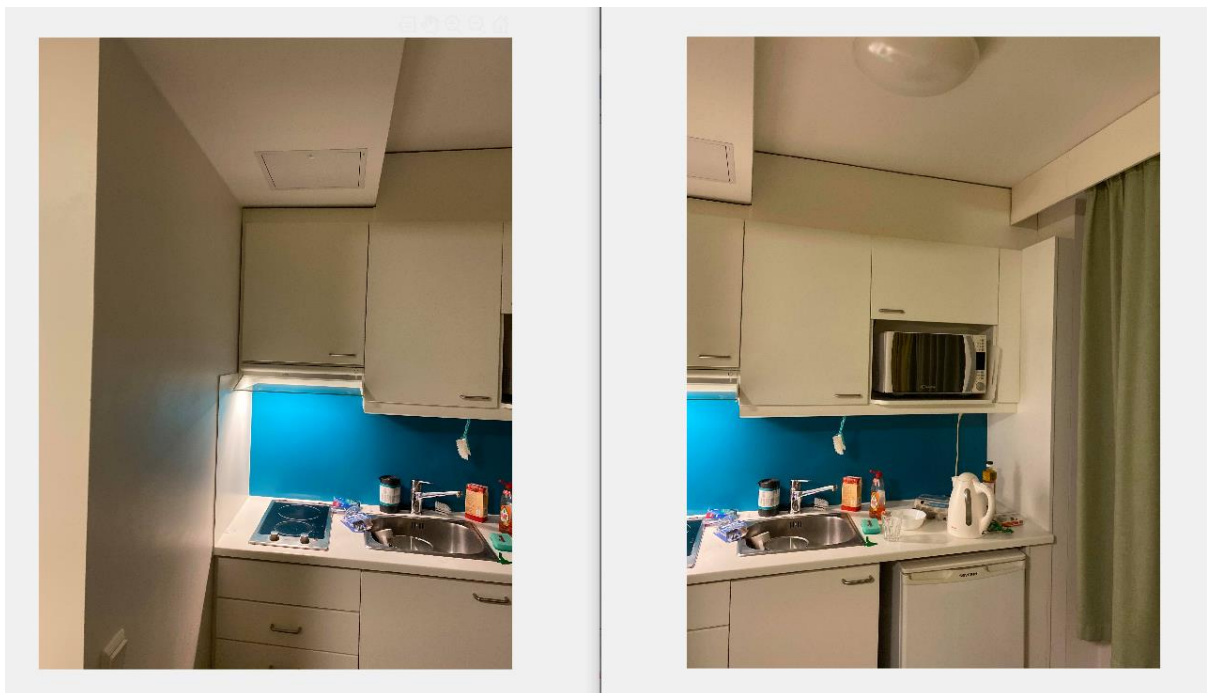


Normalised DLT

Loading the image

```
kitchen1 = imread('kitchen2.jpg');  
figure,imshow(kitchen1)  
  
kitchen2 = imread('kitchen1.jpg');  
figure,imshow(kitchen2)
```



Grayscale and finding the points that match with each other from the two images.

```
graykitchen1 = rgb2gray(kitchen1);  
graykitchen2 = rgb2gray(kitchen2);  
points1 = detectSURFFeatures(graykitchen1)  
points2 = detectSURFFeatures(graykitchen2)  
  
[features1,valid_points1] = extractFeatures(graykitchen1,points1);  
[features2,valid_points2] = extractFeatures(graykitchen2,points2);  
indexPairs = matchFeatures(features1,features2,'MaxRatio',0.1);  
  
%disp(indexPairs)  
matchedPoints1 = valid_points1(indexPairs(:,1),:);
```

```
%disp(matchedPoints1)
```

```
matchedPoints2 = valid_points2(indexPairs(:,2),:);
```

```
%disp(matchedPoints2)
```

To get the minimum amount of points needed to do the homography (four points), I randomly selected 4 points from the first image, then I first transposed the points and normalized them with the algorithm given, obtaining normalisedFourPoints at the end.

```
fourPoints1 =
```

```
3×4 single matrix
```

```
1.0e+03 *
```

2.4353	2.7005	2.8464	2.3790
2.9615	3.0575	2.8822	2.8540
0.0010	0.0010	0.0010	0.0010

```
normalisedFourPoints =
```

-0.2711	0.1928	0.4479	-0.3696
0.0397	0.2075	-0.0990	-0.1483
1.0000	1.0000	1.0000	1.0000

```
fourPoints1 = transpose(matchedPoints1([1 3 7 14],:).Location)
```

```
fourPoints1(3,:) = 1;
```

```
average = mean(fourPoints1,2)
```

```
d1 = 0;
```

```
for i = 1:size(fourPoints1,2)
```

```
    cal = sqrt((fourPoints1(1,i)-average(1,1)).^2+(fourPoints1(2,i)-average(2,1)).^2);
```

```
    d1 = d1 + cal;
```

```
end
```

```
T1 = [sqrt(2)/d1 0 -sqrt(2)*average(1,1)/d1; 0 sqrt(2)/d1 -sqrt(2)*average(2,1)/d1; 0 0 1]
```

```
normalisedFourPoints = zeros(size(fourPoints1))
```

```
for i = 1:size(fourPoints1,2)
```

```
    temp = fourPoints1(:,i)
```

```
    temp = T1*temp
```

```
    normalisedFourPoints(:,i) = temp
```

```
end
```

Doing the same thing to the **SAME** four points matched in the second image:

```
fourPoints2 = transpose(matchedPoints2([1 3 7 14],:).Location)

fourPoints2(3,:) = 1;

average2 = mean(fourPoints2,2)

d2 = 0;

for i = 1:size(fourPoints2,2)

    cal = sqrt((fourPoints2(1,i)-average2(1,1)).^2+(fourPoints2(2,i)-average2(2,1)).^2);

    d2 = d2 + cal;

end

T2 = [sqrt(2)/d2 0 -sqrt(2)*average2(1,1)/d2; 0 sqrt(2)/d2 -sqrt(2)*average2(2,1)/d2; 0 0 1]

normalisedFourPoints2 = zeros(size(fourPoints2))

for i = 1:size(fourPoints2,2)

    temp = fourPoints2(:,i)

    temp = T2*temp

    normalisedFourPoints2(:,i) = temp

end
```

After obtaining the normalisedPoints, I used the following function that I implemented called calchomography to find the homography of the points.

% Estimate the homography from a set of point correspondences.

```
function H = calchomography(points1, points2)
```

```
if size(points1, 2) < 4 || size(points2, 2) < 4
```

```
    error('no 4 points passed');
```

```
end
```

```
A = [];
```

```
for point = 1:size(points1, 2)
```

```
    x = points1(1, point);
```

```
    y = points1(2, point);
```

```
    u = points2(1, point);
```

```

v = points2(2, point);

A = [A; ...
    -x -y -1 0 0 0 x*u y*u u; ...
    0 0 0 -x -y -1 x*v y*v v];

end

% Solve A * h = 0 for h.

[U, S, V] = svd(A);
h = V(:,end);

% Reshape vectorised result back into matrix shape.
H = reshape(h, 3, 3);

% Homogeneous normalisation.
H = H ./ H(3,3);

```

Calling the function above with H and denormalise it, then saving it as mymatrix.mat:

```

H = calchomography(normalisedFourPoints,normalisedFourPoints2);
H = inv(T2)*H*T1 %Denormalising
disp(H)
save ('mymatrix.mat', 'H');

```

To show that the homography we obtained indeed works, I have created a backwardWarping algorithm that allows me to warp the first image to the second image. I didn't include the codes in **week3matlab.m** but it is at **imageAlignment.m** since backward mapping isn't necessary, the result is below, notice the sinks in both images, they line up perfectly.

```

source = im2double(imread('kitchen2.jpg'));
source2 = im2double(imread('kitchen1.jpg'));
newLeft = zeros(size(source));

% The backward warping transformation (rotation + scale about an arbitrary point).
Matrix = load('mymatrix.mat');
M = Matrix.H;
for y = 1:size(newLeft, 1)
    for x = 1:size(newLeft, 2)

```

```

% Transform source pixel location (round to pixel grid).

p = [x; y; 1];
q = inv(M) * p;
u = round(q(1) / q(3));
v = round(q(2) / q(3));

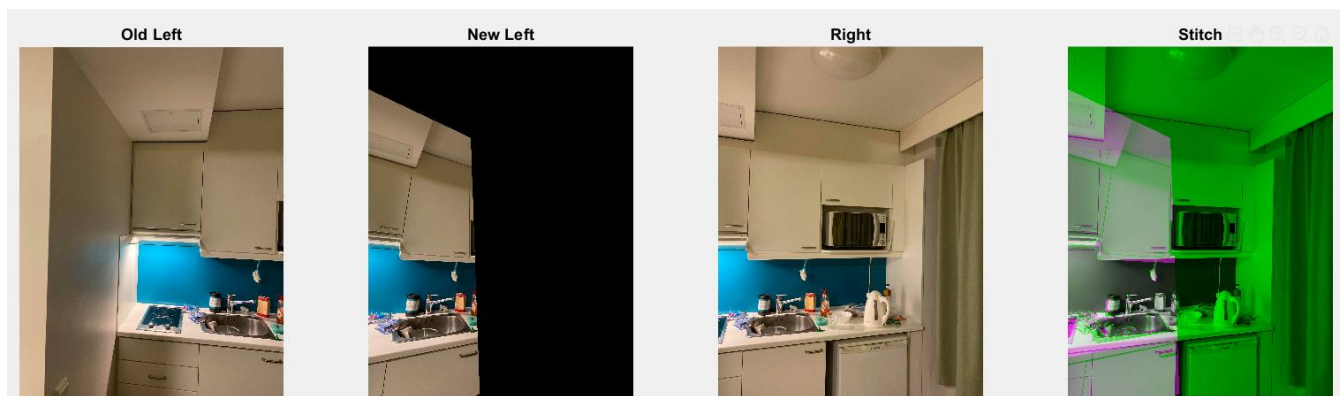
% Check if target pixel falls inside the image domain.
if (u > 0 && v > 0 && u <= size(source, 2) && v <= size(source, 1))
    % Sample the target pixel colour from the source pixel.
    newLeft(y, x, :) = source(v, u, :);
end
end
end

target = imfuse(source2,newLeft);

%overlay
subplot(141); imshow(source); title('Old Left');
subplot(142); imshow(newLeft); title('New Left');
subplot(143); imshow(source2); title('Right');
subplot(144); imshow(target); title('Stitch');

```

Results:



HoughLines detector

Reading and detecting the edge with matlab canny edge detector.

```
corridor = imread('Corridor1.jpg');  
grayCorridor = rgb2gray(corridor);  
Canny_img = edge(grayCorridor, 'Canny');  
[rows, cols] = size(Canny_img);
```

Initialising the hough space (Accumulator) to the size of theta_range and rho_range.

```
theta_maximum = 90;  
rho_maximum = hypot(rows,cols) - 1;  
theta_range = -theta_maximum:theta_maximum-1;  
rho_range = -rho_maximum:rho_maximum;  
Hough = zeros(length(rho_range), length(theta_range));
```

Converting the edge found with $r = x \cdot \cos(\theta) + y \cdot \sin(\theta)$.

```
for j = 1:rows  
    for i = 1:cols  
        if Canny_img(j, i) == 1  
            x = i - 1;  
            y = j - 1;  
            for T = theta_range  
                R = round((x * cosd(T)) + (y * sind(T)));  
                R_Index = R + rho_maximum + 1;  
                T_Index = T + theta_maximum + 1;  
                Hough(R_Index, T_Index) = Hough(R_Index, T_Index) + 1;  
            end  
        end  
    end  
end
```

Hough peak with thresholding.

```
[X,Y] = find(Hough>0.45*max(Hough(:)))
```

```
imshow(Hough,[],'XData',theta_range,'YData',rho_range,...
```

```
    'InitialMagnification','fit');
```

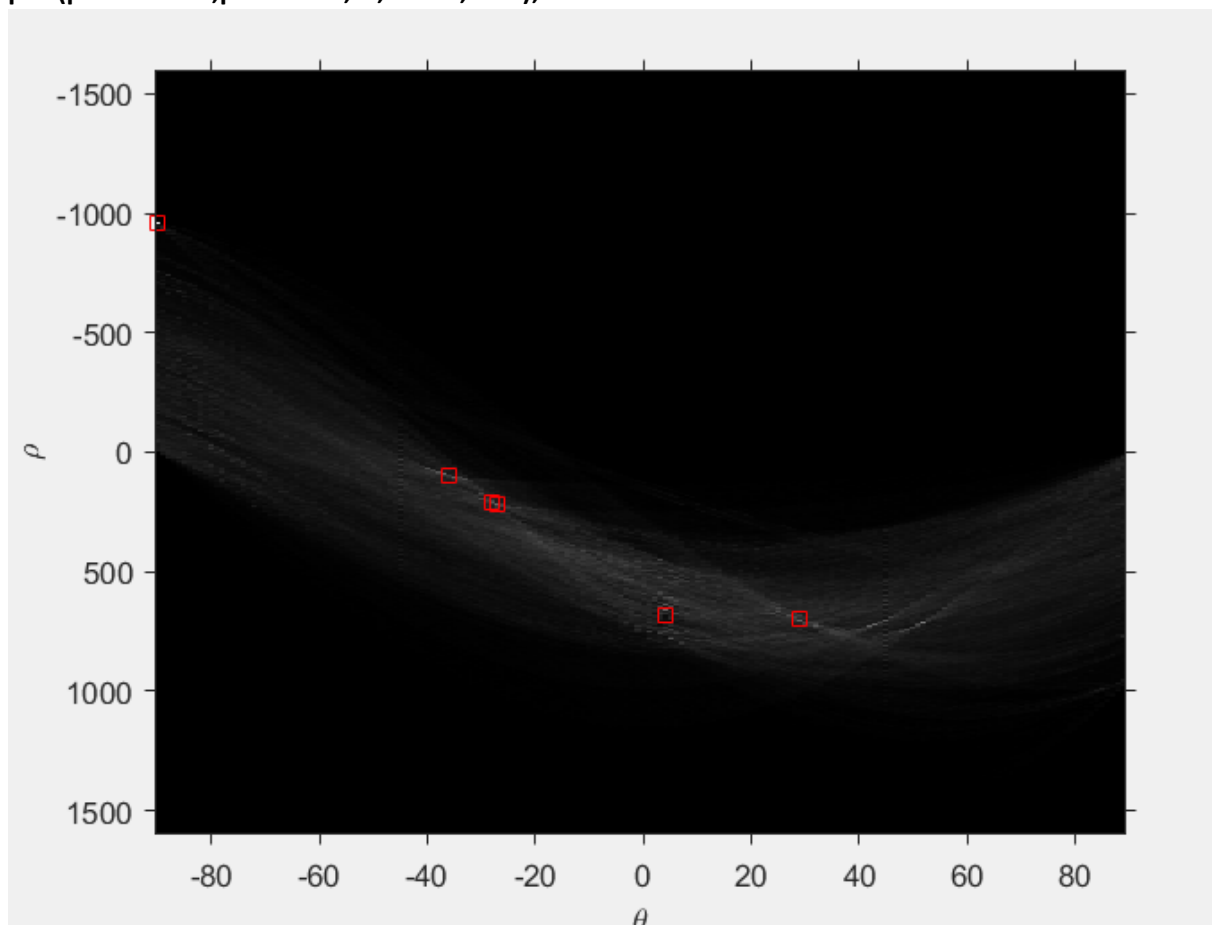
```
xlabel('\theta'), ylabel('\rho');
```

```
axis on, axis normal, hold on;
```

```
pointsRho = rho_range(X(:));
```

```
pointsTheta = theta_range(Y(:));
```

```
plot(pointsTheta,pointsRho,'s','color','red');
```



Here is my attempt at dehoughing the peaks to image space. I first get the Rho and Theta pair into a list. I then created a set of binary matrices of the size imageRow x imageColumn x number of rhotheta pair. Then I substitute all the x and y pixel coordinates into the function $x\cos(\theta) + y\sin(\theta) = \rho$, and if left - right == 0, I would mark the location of the binary matrix(x,y) as 1 and I run this for all the theta pairs against all the x and y value and store them separately in the set of binary matrices mentioned above.

In the second for loop, I just went through all of the set of binary matrices I had obtained and just aggregate them into one binary matrix.

```

for k = 1:length(pointsRho)
    for i = 1:1280
        for j = 1:960
            r = pointsRho(k);
            t = pointsTheta(k);
            threshold = abs(i*cosd(t)+j*sind(t)-r);
            if threshold < 0.5
                allLines(j,i,k) = 1;
            end
        end
    end
end

aggregatedImage = zeros(size(Canny_img));

for k = 1:length(pointsRho)
    for j = 1:1280
        for i = 1:960
            if allLines(i,j,k) == 1
                aggregatedImage(i,j) = 1;
            end
        end
    end
end
end

```



```
figure,imshow(imfuse(aggregatedImage,corridor))
```

With $\text{houghpeaks} = 0.45 * \max(\text{Hough}(:))$ we get the following lines

