

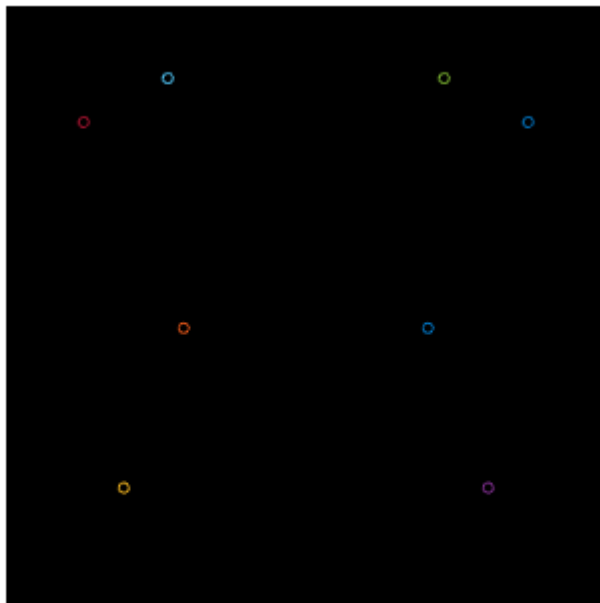
# Part I: Camera Calibration using 3D calibration object

1. Draw the image points, using small circles for each image point.

```
World_coord=[2 2 2;-2 2 2;-2 2 -2;2 2 -2; 2 -2 2;-2 -2 2;-2 -2 -2;2 -2 -2];
Image_coord=[422 323;178 323;118 483;482 483;438 73;162 73;78 117;522 117];

I=zeros(600,600);
imshow(I);
hold on;
for i=1:8
    draw_circle(Image_coord(i,1),Image_coord(i,2));
end
hold off;
snapnow;

italicP=[];
```



3. Use this Matlab function to generate 2 rows of the matrix P for each cube corner and its image and obtain a matrix with 16 rows and 12 columns. Print matrix P. italicP=[];

```
% 2. Write a Matlab function that takes as argument the homogeneous coordinates of one cube
corner and the homogeneous coordinates of its image, and returns 2 rows of the matrix P (slide 30
of the Camera Calibration pdf document).
function y = Prows(uv,xyz1)
```

```

xyz1Transpose=transpose(xyz1);
uXxyz1Transpose=-uv(1)*xyz1Transpose;
vXxyz1Transpose=-uv(2)*xyz1Transpose;
Zeros=[0 0 0 0];
y=[xyz1Transpose Zeros uXxyz1Transpose;Zeros xyz1Transpose vXxyz1Transpose];
end

```

```

for i=1:8
    capP=transpose([world_coord(i,:) 1]);
    smallp=transpose(Image_coord(i,:));
    rows=Prows(smallp,capP);
    italicP=[italicP;rows];
end
disp("P");
disp(italicP);

```

P

Columns 1 through 6

2	2	2	1	0	0
0	0	0	0	2	2
-2	2	2	1	0	0
0	0	0	0	-2	2
-2	2	-2	1	0	0
0	0	0	0	-2	2
2	2	-2	1	0	0
0	0	0	0	2	2
2	-2	2	1	0	0
0	0	0	0	2	-2
-2	-2	2	1	0	0
0	0	0	0	-2	-2
-2	-2	-2	1	0	0
0	0	0	0	-2	-2
2	-2	-2	1	0	0
0	0	0	0	2	-2

Columns 7 through 12

0	0	-844	-844	-844	-422
2	1	-646	-646	-646	-323
0	0	356	-356	-356	-178
2	1	646	-646	-646	-323
0	0	236	-236	236	-118
-2	1	966	-966	966	-483
0	0	-964	-964	964	-482
-2	1	-966	-966	966	-483
0	0	-876	876	-876	-438
2	1	-146	146	-146	-73
0	0	324	324	-324	-162
2	1	146	146	-146	-73
0	0	156	156	156	-78
-2	1	234	234	234	-117
0	0	-1044	1044	1044	-522

-2                      1                      -234                      234                      234                      -117

4. Now we need to solve the system  $Pm = 0$ . Find the singular value decomposition of matrix  $P$  using matlab `svd` function. The last column vector of  $V$  obtained by `svd(P)` should be the 12 elements in row order of the projection matrix that transformed the cube corner coordinates into their images. Print the matrix  $M$ .

```
[U,S,V]=svd(P);
Melements=transpose(V(:,end));
M=[Melements(1:4);Melements(5:8);Melements(9:12)];
disp("M");
disp(M);
```

```
M
-0.1925   -0.0283   -0.0786   -0.7346
-0.0000   -0.2044   -0.0001   -0.6120
-0.0000   -0.0001   -0.0003   -0.0024
```

5. Now we need to recover the translation vector which is a null vector of  $M$ . Find the singular value decomposition of matrix  $M = UVT$ . The 4 elements of the last column of  $V$  are the homogeneous coordinates of the position of the camera center of projection in the frame of reference of the cube (as in slide 36). Print the corresponding 3 Euclidean coordinates of the camera center in the frame of reference of the cube.

```
[U,S,V]=svd(M);
center_cam=V(:,end);
center_cam=center_cam/center_cam(end);
disp("Translation vector/camera center");
disp(center_cam(1:3));
```

```
Translation vector/camera center
-0.0000
-2.9912
-8.2695
```

6. Consider the  $3 \times 3$  matrix  $M'$  composed of the first 3 columns of matrix  $M$ . Rescale the elements of this matrix so that its element  $m_{33}$  becomes equal to 1. Print matrix  $M'$ . Now let the rotation matrices be as defined in slide 38 where the axes  $e_1, e_2, e_3$  are the  $x, y, z$  axes respectively. Therefore  $M'$  can be written as  $M_0 = KRT_z RT_y RT_x$

```
mdash=M(:,1:3);
mdash=mdash/mdash(3,3);
disp("M'");
disp(mdash);
```

M'

734.6289	107.8955	299.9999
0.0009	780.1442	0.2641
0.0000	0.3597	1.0000

7. We will perform the RQ factorization of M' in several steps. First, find a rotation matrix Rx that sets the term at position (3,2) to zero when Rx is multiplied to M'. Compute matrix N = M' \* Rx. Print Rx, theta and N.

```
cos=mdash(3,3)/sqrt((mdash(3,3)^2)+mdash(3,2)^2);
sin=-mdash(3,2)/sqrt((mdash(3,3)^2)+mdash(3,2)^2);
Rx=[1 0 0;0 cos -sin; 0 sin cos];
disp("Rx")
disp(Rx)

theta=rad2deg(atan(sin/cos));
fprintf("Theta: %f \n", theta);

N=mdash*Rx;
disp("N");
disp(N);
```

Rx

1.0000	0	0
0	0.9410	0.3384
0	-0.3384	0.9410

Theta: -19.781219

N

734.6289	-0.0000	318.8125
0.0009	734.0199	264.2723
0.0000	0	1.0627

8. The element n31 of N is small enough so that there is no need for a rotation Ry. However, element n21 is large and a rotation matrix Rz is needed to set it to zero. Compute the rotation matrix Rz. Compute the rotation angle theta\_z in degrees. This angle is actually very small.

```
cosz=mdash(2,2)/sqrt((mdash(2,2)^2)+mdash(2,1)^2);
sinz=-mdash(2,1)/sqrt((mdash(2,2)^2)+mdash(2,1)^2);
thetaz=rad2deg(atan(sinz/cosz));
fprintf("Theta: %f \n", thetaz);
Rz=[cosz -sinz 0;sinz cosz 0; 0 0 1];
```

Theta: -0.000068

9. Since we factorized out  $R_z$  we can directly compute the calibration matrix  $K$ , how? Compute  $K$  and rescale so that its element  $K_{33}$  is set to 1. Print  $K$ . What are the focal lengths of the camera in pixels? What are the pixel coordinates of the image center of the camera?

```
RxRz=transpose(Rx)*transpose(Rz);
K=mdash*inv(RxRz);
K=K/K(3,3);
disp("K");
disp(K);
disp("-----Intrinsic Parameters----- \n");
fprintf("Focal legnth----- \n");
fprintf("alpha: %f, beta: %f, gamma: %f \n", K(1,1),K(2,1),K(1,2));
fprintf("Image centers----- \n");
fprintf("u0: %f, v0: %f \n", K(1,3), K(2,3))
```

K

```
691.2796    0.0008   300.0002
 -0.0000   690.7067   248.6780
 0.0000    0.0000    1.0000
```

```
-----Intrinsic Parameters----- \n
Focal legnth-----
alpha: 691.279580, beta: -0.000000, gamma: 0.000768
Image centers-----
u0: 300.000169, v0: 248.678031
```

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## Part II: Camera Calibration using 2D calibration object

### Corner Extraction and Homography computation (10 points)

Take corner input and compute homographies

```
images=["images2.png","images9.png","images12.png","images20.png"];
max_x=270;
max_y=210;
world_coordinates=[0 max_x max_x 0; 0 0 max_y max_y;1 1 1 1];
v=[];

Hs=zeros(3,3,4);
for i=1:4
    I=imread(images(i));
    imshow(I);
    [x,y]=ginput(4);
    image_coordinates=[transpose(x);transpose(y);1 1 1 1];
    H=homography2d(world_coordinates, image_coordinates);
    fprintf("-----%s----- \n",images(i))
    disp("H:");
    disp(H);
    Hs(:,:,i)=H;
    v12=transpose(vij(1,2,H));
    v11=vij(1,1,H);
    v22=vij(2,2,H);
    v11v22=transpose(v11-v22);
    v=[v;v12;v11v22];
end
```

-----images2.png-----

H:  
-0.9679    0.0836    -53.4511  
-0.0076    -0.9024    -44.4425  
0.0000    0.0002    -0.6006

-----images9.png-----

H:  
1.0782    -0.0256    68.9334  
0.1236    0.9371    15.3774  
0.0004    -0.0001    0.5303

-----images12.png-----

H:  
0.7157    -0.0471    73.7297  
-0.1837    0.9189    60.4451  
-0.0005    -0.0002    0.6642

-----images20.png-----

H:

```
-0.8414    0.2773 -121.2006
 0.0271   -0.4126  -59.2072
 0.0001    0.0008  -0.6966
```

## Compute B

```
[b,D]=eigs(transpose(v)*v,1,'SM');
B=[b(1) b(2) b(4);b(2) b(3) b(5); b(4) b(5) b(6)];

v0=(B(1,2)*B(1,3)-B(1,1)*B(2,3))/(B(1,1)*B(2,2)-B(1,2)^2);
lambda=B(3,3)-(B(1,3)^2+v0*(B(1,2)*B(1,3)-B(1,1)*B(2,3)))/B(1,1);
alpha=sqrt(lambda/B(1,1));
beta=sqrt(lambda*B(1,1)/(B(1,1)*B(2,2)-B(1,2)^2));
gamma=-B(1,2)*(alpha^2)*beta/lambda;
u0=(gamma*v0/alpha)-(B(1,3)*(alpha^2)/lambda);

disp("B");
disp(B);

A=[alpha gamma u0;0 beta v0;0 0 1];
```

B

```
 0.0000   -0.0000   -0.0003
-0.0000    0.0000    0.0000
-0.0003    0.0000    1.0000
```

## Computing the Intrinsic and Extrinsic parameters (30 points)

### Compute R,t and R'X R for each image

```
Rs=zeros(3,3,4);
for i=1:4
    H=Hs(:,:,i);
    Ah=inv(A)*H(:,1);
    lambda=1/sqrt(transpose(Ah)*Ah);
    r1=lambda*inv(A)*H(:,1);
    r2=lambda*inv(A)*H(:,2);
    r3=cross(r1,r2);
    t=lambda*inv(A)*H(:,3);
    R=[r1 r2 r3];
    R_T=transpose(R);
    Rs(:,:,i)=R;
    fprintf("-----%s----- \n",images(i))
    disp("R:");
    disp(R);
    disp("t:");
    disp(t);
```

```
disp("R'R:");
disp(R_T*R);
end
```

-----images2.png-----

R:

-0.9998	0.1584	0.0137
-0.0089	-0.9977	0.1786
0.0153	0.1762	0.9989

t:

103.8280
-30.3413
-561.8889

R'R:

1.0000	-0.1468	0.0000
-0.1468	1.0515	0
0.0000	0	1.0299

-----images9.png-----

R:

0.9151	-0.1074	-0.3873
0.1140	0.9649	0.0720
0.3867	-0.1241	0.8953

t:

-61.6641
0.5240
462.8006

R'R:

1.0000	-0.0363	0.0000
-0.0363	0.9581	0.0000
0.0000	0.0000	0.9567

-----images12.png-----

R:

0.8558	-0.1254	0.4930
-0.1758	0.9638	0.1778
-0.4866	-0.1365	0.8028

t:

-96.2080
43.6620
590.2626

R'R:

1.0000	-0.2103	0
-0.2103	0.9633	-0.0000
0	-0.0000	0.9191

-----images20.png-----



```
R:
-0.9957    0.1476    0.0730
  0.0304   -0.5347    0.8645
  0.0880    0.8552    0.5279
```

```
t:
 67.5390
-48.7085
-727.8444
```

```
R'R:
 1.0000   -0.0880         0
-0.0880    1.0391         0
      0         0    1.0313
```

## Compute R and R' X R under constraint

```
for i=1:4
    R=RS(:, :, i);
    [U,S,V]=svd(R);
    newR=U*transpose(V);
    fprintf("-----%s----- \n",images(i))
    disp("R:");
    disp(newR)
    disp("R'R:");
    disp(transpose(newR)*newR);
end
```

-----images2.png-----

```
R:
-0.9964    0.0841    0.0135
-0.0804   -0.9811    0.1760
  0.0280    0.1743    0.9843
```

```
R'R:
 1.0000    0.0000    0.0000
 0.0000    1.0000    0.0000
 0.0000    0.0000    1.0000
```

-----images9.png-----

```
R:
 0.9136   -0.0926   -0.3960
 0.1321    0.9885    0.0736
 0.3846   -0.1196    0.9153
```

```
R'R:
 1.0000    0.0000    0.0000
 0.0000    1.0000   -0.0000
 0.0000   -0.0000    1.0000
```

-----images12.png-----

```
R:
```

0.8570	-0.0347	0.5142
-0.0716	0.9800	0.1855
-0.5104	-0.1957	0.8374

R'R:

1.0000	-0.0000	-0.0000
-0.0000	1.0000	-0.0000
-0.0000	-0.0000	1.0000

-----images20.png-----

R:

-0.9921	0.1025	0.0719
0.0075	-0.5247	0.8512
0.1250	0.8451	0.5198

R'R:

1.0000	-0.0000	0.0000
-0.0000	1.0000	0.0000
0.0000	0.0000	1.0000

## Improving accuracy

- First given the computed homographies from Section 2, compute the approximate location of each grid corner in the image. (Hint : This can be done since we know the 3d locations of the grid corners and the approximate homography. Call these points `p_approx`. Create a figure with the image and approximate grid locations. Call this Figure 1 : Projected grid corners [deliverable]

```
fprintf("-----%s----- \n",images(i))
old_Hs=Hs;
V=[];
new_Hs=zeros(3,3,4);
mean_errors=[];
old_mean_errors=[];
for i=1:4

    p_approx=[];
    H=Hs(:,:,i);
    Xs=linspace(0,max_x,10);
    Ys=linspace(0,max_y,8);
    homo_world_coordinates=[];
    I=imread(images(i));
    imshow(I);
    title("Figure 1: Projected grid corners")
    hold on;
    for x_index=1:10
        for y_index=1:8
            X2=[Xs(x_index);Ys(y_index);1];
```

```

X1=H*X2;
X1=X1/X1(3);
homo_world_coordinates=[homo_world_coordinates x2];
p_approx=[p_approx;X1(1) X1(2)];
draw_circle(X1(1),X1(2));
end
end
hold off;
snapnow;

```

**Figure 1: Projected grid corners**



**Figure 1: Projected grid corners**



**Figure 1: Projected grid corners**



**Figure 1: Projected grid corners**



- Second, using the provided Harris function detect Harris corners in the image and display them. Use the following parameter values for the Harris detection :  $\sigma = 2$ ,  $\text{thresh} = 500$ ,  $\text{radius} = 2$ . `[cim, r, c, rsubp, csubp] = harris(rgb2gray(im), sigma, thresh, radius, disp);` Here  $r$  is the y-coordinate of the Harris corner,  $c$  is the x-coordinate of the Harris corner,  $rsubp$  is the y coordinate with subpixel accuracy,  $csubp$  is the x coordinate with subpixel accuracy. Use  $rsubp$ ,  $csubp$ . Create a figure with image and overlaid Harris corners. Call this Figure 2 : Harris corners . [deliverable]

```
fprintf("-----%s----- \n",images(i))
[cim, r, c, rsubp, csubp] = harris(rgb2gray(I), 2, 500,2,0);
harris_corners=[csubp rsubp];
imshow(I);
title("Figure 2: Harris corners");
hold on;
for index=1:length(rsubp)
    draw_circle(csubp(index),rsubp(index));
end
hold off;
snapnow;
```

-----images2.png-----

**Figure 2: Harris corners**



-----images9.png-----

**Figure 2: Harris corners**



-----images12.png-----

**Figure 2: Harris corners**



-----images20.png-----

Figure 2: Harris corners



- Third, compute the closest Harris corner to each approximate grid corner. (You may find it useful to the image and `p_correct` overlaid. Call this Figure 3 : grid points . [deliverable]

```
fprintf("-----%s----- \n",images(i))
distance_matrix=dist2(harris_corners,p_approx);
[M,A]=min(distance_matrix);
p_correct=harris_corners(A,:);
imshow(I);
title("Figure 3: Grid corners");
hold on;
for index=1:length(p_correct)
    draw_circle(p_correct(index,1),p_correct(index,2));
end
hold off;

snapnow;
```

-----images2.png-----

**Figure 3: Grid corners**



-----images9.png-----

**Figure 3: Grid corners**



-----images12.png-----



**Figure 3: Grid corners**



-----images20.png-----

**Figure 3: Grid corners**



- Finally, compute a new homography from `p_correct`, print `H` [deliverable]

```
fprintf("-----%s----- \n", images(i))  
  
homo_image_coordinates=[transpose(p_correct(:,1));transpose(p_correct(:,2));ones(1,length(p_correct))];  
H=homography2d(homo_world_coordinates,homo_image_coordinates);
```

```

errors=[];
for j=1:length(p_correct)
    x2=homo_world_coordinates(:,j);
    x1=H*x2;
    x1=x1/x1(3);
    error=sqrt((x1(1)-p_correct(j,1))^2+(x1(2)-p_correct(j,2))^2);
    errors=[errors error];
end
mean_errors=[mean_errors mean(errors)];

old_errors=[];
old_H=old_Hs(:,:,i);
for j=1:length(p_correct)
    x2=homo_world_coordinates(:,j);
    x1=old_H*x2;
    x1=x1/x1(3);
    old_error=sqrt((x1(1)-p_correct(j,1))^2+(x1(2)-p_correct(j,2))^2);
    old_errors=[old_errors old_error];
end
old_mean_errors=[old_mean_errors mean(old_errors)];

disp("H:");
disp(H);

new_Hs(:,:,i)=H;
v12=transpose(vij(1,2,H));
v11=vij(1,1,H);
v22=vij(2,2,H);
v11v22=transpose(v11-v22);
V=[v;v12;v11v22];

```

-----images2.png-----

H:

0.9681	-0.0853	53.5654
0.0153	0.8942	43.0470
-0.0000	-0.0002	0.6022

-----images9.png-----

H:

1.0995	-0.0346	70.6543
0.1486	0.9483	9.4997
0.0005	-0.0001	0.5203

-----images12.png-----

H:

0.7144	-0.0565	74.9893
-0.1791	0.9014	59.7154
-0.0005	-0.0002	0.6689

-----images20.png-----

H:

```

0.8621    -0.2761   119.1720
-0.0039    0.3989    55.1261
0.0000   -0.0008    0.6757

```

end

-----images20.png-----

- Repeat this for the other three images. Then use the homographies to estimate K and R, t for each image. Report your K, R's, and t's [deliverable]. Save your results, you will need to use them in Part III

```

Hs=new_Hs;

[b,D]=eigs(transpose(V)*V,1,'SM');
B=[b(1) b(2) b(4);b(2) b(3) b(5); b(4) b(5) b(6)];

v0=(B(1,2)*B(1,3)-B(1,1)*B(2,3))/(B(1,1)*B(2,2)-B(1,2)^2);
lambda=B(3,3)-(B(1,3)^2+v0*(B(1,2)*B(1,3)-B(1,1)*B(2,3)))/B(1,1);
alpha=sqrt(lambda/B(1,1));
beta=sqrt(lambda*B(1,1)/(B(1,1)*B(2,2)-B(1,2)^2));
gamma=-B(1,2)*(alpha^2)*beta/lambda;
u0=(gamma*v0/alpha)-(B(1,3)*(alpha^2)/lambda);

A=[alpha gamma u0;0 beta v0;0 0 1];
disp("K:")
disp(A)

Rs=zeros(3,3,4);
Ts=[];
for i=1:4
    fprintf("-----%s-----\n",images(i))
    H=Hs(:,:,i);
    Ah=inv(A)*H(:,1);
    lambda=1/sqrt(transpose(Ah)*Ah);
    r1=lambda*inv(A)*H(:,1);
    r2=lambda*inv(A)*H(:,2);
    r3=cross(r1,r2);
    t=lambda*inv(A)*H(:,3);
    R=[r1 r2 r3];
    R_T=transpose(R);
    Rs(:,:,i)=R;
    disp("R:");
    disp(R);
    disp("t:");
    disp(t);
    Ts=[Ts t];
end

```

K:

838.0415	219.0149	238.9904
0	812.2935	56.5791
0	0	1.0000

-----images2.png-----

R:

0.9999	-0.2886	0.0015
0.0167	0.9686	0.1871
-0.0047	-0.1858	0.9732

t:

-96.1110  
9.5909  
522.8100

-----images9.png-----

R:

0.9011	-0.2484	-0.4062
0.1175	0.9426	-0.0057
0.4173	-0.1088	0.8786

t:

-46.1820  
-19.6612  
416.7535

-----images12.png-----

R:

0.8791	-0.2566	0.4462
-0.1523	0.9357	0.2519
-0.4516	-0.1547	0.7835

t:

-90.2868  
22.4381  
557.5697

-----images20.png-----

R:

0.9996	-0.2269	-0.0091
-0.0066	0.5377	0.8200
0.0272	-0.8265	0.5360

t:

-54.6642  
20.3374  
660.5360

- Using the new computed H, compute the errors between points in p\_correct and points you get by projecting grid corners to the image (Hint there is no need to use R, t for projecting) . Call this err\_reprojection. Report your result. [deliverable]

```
for i=1:4
    fprintf("-----%s----- \n",images(i))
    fprintf("error_reprojection: %f\n",mean_errors(i));
end
fprintf("\n")
```

```
-----images2.png-----
error_reprojection: 1.780278
-----images9.png-----
error_reprojection: 2.023190
-----images12.png-----
error_reprojection: 2.112336
-----images20.png-----
error_reprojection: 1.868357
```

- Now repeat the process using 4 images. Compare your results to your previous results and those of part 2 [deliverable].

```
for i=1:4
    fprintf("-----%s----- \n",images(i))
    fprintf("Average error before improvement: %f \n",old_mean_errors(i));
    fprintf("Average error after improvement: %f \n \n",mean_errors(i));
end
```

```
-----images2.png-----
Average error before improvement: 2.858509
Average error after improvement: 1.780278

-----images9.png-----
Average error before improvement: 5.931050
Average error after improvement: 2.023190

-----images12.png-----
Average error before improvement: 5.009999
Average error after improvement: 2.112336

-----images20.png-----
Average error before improvement: 5.645813
Average error after improvement: 1.868357
```

## Part III: Augmented Reality 101

### Augmenting an Image

```
[clipart,~,Alpha]=imread("4.png");
height=210;
width=183;
clipart_resized=imresize(clipart,[height,width]);
A_resized=imresize(Alpha,[height,width]);
for i=1:4
    fprintf("-----%s----- \n",images(i))
    H=Hs(:, :, i);
    I=imread(images(i));
    for x=1:height
        for y=1:width
            val=clipart_resized(x,y,:);
            if A_resized(x,y)~=0
                X1=[y;x;1];
                X2=H*X1;
                X2=X2/X2(3);
                I(int64(X2(2)),int64(X2(1)),:)=val;
            end
        end
    end
    imshow(I);
    snapnow;
end
```

-----images2.png-----



-----images9.png-----



-----images12.png-----



-----images20.png-----



## Augmenting an Object

```

cube=[0 210 0;90 210 0; 90 120 0;0 120 0;0 210 -90;90 210 -90;90 120 -90;0 120 -90];
edges=[1 2;1 4;2 3;3 4;5 6;5 8;6 7;7 8;1 5;2 6;3 7;4 8];
for i=1:4
    fprintf("-----%s\n",images(i))
    I=imread(images(i));
    R=RS(:,:,i);
    T=Ts(:,i);
    imshow(I)
    hold on
    points=[];
    for point_no=1:8
        point=cube(point_no,:);
        x1=[transpose(point);1];
        x2=A*[R T]*x1;
        x2=x2/x2(3);
        points=[points [x2(1);x2(2)]];
    end
    for edge_no=1:12

        edge=edges(edge_no,:);
        x=[points(1,edge(1)),points(1,edge(2))];
        y=[points(2,edge(1)),points(2,edge(2))];
        p=plot(X,Y,'-or');
        p.Linewidth =3;
    end
    snapnow;
end

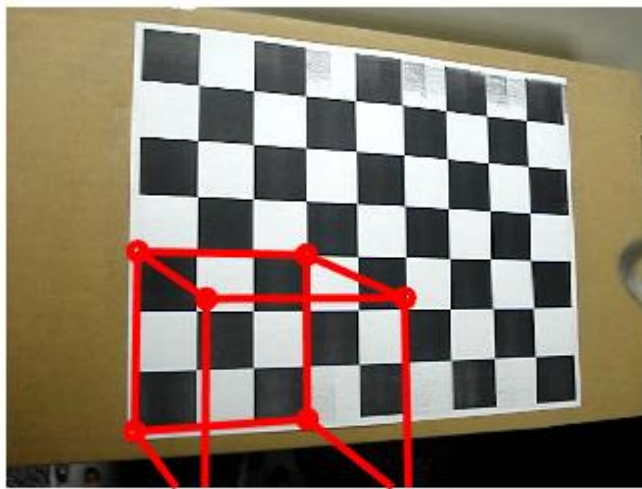
```

-----images2.png-----

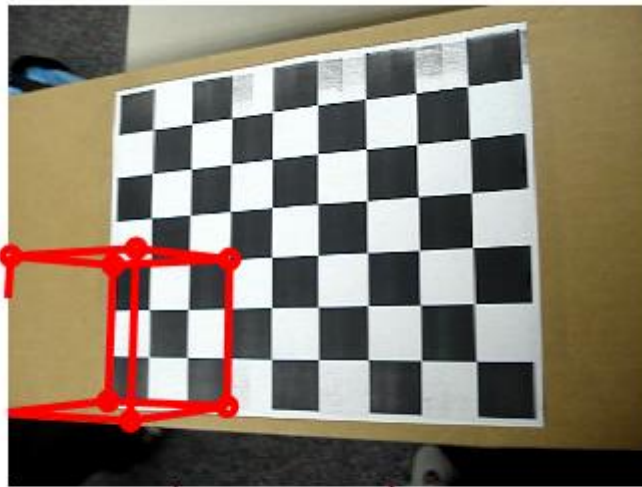




-----images9.png-----



-----images12.png-----



-----images20.png-----



Helper function to calculate V vector

```
function v=vij(i,j,H)
    t1=H(1,i)*H(1,j);
    t2=H(1,i)*H(2,j)+H(2,i)*H(1,j);
    t3=H(2,i)*H(2,j);
    t4=H(3,i)*H(1,j)+H(1,i)*H(3,j);
    t5=H(3,i)*H(2,j)+H(2,i)*H(3,j);
    t6=H(3,i)*H(3,j);
```

```
v=[t1;t2;t3;t4;t5;t6];  
end
```

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### *Extra credit*

2. If only 2 images are available, we can impose the skewless constraint  $\gamma = 0$ .