
6	550	00:21:07	93.75%	96.67%	0.1119	0.1068	0.00
7	600	00:23:01	93.75%	96.13%	0.0934	0.1060	0.00
7	650	00:24:56	98.44%	95.00%	0.0597	0.1527	0.00
8	700	00:26:50	96.88%	91.07%	0.0723	0.2467	0.00
9	750	00:28:46	96.88%	96.20%	0.1108	0.1044	0.00
9	800	00:30:40	90.62%	95.73%	0.2083	0.1258	0.00
10	850	00:32:36	100.00%	97.33%	0.0235	0.0678	0.00
10	900	00:34:31	98.44%	95.40%	0.0731	0.1331	0.00
11	950	00:36:26	96.88%	98.33%	0.0428	0.0457	0.00
11	1000	00:38:21	98.44%	97.07%	0.0459	0.0746	0.00
12	1050	00:40:15	96.88%	97.73%	0.0691	0.0832	0.00
12	1100	00:42:10	93.75%	97.87%	0.1048	0.0453	0.00
13	1150	00:44:05	100.00%	97.67%	0.0117	0.0714	0.00
13	1200	00:45:59	96.88%	97.93%	0.0905	0.0462	0.00
14	1250	00:47:54	98.44%	94.00%	0.0859	0.1623	0.00
14	1300	00:49:48	98.44%	88.00%	0.0344	0.4993	0.00
15	1350	00:51:43	100.00%	97.80%	0.0207	0.0576	0.00
16	1400	00:53:38	98.44%	97.00%	0.0279	0.0793	0.00

Training finished: Met validation criterion.



```
testpreds = classify(efficientnet_6inv_pitchangle,testing_imds);
nnz(testpreds == testing_imds.Labels)/numel(testpreds)
```

ans = 0.9759

```
confusionchart(testing_imds.Labels,testpreds);
```

0-20	166					
21-35		33	9			
36-50			41			
51-65				42		
66-80					41	
80-90						41
	0-20	21-35	36-50	51-65	66-80	80-90

Predicted Class