EECS 442 Homework #2

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Problem 1

(a)

Suppose there exists matrix H_1 , H_2 and $H = H_1H_2$ such that $\hat{M} = MH$, $\hat{M}' = M'H$. If we can solve for H_1 and H_2 respectively, then we can prove the statement is correct. Consider:

$$H_1 = \begin{bmatrix} A^{-1} & -A^{-1}b \\ 0 & 1 \end{bmatrix} \Rightarrow MH_1 = \begin{bmatrix} I & 0 \end{bmatrix}, M'H_1 = \begin{bmatrix} A'A^{-1} & -A'A^{-1}b + b' \end{bmatrix}$$

Suppose the general form of H_2 is:

$$H_2 = \begin{bmatrix} W_{3\times3} & X_{3\times1} \\ Y_{1\times3} & Z_{1\times1} \end{bmatrix}$$

Since

$$MH_1H_2 = \begin{bmatrix} I & 0 \end{bmatrix} \quad M'H_1H_2 = \begin{bmatrix} P_{2\times3} & Q_{2\times1} \\ 0 & 1 \end{bmatrix}$$

We can have:

$$W_{3\times 3} = I \quad X_{3\times 1} = 0$$

Hence, we need to prove the following relations:

$$M'H_1H_2 = \begin{bmatrix} A'A^{-1} & -A'A^{-1}b + b' \end{bmatrix} \begin{bmatrix} I & 0 \\ Y_{1\times 3} & Z_{1\times 1} \end{bmatrix} = \begin{bmatrix} P_{2\times 3} & Q_{2\times 1} \\ 0 & 1 \end{bmatrix}$$

If we represent the elements of $M'H_1H_2$ as n_{ij} , then we can deduce the following equation:

$$\begin{bmatrix} n_{3\times 1} & n_{3\times 2} & n_{3\times 3} & n_{3\times 4} \end{bmatrix} \begin{bmatrix} I & 0 \\ Y_{1\times 3} & Z_{1\times 1} \end{bmatrix} = \begin{bmatrix} P_{2\times 3} & Q_{2\times 1} \\ 0 & 1 \end{bmatrix}$$

Since we have the following relations:

$$e_3^T \left(-A'A^{-1}b + b' \right) \neq 0$$

Hence, we know that $n_{3\times 4}\neq 0$, and we can formulate H_2 as:

$$H_2 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ -\frac{n_{31}}{n_{34}} & -\frac{n_{32}}{n_{34}} & -\frac{n_{33}}{n_{34}} & \frac{1}{n_{34}} \end{bmatrix}$$

According to the derivation above, we know if we set $H = H_1H_2$, we will transform M and M' to \hat{M} and \hat{M}' respectively.

(b)

It is obvious that

$$MX = (MH) (H^{-1}X)$$
$$M'X = (M'H) (H^{-1}X)$$

Hence, if x and x' are matched points with respect to the pair of cameras (M, M'), corresponding to a 3D point, then they are also matched points with respect to the pair of cameras (MH, M'H), corresponding to the point $H^{-1}X$, that is the fundamental matrix corresponding to two pairs are the same.

(c)

According to the conclusions drawn from (b), we know the fundamental matrix corresponding to pair (M, M') and $(\hat{M}, \hat{M'})$ are the same, which is $[b]_x A$, that is:

$$F = [b]_x A = \begin{bmatrix} 0 & -1 & b_2 \\ 1 & 0 & -b_1 \\ -b_2 & b_1 & 1 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} -a_{21} & -a_{22} & -a_{23} \\ a_{11} & a_{12} & a_{13} \\ -b_2 a_{11} + b_1 a_{21} & -b_2 a_{12} + b_1 a_{22} & -b_2 a_{13} + b_1 a_{23} \end{bmatrix}$$

Hence,

$$F = a_{12} \begin{bmatrix} -\frac{a_{21}}{a_{12}} & -\frac{a_{22}}{a_{12}} & -\frac{a_{23}}{a_{12}} \\ \frac{a_{11}}{a_{12}} & 1 & \frac{a_{13}}{a_{12}} \\ -b_2 \frac{a_{11}}{a_{12}} + b_1 \frac{a_{21}}{a_{12}} & -b_2 + b_1 \frac{a_{22}}{a_{12}} & -b_2 \frac{a_{13}}{a_{13}} + b_1 \frac{a_{23}}{a_{12}} \end{bmatrix}$$

From the final expression of F, we know the seven parameters for scaled matrix F is: $-\frac{a_{21}}{a_{12}}$, $-\frac{a_{22}}{a_{12}}$, $-\frac{a_{23}}{a_{12}}$, $\frac{a_{13}}{a_{12}}$, b_1 and b_2 .

Problem 2

According to the epipolar constraints, we have: l' = Fx. Suppose x is the intersection of line l and line k on the right image plane. Then, we have following formula:

$$x = k \times l = [k]_x l$$

Combine two equations we have derived above, we have the following relations, that is:

$$l' = F[k]_x l$$

Problem 3

3.1 Fundamental Matrix

First, we apply 8 point logrithm without normalization For data set1, we have:

$$dis_{img1} = 28.026, \quad dis_{img2} = 25.163$$





For data set2, we have:

$$dis_{img1} = 9.7014, \quad dis_{img2} = 14.568$$





Secondly, we apply 8 point logrithm with normalization: For data set 1, we have:

$$dis_{img1} = 0.8844, \quad dis_{img2} = 0.82422$$





For data set2, we have:

$$dis_{img1} = 0.8914, \quad dis_{img2} = 0.89353$$





3.2 Stereo Rectification

Stereo rectification, the results are shown below: For set1,

$$H = \begin{bmatrix} 2.1065 \times 10^{-6} & -1.647 \times 10^{-6} & -3.354 \times 10^{-6} \\ 1.2335 \times 10^{-6} & -5.9127 \times 10^{-6} & -1.717 \times 10^{-3} \\ -6.2663 \times 10^{-9} & 2.013 \times 10^{-1} & 1.0212 \times 10^{-5} \end{bmatrix}$$

$$H' = \begin{bmatrix} -9.9489 \times 10^{-1} & -1.01 \times 10^{-1} & -2.8055 \times 10^2 \\ 1.01 \times 10^{-1} & -9.9489 \times 10^{-1} & -2.884 \times 10^2 \\ 9.3144 \times 10^{-4} & 9.4559 \times 10^{-5} & 1.2627 \end{bmatrix}$$

$$err = 74.279$$

For set2,

$$H = \begin{bmatrix} -7.3225 \times 10^{-7} & 2.7156 \times 10^{-6} & 2.2629 \times 10^{-3} \\ -5.4272 \times 10^{-7} & 5.2278 \times 10^{-6} & 1.4334 \times 10^{-3} \\ 2.2287 \times 10^{-9} & -9.5275 \times 10^{-11} & -7.5068 \times 10^{-6} \end{bmatrix}$$

$$H' = \begin{bmatrix} -9.9279 \times 10^{-1} & -1.1988 \times 10^{-1} & -2.8484 \times 10^{2} \\ 1.1988 \times 10^{-1} & -9.9279 \times 10^{-1} & -2.2347 \times 10^{-2} \\ 5.8878 \times 10^{-4} & 7.1092 \times 10^{-5} & 1.1689 \end{bmatrix}$$

$$err = 58.188$$

Matlab codes are attached:

```
1 function [F,dist1,dist2] = FMatrix(pathdata1,pathdata2,pathimg1,pathimg2)
2 % EECS 442 HW2 O3a
3 % fundenmantal matrix without normalization
5 % load feature points from two images. column wise data
6 % format: dimention(3) * N
7 [X1, X2] = readTextFiles(pathdata1, pathdata2);
9 % construction of W, every row is a set of corresponding points
10 W=ones(9, size(X1,2));
11 W(1,:) = X2(1,:).*X1(1,:);
12 W(2,:) = X2(1,:).*X1(2,:);
13 W(3,:) = X2(1,:);
14 \quad W(4,:) = X2(2,:).*X1(1,:);
15 W(5,:) = X2(2,:).*X1(2,:);
16 \quad W(6,:) = X2(2,:);
17 W(7,:) = X1(1,:);
18 W(8,:) = X1(2,:);
19 % calculate SVD of A
20 W=W';
[U, D, V] = svd(W);
22 f = V(:,end); % f is the last column of V
23 F = reshape(f, 3, 3);
24 F = F';
[U, D, V] = svd(F);
26 D(:,3:end) = 0; % only use the first two columns
27 F = U * D * V';
28
29 % calculate distances
30 % epipolar lines for X1
31 F = F';
32 11 = F * X2; % 3-3 * 3-n
33 % epipolar lines for X2
34 	 12 = F' * X1;
35
36 % distance between X1 and 11
37 temp = abs(11(1,:).*X1(1,:)+11(2,:).*X1(2,:)+11(3,:));
38 dist1 = temp./(l1(1,:).^2+l1(2,:).^2).^.5;
39  dist1 = mean(dist1(:));
40 % distance between X2 and 12
41 temp = abs(12(1,:).*X2(1,:)+12(2,:).*X2(2,:)+12(3,:));
42 dist2 = temp./(12(1,:).^2+12(2,:).^2).^.5;
43 dist2 = mean(dist2(:));
45 % draw feature points and epipolar lines
46 figure;
47 img1 = imread(pathimg1);
48 imshow(img1);
49 hold on;
50 for i=1:size(X1,2)
       x = X1(1,i)-50:X1(1,i)+50;
51
52
       y = -(11(1,i).*x+11(3,i))./11(2,i);
       plot(x,y,'b');
53
54 end
55 plot(X1(1,:),X1(2,:),'r*');
56 hold off;
```

```
1 function [F, dist1, dist2] = FMatrix_normalization(pathdata1, pathdata2, pathimg1, pathimg2)
2 % EECS 442 HW2 O3a
3 % fundenmantal matrix with normalization
5 % load feature points from two images. column wise data
  % format: dimention(3) * N
7 [X1, X2] = readTextFiles(pathdata1, pathdata2);
8 % normalization
9 % centroid
10 X1Trans = [-mean(X1(1,:)); -mean(X1(2,:)); 1];
11 X2Trans = [-mean(X2(1,:)); -mean(X2(2,:)); 1];
12 % Isotropic scaling
13 X1Scale = sqrt(2)./std(X1,1,2);
14  X2Scale = sqrt(2)./std(X2,1,2);
15 % transform matrix
16 transM1 = eye(3,3);
17  transM1(:,3) = X1Trans;
18 transM2 = eye(3,3);
19 transM2(:,3) = X2Trans;
20
21 \text{ scaleM1} = \text{eye}(3,3);
22 scaleM1(1,1) = X1Scale(1);
scaleM1(2,2) = X1Scale(2);
24
25 scaleM2 = eye(3,3);
26   scaleM2(1,1) = X2Scale(1);
27 \text{ scaleM2}(2,2) = X2Scale(2);
29
30 normM1 = scaleM1 * transM1;
31 normM2 = scaleM2 * transM2;
33 normX1 = normM1 * X1;
34 \text{ normX2} = \text{normM2} * \text{X2};
36 % construction of W, every row is a set of corresponding points
37 W = ones(9, size(X1, 2));
38 W(1,:) = normX2(1,:).*normX1(1,:);
39 W(2,:) = normX2(1,:).*normX1(2,:);
40 \text{ W}(3,:) = \text{normX2}(1,:);
41 W(4,:) = normX2(2,:).*normX1(1,:);
42 W(5,:) = normX2(2,:).*normX1(2,:);
43 W(6,:) = normX2(2,:);
44 W(7,:) = normX1(1,:);
45 W(8,:) = normX1(2,:);
46 % calculate SVD of A
47 W=W';
48 [U, D, V] = svd (W);
49 f = V(:,end); % f is the last column of V
50 F = reshape(f, 3, 3);
```

```
51 F = F';
[U, D, V] = svd(F);
53 D(:,3:end) = 0; % only use the first two columns
54 F = U * D * V';
55
56 % denormalization
57 F = normM2' * F * normM1;
58 F = F';
59 % calculate distances
60 % epipolar lines for X1
61 11 = F * X2; % 3-3 * 3-n
62 % epipolar lines for X2
63 12 = F' * X1;
65 % distance between X1 and 11
66 temp = abs(11(1,:).*X1(1,:)+11(2,:).*X1(2,:)+11(3,:));
67 dist1 = temp./(11(1,:).^2+11(2,:).^2).^.5;
68 dist1 = mean(dist1(:));
69 % distance between X2 and 12
70 temp = abs(12(1,:).*X2(1,:)+12(2,:).*X2(2,:)+12(3,:));
71 dist2 = temp./(12(1,:).^2+12(2,:).^2).^.5;
72  dist2 = mean(dist2(:));
74 % draw feature points and epipolar lines
75 figure;
76 img1 = imread(pathimg1);
77 imshow(img1);
78 hold on;
79    for i=1:size(X1,2)
       x = X1(1,i)-50:X1(1,i)+50;
80
81
       y = -(11(1,i).*x+11(3,i))./11(2,i);
       plot(x,y,'b');
82
83 end
84 plot(X1(1,:),X1(2,:),'r*');
85 hold off;
86
87 figure;
88 img2 = imread(pathimg2);
89 imshow(img2);
90 hold on;
91 plot(X2(1,:),X2(2,:),'ro');
92 for i=1:size(X2,2)
       x = X2(1,i)-20:X2(1,i)+20;
       y = -(12(1,i).*x+12(3,i))./12(2,i);
94
       plot(x,y,'b');
95
96 end
97 hold off;
```

```
function [HL,HR,err] = imageRect(pathdata1,pathdata2,pathimg1,pathimg2)
% get fundamental matrix

F = FMatrix_normalization(pathdata1,pathdata2,pathimg1,pathimg2);

img1 = imread(pathimg1);
img2 = imread(pathimg2);

constant img2 = imread(pathimg2);

formula = img2 = i
```

```
16 % x0 is center of the image
17 pc = [size(img2,2)/2 ; size(img2,1)/2];
   TH=[
18
       1, 0, pc(1);
       0, 1, pc(2);
20
       0, 0, 1
21
22
       ];
23 epR_prime = TH \star epR;
24 len = sqrt(epR_prime(1)^2+epR_prime(2)^2);
25 sine = epR_prime(2) / len;
26 cose = epR_prime(1) / len;
27
   RH = [
      cose, -sine, 0;
28
       sine, cose, 0;
       0, 0, 1
30
       ];
31
32 % map e_prime to the x-axis at location [length,0,1]
33 H1=RH * TH;
34 % send epipole to infinity
35 H2 = [
36
       1 0 0;
       0 1 0;
37
       -1/len 0 1
38
39
      ];
40 % transformation for one image
41 HR = H2*H1;
42
43 % find H to align epipolar lines
44 % calculate M
45 % [e'_] xM = F
exM = pinv([0, -epR(3), epR(2); epR(3), 0, -epR(1); -epR(2), epR(1), 0]);
47 M = exM * F;
48 HO = HR \star M;
49 % find HA = I + HR * epR * a'
50 [X1, X2] = readTextFiles(pathdata1, pathdata2);
X2_hat = HR * X2;
52 X1 hat = H0 * X1;
53 X1_hat = X1_hat./repmat(X1_hat(3,:),3,1);
54 X2_hat = X2_hat./repmat(X2_hat(3,:),3,1);
55 A = X1_hat';
56  b = X2_hat(1,:)';
57 % use the method solving least square parameter
18 lsp = pinv(A) * b;
59 HA = [lsp(1), lsp(2), lsp(3); 0 1 0; 0 0 1];
60 % get H
61 HL = HA * HO;
62
63 % apply transform and find errors
64 X1H = HL*X1;
65 X1H = X1H./repmat(X1H(3,:),3,1);
66 X2H = HR*X2;
67 X2H = X2H./repmat(X2H(3,:),3,1);
68 err = sum((X1H(1:2,:)-X2H(1:2,:)).^2);
69 err = mean(sqrt(err));
```