Julio Soldevilla EECS 504 Winter 2018 — Problem Set 1

Problem 1 Problem 1

Proof:

1. For this problem we can begin by just substituting the points into the equation $\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} =$

 $\begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$. If we do this, we end up having n matrices of the form

$$w \begin{bmatrix} x_i' \\ y_i' \\ 1 \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & 1 \end{bmatrix} \begin{bmatrix} x_i \\ y_i \\ 1 \end{bmatrix} \quad \text{for } i = 1, ..., n$$
 (1)

which we can rewrite to have the following matrices:

$$\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} x_i h_{11} + y_i h_{12} + h_{13} - w x_i' \\ x_i h_{21} + y_i h_{22} + h_{23} - w y_i' \\ x_i h_{31} + y_i h_{32} + 1 - w \end{bmatrix}$$
 for $i = 1, ..., n$ (2)

to find, setting the matrix of known output values $Y = \begin{bmatrix} x_1' \\ y_1' \\ 1 \\ x_2' \\ y_2' \\ 1 \\ \vdots \\ x_n' \\ y_n' \\ 1 \end{bmatrix}$ and finally setting the matrix

. THen, with this equations, we can reformulate the affine transformation estimation to minimize the standard least squares problem

$$\min \|Ax - y\|_2^2$$

with the matrices A, x and y given above.

However, it seems in the formulation above we could still do some simplification, since we don't really care about the value of w and we have a way of getting rid of it. Particularly, we can take equation 2 above and take the last entry. This entry tells us that $x_i h_{31} + y_i h_{32} + 1 - w = 0 \implies w = x_i h_{31} + y_i h_{32} + 1$ for i = 1, ..., n. Substituting this value of w into the rows above, we find that we have the relationships $x_i h_{11} + y_i h_{12} + h_{13} - x_i' x_i h_{31} - x_i' y_i h_{32} = -x_i'$ and $x_i h_{21} + y_i h_{22} + h_{23} - y_i' x_i h_{31} - y_i' y_i h_{32} = -y_i'$. This equations capture the same information as the formulation above but is just presented differently because we represented differently the relations represented the value w. Thus, we can rewrite this equation in matrix form again by

considering the following matrices $X' = \begin{bmatrix} h_{11} \\ h_{12} \\ h_{13} \\ h_{21} \\ h_{22} \\ h_{23} \\ h_{31} \\ h_{32} \end{bmatrix}$ which is the matrix of unkowns that we want

to find to find the homography transformation matrix, the matrix $Y' = \begin{bmatrix} x_1 \\ -y_1' \\ -x_2' \\ -y_2' \\ -x_3' \\ \vdots \\ -x' \end{bmatrix}$ and finally the

matrix
$$A' = \begin{bmatrix} x_1 & y_1 & 1 & 0 & 0 & 0 & -x'_1x_1 & -x'_1y_1 \\ 0 & 0 & 0 & x_1 & y_1 & 1 & -y'_1x_1 & -y'_1y_1 \\ x_2 & y_2 & 1 & 0 & 0 & 0 & -x'_2x_2 & -x'_2y_2 \\ 0 & 0 & 0 & x_2 & y_2 & 1 & -y'_2x_2 & -y'_2y_2 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x_n & y_n & 1 & 0 & 0 & 0 & -x'_nx_n & -x'_ny_n \\ 0 & 0 & 0 & x_n & y_n & 1 & -y'_nx_n & -y'_ny_n \end{bmatrix}$$
. With these matrices, we can again

broblem and just try to minimize

$$\min \|A'x' - y'\|_2^2$$

with the matrices given in the second reformulation.

2. In any case of the reformulations above, we need to solve the least squares problem min ||Ax $y|_{2}^{2} = \min(Ax - y)^{T}(Ax - y)$. In this case, we will just the second set of matrices we found

above. Recall that in this case the matrix x' is the matrix $\begin{bmatrix} h_{11} \\ h_{12} \\ h_{21} \\ h_{22} \\ h_{23} \\ h_{31} \end{bmatrix}$ and so when taking a partial $\begin{bmatrix} h_{11} \\ h_{12} \\ h_{22} \\ h_{23} \\ h_{31} \end{bmatrix}$

derivative with respect to x_i , we are taking with respect to the variable of entry x_i in this matrix. To solve this, let e = (Ax - y), then the problem to solve is min $e^T e$ which we can do by taking the first derivative of this expression, setting it equal to 0 and solving for it. Letting $f(x) = e^T e$ we have that $f'(x) = 2 \frac{\partial e}{\partial x_i} e = 0 \implies c_i^T (Ax - y) = 0$ for all i = 1, ..., n where c_i are the columns of matrix A and $\frac{\partial e}{\partial x_i} = c_i$ and this derivative comes just from writing out the product Ax - y and then differentiating every row and concatenating the resulting columns. Hence we have that the first derivative implies $c_i^T(Ax - y) = 0$, then concatenating for taking all the derivatives of the rows of the expression Ax - y with respect to all the x_i we have the expression $A^{T}(Ax - y) = 0 \implies A^{T}Ax - A^{T}y = 0 \implies A^{T}Ax = A^{T}y \implies x = 0$ $(A^TA)^{-1}A^Ty$, and so this is the result to the estimation problem from above.

I claim we need 8 points (counting both x_i and y_i) to obtain a unique solution, so we require 4 pairs of points. I think this is the answer because we want to determine the 8 entries h_{ij} in the transformation matrix. TO do this, we need to solve the system of equations with 8 equations given in the second formulation of the problem above. Since with 4 points this matrix becomes a square matrix, we have that the rank of the matrix is 8 with no relationships between the rows or columns, namely it would have full rank and so it would be invertible, allowing us to find a unique solution by just finding the inverse of A. Notice that this if we consider less than 3 pairs of points, then the matrix might not be full rank.

3. In the following we present the code used to solve the problem.

```
2 % W18 EECS 504 HW1p1 Homography Estimation
3 %
5 close all;
7 % Load image
8 inpath1 = 'football1.jpg';
9 inpath2 = 'football2.jpg';
im1 = imread(inpath1);
im2 = imread(inpath2);
14 % Display the yellow line in the first image
15 figure;
imshow(im1); title('football image 1');
17 hold on;
u = [1210, 1701]; v = [126, 939]; % marker 33
w = [942, 1294]; v = [138, 939];
plot(u, v, 'y', 'LineWidth', 2);
21 hold off;
24 % FILL BLANK HERE
25 % Specify the number of pairs of points you need.
26 n = 4;
27 %
28
29 % Get n correspondences
baseName = regexp(inpath1, '^\D+', 'match', 'once');
pointsPath = sprintf('%s_points%i.mat', baseName, n);
if exist (points Path, 'file')
     % Load saved points
     load (pointsPath);
34
 else
     % Get correspondences
36
     [XY1, XY2] = getCorrespondences(im1, im2, n);
     save(pointsPath, 'XY1', 'XY2');
39 end
40
```

```
42 % FILL YOUR CODE HERE
43 % Your code should estimate the homography and draw the
44 % corresponding yellow line in the second image.
46 %We first form matrix A:
47
_{48} A = zeros(3*n,9);
49
50 for i = 1:n
      %disp(i)
51
      k = 1 + 3*(i-1);
52
53
      %disp(k)
      for j = 1:9
54
          %disp(j)
55
           if mod(j,3) < 3 \&\& mod(j,3) \sim = 0
               % disp(XY1(i, mod(j,3)))
57
               A(k,j) = XY1(i, mod(j,3));
58
           end
59
           if \mod(j,3) == 0
               A(k,j) = 1;
61
               k = k+1;
62
           end
63
      end
64
65 end
67 % Now, we form the vector Y
68
_{69} Y = zeros(3*n,1);
70
  for i = 1:n
71
      k = 1 + 3*(i-1);
      %disp(i);
      for j = 1:3
74
           if mod(j,3) < 3 \&\& mod(j,3) \sim = 0
               Y(k,1) = XY2(i, mod(j,3));
76
               k = k + 1;
77
           end
78
           if \mod(j,3) == 0
               Y(k,1) = 1;
80
               k = k+1;
81
           end
82
      end
83
84 end
85
86 %Computing vector X (vector where every entry is one of the entries of
 %transformation matrix
Ainv = inv(A'*A);
90 H = Ainv * A' * Y;
92 %The last step is to make the entries in H form the transformation matrix
```

```
93
94 Hmatrix = vec2mat(H,3);
95
  Now we take the points u and v from the original yellow line and apply
  %transformation H to them.
  point1 = [u(1), v(1)];
  point2 = [u(2), v(2)];
100
101
  point1_transform = Hmatrix * [point1,1]';
  point2_transform = Hmatrix * [point2,1]';
103
104
u_transform = [point1_transform(1), point2_transform(1)];
  v_transform = [point1_transform(2), point2_transform(2)];
107
  % Now, we display the yellow line in the second image
  figure;
  imshow(im2); title('football image 2');
111 hold on;
w = [1210, 1701]; v = [126, 939]; % marker 33
  %u = [942, 1294]; v = [138, 939];
113
  plot(u_transform, v_transform, 'y', 'LineWidth', 2);
115
  hold off;
116
118
```

Now, we show the image that results as output of the code above in figure 1. It is important

to notice that the points obtained for the correspondences are $XY1 = \begin{bmatrix} 1134.65 & 517.9 \\ 1134.7 & 589.9 \\ 1614.2 & 561.9 \\ 966.9 & 314.2 \end{bmatrix}$

and
$$XY2 = \begin{bmatrix} 615.2638 & 613.8468 \\ 415.4766 & 689.7660 \\ 863.000 & 629.8298 \\ 231.6723 & 426.0468 \end{bmatrix}$$
 and with these vectors we get the transformation matrix
$$H = \begin{bmatrix} 0.9147 & 0.0949 & -707.4416 \\ -0.0633 & 0.9968 & 175.5697 \\ -0 & -0 & 1 \end{bmatrix}$$
:

To finish, we also present the original image with the 33 yard line in figure 2:



Figure 1: This is the image that we get as output after running the code above using the points XY1 and XY2

Problem 2 Problem 2

Proof: The issue with the given model on vertical lines is that in this case we are trying to compute the vertical distance of the points to the line, but that is impossible in this case since the line is vertical and there is no vertical distance between the data points and the line we compute (additionally, the slope of such line is $m = \infty$ and so we cannot do the substraction required in the least squares problem). So, to avoid this issue (and any issue with horizontal lines too), first we can reparameterize the line as Ax + By + C = 0. With this reformulation, we can also reformulate the least squares problem to be $E(A, B, C) = \sum_{i=1}^{n} (Ax_i + By_i + C)^2$, so the least squares problem would be

minimize
$$\sum_{i=1}^{n} (Ax_i + By_i + C)^2$$

subject to $A^2 + B^2 = 1$

where we have the last constraint to ensure that we are not considering repeated lines, or lines that are the same but the coefficients are just multiplied by some constant, i.e. lines with coefficients A, B, C and then lines with coefficients 2A, 2B, 2C. Furthermore, we can interpret the equation $Ax_i + By_i + C = d_i$ as the distance between the lime we want estimate and the points we have, then the minimization problem is to minimize the sum of this distance with respect to all the



Figure 2: Original image with 33 yard line, no transformation.

points in the data set. Notice that this reformulation of the least squares problem solves the issue with vertical and horizontal lines, since when we have vertical lines, we just set the coefficient of Y to be 0 and we will be able to minimize the horizontal distances between the points and the line we are estimating, namely minimize the expression $\sum_{i=1}^{n} (Ax_i + C)^2$ and when we have horizontal lines, the coefficient of X will just turn out to be 0, and we will be able to minimize the horizontal distances between the points and the line we are estimating, namely minimize the expression $\sum_{i=1}^{n} (By_i + C)^2$. Thus this reformulation solves the problem with the least squares estimation of horizontal and vertical lines. Now, we can also do a matrix formulation of the minimization prob-

lem by rewriting the line equation as
$$Ax + By + C = 0 \iff \begin{bmatrix} A & B & C \end{bmatrix} \begin{bmatrix} x_1 & x_2 & \dots & x_n \\ y_1 & y_2 & \dots & y_n \\ 1 & 1 & \dots & 1 \end{bmatrix} = \begin{bmatrix} x_1 & x_2 & \dots & x_n \end{bmatrix} \begin{vmatrix} x_1 & x_2 & \dots & x_n \\ x_1 & x_2 & \dots & x_n \end{vmatrix} \begin{vmatrix} x_1 & x_2 & \dots & x_n \\ x_1 & x_2 & \dots & x_n \end{vmatrix} \begin{vmatrix} x_1 & x_2 & \dots & x_n \\ x_1 & x_2 & \dots & x_n \end{vmatrix}$$

0. Then, the minimization problem would be
$$\min_{A,B,C} \left\| \begin{bmatrix} A & B & C \end{bmatrix} \begin{bmatrix} x_1 & x_2 & \dots & x_n \\ y_1 & y_2 & \dots & y_n \\ 1 & 1 & \dots & 1 \end{bmatrix} \right\|_2^2 =$$

$$\min_{A,B,C} \left(\begin{bmatrix} A & B & C \end{bmatrix} \begin{bmatrix} x_1 & x_2 & \dots & x_n \\ y_1 & y_2 & \dots & y_n \\ 1 & 1 & \dots & 1 \end{bmatrix} \right)^T \left(\begin{bmatrix} A & B & C \end{bmatrix} \begin{bmatrix} x_1 & x_2 & \dots & x_n \\ y_1 & y_2 & \dots & y_n \\ 1 & 1 & \dots & 1 \end{bmatrix} \right).$$
 Now, in this case, if we try to find the solution of this problem by finding the first derivative and setting this equa-

tion equal to 0, letting $e = \begin{bmatrix} A & B & C \end{bmatrix} \begin{bmatrix} x_1 & x_2 & \dots & x_n \\ y_1 & y_2 & \dots & y_n \\ 1 & 1 & \dots & 1 \end{bmatrix}$ we end up having that such derivative $f'(x) = 2\frac{\partial e}{\partial x_i}e = 2A[A \quad B \quad C] \begin{bmatrix} x_1 & x_2 & \dots & x_n \\ y_1 & y_2 & \dots & y_n \\ 1 & 1 & \dots & 1 \end{bmatrix} = 0$. However, if we try to solve this (therefore getting the pseudo inverse expression) we just end up having that $\begin{bmatrix} X \\ Y \\ 1 \end{bmatrix} = 0$, which is not a solution

$$f'(x) = 2\frac{\partial e}{\partial x_i}e = 2A[A \quad B \quad C]\begin{bmatrix} x_1 & x_2 & \dots & x_n \\ y_1 & y_2 & \dots & y_n \\ 1 & 1 & \dots & 1 \end{bmatrix} = 0.$$
 However, if we try to solve this (therefore

to this problem. This issue when obtaining the solution to the least squares via the first derivative implies that we **cannot** use the pseudo-inverse to solve the minimization problem.

Problem 3 Problem 3

Proof:

- 1. I claim that both A and B will observe the reflection at point x with the same strength. First of all notice that the lambertian model of reflectance $R(x) = \rho(x)^T n(x)$ is a formula that contains information about the surface normal with n(x), about the direction of incident light source $\mathcal{L}(x)$, hence it contains no information of the strength that someone will observe the reflection and thus we can concldue that the radiance emitted from a Lambertian surface is not a function of outgoing direction, hence since the formula for the Lambertian surface has no information about the strength of the reflection, it must be that both A and B perceive the strength of the reflection at x equally.
- 2. One major drawback of Lambertian surface model is that it only captures diffuse characteristics of the surface, but not ambient lighting, specular reflections or other more complex radiometric functions. One object in the real world whose reflectance is not welll modeled by the Lambertian model is a mirror or glass.

Problem 4 Problem 4

Here we present the code of the function mydemosaic.m **Proof:**

```
function I = mydemosaic(I_gray)
4 % W18 EECS 504 HW1p4 Bayer Demosaicking
6 % function mydemosaic recovers the original color image (M*N*3)
7 % from the Bayer encoded image I_gray (M*N).
```

```
11 % IMPLEMENT THE FUNCTION HERE
12
13 % Generating Bayern pattern filter
^{15} %R = [0.25, 0.5, 0.25 : 0.5, 1, 0.5: 0.25, 0.5, 0.25];
^{16} %B = [ 0.25, 0.5, 0.5; 0.5,; ];
^{17} %G = [ ; ; ];
19 % Generating the Bayer encoded image.
20 %[I, I_grey] = bayer_filter(im1);
double_I_gray = im2double(I_gray);
[M,N] = size(double_I_gray);
T = zeros(M,N,3);
24 %T(:,:,1) Red layer
^{25} %T(:,:,2) Green layer
26 %T(:,:,3) Blue layer
 for i = 2:M-1
28
      for j = 2:N-1
29
          if mod(i,2) == 0 && mod(j,2) == 0 % Green pixels on even row
30
              T(i,j,1) = (double_I_gray(i,j-1) + double_I_gray(i,j+1))/2; %Red
31
     part
              T(i,j,2) = (double_I_gray(i,j)); % Green part
              T(i,j,3) = (double_I_gray(i+1,j) + double_I_gray(i-1,j))/2; % Blue
      Part
          elseif mod(i,2) == 1 &\& mod(i,2) == 1 %Green layer on odd row
34
              T(i,j,1) = (double_I_gray(i+1,j) + double_I_gray(i-1,j))/2; %Red
35
     part
              T(i,j,2) = double_I_gray(i,j); %Green part
36
              T(i,j,3) = (double_I_gray(i,j-1) + double_I_gray(i,j+1))/2; %Blue
     part
          elseif mod(i,2) == 1 \&\& mod(j,2) == 0 \% First selecting the blue
     pixels
              T(i, j, 1) = (double_I_gray(i-1, j+1) + double_I_gray(i-1, j-1) +
39
     double_I_gray(i+1,j-1) + double_I_gray(i+1,j+1))/4; %Red layer
              T(i,j,2) = (double_I_gray(i-1,j) + double_I_gray(i+1,j) +
40
     double_I_gray(i,j-1) + double_I_gray(i,j+1))/4; %Green Layer
              T(i,j,3) = double_I_gray(i,j);%Blue layer
41
          elseif mod(i,2) == 0 \&\& mod(j,2) == 1 \% Select red pixels
              T(i,j,1) = double_I_gray(i,j); %Red Layer
43
              T(i,j,2) = (double_I_gray(i,j-1) + double_I_gray(i,j+1) +
44
     double_I_gray(i-1,j) + double_I_gray(i+1,j))/4; %Green Layer
              T(i,j,3) = (double_I_gray(i-1,j-1) + double_I_gray(i-1,j+1) +
45
     double_I_gray(i+1,j-1) + double_I_gray(i+1,j+1))/4; %Blue part
          end
      end
47
  end
49
50 % Now we correct for the borders. We divided the borders into 4 corners and
51 % 4 edges.
52
```

```
55 % Checking corners:
57 % UL corner:
  for i = [1,M]
58
       for j = [1,N]
59
           if i == 1 \&\& j == 1 \%UL corner (green corner)
               T(i,j,1) = double_I_gray(i+1,j); \% Red part
61
               T(i,j,2) = double_I_gray(i,j); % Green part
62
               T(i,j,3) = double_I_gray(i,j+1); % Blue part
63
64
               %Two possibilities for UR corner
65
               %1) for even N (blue corner)
66
           elseif i == 1 \&\& j == N \&\& \mod(N,2) == 0
67
               T(i,j,1) = double_I_gray(i+1,j-1); % Red part
68
               T(i,j,2) = (double_I_gray(i,j-1) + double_I_gray(i+1,j))/2; \%
69
      Green part
               T(i,j,3) = double_I_gray(i,j); \% Blue part
70
               %2) for odd N (green corner)
71
           elseif i == 1 \&\& j == N \&\& \mod(N,2) == 1
               T(i,j,1) = double_I_gray(i+1,j); \% Red part
73
               T(i,j,2) = double_I_gray(i,j); % Green part
74
               T(i,j,3) = double_I_gray(i,j-1); % Blue part
75
76
               %DL corner, two possibilities:
               %1) for odd M (green corner)
78
           elseif j == 1 \&\& i == M \&\& \mod(M,2) == 1
               T(i,j,1) = double_I_gray(i-1,j); \% Red part
80
               T(i,j,2) = double_I_gray(i,j); % Green part
81
               T(i,j,3) = double_I_gray(i,j+1); % Blue part
82
               %2) for even M (red corner)
83
           elseif j == 1 \&\& i == M \&\& \mod(M, 2) == 0
84
               T(i,j,1) = double_I_gray(i,j); % Red part
85
               T(i,j,2) = (double_I_gray(i-1,j) + double_I_gray(i,j+1))/2; \%
86
      Green part
               T(i,j,3) = double_I_gray(i-1,j+1); % Blue part
87
88
               %DR corner, four possibilities:
89
               %M odd and N odd (green corner)
           elseif j == N \&\& i == M \&\& \mod(N,2) == 1 \&\& \mod(M,2) == 1
91
               T(i,j,1) = double_I_gray(i-1,j); \% Red part
               T(i,j,2) = double_I_gray(i,j); % Green part
93
               T(i,j,3) = double_I_gray(i,j-1); % Blue part
94
95
               M even and N even (green corner)
           elseif j == N && i == M && mod(N,2) == 0 && mod(M,2) == 0
96
               T(i,j,1) = double_I_gray(i,j-1); \% Red part
97
               T(i,j,2) = double_I_gray(i,j); \% Green part
98
               T(i,j,3) = double_I_gray(i-1,j); % Blue part
99
               % M even and N odd (red corner)
100
           elseif j == N && i == M && mod(N,2) == 1 && mod(M,2) == 0
101
               T(i,j,1) = double_I_gray(i,j); % Red part
102
```

```
T(i,j,2) = (double_I_gray(i-1,j) + double_I_gray(i,j-1))/2; \%
103
      Green part
               T(i,j,3) = double_I_gray(i-1,j-1); % Blue part
104
               % M odd and N even (blue corner)
105
           elseif i == N \&\& i == M \&\& \mod(N,2) == 0 \&\& \mod(M,2) == 1
               T(i,j,1) = double_I_gray(i-1,j-1); \% Red part
107
               T(i,j,2) = (double_I_gray(i-1,j) + double_I_gray(i,j-1))/2; \%
108
      Green part
               T(i,j,3) = double_I_gray(i,j); % Blue part
109
           end
110
       end
111
  end
112
113
114
115
  %Edges of the picture
116
117
118 %Upper and lower edge
119
120 %Upper edge
  for i = [1,M]
121
       for j = 2:N-1
122
           if mod(j,2) == 0 \&\& i == 1 \% Blue upper pixels
123
               T(i,j,1) = (double_I_gray(i+1,j-1) + double_I_gray(i+1,j+1))/2; \%
124
      Red part
               T(i,j,2) = (double_I_gray(i,j-1) + double_I_gray(i+1,j) +
125
      double_I_gray(i,j+1))/3; % Green part
               T(i,j,3) = double_I_gray(i,j); \% Blue part
126
           elseif mod(j,2) == 1 \&\& i == 1\% Green upper pixels
127
               T(i,j,1) = (double_I_gray(i+1,j)); % Red part
128
               T(i,j,2) = double_I_gray(i,j); % Green part
129
               T(i,j,3) = (double I gray(i,j-1) + double I gray(i,j+1))/2; % Blue
130
       part
           elseif i == M &\& mod(j,2) == 0 &\& mod(M,2) == 1 % Blue lower pixels
131
      and M odd
               T(i,j,1) = (double_I_gray(i-1,j-1) + double_I_gray(i-1,j+1))/2; \%
      Red part
               T(i,j,2) = (double_I_gray(i-1,j) + double_I_gray(i,j-1) +
133
      double_I_gray(i,j+1))/3; % Green part
               T(i,j,3) = double_I_gray(i,j); \% Blue part
134
           elseif i == M \&\& mod(j,2) == 1 \&\& mod(M,2) == 1 % Green lower pixels
135
      and M odd
               T(i,j,1) = (double_I_gray(i-1,j)); % Red part
136
               T(i,j,2) = double_I_gray(i,j); % Green part
               T(i,j,3) = (double_I_gray(i,j+1) + double_I_gray(i,j-1))/2; % Blue
138
       part
           elseif i == M && mod(j,2) == 0 && mod(M,2) == 0 %Lower part green
139
      pixels and even M
               T(i,j,1) = (double_I_gray(i,j-1) + double_I_gray(i,j+1))/2; % Red
140
      part
               T(i,j,2) = double_I_gray(i,j); % Green part
141
               T(i,j,3) = (double_I_gray(i-1,j)); % Blue part
142
```

```
elseif i == M &\& mod(j,2) == 1 &\& mod(M,2) == 0 % Lower part red
143
      pixels and even M
               T(i,j,1) = double_I_gray(i,j); \% Red part
144
               T(i,j,2) = (double_I_gray(i,j-1) + double_I_gray(i,j+1) +
145
      double_I_gray(i-1,j))/3; % Green part
               T(i,j,3) = (double_I_gray(i-1,j-1) + double_I_gray(i-1,j+1))/2; \%
146
      Blue part
           end
147
      end
148
  end
149
150
  %Left and right edge
151
152
  for j = [1,N]
153
      for i = 2:M-1
154
           if j == 1 \&\& mod(i,2) == 0 \%Left edge red pixel
155
               T(i,j,1) = double_I_gray(i,j); \% Red part
156
               T(i,j,2) = (double_I_gray(i-1,j) + double_I_gray(i+1,j) +
157
      double_I_gray(i,j+1))/3; % Green part
               T(i,j,3) = (double_I_gray(i-1,j+1) + double_I_gray(i+1,j+1))/2; \%
158
      Blue part
           elseif j == 1 && mod(i,2) == 1 %Left edge green pixel
159
               T(i,j,1) = (double_I_gray(i+1,j) + double_I_gray(i-1,j))/2; \% Red
160
      part
               T(i,j,2) = double_I_gray(i,j); % Green part
161
               T(i,j,3) = double_I_gray(i,j+1); % Blue part
162
           elseif j == N \&\& mod(i,2) == 0 \&\& mod(N,2) == 1 \%Right edge, red pixel
163
       with N odd
               T(i,j,1) = double_I_gray(i,j); \% Red part
164
               T(i,j,2) = (double_I_gray(i-1,j) + double_I_gray(i+1,j) +
165
      doubl_I_gray(i, j-1))/3; % Green part
               T(i,j,3) = (double_I_gray(i-1,j-1) + double_I_gray(i+1,j-1))/2; %
166
      Blue part
           elseif j == N \&\& mod(i,2) == 1 \&\& mod(N,2) == 1 \% Right edge, green
167
      pixel with N odd
               T(i,j,1) = (double_I_gray(i+1,j) + double_I_gray(i-1,j))/2; \% Red
      part
               T(i,j,2) = double_I_gray(i,j); % Green part
169
               T(i,j,3) = double_I_gray(i,j-1); % Blue part
170
           elseif j == N \&\& mod(i,2) == 0 \&\& mod(N,2) == 0 %Right edge, green
      pixel with N even
               T(i,j,1) = double_I_gray(i,j-1); \% Red part
               T(i,j,2) = double_I_gray(i,j); % Green part
173
               T(i,j,3) = (double_I_gray(i-1,j) + double_I_gray(i+1,j))/2; % Blue
174
           elseif j == N \&\& mod(i,2) == 1 \&\& mod(N,2) == 0 \%Right edge, blue
      pixel with N even
               T(i,j,1) = (double_I_gray(i-1,j-1) + double_I_gray(i+1,j-1))/2; \%
176
      Red part
               T(i,j,2) = (double_I_gray(i-1,j) + double_I_gray(i+1,j) +
177
      double_I_gray(i, j-1))/3; % Green part
               T(i,j,3) = double_I_gray(i,j); % Blue part
178
```

```
179 end
180 end
181 end
182
183 T;
184 figure; imshow(T);
185 end
```

The original image we are considering for this problem is the following one, in figure 3:



Figure 3: Original image considered for this problem.

Then, the gray-scaled image obtained from the Bayer filter function is seen in figure 4 and in figure 5 we see the other result from the bayer filter function.

Finally, this is the image we get as output from the mydemosaic.m function we implemented, we see such image in figure 6.



Figure 4: Gray-scaled image obtained from Bayer filter function.



Figure 5: Image obtained from bayer filter (non-gray scale).



Figure 6: Output image from function mydemosaic.