

```
close all;
clear variables;
clc;
```

Read the training, validation and test partitions from the relevant

text files.
*** Adjust the file path as required. ***

```
folder = "CUB_200_2011_Subset20classes";
trainingImageNames = readtable(fullfile(folder, "train.txt"), 'ReadVariableNames', false);
trainingImageNames.Properties.VariableNames = {'index', 'imageName'};

validationImageNames = readtable(fullfile(folder, "validate.txt"), 'ReadVariableNames', false);
validationImageNames.Properties.VariableNames = {'index', 'imageName'};

testImageNames = readtable(fullfile(folder, "test.txt"), 'ReadVariableNames', false);
testImageNames.Properties.VariableNames = {'index', 'imageName'};
```

Read class info from the relevant text files

```
classNames = readtable(fullfile(folder, "classes.txt"), 'ReadVariableNames', false);
classNames.Properties.VariableNames = {'index', 'className'};

imageClassLabels = readtable(fullfile(folder, "image_class_labels.txt"), 'ReadVariableNames', false);
imageClassLabels.Properties.VariableNames = {'index', 'classLabel'};
```

Create lists of image names for training, validation and test subsets.

To be precise, we create an array of strings containing the full file path and file names for each data partition.

```
folder = "CUB_200_2011_Subset20classes/";
trainingImageList = strings(height(trainingImageNames), 1);
for iI = 1:height(trainingImageNames)
    trainingImageList(iI) = string(fullfile(folder, "images/", ...
        string(cell2mat(trainingImageNames.imageName(iI)))));
end

validationImageList = strings(height(validationImageNames), 1);
for iI = 1:height(validationImageNames)
    validationImageList(iI) = string(folder + "images/" + ...
        string(cell2mat(validationImageNames.imageName(iI)))));
end

testImageList = strings(height(testImageNames), 1);
for iI = 1:height(testImageNames)
    testImageList(iI) = string(folder + "images/" + ...
        string(cell2mat(testImageNames.imageName(iI)))));
end
```

Create image datastores for training, validation and test subsets

```
trainingImageDS = imageDatastore(trainingImageList, 'labelSource', 'foldernames', ...  
    'FileExtensions', {'.jpg'});  
trainingImageDS.ReadFcn = @readImagesIntoDatastore;  
  
validationImageDS = imageDatastore(validationImageList, 'labelSource', 'foldernames', ...  
    'FileExtensions', {'.jpg'});  
validationImageDS.ReadFcn = @readImagesIntoDatastore;  
  
testImageDS = imageDatastore(testImageList, 'labelSource', 'foldernames', ...  
    'FileExtensions', {'.jpg'});  
testImageDS.ReadFcn = @readImagesIntoDatastore;
```

The images all have different spatial resolutions (width x height), so

need to resize them to the same size. (Experiment with different sizes!)

```
targetSize = [100, 100];  
trainingImageDS_Resized = transform(trainingImageDS, @(x) imresize(x,targetSize));  
validationImageDS_Resized = transform(validationImageDS, @(x) imresize(x,targetSize));  
testImageDS_Resized = transform(testImageDS, @(x) imresize(x,targetSize));  
  
% Combine transformed datastores and labels  
labelsTraining = arrayDatastore(trainingImageDS.Labels);  
cdsTraining = combine(trainingImageDS_Resized, labelsTraining);  
labelsValidation = arrayDatastore(validationImageDS.Labels);  
cdsValidation = combine(validationImageDS_Resized, labelsValidation);  
labelsTest = arrayDatastore(testImageDS.Labels);  
cdsTest = combine(testImageDS_Resized, labelsTest);
```

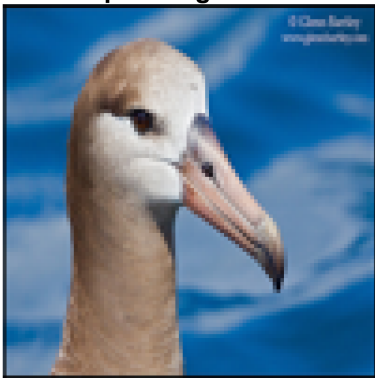
```
% %test  
% gray_img = rgb2gray(img);  
%  
% % Create a SURF object  
% surf_obj = detectSURFFeatures(gray_img);  
%  
% % Extract SURF features  
% [ztest_surf_features, ztest_valid_points] = extractFeatures(gray_img, surf_obj);  
%  
% % Compute HOG features  
% ztest_hog_features = extractHOGFeatures(gray_img, 'CellSize', [8 8]);  
% [coeff_surf, score_surf, latent_surf] = pca(ztest_surf_features);  
% [coeff_hog, score_hog, latent_hog] = pca(ztest_hog_features);  
%  
% % Use first n principal components to form feature vectors  
% n = min(size(score_surf,2), size(score_hog,2));  
% combined_features = [score_surf(:,1:n), score_hog(:,1:n)];
```

```
img = cdsTraining.read{1};
```

```
cellSize = [32 32];      % Set cell size to 16x16  
[hog_8x8, vis8x8] = extractHOGFeatures(img, 'CellSize', cellSize);  
numImagesTrain = numel(trainingImageDS_Resized.UnderlyingDatastores{1, 1}.Files);  
hogFeatureSize = length(hog_8x8);
```

```
[hog_8x8, vis8x8] = extractHOGFeatures(img, 'CellSize', cellSize);  
subplot(1, 2, 1);  
imshow(img);  
title('Sample Image Resized');  
subplot(1, 2, 2);  
imshow(rgb2gray(img)); hold on;  
title('SIFT Feature Points');  
plot(vis8x8);  
hold off;  
% noise reduction  
J = imnoise(im2gray(img), 'gaussian', 0, 0.025);  
K = wiener2(J, [5 5]);  
imshow(K);
```

Sample Image Resized



```
img_equalised = histeq(im2gray(img));
```

```
[hog_eq, vis_eq] = extractHOGFeatures(img_equalised,"CellSize", cellSize);
hog_8x8(1, 145:288) = hog_eq;
```

Extract HOG features

```
% %fitsvm
% %surf
% trainingFeatures = zeros(numImagesTrain, 2200, 'single');
% reset(cdsTraining); % Make sure we start with the first image
% for i = 1:numImagesTrain
%     imgFromDS = read(cdsTraining); % Get item from datastore. Note, this returns a cell array
%     imgGray = im2gray(imgFromDS{1}); % Ensure images are grayscale
%     points = detectSURFFeatures(imgGray);
%     features = extractFeatures(imgGray, points);
%     sz = size(features);
%     trainingFeatures(i, 1:sz(1)*sz(2)) = reshape(features, 1, []);
% end
```

```
% % surf test
%
% numImagesTest = numel(testImageDS_Resized.UnderlyingDatastores{1, 1}.Files);
%
% testFeatures = zeros(numImagesTest, 2200, 'single');
% reset(cdsTraining); % Make sure we start with the first image
% for i = 1:numImagesTest
%     imgFromDS = read(cdsTraining); % Get item from datastore. Note, this returns a cell array
%     imgGray = im2gray(imgFromDS{1}); % Ensure images are grayscale
%     points = detectSURFFeatures(imgGray);
%     features = extractFeatures(imgGray, points);
%     sz = size(features);
%     testFeatures(i, 1:sz(1)*sz(2)) = reshape(features, 1, []);
% end
```

```
% surfPoints = detectSURFFeatures(rgb2gray(img));
% subplot(1, 2, 1);
% imshow(img);
% subplot(1, 2, 2);
% imshow(rgb2gray(img)); hold on;
% plot(surfPoints);
% hold off;
```

```
% HOG
numImagesTrain = numel(trainingImageDS_Resized.UnderlyingDatastores{1, 1}.Files);
trainingFeaturesHOG = zeros(numImagesTrain, hogFeatureSize, 'single');
reset(cdsTraining); % Make sure we start with the first image
for i = 1:numImagesTrain
    imgFromDS = read(cdsTraining); % Get item from datastore. Note, this returns a cell array
    imgGray = im2gray(imgFromDS{1}); % Ensure images are grayscale
```

```

    %imgEqul = histeq(imgGray);
    %imgflip = flip(imgGray,2);
    temp = imnoise(imgGray, "gaussian", 0, 0.025);
    reduceNoise = wiener2(temp, [5 5]);
    trainingFeaturesHOG(i, 1:144) = extractHOGFeatures(imgGray, 'CellSize', cellSize);
    %trainingFeatures(i, 145:288) = extractHOGFeatures(imgEqul, 'CellSize', cellSize);
    trainingFeaturesHOG(i, 145:288) = extractHOGFeatures(reduceNoise, 'CellSize', cellSize);
end

```

```

% HOG test
numImagesTest = numel(testImageDS_Resized.UnderlyingDatastores{1, 1}.Files);
testFeaturesHOG = zeros(numImagesTest, hogFeatureSize, 'single');
reset(cdsTest); % Make sure we start with the first image
for i = 1:numImagesTest
    imgFromDS = read(cdsTest); % Get item from datastore
    imgGray = im2gray(imgFromDS{1}); % Ensure images are grayscale
    %imgEqul = histeq(imgGray);
    %imgflip = flip(imgGray,2);
    temp = imnoise(imgGray, "gaussian", 0, 0.025);
    reduceNoise = wiener2(temp, [10 10]);
    testFeaturesHOG(i, 1:144) = extractHOGFeatures(imgGray, 'CellSize', cellSize);
    %testFeatures(i, 145:288) = extractHOGFeatures(imgEqul, 'CellSize', cellSize);
    testFeaturesHOG(i, 145:288) = extractHOGFeatures(reduceNoise, 'CellSize', cellSize);
end

```

```

% %surf
% % imset = imageSet(cdsTraining.UnderlyingDatastores{1, 1}.UnderlyingDatastores{1, 1}.Files);
% imset = imageSet(trainingImageDS.Files);
% surfMetricThreshold = 1000;
% surfNumOctaves = 3;
% surfNumScaleLevels = 4;
% % Extract SURF features from the image set
% [surfFeatures, surfLocations] = helperExtractSURFFeaturesFromImageSet(trainingImageDS, surfMetricThreshold);

```

```

% %sift features
% numFeatures = 50;
% maxFeatures = 100;
%
% % Start by extracting features from the training set via the helper function.
% % These features will be used to train the classifier.
% [trainingFeaturesSIFT, trainingLabelsSIFT] = ...
%     helperExtractSIFTFeaturesFromImageSet(trainingImageDS_Resized.UnderlyingDatastores{1, 1}, numFeatures, maxFeatures);

```

```

% %sift test
% [testFeaturesSIFT, testLabelsSIFT] = ...
%     helperExtractSIFTFeaturesFromImageSet(testImageDS_Resized.UnderlyingDatastores{1, 1}, numFeatures, maxFeatures);

```

Check if we have a GPU available and clear any old data from it

```
if (gpuDeviceCount() > 0)
    disp('Found GPU:');
    disp(gpuDeviceTable);
    device = gpuDevice(1);
    reset(device); % Clear previous values that might still be on the GPU
end
```

Found GPU:

Index	Name	ComputeCapability	DeviceAvailable	DeviceSelected
1	"GRID T4-8Q"	"7.5"	true	true

```
% combinedFeatures = [trainingFeatures trainingFeaturesHOG];
% combinedFeatures = [trainingFeatures trainingFeaturesHOG trainingFeaturesSIFT];
% combinedFeatures = [trainingFeatures trainingFeaturesSIFT];
```

Train a multi-class SVM

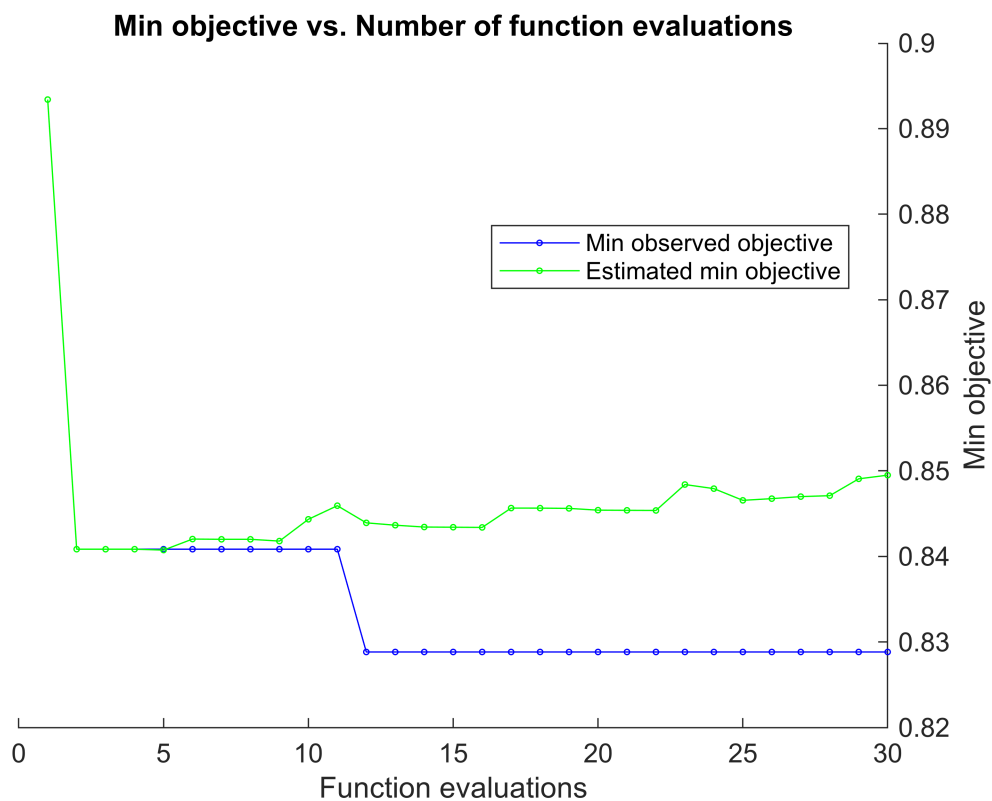
```
t = templateLinear('Solver', 'dual');
options = struct('UseParallel', true);

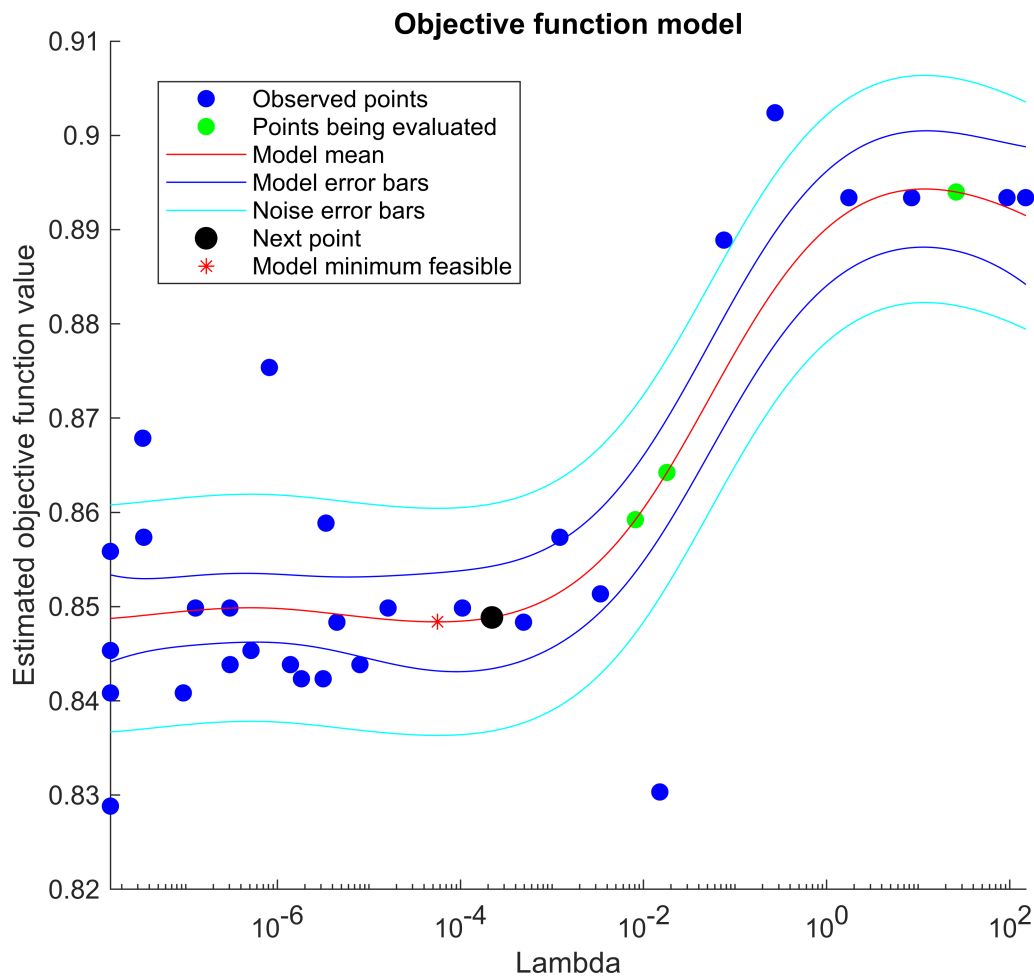
% fitcecoc uses multiple SVM learners and a 'One-vs-One' encoding scheme.
% Classifier = fitcecoc(trainingFeatures, trainingImageDS.Labels, 'Learners', t, 'Coding', 'one-vs-one');
% 'OptimizeHyperparameters', {'Lambda'}, ...
% 'HyperparameterOptimizationOptions', options);
Classifier = fitcecoc(trainingFeaturesHOG, trainingImageDS.Labels, 'Learners', t, ...
    'OptimizeHyperparameters', {'Lambda'}, ...
    'HyperparameterOptimizationOptions', options);
```

Copying objective function to workers...
Done copying objective function to workers.

Iter	Active workers	Eval result	Objective	Objective runtime	BestSoFar (observed)	BestSoFar (estim.)	Lambda
1	4	Best	0.89339	11.188	0.89339	0.89339	93.217
2	2	Best	0.84084	11.41	0.84084	0.84084	9.398e-08
3	2	Accept	0.88889	11.635	0.84084	0.84084	0.075455
4	2	Accept	0.84384	11.548	0.84084	0.84084	7.9845e-06
5	4	Accept	0.84985	11.455	0.84084	0.84074	1.6191e-05
6	4	Accept	0.84835	11.814	0.84084	0.84203	0.00048873
7	3	Accept	0.84535	11.885	0.84084	0.842	5.1564e-07
8	3	Accept	0.9024	12.217	0.84084	0.842	0.27303
9	4	Accept	0.84084	12.042	0.84084	0.84179	1.5041e-08
10	4	Accept	0.84985	12.158	0.84084	0.84434	1.2761e-07
11	4	Accept	0.85586	11.668	0.84084	0.84593	1.5016e-08
12	4	Best	0.82883	11.956	0.82883	0.84392	1.5023e-08
13	4	Accept	0.84535	13.33	0.82883	0.84365	1.5045e-08
14	4	Accept	0.85135	12.986	0.82883	0.84343	0.0033783
15	4	Accept	0.84384	13.096	0.82883	0.84341	1.3896e-06
16	4	Accept	0.84985	13.494	0.82883	0.84339	0.00010517
17	4	Accept	0.85736	11.644	0.82883	0.84565	3.4766e-08
18	4	Accept	0.89339	11.689	0.82883	0.84564	8.4752

19	4	Accept	0.84234	11.463	0.82883	0.84561	3.1662e-06
20	4	Accept	0.86787	11.292	0.82883	0.8454	3.3915e-08
=====							
Iter	Active workers	Eval result	Objective	Objective runtime	BestSoFar (observed)	BestSoFar (estim.)	Lambda
=====							
21	4	Accept	0.85736	11.117	0.82883	0.84539	0.0012199
22	4	Accept	0.89339	11.045	0.82883	0.84537	1.7487
23	4	Accept	0.85886	11.32	0.82883	0.8484	3.3985e-06
24	4	Accept	0.84384	11.514	0.82883	0.84792	3.0475e-07
25	4	Accept	0.84234	11.857	0.82883	0.84655	1.8324e-06
26	4	Accept	0.84835	11.755	0.82883	0.84675	4.472e-06
27	4	Accept	0.83033	11.713	0.82883	0.84699	0.015063
28	4	Accept	0.84985	11.833	0.82883	0.8471	3.0316e-07
29	4	Accept	0.87538	11.833	0.82883	0.84907	8.183e-07
30	4	Accept	0.89339	11.436	0.82883	0.84949	149.12





Optimization completed.
 MaxObjectiveEvaluations of 30 reached.
 Total function evaluations: 30
 Total elapsed time: 104.0134 seconds
 Total objective function evaluation time: 355.393

Best observed feasible point:
Lambda

1.5023e-08

Observed objective function value = 0.82883
 Estimated objective function value = 0.84875
 Function evaluation time = 11.9557

Best estimated feasible point (according to models):
Lambda

9.398e-08

Estimated objective function value = 0.84949
 Estimated function evaluation time = 11.8312


```
% combineTestFeature = [testFeatures testFeaturesHOG];
% combineTestFeature = [testFeatures testFeaturesHOG testFeaturesSIFT];
% combineTestFeature = [testFeatures testFeaturesSIFT];
```

Test the accuracy on the test partition

```
YPred = predict(Classifier, testFeaturesHOG);
YTest = testImageDS.Labels;

% Calculate overall accuracy
accuracy = sum(YPred == YTest)/numel(YTest) % Output on command line
```

```
accuracy = 0.1667
```

```
% Show confusion matrix in figure
[m, order] = confusionmat(YTest, YPred);
figure(2);
cm = confusionchart(m, order, ...
    'ColumnSummary','column-normalized', ...
    'RowSummary','row-normalized');
title("Overall Accuracy (HOG): "+ string(round(accuracy*100, 1)) + "%");
```

