

Image Stitching

Image Processing Lecture 2

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Here are three images captured simply by rotating the camera.



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Image Stitching

The process of combining multiple images into a single image.



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Homography

Recall that a Homography is a square matrix applied to homogeneous coordinates.

We are now going to use Homographies to stitch images together.

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Theorem

Two images captured by a rotated camera are related by a Homography.

$$I_2(x) = I_1(Hx)$$

I_1, I_2 are the first and second image.

H is the Homography

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Theorem

Two images of a flat plane are related by a Homography.

$$I_2(x) = I_1(Hx)$$

I_1, I_2 are the first and second image.

H is the Homography

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Big Question:

How do we find the Homography that relates two images together?

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Picking Matching Points

We will use matched points in the two images.



At first, we will match points by hand. Later, we'll see how to have the computer match points.

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We know all the x s and the y s. We have to find the H .

$$\begin{aligned}y_1 &= Hx_1 \\y_2 &= Hx_2 \\&\vdots \\y_N &= Hx_N\end{aligned}$$

Note: the equality above is in the Homogeneous sense. That is, the vector y_i is a scalar multiple of Hx_i .

Since the points are all in a two dimensional projective space, they are all three element vectors.

Since they are equal in a projective sense, the components of the vectors are scalar multiples of each other.

If we think of a pair of matching three element vectors as 3D Euclidean vectors, then they would be parallel.

This implies the following:

$$y_i \times (Hx_i) = 0$$

Recall that we can rewrite $H x_i$ as the following

$$H x_i = \begin{bmatrix} r_1^T x_i \\ r_2^T x_i \\ r_3^T x_i \end{bmatrix}$$

where r_i^T is the i^{th} row of H . Let $x'_i = H x_i$. Recall that the cross product can be rewritten as

$$y \times x'_i = \begin{bmatrix} y_{i,2} x'_{i,3} - x'_{i,2} y_{i,3} \\ x'_{i,1} y_{i,3} - x'_{i,3} y_{i,1} \\ x'_{i,1} y_{i,2} - y_{i,2} x'_{i,1} \end{bmatrix}$$

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Combining the previous equations

$$y_i \times H x_i = \begin{bmatrix} y_{i,2} (r_3^T x_i) - (r_2^T x_i) y_{i,3} \\ (r_1^T x_i) y_{i,3} - (r_3^T x_i) y_{i,1} \\ (r_1^T x_i) y_{i,2} - y_{i,2} (r_1^T x_i) \end{bmatrix} = 0$$

We're going to use a neat trick. We're going to isolate the rows of H by rewriting the above equation as

$$\begin{bmatrix} 0^T & -y_{i,3} x_i^T & y_{i,2} x_i^T \\ y_{i,3} x_i^T & 0^T & -y_{i,1} x_i^T \\ -y_{i,2} x_i^T & y_{i,1} x_i^T & 0^T \end{bmatrix} \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix} = 0$$

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$$\begin{bmatrix} 0^T & -y_{i,3} x_i^T & y_{i,2} x_i^T \\ y_{i,3} x_i^T & 0^T & -y_{i,1} x_i^T \\ -y_{i,2} x_i^T & y_{i,1} x_i^T & 0^T \end{bmatrix} \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix} = 0$$

This is a system of equations! Each pair of matched points gives us 3 equations with 9 unknowns.

Unfortunately, one of the equations is redundant. (The third equation does not provide any more information.) So we only have two equations with 9 unknowns.

$$\begin{bmatrix} 0^T & -y_{i,3} x_i^T & y_{i,2} x_i^T \\ y_{i,3} x_i^T & 0^T & -y_{i,1} x_i^T \end{bmatrix} \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix} = 0$$

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$$\begin{bmatrix} 0^T & -y_{i,3} x_i^T & y_{i,2} x_i^T \\ y_{i,3} x_i^T & 0^T & -y_{i,1} x_i^T \end{bmatrix} \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix} = 0$$

We'll rewrite this as $A_i h = 0$

where $h = \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix}$

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We can combine the equations from all matched points into a single system

$$A h = \begin{bmatrix} A_1 \\ A_2 \\ \vdots \\ A_N \end{bmatrix} h = 0$$

We can find h , and therefore H by finding a non-zero vector in the null space of A !

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Finding a Vector in the Null Space

Compute the SVD of the matrix A :

$$[u,s,v] = \text{svd}(A);$$

Find the smallest singular value:

$$[\text{minS}, \text{minIdx}] = \min(\text{diag}(s));$$

The null vector is the column of v corresponding to the minimum singular value:

$$\text{nullVector} = v(:, \text{minIdx});$$

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Question:

**Now that we have the
Homography, how do we stitch the
images together?**

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Finding the Range

Image 1

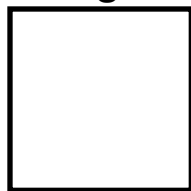
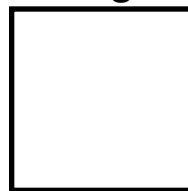


Image 2



Projected 1

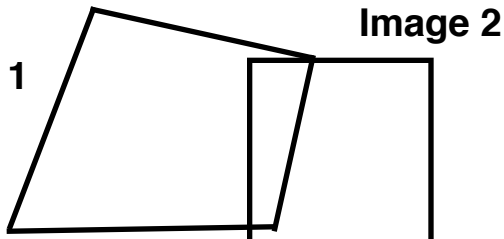
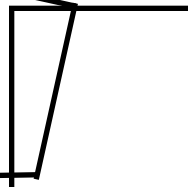
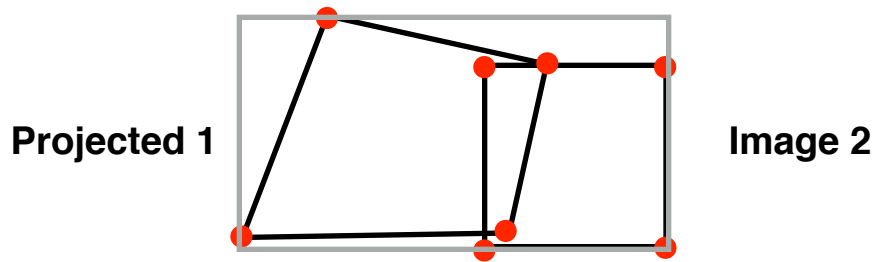


Image 2



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The min and max coordinates of the red dots tell you how big to make your stitched image.

They tell you the range of the projective space!

Note: You'll have to properly place Image 2. If it comes out as shown above, you're going to have to shift Image 2 to the right.

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Now that you have the range, pull Image 1 into the projected space.

Iterate over the pixels in the projected space.

For each pixel, apply H^{-1} to see which pixel in Image 1 belongs at that location.

Populate the value in the stitched image with the intensity from Image 1.

$$I_P(x) = I_1(H^{-1} x)$$

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Does this picture make more sense now?

