

```
addpath('./data');
```

## Load the Dataset

```
load('trunk12-patches.mat');

% Extract training and testing data and labels
trainData = trunk.train.data;
trainLabels = trunk.train.labels;
testData = trunk.test.data;
testLabels = trunk.test.labels;

% Normalize the images to the range [0, 1]
trainData = double(trainData) / 255;
testData = double(testData) / 255;
```

## Basic CNN Architecture

```
layers = [
    imageInputLayer([40 40 3])

    convolution2dLayer(3, 8, 'Padding', 'same')

    maxPooling2dLayer(2, 'Stride', 2)

    convolution2dLayer(3, 16, 'Padding', 'same')

    maxPooling2dLayer(2, 'Stride', 2)

    fullyConnectedLayer(64)

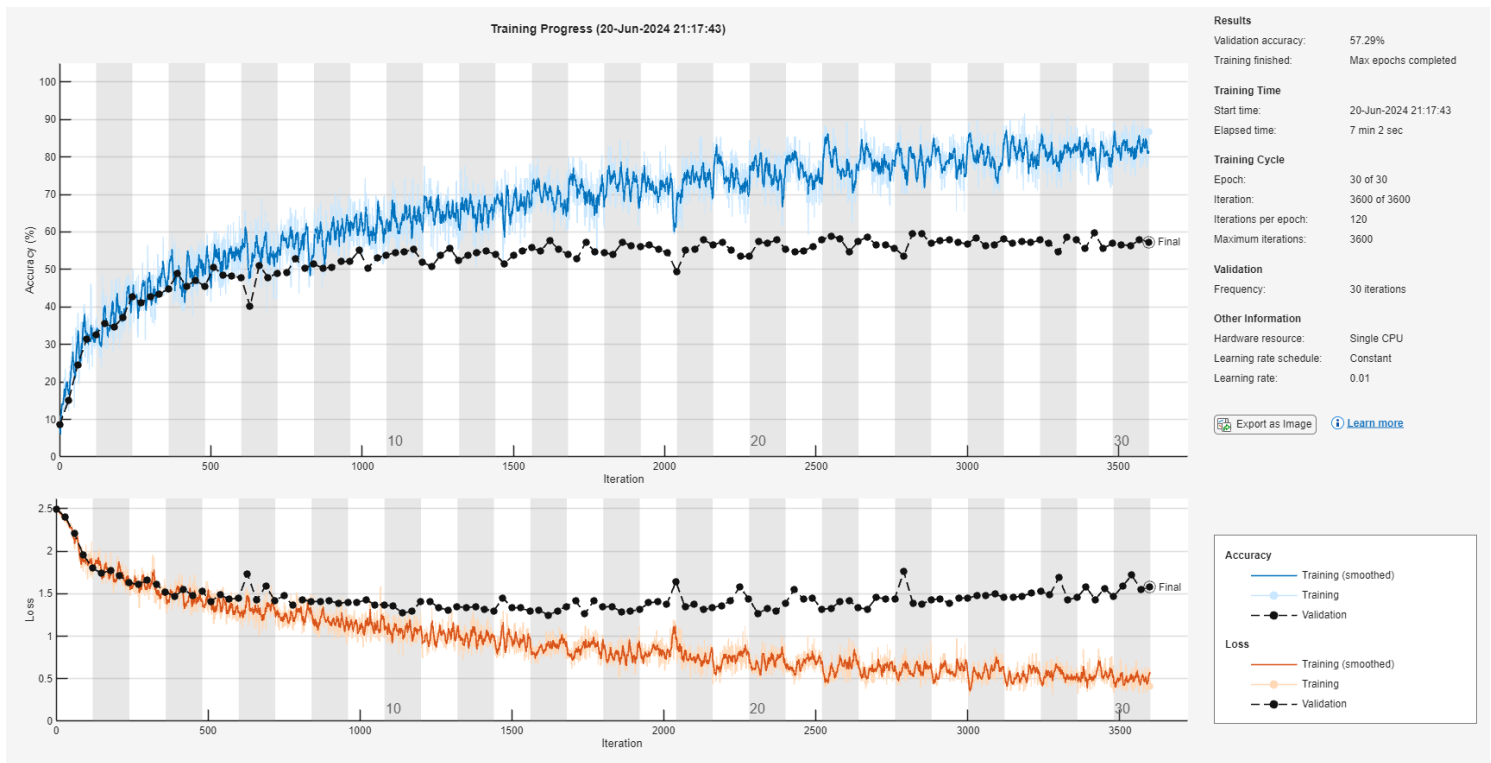
    fullyConnectedLayer(12)
    softmaxLayer
    classificationLayer
];
```

## Specify Training Options

```
options = trainingOptions('sgdm', ...
    'InitialLearnRate',0.01, ...
    'MaxEpochs',30, ...
    'Shuffle','every-epoch', ...
    'ValidationData',{testData, testLabels}, ...
    'ValidationFrequency',30, ...
    'Verbose',false, ...
    'Plots','training-progress');
```

## Model Training

```
net = trainNetwork(trainData, trainLabels, layers, options);
```



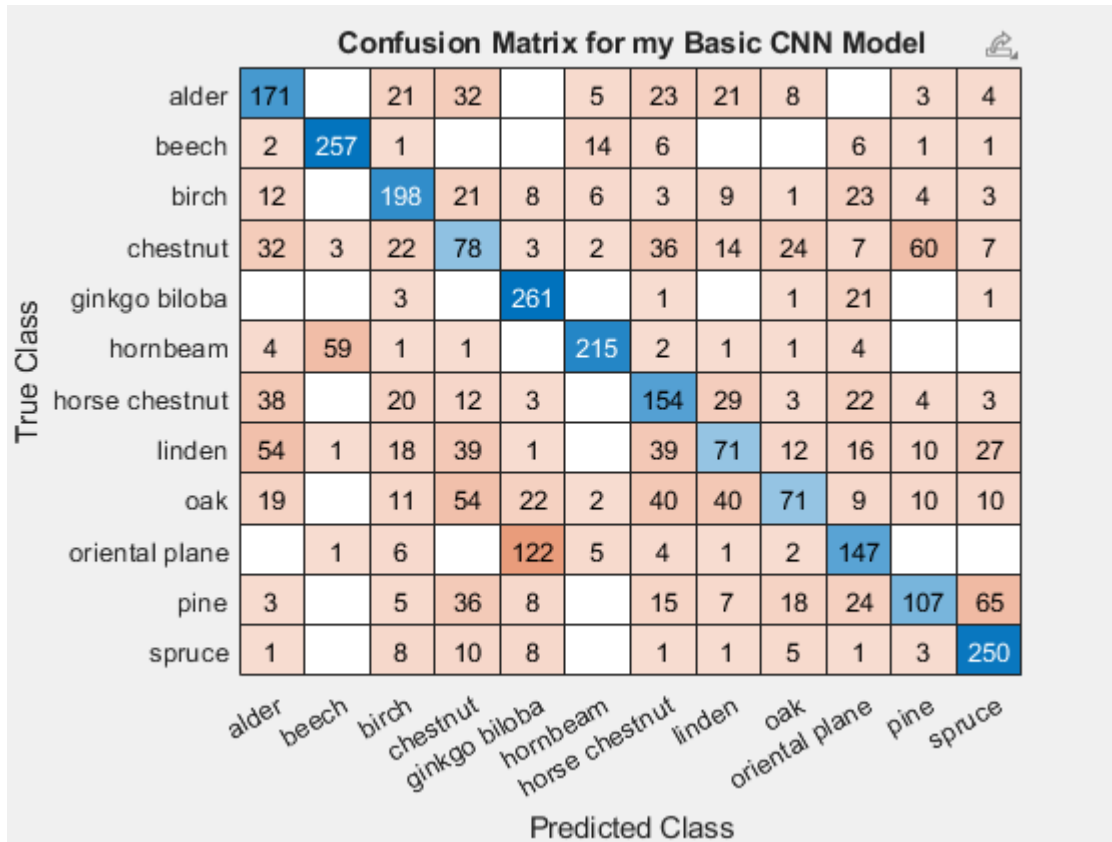
## Evaluate the Model

```
predictedLabels = classify(net, testData);  
accuracy = sum(predictedLabels == testLabels) / numel(testLabels);  
fprintf('Test accuracy: %.2f%%\n', accuracy * 100);
```

Test accuracy: 57.29%

## Generate confusion matrix

```
figure('Visible', 'on');  
confusionchart(testLabels, predictedLabels);  
title('Confusion Matrix for my Basic CNN Model');
```



drawnow;

```
% Select a subset of test images for visualization
numSamples = 10; % Number of samples to visualize
sampleIndices = randperm(numel(testLabels), numSamples);

figure('Visible', 'on');
for i = 1:numSamples
    index = sampleIndices(i);
    img = testData(:,:,index);
    trueLabel = testLabels(index);
    predictedLabel = predictedLabels(index);

    subplot(2, numSamples/2, i);
    imshow(img, 'InitialMagnification', 'fit');

    if trueLabel == predictedLabel
        borderColor = 'g'; % Green for correct classification
        titleText = sprintf('Correct: %s', string(trueLabel));
    else
        borderColor = 'r'; % Red for incorrect classification
        titleText = sprintf('Wrong: %s (Pred: %s)', string(trueLabel),
string(predictedLabel));
    end
end
```

```

rectangle('Position', [0.5, 0.5, size(img, 2), size(img, 1)], 'EdgeColor',
borderColor, 'LineWidth', 3);
title(titleText);
end

```



```
drawnow;
```

Improving by adding pre-processing (Normalisation)

```

% Normalize the images to mean 0 and variance 1
meanTrain = mean(trainData(:));
stdTrain = std(trainData(:));
trainData = (trainData - meanTrain) / stdTrain;
testData = (testData - meanTrain) / stdTrain; % Use the same mean and std as train
data

```

```

% Visualization of the original and normalized images
numSamples = 5; % Number of samples to visualize
sampleIndices = randperm(size(trunk.train.data, 4), numSamples);
sampleData = trunk.train.data(:,:,:,sampleIndices);
normalizedSampleData = double(sampleData) / 255;
normalizedSampleData = (normalizedSampleData - meanTrain) / stdTrain;

```

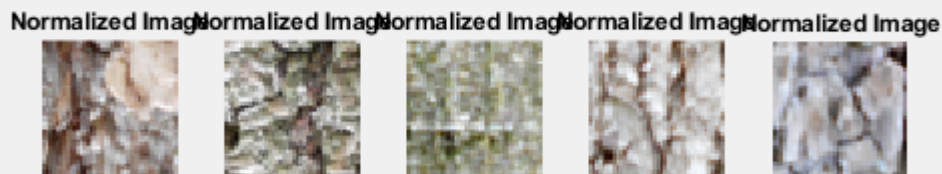
```
figure('Visible', 'on');
```

```

for i = 1:numSamples
    % Original image
    subplot(2, numSamples, i);
    imshow(uint8(sampleData(:,:,i)));
    title('Original Image');

    % Normalized image
    subplot(2, numSamples, numSamples + i);
    imshow(mat2gray(normalizedSampleData(:,:,i))); % mat2gray scales the image to
[0, 1] for display
    title('Normalized Image');
end

```



```

layers = [
    imageInputLayer([40 40 3])

    convolution2dLayer(3, 8, 'Padding', 'same')

    maxPooling2dLayer(2, 'Stride', 2)

    convolution2dLayer(3, 16, 'Padding', 'same')

```

```

maxPooling2dLayer(2, 'Stride', 2)

fullyConnectedLayer(64)

fullyConnectedLayer(12)
softmaxLayer
classificationLayer
];

```

```

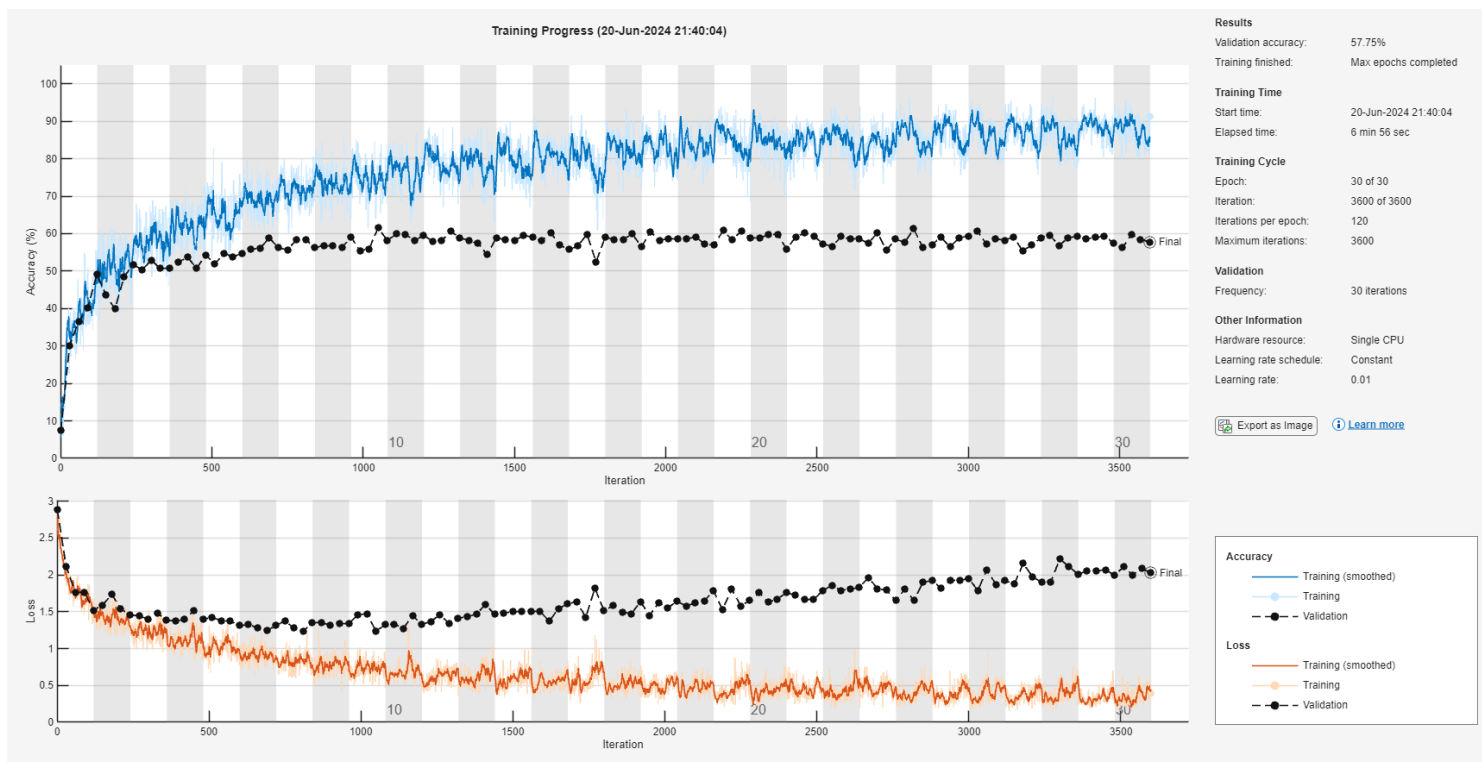
options = trainingOptions('sgdm', ...
    'InitialLearnRate',0.01, ...
    'MaxEpochs',30, ...
    'Shuffle','every-epoch', ...
    'ValidationData',{testData, testLabels}, ...
    'ValidationFrequency',30, ...
    'Verbose',false, ...
    'Plots','training-progress');

```

```

netImproved = trainNetwork(trainData, trainLabels, layers, options);

```



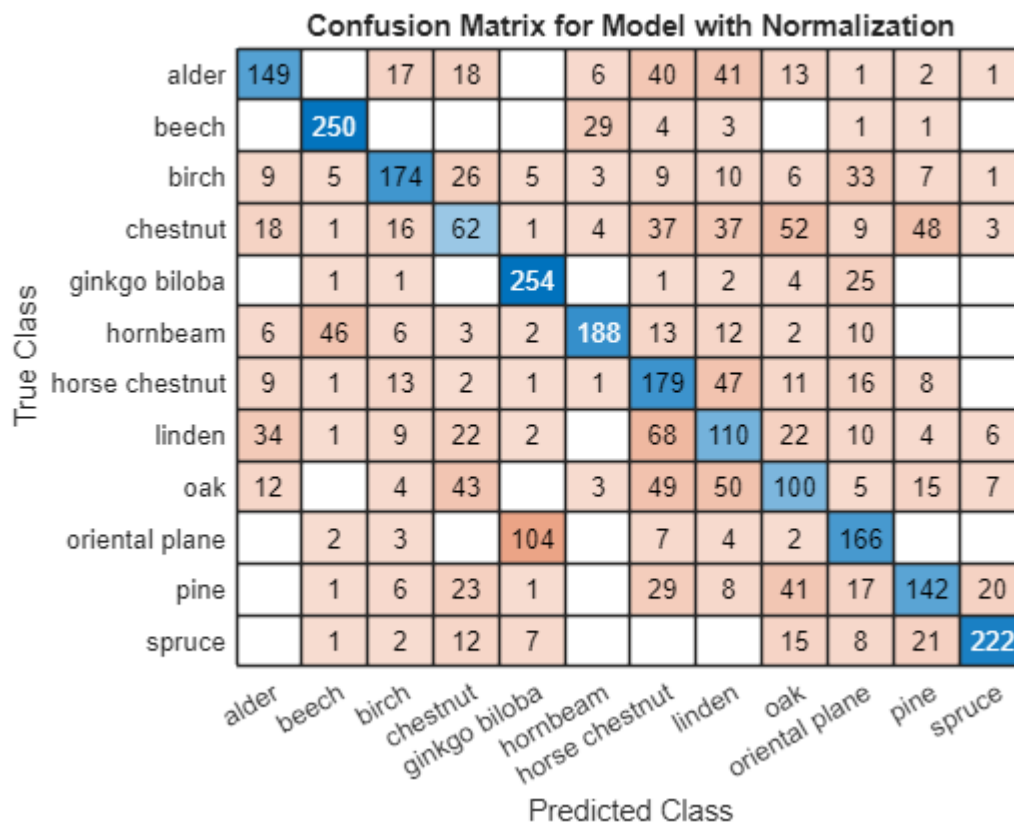
```

predictedLabels = classify(netImproved, testData);
accuracy = sum(predictedLabels == testLabels) / numel(testLabels);
fprintf('Test accuracy: %.2f%%\n', accuracy * 100);

```

Test accuracy: 57.75%

```
figure;
confusionchart(testLabels, predictedLabels);
title('Confusion Matrix for Model with Normalization');
```



```
numSamples = 10;
sampleIndices = randperm(numel(testLabels), numSamples);

figure('Visible', 'on');
for i = 1:numSamples
    index = sampleIndices(i);
    img = testData(:,:,index);
    trueLabel = testLabels(index);
    predictedLabel = predictedLabels(index);

    subplot(2, numSamples/2, i);
    imshow(img, 'InitialMagnification', 'fit');

    if trueLabel == predictedLabel
        borderColor = 'g';
        titleText = sprintf('Correct: %s', string(trueLabel));
    else
        borderColor = 'r';
        titleText = sprintf('Wrong: %s (Pred: %s)', string(trueLabel),
string(predictedLabel));
    end
end
```

```

rectangle('Position', [0.5, 0.5, size(img, 2), size(img, 1)], 'EdgeColor',
borderColor, 'LineWidth', 3);
title(titleText);
end

```



```

drawnow;

```

Not Normalised Data with Enhanced CNN model

```

trainData_not_normalised = trunk.train.data;
trainLabels = trunk.train.labels;
testData_not_normalised = trunk.test.data;
testLabels = trunk.test.labels;

```

```

layers = [
    imageInputLayer([40 40 3], 'Normalization', 'none')

    convolution2dLayer(3, 32, 'Padding', 'same')
    batchNormalizationLayer
    reluLayer
    maxPooling2dLayer(2, 'Stride', 2)

```



```

convolution2dLayer(3, 64, 'Padding', 'same')
batchNormalizationLayer
reluLayer
maxPooling2dLayer(2, 'Stride', 2)

convolution2dLayer(3, 128, 'Padding', 'same')
batchNormalizationLayer
reluLayer
maxPooling2dLayer(2, 'Stride', 2)

fullyConnectedLayer(256)
dropoutLayer(0.5) % Add dropout to prevent overfitting
reluLayer

fullyConnectedLayer(numel(unique(trainLabels)))
softmaxLayer
classificationLayer
];

```

```

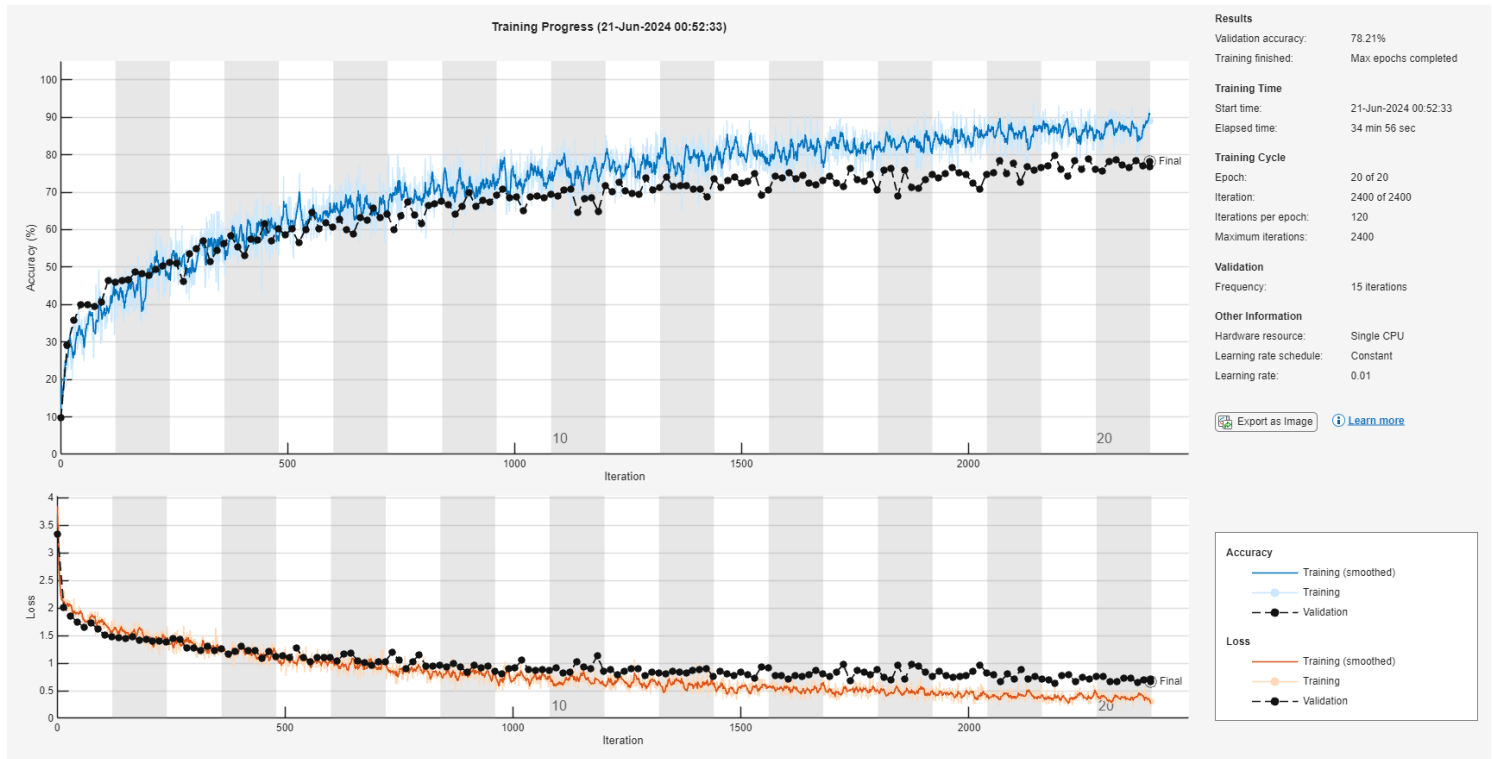
options = trainingOptions('sgdm', ...
    'InitialLearnRate',0.01, ...
    'MaxEpochs',20, ...
    'Shuffle','every-epoch', ...
    'ValidationData',{testData_not_normalised, testLabels}, ...
    'ValidationFrequency',15, ...
    'Verbose',false, ...
    'Plots','training-progress');

```

```

% Train the network
netImproved2 = trainNetwork(trainData_not_normalised, trainLabels, layers, options);

```



```
predictedLabels = classify(netImproved2, testData_not_normalised);
accuracy = sum(predictedLabels == testLabels) / numel(testLabels);
fprintf('Test accuracy: %.2f%%\n', accuracy * 100);
```

Test accuracy: 78.21%

```
figure;
confusionchart(testLabels, predictedLabels);
title('Confusion Matrix for Model with Batch Normalization, RELU, and Drop out
layer without normalised Data');
```

### on Matrix for Model with Batch Normalization, RELU, and Drop out layer without

True Class	alder	240	1	14	11		11	2	2	6		1	
	beech		286				2						
	birch	6	5	233	6	2	1	2		1	26	4	2
	chestnut	23	3	7	124			12		45	1	70	3
	ginkgo biloba					277	1	1		1	8		
	hornbeam		47		2		233				6		
	horse chestnut	5		17	1		7	217	22	1	5	12	1
	linden	27	1	10	18			51	155	10	8	4	4
	oak	12		2	9	2	2	16	16	193	4	17	15
	oriental plane		2	1		41	8	1			235		
	pine			2	6			19		3	1	223	34
	spruce					1							287
		alder	beech	birch	chestnut	ginkgo biloba	hornbeam	horse chestnut	linden	oak	oriental plane	pine	spruce
		Predicted Class											

```

numSamples = 10;
sampleIndices = randperm(numel(testLabels), numSamples);

figure('Visible', 'on');
for i = 1:numSamples
    index = sampleIndices(i);
    img = testData(:,:,index);
    trueLabel = testLabels(index);
    predictedLabel = predictedLabels(index);

    subplot(2, numSamples/2, i);
    imshow(img, 'InitialMagnification', 'fit');

    if trueLabel == predictedLabel
        borderColor = 'g'; % Green for correct classification
        titleText = sprintf('Correct: %s', string(trueLabel));
    else
        borderColor = 'r'; % Red for incorrect classification
        titleText = sprintf('Wrong: %s (Pred: %s)', string(trueLabel),
string(predictedLabel));
    end

    rectangle('Position', [0.5, 0.5, size(img, 2), size(img, 1)], 'EdgeColor',
borderColor, 'LineWidth', 3);
    title(titleText);

```

end

Wrong: chestnut (Pred: pine)    Correct: pine    Wrong: oak (Pred: pine)    Wrong: oak (Pred: chestnut)    Correct: linden



Wrong: linden (Pred: horse chestnut)    Correct: horse chestnut    Correct: horse chestnut    Wrong: linden (Pred: spruce)    Wrong: oak (Pred: linden)

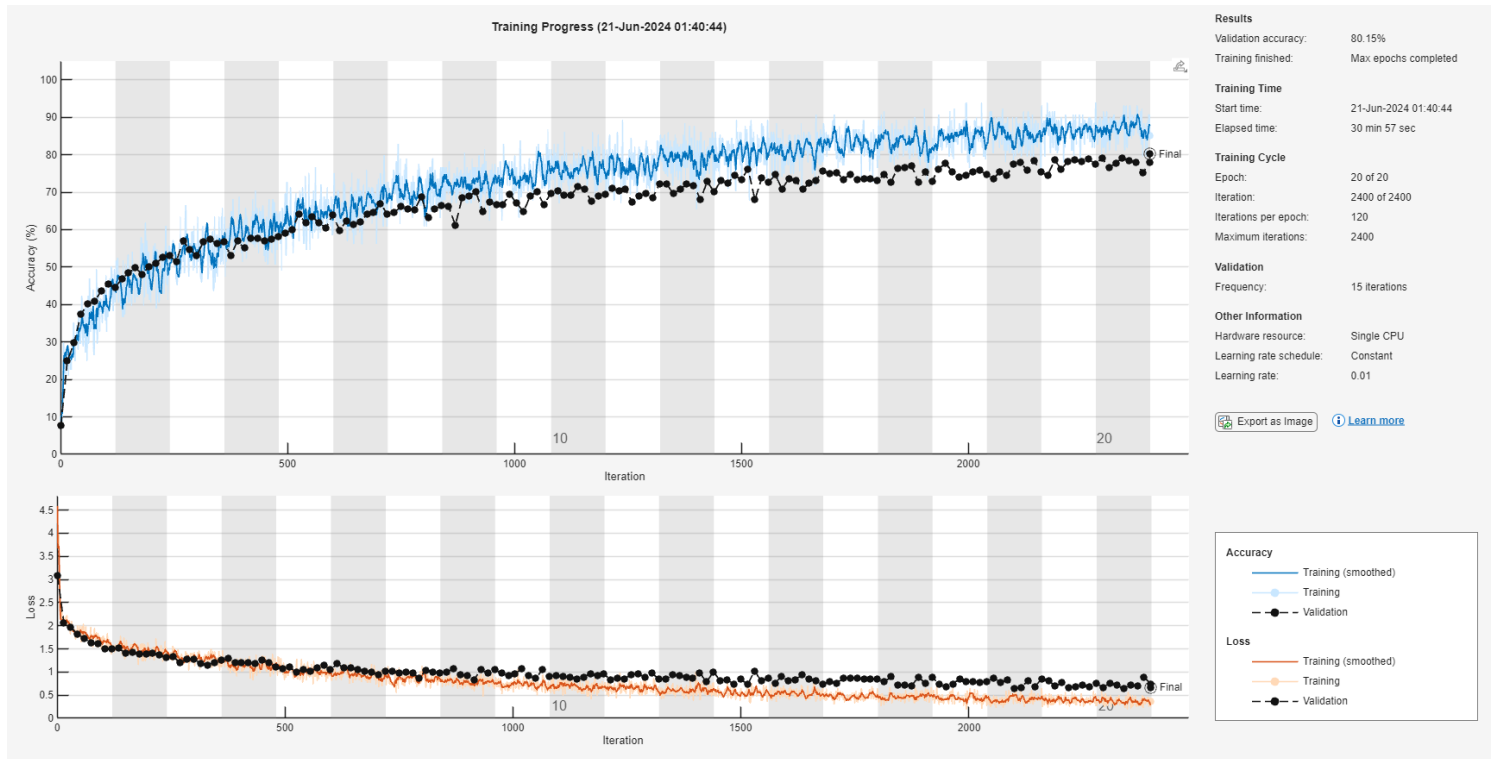


drawnow;

Normalised Data with Enhanced CNN model

```
options = trainingOptions('sgdm', ...  
    'InitialLearnRate',0.01, ...  
    'MaxEpochs',20, ...  
    'Shuffle','every-epoch', ...  
    'ValidationData',{testData, testLabels}, ...  
    'ValidationFrequency',15, ...  
    'Verbose',false, ...  
    'Plots','training-progress');
```

```
% Train the network  
netImproved3 = trainNetwork(trainData, trainLabels, layers, options);
```



```
predictedLabels = classify(netImproved3, testData);
accuracy = sum(predictedLabels == testLabels) / numel(testLabels);
fprintf('Test accuracy: %.2f%%\n', accuracy * 100);
```

Test accuracy: 80.15%

```
figure;
confusionchart(testLabels, predictedLabels);
title('Confusion Matrix for Model with Batch Normalization, RELU, and Drop out
layer with Normalised Data');
```

### Confusion Matrix for Model with Batch Normalization, RELU, and Drop out layer with N

True Class	alder	262		5	5		2	6	5	3			
	beech		280				8						
	birch	10		243		3	2	4	3		16	5	2
	chestnut	36		12	140		1	11	2	17		68	1
	ginkgo biloba					280				1	6		1
	hornbeam		32		2	2	237		2		13		
	horse chestnut	9		18	1		4	218	19	1	7	11	
	linden	29		5	4			28	192	14	5	9	2
	oak	28		2	21	2	2	4	21	193		10	5
	oriental plane		1			98	1		2		186		
	pine				4			4	6	4	1	252	17
	spruce									1			287
		alder	beech	birch	chestnut	ginkgo biloba	hornbeam	horse chestnut	linden	oak	oriental plane	pine	spruce
		Predicted Class											

```

numSamples = 10;
sampleIndices = randperm(numel(testLabels), numSamples);

figure('Visible', 'on');
for i = 1:numSamples
    index = sampleIndices(i);
    img = testData(:,:,index);
    trueLabel = testLabels(index);
    predictedLabel = predictedLabels(index);

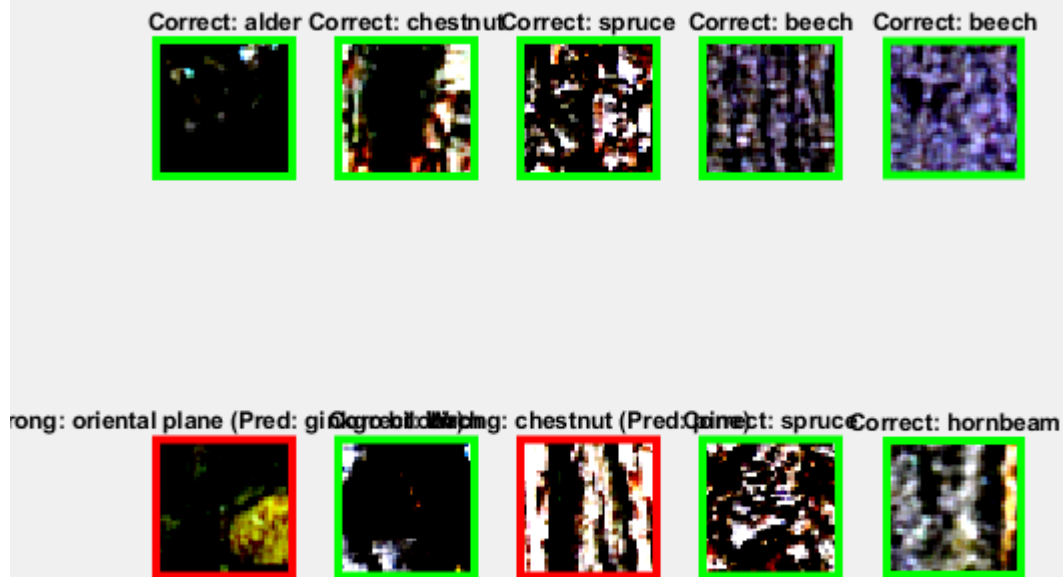
    subplot(2, numSamples/2, i);
    imshow(img, 'InitialMagnification', 'fit');

    if trueLabel == predictedLabel
        borderColor = 'g'; % Green for correct classification
        titleText = sprintf('Correct: %s', string(trueLabel));
    else
        borderColor = 'r'; % Red for incorrect classification
        titleText = sprintf('Wrong: %s (Pred: %s)', string(trueLabel),
string(predictedLabel));
    end

    rectangle('Position', [0.5, 0.5, size(img, 2), size(img, 1)], 'EdgeColor',
borderColor, 'LineWidth', 3);
    title(titleText);

```

end



drawnow;

Remove ReLU Activation from Enhanced Model with Normalised Data

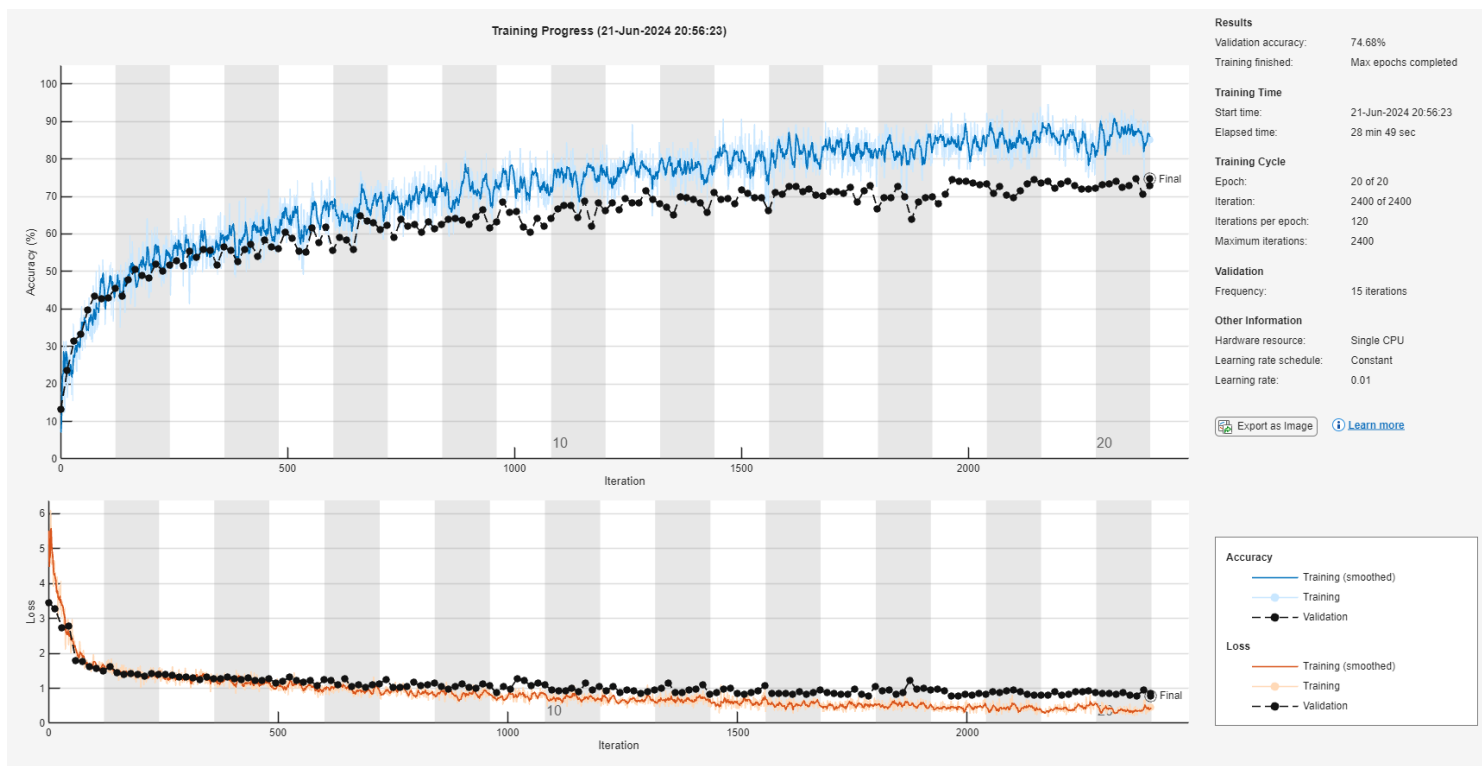
```
layers = [  
    imageInputLayer([40 40 3], 'Normalization', 'none')  
  
    convolution2dLayer(3, 32, 'Padding', 'same')  
    batchNormalizationLayer  
    maxPooling2dLayer(2, 'Stride', 2)  
  
    convolution2dLayer(3, 64, 'Padding', 'same')  
    batchNormalizationLayer  
    maxPooling2dLayer(2, 'Stride', 2)  
  
    convolution2dLayer(3, 128, 'Padding', 'same')  
    batchNormalizationLayer  
    maxPooling2dLayer(2, 'Stride', 2)  
  
    fullyConnectedLayer(256)  
    dropoutLayer(0.5) % Add dropout to prevent overfitting  
  
    fullyConnectedLayer(numel(unique(trainLabels)))  
    softmaxLayer
```

```
classificationLayer
];
```

```
options = trainingOptions('sgdm', ...
    'InitialLearnRate',0.01, ...
    'MaxEpochs',20, ...
    'Shuffle','every-epoch', ...
    'ValidationData',{testData, testLabels}, ...
    'ValidationFrequency',15, ...
    'Verbose',false, ...
    'Plots','training-progress');
```

```
% Train the network
```

```
netImproved4 = trainNetwork(trainData, trainLabels, layers, options);
```



```
predictedLabels = classify(netImproved4, testData);
accuracy = sum(predictedLabels == testLabels) / numel(testLabels);
fprintf('Test accuracy: %.2f%%\n', accuracy * 100);
```

```
Test accuracy: 74.68%
```

```
figure;
confusionchart(testLabels, predictedLabels);
title('Confusion Matrix for Removed ReLU Activation from Enhanced Model with Normalised Data');
```



**Confusion Matrix for Removed ReLU Activation from Enhanced Model with Normalization**

True Class	alder	227		7	28		1	15	3	6		1	
	beech		282				4		2				
	birch	2	1	204	29	2	5	2	3	5	27	7	1
	chestnut	16		4	190		1	1		34	1	41	
	ginkgo biloba					282	1			2	2	1	
	hornbeam	1	45	2	2	1	232			2	3		
	horse chestnut	16		11	6		2	196	30	3	1	21	2
	linden	20	1	4	36			36	172	11	2	5	1
	oak	41			47	13	11	15	38	105		9	9
	oriental plane					66	6	6			210		
	pine	1	1	1	18	2		13	1	18	5	196	32
	spruce					1				1		1	285
		alder	beech	birch	chestnut	ginkgo biloba	hornbeam	horse chestnut	linden	oak	oriental plane	pine	spruce
		Predicted Class											

```

numSamples = 10;
sampleIndices = randperm(numel(testLabels), numSamples);

figure('Visible', 'on');
for i = 1:numSamples
    index = sampleIndices(i);
    img = testData(:,:,index);
    trueLabel = testLabels(index);
    predictedLabel = predictedLabels(index);

    subplot(2, numSamples/2, i);
    imshow(img, 'InitialMagnification', 'fit');

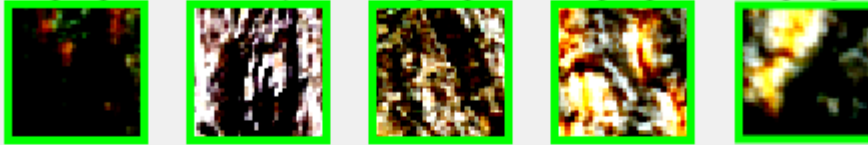
    if trueLabel == predictedLabel
        borderColor = 'g';
        titleText = sprintf('Correct: %s', string(trueLabel));
    else
        borderColor = 'r';
        titleText = sprintf('Wrong: %s (Pred: %s)', string(trueLabel),
string(predictedLabel));
    end

    rectangle('Position', [0.5, 0.5, size(img, 2), size(img, 1)], 'EdgeColor',
borderColor, 'LineWidth', 3);
    title(titleText);

```

end

Correct: ginkgo biloba Correct: pin Correct: ginkgo biloba Correct: ginkgo biloba Correct: ginkgo biloba



Correct: hornbeam Correct: chestnut Wrong: oak (Pred: alder) Correct: oak Correct: alder



drawnow;

Removed dropout layer from Enhanced Model with Normalised Data

```
layers = [  
    imageInputLayer([40 40 3], 'Normalization', 'none')  
  
    convolution2dLayer(3, 32, 'Padding', 'same')  
    batchNormalizationLayer  
    reluLayer  
    maxPooling2dLayer(2, 'Stride', 2)  
  
    convolution2dLayer(3, 64, 'Padding', 'same')  
    batchNormalizationLayer  
    reluLayer  
    maxPooling2dLayer(2, 'Stride', 2)  
  
    convolution2dLayer(3, 128, 'Padding', 'same')  
    batchNormalizationLayer  
    reluLayer  
    maxPooling2dLayer(2, 'Stride', 2)  
  
    fullyConnectedLayer(256)  
    reluLayer
```

```

fullyConnectedLayer(numel(unique(trainLabels)))
softmaxLayer
classificationLayer
];

```

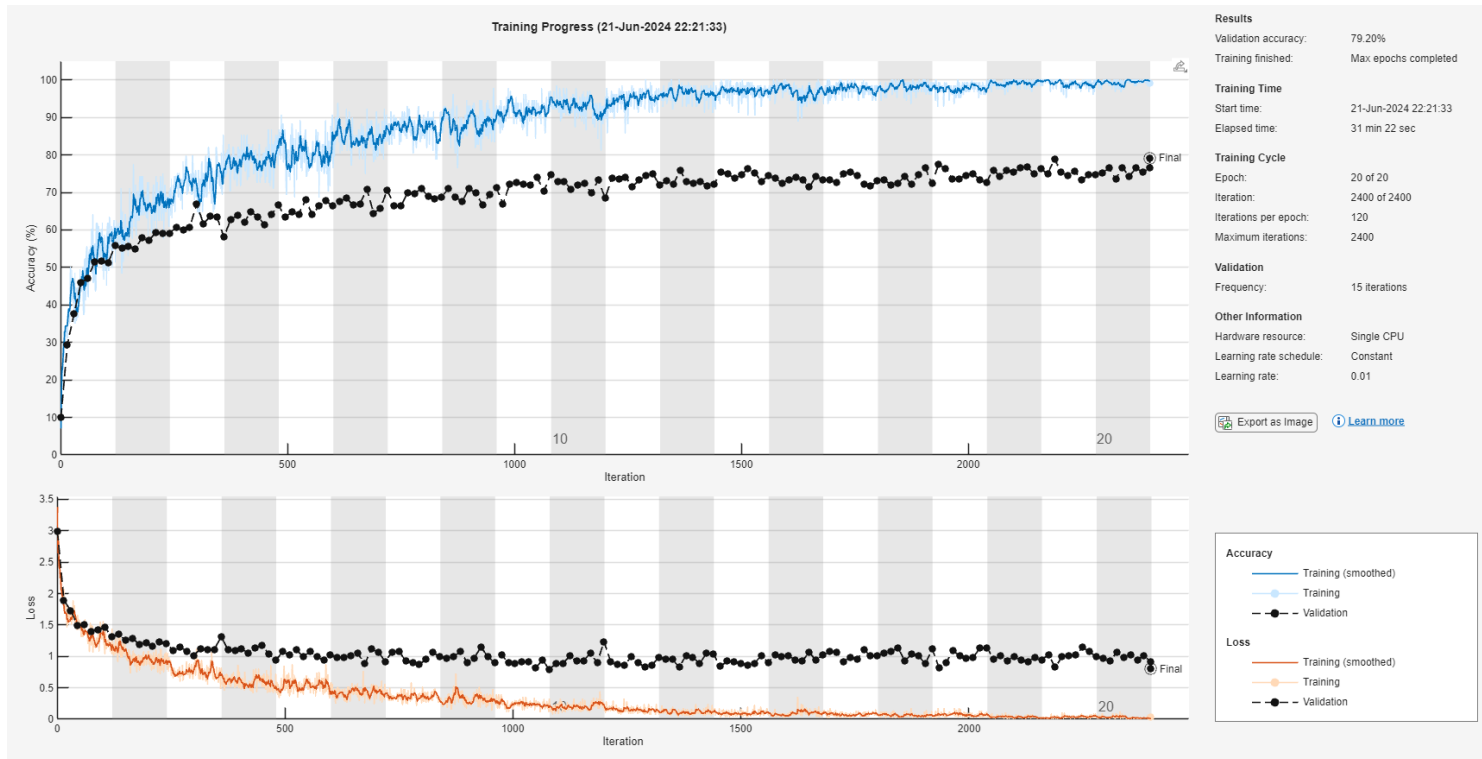
```

options = trainingOptions('sgdm', ...
    'InitialLearnRate',0.01, ...
    'MaxEpochs',20, ...
    'Shuffle','every-epoch', ...
    'ValidationData',{testData, testLabels}, ...
    'ValidationFrequency',15, ...
    'Verbose',false, ...
    'Plots','training-progress');

```

% Train the network

```
netImproved5 = trainNetwork(trainData, trainLabels, layers, options);
```



```

predictedLabels = classify(netImproved5, testData);
accuracy = sum(predictedLabels == testLabels) / numel(testLabels);
fprintf('Test accuracy: %.2f%%\n', accuracy * 100);

```

Test accuracy: 79.20%

```

figure;
confusionchart(testLabels, predictedLabels);

```

```
title('Confusion Matrix for Removed dropout layer from Enhanced Model with Normalised Data');
```

**Confusion Matrix for Removed dropout layer from Enhanced Model with Normali**

True Class	alder	237		5	14		3	5	6	17			1
	beech		285				3						
	birch	9		235	9		1	4	3		20	6	1
	chestnut	15	1	20	143		3	6	7	34	1	56	2
	ginkgo biloba			1		282					5		
	hornbeam		47	1	3	1	233				3		
	horse chestnut	12	2	14	8		5	198	30	2	5	12	
	linden	29	3	8	13		1	22	170	18	12	10	2
	oak	11			34	6	3	9	19	193	4	7	2
	oriental plane		2		6	25	2		2		251		
	pine			2	5			10	5	8	5	224	29
	spruce				1					1			286
		alder	beech	birch	chestnut	ginkgo biloba	hornbeam	horse chestnut	linden	oak	oriental plane	pine	spruce
Predicted Class													

```
numSamples = 10;
sampleIndices = randperm(numel(testLabels), numSamples);

figure('Visible', 'on');
for i = 1:numSamples
    index = sampleIndices(i);
    img = testData(:,:, :, index);
    trueLabel = testLabels(index);
    predictedLabel = predictedLabels(index);

    subplot(2, numSamples/2, i);
    imshow(img, 'InitialMagnification', 'fit');

    if trueLabel == predictedLabel
        borderColor = 'g';
        titleText = sprintf('Correct: %s', string(trueLabel));
    else
        borderColor = 'r';
        titleText = sprintf('Wrong: %s (Pred: %s)', string(trueLabel),
string(predictedLabel));
    end
end
```

```

rectangle('Position', [0.5, 0.5, size(img, 2), size(img, 1)], 'EdgeColor',
borderColor, 'LineWidth', 3);
title(titleText);
end

```



```

drawnow;

```

Remove increased convolutional filters (original filters)

```

layers = [
    imageInputLayer([40 40 3], 'Normalization', 'none')

    convolution2dLayer(3, 32, 'Padding', 'same')
    batchNormalizationLayer
    reluLayer
    maxPooling2dLayer(2, 'Stride', 2)

    convolution2dLayer(3, 64, 'Padding', 'same')
    batchNormalizationLayer
    reluLayer
    maxPooling2dLayer(2, 'Stride', 2)

    fullyConnectedLayer(256)
    dropoutLayer(0.5) % Add dropout to prevent overfitting
    reluLayer

```

```

fullyConnectedLayer(numel(unique(trainLabels)))
softmaxLayer
classificationLayer
];

```

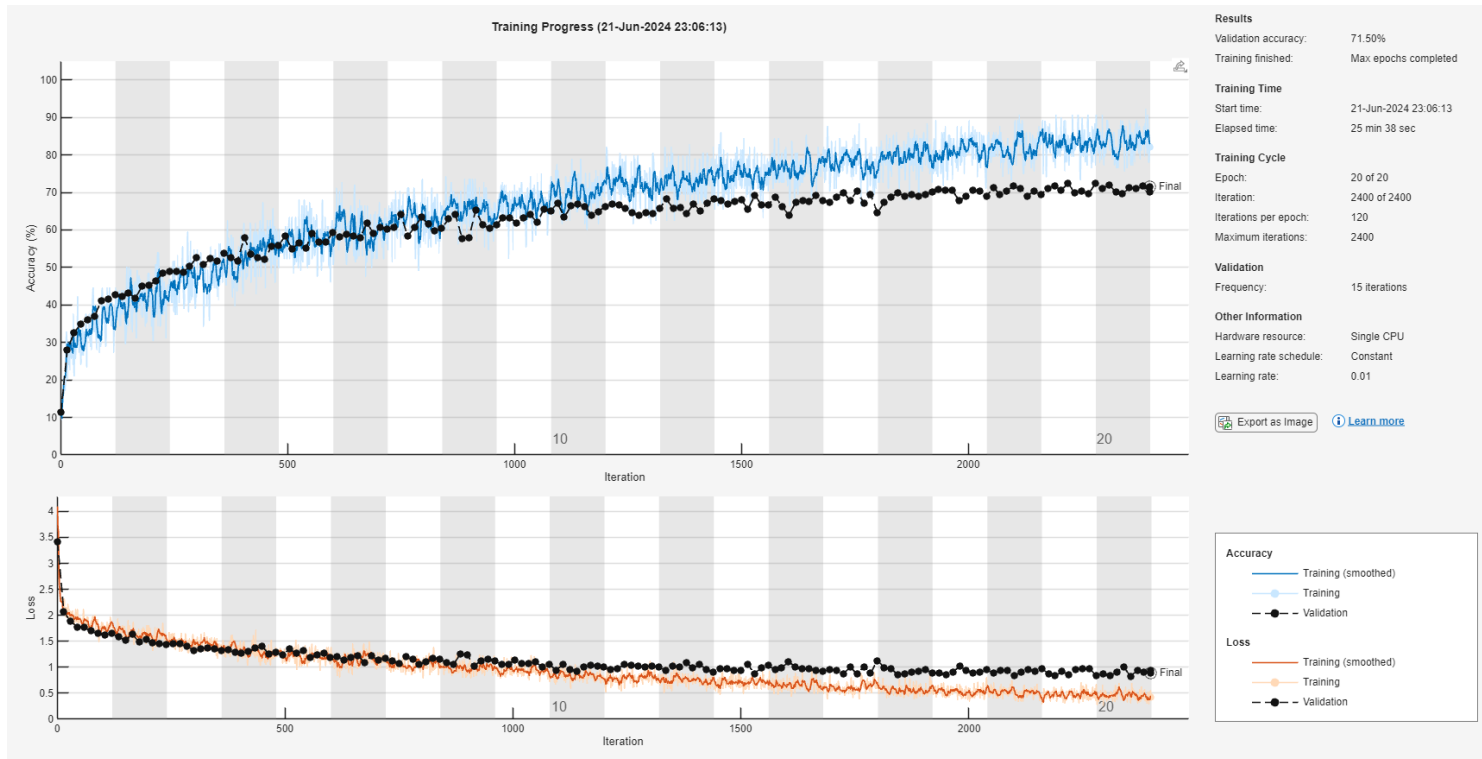
```

options = trainingOptions('sgdm', ...
    'InitialLearnRate',0.01, ...
    'MaxEpochs',20, ...
    'Shuffle','every-epoch', ...
    'ValidationData',{testData, testLabels}, ...
    'ValidationFrequency',15, ...
    'Verbose',false, ...
    'Plots','training-progress');

```

```
% Train the network
```

```
netImproved6 = trainNetwork(trainData, trainLabels, layers, options);
```



```

predictedLabels = classify(netImproved6, testData);
accuracy = sum(predictedLabels == testLabels) / numel(testLabels);
fprintf('Test accuracy: %.2f%%\n', accuracy * 100);

```

```
Test accuracy: 71.50%
```

```

figure;
confusionchart(testLabels, predictedLabels);

```

```
title('Confusion Matrix for Removed increased convolutional filters (original filters) from Enhanced Model with Normalised Data');
```

for Removed increased convolutional filters (original filters) from Enhanced Model

True Class	alder	189		6	27		7	16	33	9		1	
	beech		282				6						
	birch	10	4	217	13	1	6	8	2		12	11	4
	chestnut	19	1	5	73		1	30	10	48		99	2
	ginkgo biloba			1		276					11		
	hornbeam		63		4		217				4		
	horse chestnut	2	3	6	4			189	50	2	5	27	
	linden	3	1	4	16			53	177	3	4	20	7
	oak	7			48	2	2	17	26	160	3	14	9
	oriental plane		3		14	73	10	6	4	3	175		
	pine			2	7			12	2	3		231	31
	spruce									1		2	285
		alder	beech	birch	chestnut	ginkgo biloba	hornbeam	horse chestnut	linden	oak	oriental plane	pine	spruce
		Predicted Class											

```
numSamples = 10;
sampleIndices = randperm(numel(testLabels), numSamples);

figure('Visible', 'on');
for i = 1:numSamples
    index = sampleIndices(i);
    img = testData(:,:,index);
    trueLabel = testLabels(index);
    predictedLabel = predictedLabels(index);

    subplot(2, numSamples/2, i);
    imshow(img, 'InitialMagnification', 'fit');

    if trueLabel == predictedLabel
        borderColor = 'g';
        titleText = sprintf('Correct: %s', string(trueLabel));
    else
        borderColor = 'r';
        titleText = sprintf('Wrong: %s (Pred: %s)', string(trueLabel),
string(predictedLabel));
    end
end
```

```
    rectangle('Position', [0.5, 0.5, size(img, 2), size(img, 1)], 'EdgeColor',  
borderColor, 'LineWidth', 3);  
    title(titleText);  
end  
  
drawnow;
```