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

Reading training, validation and test image names from text files

*** Please adjust the file path as required ***

```
folder = "CUB_200_2011";
trainingImageNames = readtable(fullfile(folder, "train200.txt"),'ReadVariableNames', false);
trainingImageNames.Properties.VariableNames = {'index', 'imageName'};

validationImageNames = readtable(fullfile(folder, "validate200.txt"),'ReadVariableNames', false,
validationImageNames.Properties.VariableNames = {'index', 'imageName'};

testImageNames = readtable(fullfile(folder, "test200.txt"),'ReadVariableNames', false);
testImageNames.Properties.VariableNames = {'index', 'imageName'};
```

Reading class names and image class labels from 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'};
```

```
% bounding box
boundingBox = readtable(fullfile(folder, "bounding_boxes.txt"), 'ReadVariableNames', false);
boundingBox.Properties.VariableNames = {'index', 'x', 'y', 'w', 'h'};
```

Creating training, validation and test subset lists of image names

Full file path names are created in an array format.

```
folder = "CUB_200_2011/";
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)
```

```
% mapping bounding boxes
trainingBox = return_bounding_box_mapping(trainingImageNames, boundingBox);
validationBox = return_bounding_box_mapping(validationImageNames, boundingBox);
testBox = return_bounding_box_mapping(testImageNames, boundingBox);
```

Creating training, validation and test subset image datastores

```
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;
```

countEachLabel(trainingImageDS)

200		2002	+-67	_
alis	=	200×2	Labi	2

	Label	Count
1	001	36
2	002	36
3	003	35
4	004	36
5	005	26
6	006	25
7	007	30
8	008	29
9	009	35
10	010	36
11	011	36
12	012	34
13	013	36
14	014	36
15	015	35

		2
40	Label	Count
16	016	35
17	017	34
18	018	26
19	019	35
20	020	35
21	021	36
22	022	34
23	023	35
24	024	31
25	025	36
26	026	36
27	027	36
28	028	35
29	029	36
30	030	36
31	031	36
32	032	32
33	033	35
34	034	35
35	035	36
36	036	36
37	037	35
38	038	36
39	039	35
40	040	36
41	041	36
42	042	36
43	043	35
44	044	36
45	045	36
46	046	36
47	047	36
48	048	36

	Label	Count
49	049	36
50	050	36
51	051	36
52	052	36
53	053	36
54	054	36
55	055	36
56	056	36
57	057	36
58	058	35
59	059	36
60	060	35
61	061	36
62	062	36
63	063	36
64	064	36
65	065	30
66	066	36
67	067	36
68	068	36
69	069	36
70	070	36
71	071	36
72	072	36
73	073	36
74	074	36
75	075	34
76	076	36
77	077	36
78	078	35
79	079	36
80	080	36
81	081	36

	Label	Count
82	082	36
83	083	36
84	084	29
85	085	36
86	086	36
87	087	36
88	088	36
89	089	36
90	090	36
91	091	36
92	092	36
93	093	36
94	094	36
95	095	36
96	096	36
97	097	35
98	098	36
99	099	36
100	100	36

countEachLabel(validationImageDS)

ans = 200×2 table

	Label	Count
1	001	12
2	002	12
3	003	12
4	004	12
5	005	9
6	006	8
7	007	12
8	008	10
9	009	12

	Label	Count
10	010	12
11	011	12
12	012	11
13	013	12
14	014	12
15	015	12
16	016	12
17	017	12
18	018	10
19	019	12
20	020	12
21	021	12
22	022	11
23	023	12
24	024	11
25	025	12
26	026	12
27	027	12
28	028	12
29	029	12
30	030	12
31	031	12
32	032	11
33	033	12
34	034	12
35	035	12
36	036	12
37	037	12
38	038	12
39	039	12
40	040	12
41	041	12
42	042	12

43 043 12 44 044 12 45 045 12 46 046 12 47 047 12 48 048 12 49 049 12 50 050 12 51 051 12 52 052 12 53 053 12 54 054 12 55 055 12 56 056 12 57 057 12 58 058 12 59 059 12 60 060 12 61 061 12 62 062 12 63 063 12 64 064 12 65 065 10 66 066 12 67 067 12 68 068 12 70 070 12 71 <td< th=""><th></th><th>Label</th><th>Count</th></td<>		Label	Count
44 044 12 45 045 12 46 046 12 47 047 12 48 048 12 49 049 12 50 050 12 51 051 12 52 052 12 53 053 12 54 054 12 55 055 12 56 056 12 57 057 12 58 058 12 59 059 12 60 060 12 61 061 12 62 062 12 63 063 12 64 064 12 65 065 10 66 066 12 67 067 12 68 068 12 70 070 12 71 071 12 72 <td< th=""><th>43</th><th></th><th></th></td<>	43		
45 045 12 46 046 12 47 047 12 48 048 12 49 049 12 50 050 12 51 051 12 52 052 12 53 053 12 54 054 12 55 055 12 56 056 12 57 057 12 58 058 12 59 059 12 60 060 12 61 061 12 62 062 12 63 063 12 64 064 12 65 065 10 66 066 12 67 067 12 68 068 12 69 069 12 70 070 12 71 071 12 72 <td< th=""><th>44</th><th></th><th></th></td<>	44		
46 046 12 47 047 12 48 048 12 49 049 12 50 050 12 51 051 12 52 052 12 53 053 12 54 054 12 55 055 12 56 056 12 57 057 12 58 058 12 59 059 12 60 060 12 61 061 12 62 062 12 63 063 12 64 064 12 65 065 10 66 066 12 67 067 12 68 068 12 69 069 12 70 070 12 71 071 12 72 072 12 74 <td< th=""><th>45</th><th></th><th></th></td<>	45		
47 047 12 48 048 12 49 049 12 50 050 12 51 051 12 52 052 12 53 053 12 54 054 12 55 055 12 56 056 12 57 057 12 58 058 12 59 059 12 60 060 12 61 061 12 62 062 12 63 063 12 64 064 12 65 065 10 66 066 12 67 067 12 68 068 12 70 070 12 71 071 12 72 072 12 73 073 12 74 074 12			
48 048 12 49 049 12 50 050 12 51 051 12 52 052 12 53 053 12 54 054 12 55 055 12 56 056 12 57 057 12 58 058 12 59 059 12 60 060 12 61 061 12 62 062 12 63 063 12 64 064 12 65 065 10 66 066 12 67 067 12 68 068 12 69 069 12 70 070 12 71 071 12 72 072 12 74 074 12			
49 049 12 50 050 12 51 051 12 52 052 12 53 053 12 54 054 12 55 055 12 56 056 12 57 057 12 58 058 12 59 059 12 60 060 12 61 061 12 62 062 12 63 063 12 64 064 12 65 065 10 66 066 12 67 067 12 68 068 12 69 069 12 70 070 12 71 071 12 72 072 12 74 074 12			
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51 051 12 52 052 12 53 053 12 54 054 12 55 055 12 56 056 12 57 057 12 58 058 12 59 059 12 60 060 12 61 061 12 62 062 12 63 063 12 64 064 12 65 065 10 66 066 12 67 067 12 68 068 12 69 069 12 70 070 12 71 071 12 72 072 12 73 073 12 74 074 12			
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64 064 12 65 065 10 66 066 12 67 067 12 68 068 12 69 069 12 70 070 12 71 071 12 72 072 12 73 073 12 74 074 12		062	12
65 065 10 66 066 12 67 067 12 68 068 12 69 069 12 70 070 12 71 071 12 72 072 12 73 073 12 74 074 12			12
66 066 67 067 68 068 69 069 70 070 71 071 72 072 73 073 74 074 12 13 14 15 16 17 18 19 10 12 12 12 12 12		064	12
67 067 12 68 068 12 69 069 12 70 070 12 71 071 12 72 072 12 73 073 12 74 074 12	65	065	10
68 068 12 69 069 12 70 070 12 71 071 12 72 072 12 73 073 12 74 074 12		066	12
69 069 70 070 71 071 72 072 73 073 74 074 12 73 074 12	67	067	12
70 070 12 71 071 12 72 072 12 73 073 12 74 074 12	68	068	12
71 071 12 72 072 12 73 073 12 74 074 12	69	069	12
72 072 12 73 073 12 74 074 12	70	070	12
73 073 12 74 074 12	71	071	12
74 074 12	72	072	12
12	73	073	12
⁷⁵ 075 12	74	074	12
	75	075	12

	Label	Count
76	076	12
77	077	12
78	078	12
79	079	12
80	080	12
81	081	12
82	082	12
83	083	12
84	084	12
85	085	12
86	086	12
87	087	12
88	088	12
89	089	12
90	090	12
91	091	12
92	092	12
93	093	12
94	094	12
95	095	12
96	096	12
97	097	12
98	098	12
99	099	12
100	100	12

countEachLabel(testImageDS)

ans = 200×2 table

	Label	Count
1	001	12
2	002	12
3	003	11

4 004 12 5 005 9 6 006 8 7 007 12 8 008 9 9 009 12 10 010 12 11 011 12 12 012 11 13 013 12 14 014 12 15 015 11 16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 30 030 12 31 031 12 32 032		Label	Count
5 005 9 6 006 8 7 007 12 8 008 9 9 009 12 10 010 12 11 011 12 12 012 11 13 013 12 14 014 12 15 015 11 16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 34 034	4		
6 006 8 7 007 12 8 008 9 9 009 12 10 010 12 11 011 12 12 012 11 13 013 12 14 014 12 15 015 11 16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 <th>5</th> <th></th> <th></th>	5		
7 007 12 8 008 9 9 009 12 10 010 12 11 011 12 12 012 11 13 013 12 14 014 12 15 015 11 16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035<	6		
8 008 9 9 009 12 10 010 12 11 011 12 12 012 11 13 013 12 14 014 12 15 015 11 16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12	7		
10 010 12 11 011 12 12 012 11 13 013 12 14 014 12 15 015 11 16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	8		
10 010 12 11 011 12 12 012 11 13 013 12 14 014 12 15 015 11 16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	9	009	12
12 012 11 13 013 12 14 014 12 15 015 11 16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	10	010	12
13 013 12 14 014 12 15 015 11 16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	11	011	12
14 014 12 15 015 11 16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	12	012	11
15 015 11 16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	13	013	12
16 016 11 17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	14	014	12
17 017 11 18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	15	015	11
18 018 9 19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	16	016	11
19 019 12 20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	17	017	11
20 020 12 21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	18	018	9
21 021 12 22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	19	019	12
22 022 11 23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	20	020	12
23 023 12 24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	21	021	12
24 024 10 25 025 12 26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	22	022	11
25 025 26 026 27 027 28 028 29 029 30 030 31 031 32 032 33 033 34 034 35 035 12 12 12 12 12	23	023	12
26 026 12 27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	24	024	10
27 027 12 28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	25	025	12
28 028 12 29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	26	026	12
29 029 12 30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	27	027	12
30 030 12 31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	28	028	12
31 031 12 32 032 10 33 033 12 34 034 12 35 035 12	29	029	12
32 032 10 33 033 12 34 034 12 35 035 12	30	030	12
33 033 12 34 034 12 35 035 12	31	031	12
34 034 12 35 035 12	32	032	10
35 035 12	33	033	12
12	34	034	12
36 036 12	35	035	12
	36	036	12

	Label	Count
37	037	12
38		
39	038	12
40	039	12
41	040	12
	041	12
42	042	12
43	043	12
44	044	12
45	045	12
46	046	12
47	047	12
48	048	12
49	049	12
50	050	12
51	051	12
52	052	12
53	053	12
54	054	12
55	055	12
56	056	12
57	057	12
58	058	11
59	059	12
60	060	12
61	061	12
62	062	12
63	063	12
64	064	12
65	065	10
66	066	12
67	067	12
68	068	12
69	069	12
	1	L

	Label	Count
70	070	12
71	071	12
72	072	12
73	073	12
74	074	12
75	075	11
76	076	12
77	077	12
78	078	12
79	079	12
80	080	12
81	081	12
82	082	12
83	083	12
84	084	12
85	085	12
86	086	12
87	087	12
88	088	12
89	089	12
90	090	12
91	091	12
92	092	12
93	093	12
94	094	12
95	095	12
96	096	12
97	097	12
98	098	12
99	099	12
100	100	12

:

```
% apply bounding box
trainingImageDS.ReadFcn = @(file_name) read_bounding_box_image_to_datastore(file_name, training
validationImageDS.ReadFcn = @(file_name) read_bounding_box_image_to_datastore(file_name, validationImageDS.ReadFcn = @(file_name) read_bounding_box_image_to_datastore(file_name, testBox);
```

Standardising the image sizes for ease of analysis and comparison

```
% target_size = [100, 100];
target_size = [224, 224];

% resizing using transform operation
training_image_datastore_resized = transform(trainingImageDS, @(image_i) imresize(image_i, target_size)
validation_image_datastore_resized = transform(validationImageDS, @(image_i) imresize(image_i, target_size)

% Combine transformed datastores and labels
training_labels = arrayDatastore(trainingImageDS.Labels);
training_combined_datastore = combine(training_image_datastore_resized, training_labels);

validation_labels = arrayDatastore(validationImageDS.Labels);
validation_combined_datastore = combine(validation_image_datastore_resized, validation_labels)

test_labels = arrayDatastore(testImageDS.Labels);
test_combined_datastore = combine(test_image_datastore_resized, test_labels);
```

```
trainedNetwork = resnet50;
analyzeNetwork(trainedNetwork);
```

```
% when we analyze the network the first layer will mention the input size
% for resnet 50 it's 224 224
lgraph = layerGraph(trainedNetwork);
deltafc1000 = fullyConnectedLayer(200,'Name','dfc1000');
deltaClassificationfc1000 = classificationLayer('Name', 'dcfc1000', 'Classes', 'auto');
lgraph = replaceLayer(lgraph,'fc1000',deltafc1000);
lgraph = replaceLayer(lgraph,'ClassificationLayer_fc1000',deltaClassificationfc1000);
```

Checking if a GPU is available and clearing any old data from it

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

Found GPU:

Index Name ComputeCapability DeviceAvailable DeviceSelected

Training a multi-class SVM

1

```
% learning_rate = 0.01;
learning_rate = 0.001;
% learning_rate = 0.0001;
% batch_size = 8;
batch_size = 16;
% batch_size = 32;
epochs = 5;
% epochs = 10;
% epochs = 20;
options = trainingOptions('sgdm', ...
        'InitialLearnRate', learning_rate, ...
        'MiniBatchSize', batch_size, ...
        'MaxEpochs', epochs, ...
        'Verbose', true, ...
        'Shuffle', 'every-epoch', ...
        'VerboseFrequency', 1, ...
        'ValidationData', validation_combined_datastore, ...
        'Plots', 'training-progress');
myCNN = trainNetwork(training_combined_datastore, lgraph, options);
```

Training on single GPU.

Initializing input data normalization.

=======	=========	==========	:========	:========		=========	
Epoch	Iteration	Time Elapsed	Mini-batch	Validation	Mini-batch	Validation	Base Learni
	1	(hh:mm:ss)	Accuracy	Accuracy	Loss	Loss	Rate
1 l	1	00:01:02	 0.00%	0.51%	 5.9187	======================================	 0.0
1 1	1 I	00:01:04	0.00%	0.51%	5.8983	3.7234 	0.0 0.0
1 1	2	00:01:04	6.25%		6.0086	 	0.0 0.0
1 I	ا ار ا	00:01:05	0.00%		5.6243	 	0.0
1 I	4	00:01:06	0.00%		5.6727	 	0.0
1 I	5 I	00:01:06	0.00%		5.6228	 	0.0
1	7	00:01:00	0.00%		5.6216	 	0.0
1	8	00:01:07	0.00%		5.2045	! 	0.0
1	9	00:01:08	0.00%		5.6365	! 	0.0
1	10	00:01:09	0.00%		5.9556	! 	0.0
- i	11	00:01:09	0.00%		5.4662	İ	0.0
- 1	12	00:01:10	0.00%		5.9389	İ	0.0
1	13	00:01:10	0.00%		5.6600	İ	0.0
1 İ	14	00:01:11	0.00%		5.4044	İ	0.0
1	15	00:01:11	0.00%		5.5680	İ	0.0
1	16	00:01:12	0.00%		5.6358	İ	0.0
1	17	00:01:13	0.00%		5.2918	İ	0.0
1	18	00:01:13	0.00%		5.3203		0.0
1	19	00:01:14	6.25%		5.4539		0.0
1	20	00:01:15	0.00%		5.4192		0.0
1	21	00:01:15	0.00%		5.4304		0.0

1	22	00:01:16	6.25%		5.4154	
j 1	23	00:01:16	0.00%	j	5.5791	j
j 1	24	00:01:17	0.00%		5.4018	i
j 1	25	00:01:18	12.50%		5.1139	i
1	26	00:01:18	0.00%		5.6352	i
i 1	27	00:01:19	0.00%		5.1723	i
i 1	28	00:01:19	6.25%		5.4986	i
1	29	00:01:20	0.00%		5.1920	i
1	30	00:01:21	0.00%	i	5.3908	i
1	31	00:01:21	0.00%		5.5923	i
1	32	00:01:22	0.00%		5.5744	i
1	33	00:01:22	6.25%		5.2683	i
1	34	00:01:23	0.00%		5.1938	i
1	35	00:01:24	0.00%	! 	5.1470	i
1	36	00:01:24	0.00%	! 	5.6528	i
1	37	00:01:25	0.00%		5.5139	i
1	38	00:01:25	0.00%	! 	5.4737	i
1 1	39	00:01:26	0.00%	! 	4.9971	i
1 1	40	00:01:27	0.00%	! 	5.4714	i
1 1	41	00:01:27	6.25%	 	5.3312	i
1 1	42	00:01:28	6.25%	 	5.0594	i
	43	00:01:28	0.00%	 	5.2599	i
	44	00:01:29	0.00%	 	5.3290	i
	45	00:01:29	0.00%	 	5.2744	
	45	00:01:30	6.25%	 	4.9560	<u> </u>
	47	00:01:31	6.25%	 	5.4913	
	47	00:01:31	6.25%	 	4.9101	<u> </u>
	48	00:01:32	0.00%	 	5.2027	<u> </u>
	50	00:02:04	0.00%	 3.25%	5.4665	5.1792
	51	00:02:05	0.00%	J.25% 	5.3335	3.1792
	52	00:02:05	0.00%	 	5.3663	ŀ
	53	00:02:06	12.50%	 	5.1620	<u> </u>
	54	00:02:07	0.00%	 	5.0603	<u> </u>
:	!	•	•	 		ŀ
1	55	00:02:07	6.25%	 	5.3479	ŀ
1	56 57	00:02:08	12.50% 6.25%	 	4.8567 4.8239	
1 1	58	00:02:09	6.25%	 	4.7928	<u> </u>
	59	00:02:10	0.00%	 	5.2599	
:	60	00:02:10	12.50%	 	4.9608	<u> </u>
1	61	00:02:10	6.25%	 	5.0695	ŀ
1 1	62	00:02:11	6.25%	 	3.0093 4.7597	ŀ
	63	00:02:11	0.00%	 	5.0683	
	64	00:02:12	6.25%	 	5.2108	<u> </u>
1	65	00:02:12	0.00%	 	5.2371	
	66	00:02:14	0.00%	 	5.2699	<u> </u>
	67	00:02:14	6.25%	 	4.7828	
	68	00:02:14	6.25%	 	5.1864	
	69	00:02:15	6.25%	 	4.6722	
1	70	00:02:16	12.50%	 	5.0193	i
	70	00:02:17	18.75%	 	4.5268	i
	72	00:02:17	6.25%	 	5.2521	i
1	73	00:02:17	12.50%	! 	4.9102	i
1 1	73	00:02:18	0.00%	! 	4.9958	i
	75	00:02:19	31.25%	 	4.2000	i
1	75	00:02:19	12.50%	 	4.6722	i
	76	00:02:19	6.25%	 	5.1061	
1	77	00:02:21	12.50%	 	4.7942	
1	79	00:02:21	6.25%	 	4.7942 4.5445	
1	80	00:02:22	6.25%	 	4.9980	
	81	00:02:22	12.50%	 	4.9447	
	82	00:02:23	0.00%	 	5.2966	
	83	00:02:24	0.00%	 	4.8496	
	84	00:02:24	0.00%	 	4.7065	
	85	00:02:25	12.50%	 	4.5940	
1 1	1 00	1 00.02.23	1 12.30%	ı	4.3340	I

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	1	86	00:02:25	18.75%	1	4.5215	1
	1	87	00:02:26	6.25%	ĺ	5.0951	
	1	88	00:02:27	6.25%	1	4.6846	
	1	89	00:02:27	12.50%	1	4.7272	
	1	90	00:02:28	6.25%	1	4.5818	
	1	91	00:02:28	18.75%		4.0724	
	1	92	00:02:29	6.25%		4.8495	
	1	93	00:02:30	18.75%	ļ	4.3743	
	1	94	00:02:30	0.00%	ļ	4.9975	
ļ	1	95	00:02:31	6.25%	ļ	4.9611	ļ
ļ	1	96	00:02:31	6.25%	ļ	4.6548	
ļ	1	97	00:02:32	12.50%	ļ	4.3961	
ļ	1	98	00:02:32	25.00%	ļ	4.4095	
	1	99	00:02:33	12.50%	44.05%	4.6027	4 5660
ļ	1	100	00:03:05	6.25%	11.05%	4.6012	4.5668
	1	101	00:03:06	6.25%	ļ	4.5638	
	1	102	00:03:07	12.50%	ļ	4.7994	
	1	103	00:03:07	0.00%	ļ	4.9787	
-	1	104	00:03:08	6.25%		4.6481	
-	1	105	00:03:08	6.25%	ļ	4.5933	
i i	1 1	106 107	00:03:09 00:03:10	12.50% 6.25%	ļ	4.5933 4.8264	
-	1	107	00:03:10	12.50%		4.8203	
-	1	100	00:03:10	0.00%		5.0398	
-	1	110	00:03:11	18.75%		4.3814	
	1	111	00:03:12	12.50%	ł	4.1887	
i	1	112	00:03:12	25.00%	ł	4.2959	
i	1	113	00:03:12	12.50%	I I	4.6197	
i	1	114	00:03:14	18.75%	i	4.4419	
i	1	115	00:03:14	12.50%	i	4.2552	
i	1	116	00:03:15	12.50%	i	4.0633	
i	1	117	00:03:15	6.25%	i	4.4600	
i	1 İ	118	00:03:16	25.00%	į	4.2425	į
i	1 j	119	00:03:17	12.50%	į	4.7657	į
i	1 İ	120	00:03:17	12.50%	į	4.3780	j
i	1	121	00:03:18	12.50%	į	4.6669	j
İ	1	122	00:03:18	25.00%	İ	4.0758	į
j	1	123	00:03:19	18.75%	į	4.2744	į
	1	124	00:03:20	18.75%	[4.0031	
	1	125	00:03:20	25.00%	1	3.9880	
	1	126	00:03:21	12.50%	1	3.9796	
ļ	1	127	00:03:22	12.50%		4.5148	
ļ	1	128	00:03:22	6.25%	1	4.6131	
ļ	1	129	00:03:23	25.00%	j	4.0403	ļ
ļ	1	130	00:03:24	25.00%	ļ	4.0014	ļ
-	1	131	00:03:25	12.50%	Į.	4.1259	ļ
-	1	132	00:03:25	25.00%	ļ	3.8110	
-	1	133	00:03:26	12.50%	!	4.2261	ļ
	1	134	00:03:27	18.75%	ļ	3.7621	
	1	135	00:03:27	25.00%	ļ	4.1458	ļ
-	1	136	00:03:28	37.50%	ļ	3.8040	l I
-	1	137	00:03:28	6.25%		4.5609	
-	1	138	00:03:29	0.00%		4.3490	
-	1	139	00:03:30	18.75%	ļ	4.2154	
l I	1	140 141	00:03:30	18.75% 12.50%	ļ	4.2468 3.7304	
	1	141 142	00:03:31	12.50%	ļ	:	l I
l I	1	142	00:03:32	6.25% 31.25%	ļ	4.1653	
	1 1	143	00:03:33	31.25% 6.25%	ļ	3.9980 4.1665	l I
-	1	144	00:03:33 00:03:34	12.50%		4.1911	
-	1	145	00:03:35	6.25%	· ·	4.1911	
1	1	147	00:03:35	12.50%	-	3.9926	
i	1	148	00:03:36	18.75%	1	4.3847	
i	1	149	00:03:37	6.25%		4.3133	
ı	- 1	±-7	00.05.57	0.23/0	I	4.2122	1

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1	150	00:04:11	31.25%	19.39%	3.5636	3.8998
1	151	00:04:12	12.50%	ļ	4.0147	ļ
1	152	00:04:13	18.75%	į.	4.0608	ļ
1	153	00:04:14	25.00%	ļ	3.8983	ļ
1	154	00:04:14	6.25%	ļ	3.9110	ļ
1	155	00:04:15	25.00%	ļ	3.4004	ļ
1	156	00:04:15	25.00%	ļ	3.7995	ļ
1	157	00:04:16	37.50%	ļ	3.5339	ļ
1	158	00:04:17	25.00%	ļ	3.6055	ļ
1	159	00:04:17	25.00%	ļ	3.6335	ļ
1	160	00:04:18	18.75%	ļ	4.0077	ļ
1	161	00:04:18	18.75%	!	4.3082	ļ
1	162	00:04:19	25.00%	ļ	3.4325	ļ
1	163	00:04:20	12.50%	ļ	3.8178	ļ
1	164	00:04:20	25.00%	ļ	3.8211	ļ
1	165	00:04:21	37.50%	!	3.3951	ļ
1	166	00:04:21	25.00%	ļ	4.1166	ļ
1	167	00:04:22	18.75%	!	3.6986	ļ
1	168	00:04:23	31.25%	ļ	3.1553	ļ
1	169	00:04:23	18.75%	ļ	3.8857	ļ
1	170	00:04:24	6.25%	ļ	3.7885	
1	171	00:04:24	25.00%	ļ	3.5498	
1	172	00:04:25	6.25%	ļ	3.7351	ļ
1	173	00:04:26	31.25%	ļ	3.8702	ļ
1	174	00:04:26	31.25%	ļ	3.7440	ļ
1	175	00:04:27	37.50%	ļ	3.2095	ļ
1	176	00:04:27	12.50%	ļ	3.9718	ļ
1	177	00:04:28	18.75%	ļ	3.7505	ļ
1	178	00:04:28	25.00%	ļ	3.7649	ļ
1	179	00:04:29	18.75%	ļ	3.8280	ļ
1	180	00:04:30	12.50%	ļ	4.0190	ļ
1	181	00:04:30	12.50%	ļ	3.4879	ļ
1	182	00:04:31	25.00%	ļ	3.6562	ļ
1	183	00:04:31	25.00%	ļ	3.7005	ļ
1	184	00:04:32	12.50%	ļ	3.8112	ļ
1	185	00:04:33	25.00%	ļ	3.4643	ļ
1	186	00:04:33	31.25%	ļ	2.7868	ļ
1	187	00:04:34	43.75%	ļ	3.1994	ļ
1	188	00:04:34	31.25%	ļ	3.5915	ļ
1	189	00:04:35	12.50%	!	3.2605	ļ
1	190	00:04:35	25.00%	ļ	3.3942	ļ
1	191	00:04:36	25.00%	ļ	3.8871	ļ
1	192	00:04:37	31.25%	ļ	3.5863	ļ
1	193	00:04:37	37.50%	ļ	3.1359	ļ
1	194	00:04:38	25.00% 31.25%	ļ	3.7947	l I
1	195	00:04:38	31.25%	ļ	3.4446	l I
1	196 197	00:04:39	25.00%	ļ	3.6012	l I
1	197	00:04:40	37.50%	ļ	3.2215	l I
1 1	198	00:04:40	31.25%	l I	3.4553	l I
1	199	00:04:41	25.00% 18.75%	29.93%	3.7088	3.2984
1 1	200 201	00:05:14	18.75% 31.25%	43.33/6 I	3.8485 3.0359	J. 2704
1	201 202	00:05:15	31.25% 18.75%	ļ	:	l I
1 1	203	00:05:15	18.75% 31.25%		3.6088	
1 1	204	00:05:16 00:05:16	25.00%	ļ	3.3326 3.5813	l I
1 1 1	204 205	00:05:16	31.25%	ļ	3.2417	l I
1 1	206	00:05:17	56.25%		2.6941	l I
: :	:					
1 1	207	00:05:18 00:05:19	25.00% 43.75%		3.5782 2.9997	l I
1 1	208	00:05:19	43.75% 18.75%			
1 1 1	209 210	00:05:19	18.75% 12.50%	ļ	3.4501 3.7943	l I
1 1	210	00:05:20	31.25%		3.1925	l I
1 1	211 212	00:05:21	31.25%		3.0883	l I
1 1	212 213	00:05:21	37.50%		3.1339	
1 1	213	00.03.22	3/.36%	I	3.1333	I

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1	214	00:05:22	6.25%	I	4.2067	
1	215	00:05:23	12.50%	İ	3.8118	İ
j 1	216	00:05:24	31.25%	j	2.9313	İ
1	•	00:05:24	25.00%	j	3.3887	İ
j 1		00:05:25	25.00%	į	3.1834	İ
j 1	:	00:05:25	43.75%	į	2.8863	İ
j 1	220	00:05:26	18.75%	į	3.4249	İ
j 1	221	00:05:27	31.25%	į	3.1464	i
1	•	00:05:27	31.25%	i	3.0136	i
1		00:05:28	25.00%	i	3.4325	i
1	:	00:05:29	37.50%	i	3.1257	i
1	225	00:05:29	31.25%	i	3.0148	i
1	226	00:05:30	25.00%	i	3.5078	i
1	•	00:05:30	43.75%	i	2.8660	i
1		00:05:31	25.00%	i	2.9265	i
1	:	00:05:31	18.75%	i	3.0684	i
	230	00:05:32	43.75%	i i	2.8333	i
	231	00:05:32	37.50%		3.0113	i
	•	00:05:33	25.00%	¦	3.2099	i
		00:05:34	25.00%	¦	3.4076	i
	:	00:05:34	37.50%	l l	2.4519	i
	235	00:05:35	31.25%	¦	3.2382	i
	236	00:05:36	50.00%		2.7888	i
	•	00:05:36	18.75%	l I	3.2870	ł
		00:05:37	25.00%	 	3.2701	ł
	:	00:05:37	37.50%	ļ	3.0343	ł
	239	00:05:38	12.50%	l I	3.2683	ł
	241	00:05:39	56.25%	l I	2.9233	ł
	•	!		l I		ł
1	•	00:05:39	37.50%	l I	2.9329	ł
1 1	:	00:05:40 00:05:40	37.50% 37.50%	l I	3.0893 2.9909	ł
:	!	: :		l I		ł
1	245	00:05:41	31.25%	l I	2.9711	ł
1	246	00:05:42	31.25%	l I	3.0290	ł
1	•	00:05:42	31.25%	ļ	3.1100	ļ
1	:	00:05:43	43.75%	l I	3.2360 2.8478	ł
1	249	00:05:43	50.00%	ا عد عم% ا		2 9570
1	!	00:06:17	43.75%	36.30%	2.8043	2.8570
1	251	00:06:17	31.25%	l I	3.0863	ł
1	:	00:06:18 00:06:18	25.00%	l I	2.8698 2.4779	ł
1 1	:	00:06:18	50.00% 56.25%	ļ	2.4779 2.6818	ł
:	!	: :		l I		ł
1	255	00:06:20	25.00%	l I	3.6401	ł
1		00:06:20	43.75%	l I	2.7446	ł
1		00:06:21	43.75%	l I	2.6126	ł
1		00:06:21 00:06:22	12.50%	ļ	3.1337 2.8762	ł
1		: :	31.25%	l I	:	ł
1		00:06:23 00:06:23	31.25% 62.50%	 	2.7000	ł
1 1		00:06:24		l I	2.4380	ł
	:	: :	50.00%	l I	2.3738 3.1464	ł
1		00:06:24	18.75%	ļ		ł
1		00:06:25	31.25%	l I	3.2211	ł
1	266	00:06:26	18.75%	l I	2.8631	ł
1		00:06:26	50.00%	l I	2.9217	ł
1	:	00:06:27	37.50%	l I	2.9646	ł
1		00:06:27	56.25%	ļ	2.9800	ł
1		00:06:28	37.50%	l I	2.7413	
1		00:06:28	43.75%	l I	3.1228	
1	271	00:06:29	56.25%	l I	2.2509	
1	:	00:06:30	31.25%	ļ	3.1931	
1		00:06:30	37.50%	ļ	2.7310	
1		00:06:31	50.00%	ļ	2.1911	
1	•	00:06:31	31.25%	ļ	2.7648	
1	•	00:06:32	37.50%	ļ	2.7878	
1	277	00:06:33	31.25%	I	3.4135	I

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- 1	1	278	00:06:33	31.25%	1	3.0111	1
i	1 İ	279	00:06:34	43.75%	i	2.7879	i
i	1	280	00:06:34	37.50%	i	2.4788	į
i	1	281	00:06:35	25.00%	i	2.8816	į
i	1	282	00:06:35	31.25%	i	2.7247	i
i	1	283	00:06:36	62.50%	i	2.3985	i
i	1	284	00:06:37	50.00%	i	2.4379	i
i	1	285	00:06:37	37.50%	i	2.6064	į
i	1	286	00:06:38	43.75%	i	2.7486	i
i	1	287	00:06:39	25.00%	i	2.7217	i
i	1	288	00:06:39	43.75%	i	2.3057	i
i	1	289	00:06:40	37.50%	i	2.5321	i
i	1	290	00:06:40	56.25%	i	2.5203	i
i	1	291	00:06:41	50.00%	i	2.3202	i
i	1	292	00:06:41	43.75%	i	2.7573	i
i	1	293	00:06:42	31.25%	i	2.8282	
i	1	294	00:06:43	31.25%	i	2.2542	i
i	1	295	00:06:43	31.25%	i	2.4678	i
i	1	296	00:06:44	25.00%	i	3.2303	
i	1	297	00:06:44	50.00%	i	2.1767	
H	1	298	00:06:45	25.00%	ŀ	2.7121	
-	1	299	00:06:46	43.75%		2.6946	
-	1	300	00:07:19	50.00%	43.13%	2.4762	2.4617
-	1	301	00:07:19	50.00%	45.15%	2.5055	2.4017
-	1	302	00:07:20	37.50%	-	2.6482	
-	1	303	00:07:21	50.00%	-	2.1761	
-	1	304	00:07:21	37.50%		3.0265	
-	1	305	00:07:21	37.50%		2.4347	
-	1	306	00:07:22	37.50%	-	2.6600	
-	1	307	00:07:23	43.75%	-	2.5143	
-	1	308	00:07:24	31.25%	-	2.8697	
-	1	309	00:07:24	68.75%	-	1.9936	
-	1	310	00:07:25	43.75%	-	2.6525	
-	1	311	00:07:26	56.25%		2.1295	l I
-	1	312	00:07:26	37.50%	-	2.6965	
-	1	313	00:07:27	43.75%		2.5421	
-	1	314	00:07:27	37.50%		2.4427	
-	1	315	00:07:27	37.50%	-	2.3331	
-	1	316	00:07:29	37.50%		2.9232	
-	1	317	00:07:29	56.25%		2.2447	
-	1	318	00:07:30	43.75%		2.4829	ļ
i	1	319	00:07:30	43.75%	i	2.6187	
i	1	320	00:07:31	25.00%	i	2.5963	i
-	1	321	00:07:31	62.50%	!	1.7638	
	1	322	00:07:32	50.00%		2.3391	
-	1	323	00:07:32	50.00%		2.6016	
i	1	324	00:07:33	56.25%	i	2.4498	
¦	1	325	00:07:34	43.75%	i	2.2350	
i	1	326	00:07:35	37.50%	i	2.2209	
¦	1	327	00:07:35	62.50%	i	1.9736	
-	1	328	00:07:36	37.50%		2.4075	
i	1	329	00:07:36	56.25%	i	1.7766	i
i	1	330	00:07:37	50.00%	i	2.7410	i
i	1	331	00:07:38	37.50%	i	2.8329	i
l	1	332	00:07:38	31.25%	i	2.3726	
l	1	333	00:07:30	62.50%	i	2.0422	
i	1	334	00:07:39	56.25%	i	2.4614	
l	1	335	00:07:40	37.50%	i	2.3335	
i	1	336	00:07:41	62.50%	i	2.3081	
¦	1	337	00:07:41	68.75%	i	2.1821	
	1	338	00:07:41	37.50%	i	2.2911	
¦	1	339	00:07:42	37.50%	i	2.3740	
i	1	340	00:07:42	18.75%	i	2.3571	
i	1	341	00:07:44	50.00%	i	2.1405	
ı	- 1	J-1	00.07.77	30.00%	I	2.1-02	į.

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ı	1	342	00:07:44	56.25%		2.2470	1
i	1	343	00:07:45	56.25%		1.9256	i
i	1	344	00:07:45	50.00%		2.3094	i
i	1	345	00:07:46	43.75%		2.3637	i
ŀ	1	346	00:07:46	56.25%		2.0322	ł
ŀ	1	347	00:07:47	43.75%		2.4647	ł
ŀ		347	00:07:48	50.00%		2.4047	ł
ļ	1		: :				ł
- !	1	349	00:07:48	43.75%	F2 440/	2.1217	2 1260
ļ	1	350	00:08:22	37.50%	52.11%	2.6239	2.1269
- !	1	351	00:08:22	50.00%		2.2673	ļ
ļ	1	352	00:08:23	75.00%		1.5958	ļ
ļ	1	353	00:08:23	56.25%		2.3256	ļ
ļ	1	354	00:08:24	56.25%		2.5484	ļ.
ļ	1	355	00:08:25	43.75%		2.3988	ļ ļ
ļ	1	356	00:08:25	50.00%		2.0104	ļ .
ļ	1	357	00:08:26	56.25%		2.1555	ļ.
	1	358	00:08:26	56.25%		1.9481	l
	1	359	00:08:27	37.50%		2.7694	l
	1	360	00:08:27	56.25%		2.2966	1
	1	361	00:08:28	43.75%		2.0413	I
	1	362	00:08:29	43.75%		1.9109	I
	1	363	00:08:29	62.50%		1.8292	1
	1	364	00:08:30	62.50%		1.8537	1
İ	1	365	00:08:30	50.00%		2.0819	İ
i	1	366	00:08:31	56.25%		2.1681	İ
i	1	367	00:08:32	50.00%		1.9098	j
i	1	368	00:08:32	25.00%		3.0785	i
i	1	369	00:08:33	56.25%		1.8956	i
i	1	370	00:08:33	56.25%		2.1392	i
i	1	371	00:08:34	75.00%		1.6263	i
ŀ	1	372	00:08:35	43.75%		2.0450	i
i	1	372	00:08:35	43.75%		2.3308	i
ŀ	1	373	00:08:36	50.00%		2.2639	ł
ŀ	1	374	00:08:36	62.50%		2.1840	ł
ŀ						2.1840	ł
ŀ	1 1	376 377	00:08:37 00:08:38	37.50%		2.5134	ł
ŀ			: :	37.50% 62.50%			ł
ļ	1	378	00:08:38			1.8210	ł
ŀ	1	379	00:08:39	62.50%		1.9070	ł
- !	1	380	00:08:39	31.25%		2.1476	ļ
ŀ	1	381	00:08:40	50.00%		2.3430	ļ
ļ	1	382	00:08:41	31.25%		2.4695	ļ
ļ	1	383	00:08:41	56.25%		1.9524	ļ
!	1	384	00:08:42	68.75%		1.8733	ļ.
ļ	1	385	00:08:42	62.50%		1.9101	ļ
ļ	1	386	00:08:43	31.25%		2.4549	ļ
ļ	1	387	00:08:43	50.00%		2.2013	ļ
ļ	1	388	00:08:44	43.75%		2.2063	ļ
ļ	1	389	00:08:45	50.00%		2.5019	ļ .
ļ	1	390	00:08:45	56.25%		2.1069	ļ .
ļ	1	391	00:08:46	43.75%		2.1641	ļ .
ļ	1	392	00:08:46	68.75%		1.6799	ļ
	1	393	00:08:47	43.75%		1.7988	l
	1	394	00:08:48	43.75%		2.2483	
	1	395	00:08:48	62.50%		1.5983	1
	1	396	00:08:49	68.75%		1.6497	I
ĺ	1	397	00:08:49	50.00%	ĺ	2.1205	
	1	398	00:08:50	50.00%		1.9111	
İ	1	399	00:08:51	75.00%	İ	1.5282	j
İ	1	400	00:09:22	56.25%	55.27%	1.8448	1.9441
į	1	401	00:09:23	56.25%	İ	1.9880	j
i	1	402	00:09:24	37.50%		1.8849	j
i	1	403	00:09:24	56.25%		2.4594	j
i	1	404	00:09:25	43.75%		1.9002	j
i	1	405	00:09:25	50.00%		1.9310	j
- 1	-					[ı

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1	406	00:09:26	56.25%	I	1.8238	
1	407	00:09:27	68.75%	ĺ	2.1486	İ
1	408	00:09:27	56.25%	ĺ	2.1872	İ
1	409	00:09:28	93.75%	j	1.3730	į
1	410	00:09:28	37.50%	j	2.2776	į
1	411	00:09:29	50.00%	j	1.7601	İ
1	412	00:09:30	75.00%	į	1.2008	į
1	413	00:09:30	56.25%	į	1.8730	į
1	414	00:09:31	50.00%	į	2.0773	į
1	415	00:09:31	50.00%	į	2.0051	j
1	416	00:09:32	56.25%	İ	2.4787	j
1	417	00:09:33	56.25%	i	1.7974	j
1	418	00:09:33	87.50%	i	1.3973	İ
1	419	00:09:34	68.75%	i	2.0603	j
1	420	00:09:34	37.50%	i	1.8954	İ
1	421	00:09:35	37.50%	i	2.4968	
1	422	00:09:36	50.00%	i	2.5792	
1	423	00:09:36	62.50%	i	2.0210	
1	424	00:09:37	50.00%	i	1.6667	
1	425	00:09:37	62.50%	¦	1.8758	
1	426	00:09:38	62.50%	i i	1.6443	
1	427	00:09:30	43.75%	¦	1.9659	
1	428	00:09:39	56.25%	¦	1.6414	
1 1	428	00:09:40	62.50%	¦	1.8527	
1 1	430	00:09:40	62.50%	 	1.6689	
	430	00:09:40	56.25%	¦	2.1157	
1 1	431	00:09:41	56.25%	 	1.7120	
1 1	432	00:09:41	50.00%	 	1.9258	
	433 434			 		
1		00:09:43	43.75%	 	2.1277 1.7186	
1	435	00:09:43	68.75%	l I		
1	436	00:09:44	50.00%	 	1.6660	
1	437	00:09:44	50.00%		1.9487	
1	438	00:09:45	68.75%		1.7285	
1	439	00:09:46	62.50%	!	1.7411	
1	440	00:09:46	43.75%	l I	2.0628	
1	441	00:09:47	37.50%		2.4177	
2	442	00:09:47	81.25%		1.0658	
2	443	00:09:48	68.75%		1.3992	
2	444	00:09:49	62.50%		1.2581	
2	445	00:09:49	87.50%	!	1.1456	
2	446	00:09:50	62.50%		1.7160	
2	447	00:09:50	81.25%	!	1.0083	
2	448	00:09:51	87.50%		0.8574	
2	449	00:09:52	62.50%	EC 00%	1.5601	1 7774
2	450 451	00:10:24 00:10:25	62.50%	56.96%	1.5591	1.7774
2 2	451 452		62.50% 62.50%	ļ	1.4852	
2	452 453	00:10:25	62.50% 81.25%	ļ	1.8060	
	453 454	00:10:26 00:10:26	81.25% 75.00%	ļ	0.8540 1.3703	
2	454 455	00:10:26	75.00% 75.00%	 	1.3703 1.1199	
2 2	455	00:10:27	50.00%	ļ	1.1199	
2	450 457			¦		
2	457	00:10:28	62.50%		1.2486	
		00:10:29	75.00%	 	1.5471	
2	459	00:10:29	62.50%	 	1.4396	
2	460 461	00:10:30 00:10:30	68.75%		1.5040 1.5715	
2	461		68.75%	 	:	
2	462	00:10:31	62.50%		1.6374	
2	463	00:10:32	75.00% 69.75%		1.5659	
2	464	00:10:32	68.75%		1.7831	
2	465	00:10:33	56.25% 81.25%	ļ	1.4247	
2	466 467	00:10:33	81.25% 81.25%		1.1639	
2	467	00:10:34	81.25%		1.2343	
2	468	00:10:35	87.50%		0.9855 1.1922	
2	469	00:10:35	68.75%	ı	1.1922	I

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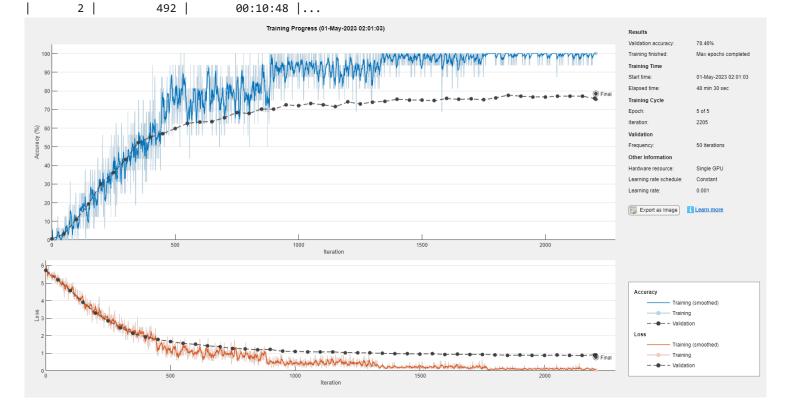
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	2	470	00:10:36	75.00%	1.0807	
	2	471	00:10:36	75.00%	1.6256	
	2	472	00:10:37	75.00%	1.2880	
ĺ	2	473	00:10:37	75.00%	1.6139	İ
ĺ	2	474	00:10:38	62.50%	1.6019	İ
ĺ	2	475	00:10:39	56.25%	1.3488	ĺ
ĺ	2	476	00:10:39	62.50%	1.6224	İ
ĺ	2	477	00:10:40	81.25%	1.1712	İ
ĺ	2	478	00:10:40	81.25%	1.2906	İ
ĺ	2	479	00:10:41	81.25%	1.2169	İ
ĺ	2	480	00:10:42	93.75%	0.8137	ĺ
ĺ	2	481	00:10:42	75.00%	1.3730	İ
ĺ	2	482	00:10:43	87.50%	1.0668	İ
	2	483	00:10:43	81.25%	1.6147	
ĺ	2	484	00:10:44	81.25%	1.0411	İ
ĺ	2	485	00:10:44	87.50%	0.8098	ĺ
	2	486	00:10:45	87.50%	0.7533	
	2	487	00:10:46	75.00%	1.4708	
	2	488	00:10:46	56.25%	1.9485	
	2	489	00:10:47	81.25%	1.1207	
ĺ	2	490	00:10:47	62.50%	1.4242	ĺ
	2	491	00:10:48	93.75%	0.9144	I
		400	1 00 40 40	I		

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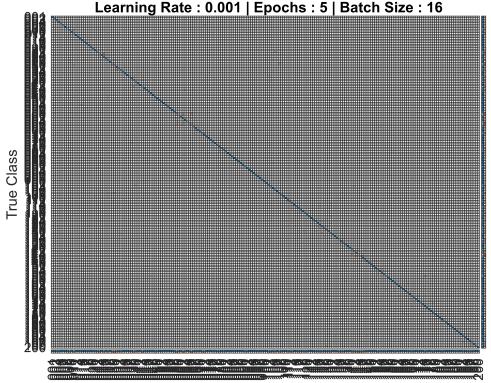
Test the accuracy on the test partition

```
YPred = classify(myCNN, test_image_datastore_resized);
YTest = testImageDS.Labels;

% Calculate overall accuracy
overall_accuracy = sum(YPred == YTest)/numel(YTest) % Output on command line
overall_accuracy = 0.7873
```

```
% Show confusion matrix in figure
[matrix, order] = confusionmat(YTest, YPred);
figure(2);
cm = confusionchart(matrix, order, ...
    'ColumnSummary','column-normalized', ...
    'RowSummary','row-normalized');
title({"Resnet50 with Bounding Box: Overall Accuracy " + string(round(overall_accuracy*100, 1))
    " | Image Size : " + target_size(1) + " x " + target_size(1); ...
    "Learning Rate : " + learning_rate + " | Epochs : " + epochs + " | Batch Size : " + batch_string
```

Resnet50 with Bounding Box: Overall Accuracy 78.7% | Image Size : 224 x 224



Predicted Class

```
class_wise_correct_recognition_rates = zeros(height(order), 1);
samples_per_row = sum(matrix, 2);
for i = 1:height(order)
    class_wise_correct_recognition_rates(i) = round(100 * matrix(i, i) / samples_per_row(i), 1
end
class_name_labels = table2array(classNames(:,2));

class_wise_recognition_rates = table(class_name_labels, ...
    class_wise_correct_recognition_rates, ...
    'VariableNames',["Class Name", "Correct Recognition Rate (%)"]);

disp("Class Weighted Average Overall Accuracy is " + string(round(overall_accuracy*100, 2)) + 1
```

Class Weighted Average Overall Accuracy is 78.73%

```
disp(class_wise_recognition_rates);
```

{'001.Black_footed_Albatross'	^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^	83.3
{'002.Laysan_Albatross'	}	75
{'003.Sooty_Albatross'	}	81.8
{'004.Groove_billed_Ani'	}	83.3
{'005.Crested_Auklet'	}	77.8
{'006.Least_Auklet'	}	75
{'007.Parakeet_Auklet'	}	100
{'008.Rhinoceros_Auklet'	}	55.6
{'009.Brewer_Blackbird'	}	50
{'010.Red_winged_Blackbird'	}	91.7
{'011.Rusty_Blackbird'	j.	41.7
{'012.Yellow_headed_Blackbird'	ì	100
{'013.Bobolink'	í	91.7
{'014.Indigo_Bunting'	}	91.7
{'015.Lazuli_Bunting'	}	100
{'016.Painted_Bunting'	}	81.8
{'017.Cardinal'	, ,	90.9
{'018.Spotted_Catbird'	, ,	100
{'019.Gray_Catbird'	J	75
{'020.Yellow_breasted_Chat'	J	83.3
	}	
{'021.Eastern_Towhee'	}	83.3
{'022.Chuck_will_Widow'	}	54.5
{'023.Brandt_Cormorant'	}	33.3
{'024.Red_faced_Cormorant'	}	80
{'025.Pelagic_Cormorant'	}	83.3
{'026.Bronzed_Cowbird'	}	83.3
{'027.Shiny_Cowbird'	}	50
{'028.Brown_Creeper'	}	100
{'029.American_Crow'	}	33.3
{'030.Fish_Crow'	}	33.3
<pre>{'031.Black_billed_Cuckoo'</pre>	}	91.7
{'032.Mangrove_Cuckoo'	}	70
{'033.Yellow_billed_Cuckoo'	}	58.3
<pre>{'034.Gray_crowned_Rosy_Finch'</pre>	}	83.3
{'035.Purple_Finch'	}	91.7
{'036.Northern_Flicker'	}	91.7
{'037.Acadian_Flycatcher'	}	41.7
{'038.Great_Crested_Flycatcher'	}	83.3
{'039.Least_Flycatcher'	}	50
{'040.0live sided Flycatcher'	}	58.3
{'041.Scissor_tailed_Flycatcher'	}	83.3
{'042.Vermilion_Flycatcher'		91.7
{'043.Yellow_bellied_Flycatcher'	í	25
{'044.Frigatebird'	}	83.3
{'045.Northern Fulmar'	}	58.3
{'046.Gadwall'	}	83.3
{'047.American Goldfinch'	}	100
{'048.European_Goldfinch') }	100
{'049.Boat_tailed_Grackle'	J	75
{'050.Eared_Grebe') }	100
	\frac{1}{2}	
{'051.Horned_Grebe'	}	75
{'052.Pied_billed_Grebe'	}	91.7
{'053.Western_Grebe'	}	100
{'054.Blue_Grosbeak'	}	75
{'055.Evening_Grosbeak'	}	91.7
{'056.Pine_Grosbeak'	}	91.7
{'057.Rose_breasted_Grosbeak'	}	91.7
{'058.Pigeon_Guillemot'	}	90.9
{'059.California_Gull'	}	41.7
{'060.Glaucous_winged_Gull'	}	16.7
{'061.Heermann_Gull'	}	100

<pre>{'062.Herring_Gull' {'063.Ivory_Gull' {'064.Ring_billed_Gull'</pre>	<pre>} } } } } } } } } } } } } } } } } } }</pre>	66.7 91.7 41.7
{'065.Slaty_backed_Gull'	}	30
{'066.Western_Gull'	}	33.3
<pre>{'067.Anna_Hummingbird' {'068.Ruby_throated_Hummingbird'</pre>	} ``	66.7 83.3
{ '069.Rufous_Hummingbird'	} l	83.3
{'070.Green_Violetear'	}	91.7
{'071.Long_tailed_Jaeger'	}	41.7
{'072.Pomarine_Jaeger'	}	83.3
{'073.Blue_Jay'	}	91.7
{'074.Florida_Jay'	}	100
{'075.Green_Jay'	}	100
{'076.Dark_eyed_Junco'	}	91.7
{'077.Tropical_Kingbird'	}	91.7
{'078.Gray_Kingbird'	}	66.7
<pre>{'079.Belted_Kingfisher' {'080.Green_Kingfisher'</pre>	} `	91.7 66.7
{'081.Pied_Kingfisher'	} \	75
{'082.Ringed_Kingfisher'	} }	66.7
{'083.White_breasted_Kingfisher'	}	100
{'084.Red_legged_Kittiwake'	}	83.3
{'085.Horned_Lark'	}	91.7
{'086.Pacific_Loon'	}	100
{'087.Mallard'	}	91.7
{'088.Western_Meadowlark'	}	100
{'089.Hooded_Merganser'	}	91.7
{'090.Red_breasted_Merganser'	}	91.7
{'091.Mockingbird'	}	66.7
{'092.Nighthawk'	}	83.3
{'093.Clark_Nutcracker'	}	100
<pre>{'094.White_breasted_Nuthatch' {'095.Baltimore_Oriole'</pre>	}	100 100
{'096.Hooded_Oriole'	}	83.3
{'097.Orchard_Oriole'	}	91.7
{'098.Scott_Oriole'	}	91.7
['099.0venbird'	}	91.7
{'100.Brown_Pelican'	}	91.7
{'101.White_Pelican'	}	100
{'102.Western_Wood_Pewee'	}	66.7
{'103.Sayornis'	}	50
{'104.American_Pipit'	}	100
{'105.Whip_poor_Will'	}	70
<pre>{'106.Horned_Puffin' {'107.Common_Raven'</pre>	} l	100 50
{'108.White_necked_Raven'	}	75
{'109.American_Redstart'	}	91.7
{'110.Geococcyx'	ì	100
{'111.Loggerhead_Shrike'	}	83.3
{'112.Great_Grey_Shrike'	}	66.7
{'113.Baird_Sparrow'	}	80
{'114.Black_throated_Sparrow'	}	100
{'115.Brewer_Sparrow'	}	58.3
{'116.Chipping_Sparrow'	}	66.7
{'117.Clay_colored_Sparrow'	} `	75
{'118.House_Sparrow' {'119.Field_Sparrow'	} \	91.7 58.3
{'120.Fox_Sparrow') }	83.3
{'121.Grasshopper_Sparrow'	}	75
{'122.Harris_Sparrow'	}	100
{'123.Henslow_Sparrow'	}	66.7
{'124.Le_Conte_Sparrow'	} } } } } } } } } }	66.7
{'125.Lincoln_Sparrow'	}	83.3

{'126.Nelson_Sharp_tailed_Sparrow'	ı	91.7
	}	
{'127.Savannah_Sparrow'	}	75
{'128.Seaside_Sparrow'	}	58.3
{'129.Song_Sparrow'	}	75
{'130.Tree_Sparrow'	}	66.7
{'131.Vesper_Sparrow'	}	58.3
<pre>{'132.White_crowned_Sparrow'</pre>	}	91.7
{'133.White_throated_Sparrow'	}	91.7
{'134.Cape_Glossy_Starling'	}	100
{'135.Bank_Swallow'	, }	33.3
{'136.Barn_Swallow'	}	75
{'137.Cliff_Swallow'	`` ``	58.3
	J	83.3
{'138.Tree_Swallow') }	
{'139.Scarlet_Tanager'	}	100
{'140.Summer_Tanager'	}	83.3
{'141.Artic_Tern'	}	63.6
{'142.Black_Tern'	}	50
{'143.Caspian_Tern'	}	75
{'144.Common_Tern'	}	16.7
{'145.Elegant_Tern'	}	25
{'146.Forsters_Tern'	}	91.7
{'147.Least_Tern'	}	66.7
{'148.Green_tailed_Towhee'	`` ``	83.3
{'149.Brown Thrasher') l	75
) }	
{'150.Sage_Thrasher'	}	91.7
{'151.Black_capped_Vireo'	}	90
{'152.Blue_headed_Vireo'	}	91.7
{'153.Philadelphia_Vireo'	}	66.7
{'154.Red_eyed_Vireo'	}	83.3
{'155.Warbling_Vireo'	}	83.3
{'156.White_eyed_Vireo'	}	75
{'157.Yellow_throated_Vireo'	`^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^ ^	33.3
{'158.Bay_breasted_Warbler'	, }	100
{'159.Black_and_white_Warbler'	}	100
{'160.Black_throated_Blue_Warbler') \	91.7
{'161.Blue_winged_Warbler') l	83.3
) }	
{'162.Canada_Warbler'	}	83.3
{'163.Cape_May_Warbler'	}	91.7
{'164.Cerulean_Warbler'	}	83.3
{'165.Chestnut_sided_Warbler'	}	75
{'166.Golden_winged_Warbler'	}	83.3
{'167.Hooded_Warbler'	}	75
{'168.Kentucky_Warbler'	}	100
{'169.Magnolia_Warbler'	}	50
{'170.Mourning_Warbler'	į.	83.3
{'171.Myrtle_Warbler'	}	75
{'172.Nashville_Warbler'	}	91.7
{'173.Orange_crowned_Warbler') }	58.3
{'174.Palm_Warbler'	J	75
) }	
{'175.Pine_Warbler'	}	91.7
{'176.Prairie_Warbler'	}	91.7
{'177.Prothonotary_Warbler'	}	91.7
{'178.Swainson_Warbler'	}	100
{'179.Tennessee_Warbler'	}	75
{'180.Wilson_Warbler'	}	83.3
{'181.Worm_eating_Warbler'	}	91.7
{'182.Yellow_Warbler'	}	100
{'183.Northern_Waterthrush'	į.	91.7
{'184.Louisiana_Waterthrush'	<pre>} } } } } } } } } } } </pre>	41.7
{'185.Bohemian_Waxwing'	ì	83.3
{'186.Cedar_Waxwing'	, }	100
{'187.American_Three_toed_Woodpeck	er'}	
	-	80
{'188.Pileated_Woodpecker'	}	91.7
{'189.Red_bellied_Woodpecker'	}	100

{'190.Red_cockaded_Woodpecker'	}	100
{'191.Red_headed_Woodpecker'	}	91.7
{'192.Downy_Woodpecker'	}	100
{'193.Bewick_Wren'	}	66.7
{'194.Cactus_Wren'	}	91.7
{'195.Carolina_Wren'	}	100
{'196.House_Wren'	}	83.3
{'197.Marsh_Wren'	}	50
{'198.Rock_Wren'	}	83.3
{'199.Winter_Wren'	}	50
{'200.Common_Yellowthroat'	}	100