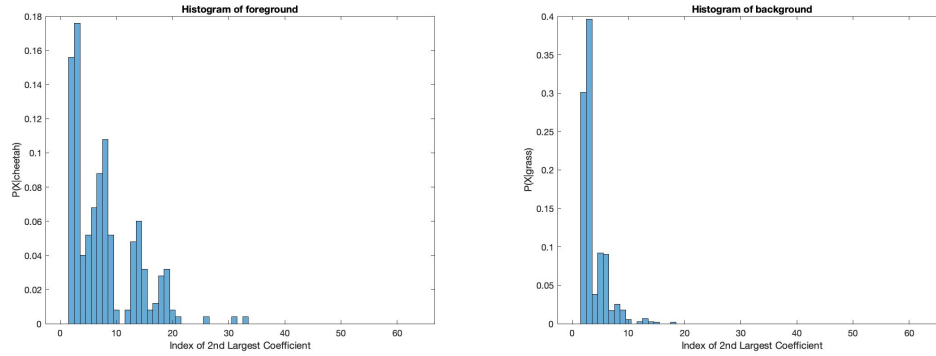


Problem 5

1. The training data consists of 250 examples of cheetah and 1053 examples of grass. The prior probability can be calculated as follows:

$$P(Y|cheetah) = \frac{250}{250+1053} = 0.1918 \quad P(Y|grass) = \frac{1053}{250+1053} = 0.8081$$

2. In order to plot the index histograms, I sorted the rows of the training samples and store the index of the second largest energy value in each row. After which, I plotted the histogram of the normalised indices for foreground and background training samples. The plot on the left indicates the class-conditional of $P_{X|Y}(x|cheetah)$ and the right indicates $P_{X|Y}(x|grass)$ respectively.



3. The mask was calculated as follows: First, I padded the image using MATLAB's padarray function, using replicate along the last array dimension (post). Then, using a sliding window approach, I computed the DCT of a 8x8 block along the row and stored the index of the 2nd largest coefficient as listed by the provided zig-zag pattern. Post this, using the prior probabilities and Bayes Decision Rule, each pixel was classified into Foreground (1) or Background (0) classes respectively. The resulting mask A is shown below.

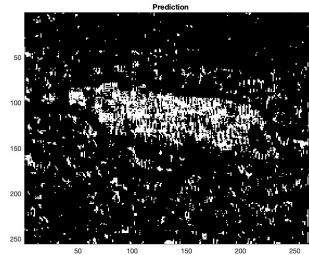


Figure 1: Calculated mask

4. Using the mask generated above, the probability of error can be calculated by summing the product of individual class error, computed by comparing mask A with *cheetah_mask.bmp*, and prior probability. Using this formula the probability of error is

$$\begin{aligned} P(\text{Error}) &= P_{X|Y}(\text{grass}|\text{cheetah}) * P_Y(\text{Cheetah}) + P_{X|Y}(\text{cheetah}|\text{grass}) * P_Y(\text{grass}) \\ &= 0.6760 * 0.1919 + 0.0559 * 0.8081 \\ &= 0.1749 \end{aligned}$$

MATLAB code

```
1 clear
2 clc
3
4 trainSample = load('TrainingSamplesDCT_8.mat');
5 fgSamples = trainSample.TrainsampleDCT_FG;
6 bgSamples = trainSample.TrainsampleDCT_BG;
7
8 fgSamplesDim = size(fgSamples);
9 bgSamplesDim = size(bgSamples);
10
11 % fgSamples has 250 training examples of 64 features each
12 % bgSamples has 1053 training examples of 64 features each
13
14 priorYCheetah = fgSamplesDim(1) / (fgSamplesDim(1) +
    bgSamplesDim(1));
15 priorYGrass = bgSamplesDim(1) / (fgSamplesDim(1) +
    bgSamplesDim(1));
16
17 disp('prior probability of Cheetah');
18 disp(priorYCheetah);
19 disp('prior probability of Grass');
20 disp(priorYGrass);
21
22 % histogram plotting for CCD
23
24 fgScalar = zeros(fgSamplesDim(1), 1);
25 bgScalar = zeros(bgSamplesDim(1), 1);
```

```
26
27 for i = 1:fgSamplesDim(1)
28     [value, position] = sort(abs(fgSamples(i,:)), 'descend');
29     fgScalar(i) = position(2); % take position of the second
        highest
30 end
31
32 for i = 1:bgSamplesDim(1)
33     [value, position] = sort(abs(bgSamples(i,:)), 'descend');
34     bgScalar(i) = position(2); % take position of the second
        highest
35 end
36
37 binRange = 0.5 : 1 : 63.5;
38
39 fgCount = histcounts(fgScalar, binRange);
40 bgCount = histcounts(bgScalar, binRange);
41 fgProb = fgCount / sum(fgCount); % normalised CCD
42 bgProb = bgCount / sum(bgCount); % normalised CCD
43
44 figure;
45 h2 = histogram('BinCounts', fgProb, 'BinEdges', binRange);
46 xlabel('Index of 2nd Largest Coefficient')
47 ylabel('P(X|cheetah)')
48 title('Histogram of foreground')
49 figure;
50 h3 = histogram('BinCounts', bgProb, 'BinEdges', binRange);
51 xlabel('Index of 2nd Largest Coefficient')
52 ylabel('P(X|grass)')
53 title('Histogram of background')
54
55 original_Image = imread('cheetah.bmp');
56 pad_Image = padarray(original_Image, [7 7], 'replicate', 'post
    ');
57 imageModified = im2double(pad_Image);
58 [image_row, image_col] = size(imageModified);
59
60 % create feature vector
61 zigzagPattern = load('Zig-Zag Pattern.txt');
62 zigzagPattern = zigzagPattern + 1; % 1 indexing in MATLAB
63
64 featureVector = zeros(image_row - 7, image_col - 7);
```

```
65 for i = 1:image_row - 7
66     for j = 1:image_col - 7
67         block = imageModified(i:i+7, j: j+7);
68         dctOutput = dct2(block);
69         orderedDCTOutput(zigzagPattern(:)) = dctOutput(:);
70         [value, sortedDCTOutput] = sort(abs(orderedDCTOutput),
71             'descend');
72         % disp(sortedDCTOutput);
73         featureVector(i, j) = sortedDCTOutput(2);
74     end
75 end
76 A = zeros(image_row - 7, image_col - 7);
77 for i = 1:image_row - 7
78     for j = 1:image_col - 7
79         if fgProb(1, featureVector(i, j)) * priorYCheetah >
80             bgProb(1, featureVector(i, j)) * priorYGrass
81             A(i, j) = 1;
82         else
83             A(i, j) = 0;
84         end
85     end
86 end
87
88 figure;
89 imagesc(A);
90 title('Prediction');
91 colormap(gray(255));
92
93 groundTruth = imread('cheetah_mask.bmp');
94 groundTruthModified = im2double(groundTruth);
95
96 groundTruthFGCount = 0;
97 groundTruthBGCount = 0;
98 for i = 1 : image_row - 7
99     for j = 1 : image_col - 7
100         if groundTruthModified(i, j) == 1
101             groundTruthFGCount = groundTruthFGCount + 1;
102         else
103             groundTruthBGCount = groundTruthBGCount + 1;
104         end
105     end
106 end
```

```
105     end
106 end
107
108 errorFGCount = 0; % false negative
109 errorBGCount = 0; % false positive
110 for i = 1:image_row - 7
111     for j = 1:image_col - 7
112         if A(i,j) == 0 && groundTruthModified(i, j) == 1 % P(
            grass | cheetah)
113             errorFGCount = errorFGCount + 1;
114         elseif A(i,j) == 1 && groundTruthModified(i, j) == 0 %
            P(cheetah | grass)
115             errorBGCount = errorBGCount + 1;
116         end
117     end
118 end
119
120 fgError = errorFGCount / groundTruthFGCount;
121 bgError = errorBGCount / groundTruthBGCount;
122
123 probError = (fgError * priorYCheetah) + (bgError * priorYGrass
    );
124
125 disp('Probability of Error');
126 disp(probError);
```