Flower Classification

Dataset

path to flower photos

```
% load from saved workspace. to not run again
% load workspace_variables\workspace_var

% here its in same directory as the matlab file, if not provide the path (eg:
C:\...)
path = "Data\flower_photos";
```

datastore creation

```
ds = imageDatastore(path, "IncludeSubfolders", true, "LabelSource", "foldernames");
```

Split into randomized training and testing sets

```
rng default
[trainImgs,testImgs] = splitEachLabel(ds,0.6,"randomized");
```

Preprocessing

Get Network and its layers

```
net = googlenet

net =
   DAGNetwork with properties:

        Layers: [144×1 nnet.cnn.layer.Layer]
   Connections: [170×2 table]
        InputNames: {'data'}
   OutputNames: {'output'}
```

input layer properties, getting the requirement

```
ly = net.Layers;
ly1 = ly(1).InputSize;
inputSZ = ly1(1:2);
```

updating datastore size

```
trainImgs = augmentedImageDatastore(inputSZ, trainImgs);
testImgs = augmentedImageDatastore(inputSZ, testImgs);
```

Number of categories

```
categos = numel(categories(ds.Labels));
```

New Network

modify GoogLeNet

layers

```
layers = layerGraph(net);
```

Modify the classification and output layers

```
% fully connected layer
newFc = fullyConnectedLayer(categos, "Name", "fc_nw");
layers = replaceLayer(layers, "loss3-classifier", newFc);

% output layer
newOut = classificationLayer("Name", "out_nw");
layers = replaceLayer(layers, "output", newOut);
```

Training algorithm

Training algorithm is set to SGDM

```
% learning rate set to 0.001
options = trainingOptions("sgdm", ...
    'MaxEpochs',20,...
    'MiniBatchSize', 64, ...
    "InitialLearnRate", 0.01, ...
    "LearnRateSchedule", "piecewise", ...
    "LearnRateDropFactor", 0.2, ...
    "LearnRateDropPeriod", 5, ...
    "Momentum", 0.9, ...
    "Shuffle", "once", ...
    'Plots', 'training-progress');
```

Perform training

```
[flowernet,info] = trainNetwork(trainImgs, layers, options);
```

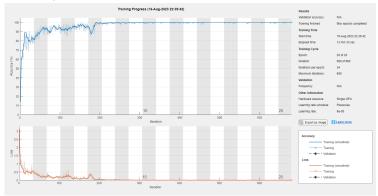
Training on single GPU.

Initializing input data normalization.

======	========	===========			============	
Epoch	Iteration	Time Elapsed	Mini-batch	Mini-batch	Base Learning	
		(hh:mm:ss)	Accuracy	Loss	Rate	
1	1	00:00:03	14.06%	3.1381	0.0100	
2	50	00:00:56	89.06%	0.2738	0.0100	
3	100	00:02:01	93.75%	0.1722	0.0100	
5	150	00:03:04	98.44%	0.0885	0.0100	
6	200	00:04:08	100.00%	0.0145	0.0020	
8	250	00:05:05	100.00%	0.0065	0.0020	
9	300	00:06:03	100.00%	0.0087	0.0020	
11	350	00:06:56	100.00%	0.0048	0.0004	
12	400	00:07:53	100.00%	0.0012	0.0004	
14	450	00:08:48	100.00%	0.0124	0.0004	
15	500	00:09:48	100.00%	0.0130	0.0004	
17	550	00:10:52	100.00%	0.0044	8.0000e-05	
18	600	00:11:48	100.00%	0.0044	8.0000e-05	
20	650	00:12:54	100.00%	0.0015	8.0000e-05	
20	680	00:13:33	100.00%	0.0029	8.0000e-05	

|-----|

Training finished: Max epochs completed.



Save the new Network

save workspace_variables\flowernet.mat flowernet;

Test the model

```
preds = classify(flowernet, testImgs);
```

Evaluate the results

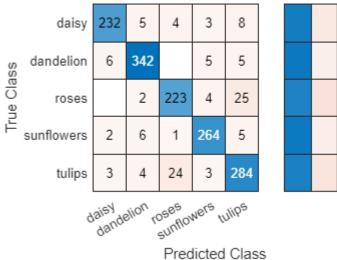
Accuracy

```
categ = extractBetween(testImgs.Files, "flower_photos\", "\");
accuracy = nnz(preds == categ) / numel(preds)
```

accuracy = 0.9212

Confusion Matrix

confusionchart(categorical(categ), preds, "RowSummary", "row-normalized");



Analyze misclassifications

Confusion matrix

```
% file of all the test images test set
paths = string(testImgs.Files);

% actual category of the flowers in test set
trueCat = extractBetween(paths, "flower_photos\", "\");

% original
org = categorical(trueCat);

% confusion matrix
cm = confusionmat(org, preds);

% Split into correct and incorrect classifications
% yes is diagonal elements i.e. predicted correctly
yes = diag(cm);

% elements other than diagonal, i.e. wrongly predicted
no = cm - diag(yes);
```

Misclassification rate for each letter

```
misratebyflwrs = sum(no,2) ./ sum(cm,2);
```

Table with letter names and misclassification rate

```
flwrs = categories(ds.Labels);

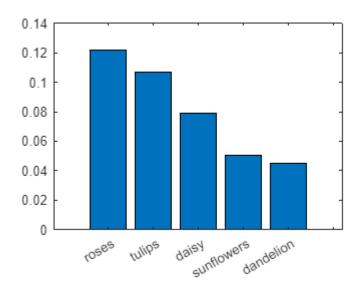
misratebyflwrs = table(flwrs,misratebyflwrs,'VariableNames',
   ["Flowers","MisClassRate"]);

% Sort by worst misclassification
   misratebyflwrs = sortrows(misratebyflwrs,"MisClassRate","descend")
```

$misratebyflwrs = 5 \times 2 table$

	Flowers	MisClassRate		
1	'roses'	0.1220		
2	'tulips'	0.1069		
3	'daisy'	0.0794		
4	'sunflowers'	0.0504		
5	'dandelion'	0.0447		

```
bar(misratebyflwrs.MisClassRate)
xticks(1:26)
xticklabels(misratebyflwrs.Flowers)
```



Individual Category misclassification

```
flw = "roses";
```

True class that were misclassified as something else

```
% the flower category selected (in drop down menu),
% and select the misclassified among that category
% i.e. category 'flwr' and wrong predcited
misclassidx = (trueCat == flw) & (string(preds) ~= trueCat);
```

Table of the misclassified observations, with the predicted letter

```
% name of the wrong category predicted, instead of category 'flw'
wrongPred = string(preds(misclassidx));

% files of wrong predicted flowers of category 'flw'
files_categoryPred = paths(misclassidx);
```

Plot

```
num = numel(wrongPred);
k = 1;
wr_pred_img = files_categoryPred(k);
imshow(wr_pred_img);
title("Predicted: "+wrongPred(k)+" -> Actual: "+flw)
```

Predicted: tulips -> Actual: roses



Saving all the workspace variables to .mat file

save workspace_variables\workspace_var;