Laplacian Pyramid for Image Encoding: Part 1

In this project you will use the Laplacian pyramid algorithm for image encoding/compression.

The motivation behind this application, lies in the fact that the Laplacian pyramid stores image differences at high resolutions which are quite predictable, in the sense that the variety of the values is pretty small, with most pixels being concentrated around zero. To capitalize on that, it was proposed by Burt and Adelson to encode these values in a more compact representation, by quantizing the values, requiring essentially less bits of information to represent each one of those. This is relevant to the Image Approximation of the second lab, where we were able to compress the image by cutting the high frequencies of the image and storing a low frequency representation. Here the motivation is the same, since we again attempt to compress the image, however, the approach is different, where we do not necessarily cut down the high frequencies, so details of the image can still be visible. Take a look at the result below, where we attempt to compress the image with the two techniques, for approximately the same compression ratio.







For the first part of this project, you will quantize the values of the images (the differences and the low resolution blurred version) stored in the Laplacian pyramid using uniform quantization. While for most images, we assume that the pixel values lie in the interval [0, 255], here we will assume only a small discrete set of these values are available, e.g. {0, 63, 127, 191, 255}, and each pixel will be assigned a value from this discrete set only. Since we limit ourselves in this small set, we lose some accuracy in our values, but we need much less space to represent the values of each pixel. In the above example, each pixel can take only one of the 5 values, instead of the 256 values that are originally available.

In the next script, follow the instruction to complete the function encoding, that takes as input a Laplacian pyramid and encodes it by uniformly quantizing the values of the pixels for the different images. You can find the all the relevant material for this lab by downloading the zip file Project1.zip (http://courses.edx.org/asset-v1:PennX+ROBO2x+2T2017+type@asset+block@Project1.zip).

Your Script





C Reset MATLAB Documentation (https://www.mathworks.com/help/)

```
1 % load the image we will experiment with
 2 I = imresize(double(rgb2gray(imread('lena.png'))),[256 256]);
\mathbf{4}\,|\,\% build the Laplacian pyramid of this image with 6 levels
 5 depth = 6;
 6 L = laplacianpyr(I,depth);
 8 % compute the quantization of the Laplacian pyramid
9 bins = [16,32,64,128,128,256]; % number of bins for each pyramid level
10 LC = encoding(L,bins);
11
12 function LC = encoding(L, bins)
13
      % Input:
14
      % L: the Laplacian pyramid of the input image
15
      % bins: [an array of ints, position i representing the]
16
      % number of bins used for discretization of each pyramid level
17
18
      % Output:
19
      % LC: the quantized version of the image stored in the Laplacian pyramid
```

```
20
21
      % Please follow the instructions to fill in the missing commands.
22
       depth = numel(bins);
23
       LC = cell(1,depth);
24
25
       for i = 1:depth
26
27
28
           % 1) Compute the edges of the bins we will use for discretization
29
           % (MATLAB command: linspace)
           % For level i, the linspace command will give you a row vector
30
           % with bins(i) linearly spaced points between [X1,X2].
31
           % Remember that the range [X1,X2] depends on the level of the
32
           % pyramid. The difference images (levels 1 to depth-1) are in
33
           % the range of [-128,128], while the blurred image is in the
34
35
           % range of [0,256]
36
           if i == depth % blurred image in range [0, 256]
37
               edges = linspace(0,256,bins(i));
           else % difference image in range [-128,128]
38
               edges = linspace(-128,128,bins(i));
39
40
           end
41
42
           % 2) Compute the centers that correspond to the above edges
43
           % The 1st center -> (1st edge + 2nd edge)/2
44
           % The 2nd center -> (2nd edge + 3rd edge)/2 and so on
           half_lng_int = (edges(2)-edges(1))/2;
45
46
           centers = edges + half_lng_int .* ones(1,bins(i));
47
           centers(end) = [];
48
           % 3) Discretize the values of the image at this level of the
49
           % pyramid according to edges (MATLAB command: discretize)
50
           % Hint: use 'centers' as the third argument of the discretize
51
           % command to get the value of each pixel instead of the bin index.
52
53
           LC{i} = discretize(L{i}, edges, centers);
54
55
       end
56
57
58 end
59
60
61
62
63
64
65
66
67
68 function L = laplacianpyr(I,depth)
69
70
      % Add your code from the previous step
       L = cell(1,depth);
71
72
73
      % 1) Create a Gaussian pyramid
74
      % Use the function you already created.
75
      G = gausspyr(I,depth);
76
      % 2) Create a pyramid, where each level is the corresponding level of
77
78
      % the Gaussian pyramid minus the expanded version of the next level of
79
       % the Gaussian pyramid.
      % Remember that the last level of the Laplacian pyramid is the same as
80
      % the last level of the Gaussian pyramid.
81
      for i = 1:depth
82
           if i < depth</pre>
83
               % same level of Gaussian pyramid minus the expanded version of next level
84
85
               L{i} = G{i} - expand(G{i+1});
86
           else
87
               % same level of Gaussian pyramid
               1541 - 6541.
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88
89
            end
90
        end
91
92 end
93
   function G = gausspyr(I,depth)
94
95
96
       % Add your code from the previous step
97
       G = cell(1,depth);
98
99
        % 1) Create a pyramid, where the first level is the original image
        % and every subsequent level is the reduced version of the previous level
100
        for i = 1:depth
101
            if i == 1
102
103
                G{i} = I; % original image
104
            else
                G\{i\} = reduce(G\{i-1\}); % reduced version of the previous level
105
            end
106
        end
107
108
109 end
110
111 function g = reduce(I)
112
       % Add your code from the previous step
113
114
       Gauss = fspecial('gaussian',5,1);
115
       % 2) Convolve the input image with the filter kernel (MATLAB command imfilter)
116
       % Tip: Use the default settings of imfilter
117
        I = im2double(I);
118
        im_filtered = imfilter(I,Gauss);
119
120
121
       % 3) Subsample the image by a factor of 2
122
       % i.e., keep only 1st, 3rd, 5th, .. rows and columns
        g = im_filtered(1:2:end, 1:2:end,:);
123
124
125 end
126
   function g = expand(I)
127
128
       % Add your code from the previous step
129
        I = im2double(I);
130
131
        [m,n,clr] = size(I);
132
        I_{exp} = zeros(2*m, 2*n, clr);
133
        % note: 1:2 gives odd indices
        I_{exp}(1:2:2*m, 1:2:2*n,:) = I(1:m, 1:n,:);
134
135
136
       % 2) Create a Gaussian kernel of size 5x5 and
       % standard deviation equal to 1 (MATLAB command fspecial)
137
       Gauss = fspecial('gaussian',5,1);
138
139
       % 3) Convolve the input image with the filter kernel (MATLAB command imfilter)
140
       % Tip: Use the default settings of imfilter
141
142
       % Remember to multiply the output of the filtering with a factor of 4
143
        g = 4*imfilter(I_exp,Gauss);
144
145
146
147
148
```