

An aerial photograph of the Colorado School of Mines campus. In the foreground, there are several large, modern buildings with light-colored facades and flat roofs. A large green field, likely a sports field, is visible in the middle ground. In the background, there are rolling hills with sparse vegetation and some residential houses. The sky is blue with a few wispy clouds.

Colorado School of Mines

Image and Multidimensional Signal Processing

Professor William Hoff

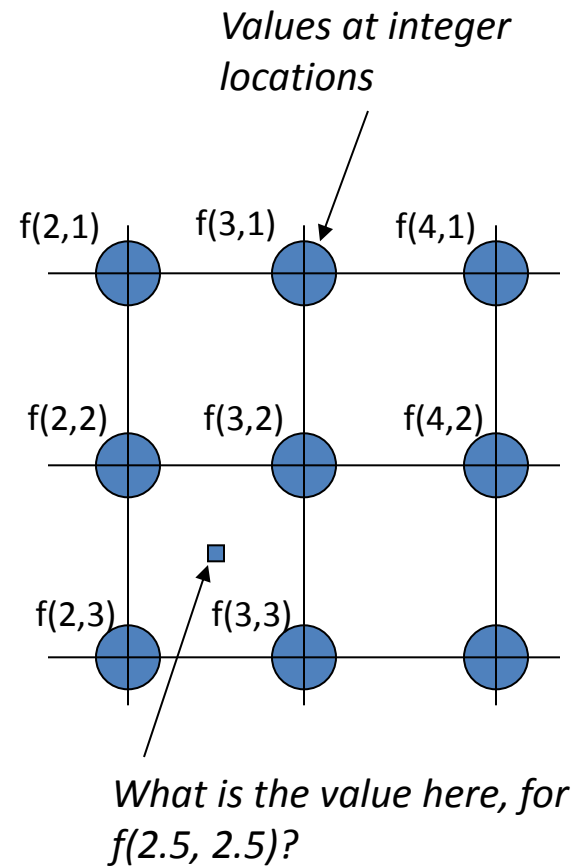
Dept of Electrical Engineering & Computer Science

<http://inside.mines.edu/~whoff/>

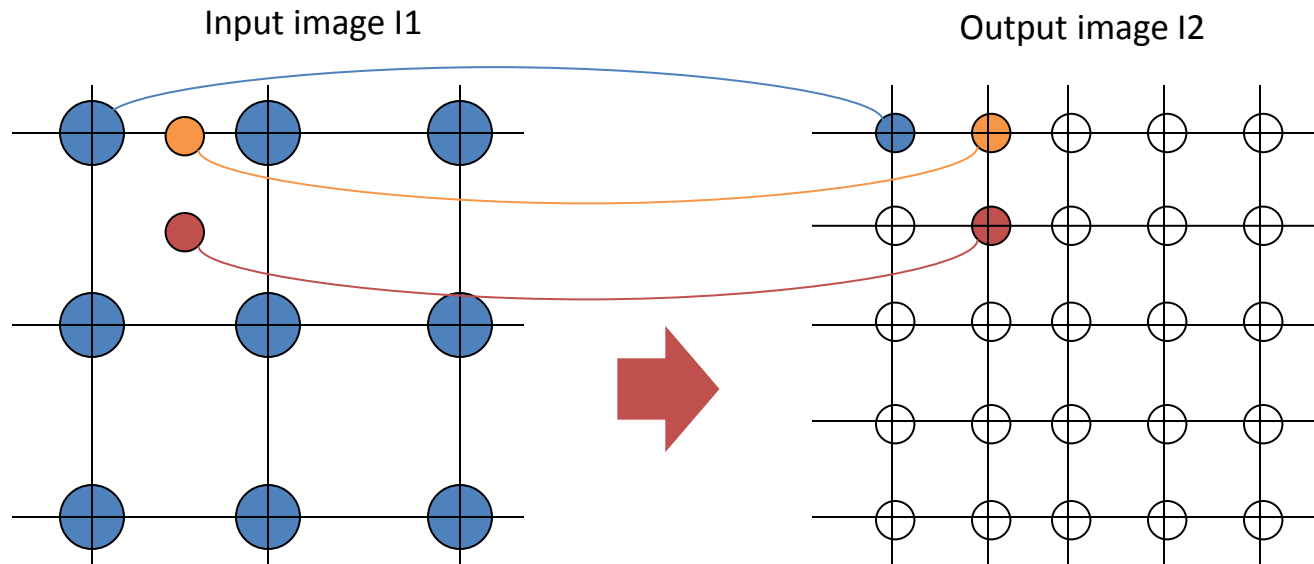
Interpolation and Spatial Transformations

Image Interpolation (Sect 2.4.4)

- We often need to estimate (interpolate) the value of an image at a non-integer location
- Example: expanding a 500x500 image to 1000x1000 – pixel spacing is different
- We can do this by looking at the values nearby. Methods include:
 - Nearest neighbor – just take the value of the closest neighbor
 - Bilinear – take a combination of the four closest neighbors
 - Bicubic – use the closest 16 neighbors (most computationally expensive, but best results)



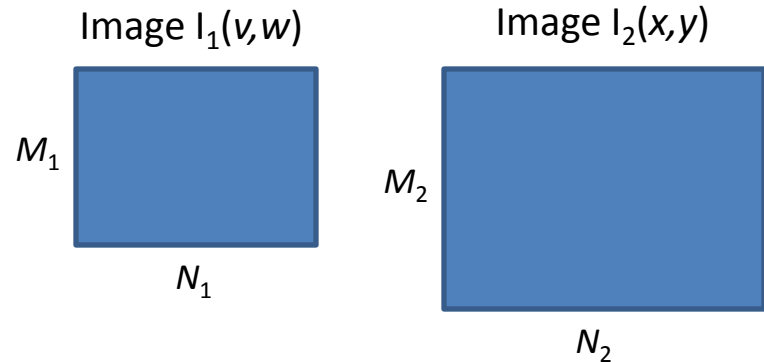
Application: Resizing an Image



- We need to assign values to each pixel in the output image
- So scan through the output image; at each pixel calculate the value from the input image at the corresponding location
- If not at an integer position in the input image, we need to interpolate

Mapping function

- Assume
 - Image I_1 has size $M_1 \times N_1$
 - Image I_2 has size $M_2 \times N_2$
- Then
 - $x = c_x v$
 - $y = c_y w$
 - where c_x, c_y are the scaling factors:
 - $c_x = N_2/N_1$
 - $c_y = M_2/N_1$
- We actually want the inverse function
 - $v = x/c_x$
 - $w = y/c_y$



Matlab example – Nearest Neighbor Interpolation

```
clear all
close all

I1 = imread('cameraman.tif');
M1 = size(I1,1);      % Number of rows in I
N1 = size(I1,2);      % Number of columns in I

% Pick size of output image
M2 = 300;
N2 = 100;
I2 = zeros(M2,N2);    % Allocate output image

cx = N2/N1;           % Scale in x
cy = M2/M1;           % Scale in y
for x=1:N2
    for y=1:M2
        % Calculate position in input image
        v = x/cx;
        w = y/cy;

        % We'll just pick the nearest neighbor to (v,w)
        v = round(v);
        w = round(w);

        I2(y,x) = I1(w,v);
    end
end
```

Bilinear interpolation

- The value at (x,y) is

$$f(x,y) = ax + by + cxy + d$$

where a,b,c,d are coefficients determined by the four closest neighbors

- Equivalent to doing linear interpolation in one dimension, then the other

- To find the value at P , if we know values at the Q 's

- First interpolate horizontally to find values at R_1, R_2

$$f(R_1) \approx \frac{x_2 - x}{x_2 - x_1} f(Q_{11}) + \frac{x - x_1}{x_2 - x_1} f(Q_{21})$$

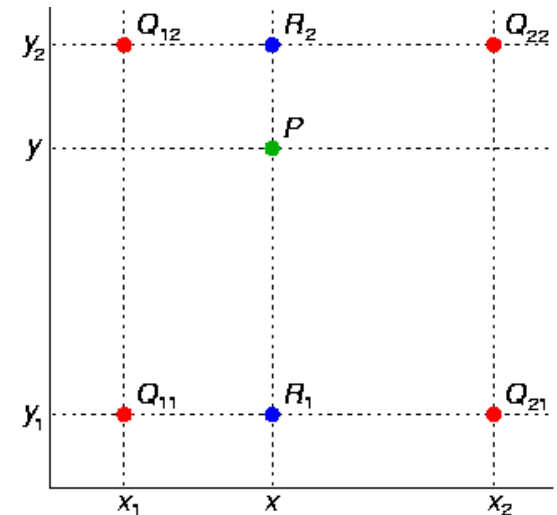
$$f(R_2) \approx \frac{x_2 - x}{x_2 - x_1} f(Q_{12}) + \frac{x - x_1}{x_2 - x_1} f(Q_{22})$$

- Then interpolate vertically

$$f(P) \approx \frac{y_2 - y}{y_2 - y_1} f(R_1) + \frac{y - y_1}{y_2 - y_1} f(R_2).$$

- If corners are at $(0,0)$, $(0,1)$, $(1,0)$, $(1,1)$

$$f(x,y) \approx f(0,0)(1-x)(1-y) + f(1,0)x(1-y) + f(0,1)(1-x)y + f(1,1)xy.$$



http://en.wikipedia.org/wiki/Bilinear_interpolation

Geometric Transformations (2.6.5)

- Example applications
 - Rotation
 - De-warp to undo lens distortion
 - Register two images (e.g., satellite images of the same patch of ground)
 - Also called “rubber sheet” transformations
- A transformation is a function that maps pixel coordinates
$$(x,y) = T(v,w)$$
 - where
 - (v,w) are pixel coordinates in the original image
 - (x,y) are the corresponding pixel coordinates in the transformed image
 - Example: $T(x,y) = (v/2, w/2)$ shrinks the image by a factor of 2
- Remember that we actually need $(v,w) = T^{-1}(x,y)$

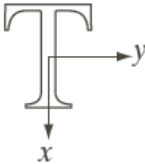

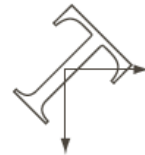
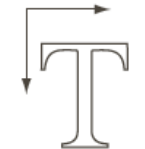


Matrix Representation

- We can often represent the transformation T as a matrix
- It is convenient to append a “1” to the pixel coordinates
- Example: A common transformation is the “affine transform”

$$\begin{bmatrix} x & y & 1 \end{bmatrix} = \begin{bmatrix} u & v & 1 \end{bmatrix} \mathbf{T} = \begin{bmatrix} u & v & 1 \end{bmatrix} \begin{bmatrix} t_{11} & t_{12} & 0 \\ t_{21} & t_{22} & 0 \\ t_{31} & t_{32} & 1 \end{bmatrix}$$

TABLE 2.2

Affine transformations based on Eq. (2.6.–23).

Transformation Name	Affine Matrix, T	Coordinate Equations	Example
Identity	$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$\begin{aligned} x &= v \\ y &= w \end{aligned}$	
Scaling	$\begin{bmatrix} c_x & 0 & 0 \\ 0 & c_y & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$\begin{aligned} x &= c_x v \\ y &= c_y w \end{aligned}$	
Rotation	$\begin{bmatrix} \cos \theta & \sin \theta & 0 \\ -\sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$\begin{aligned} x &= v \cos \theta - w \sin \theta \\ y &= v \sin \theta + w \cos \theta \end{aligned}$	
Translation	$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ t_x & t_y & 1 \end{bmatrix}$	$\begin{aligned} x &= v + t_x \\ y &= w + t_y \end{aligned}$	
Shear (vertical)	$\begin{bmatrix} 1 & 0 & 0 \\ s_v & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$\begin{aligned} x &= v + s_v w \\ y &= w \end{aligned}$	
Shear (horizontal)	$\begin{bmatrix} 1 & s_h & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$\begin{aligned} x &= v \\ y &= s_h v + w \end{aligned}$	

Estimating parameters of the transformation

- If you want to register one image to another, you need to estimate the parameters of the transformation function
- You can do this if you know the correspondences for a few control points (“tiepoints”) in each image
 - I.e., (x_1, y_1) corresponds to (v_1, w_1) , (x_2, y_2) corresponds to (v_2, w_2) , etc

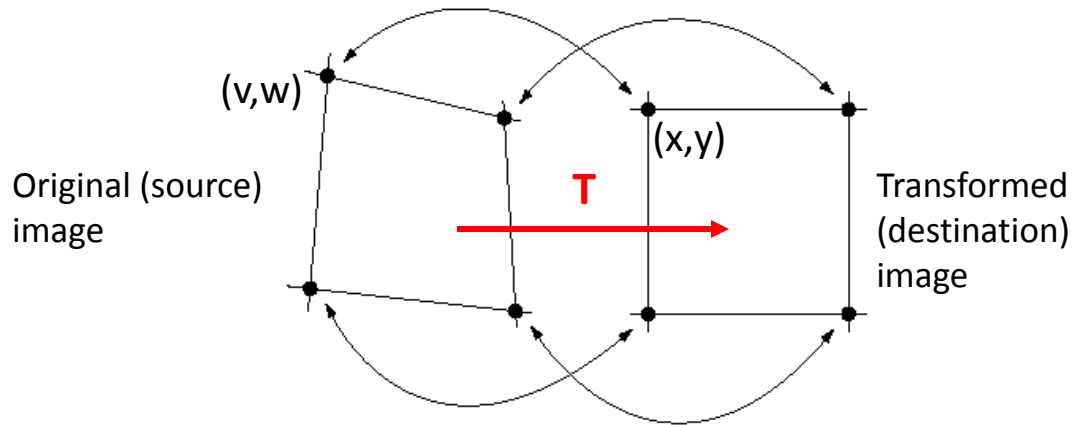


FIGURE 5.32
Corresponding tiepoints in two image segments.

- For the affine transformation:
$$\begin{bmatrix} x & y & 1 \end{bmatrix} = \begin{bmatrix} v & w & 1 \end{bmatrix} \mathbf{T} = \begin{bmatrix} v & w & 1 \end{bmatrix} \begin{bmatrix} t_{11} & t_{12} & 0 \\ t_{21} & t_{22} & 0 \\ t_{31} & t_{32} & 1 \end{bmatrix}$$
 - There are six unknowns – we need at least three pairs of tiepoints (each yields two equations)

Geometric Transformations

- Recall that we actually need the inverse mapping, from transformed image to source image

$$(v,w) = T^{-1}(x,y)$$

- This is because what we really need is the values of the pixels at the integer locations in the transformed image

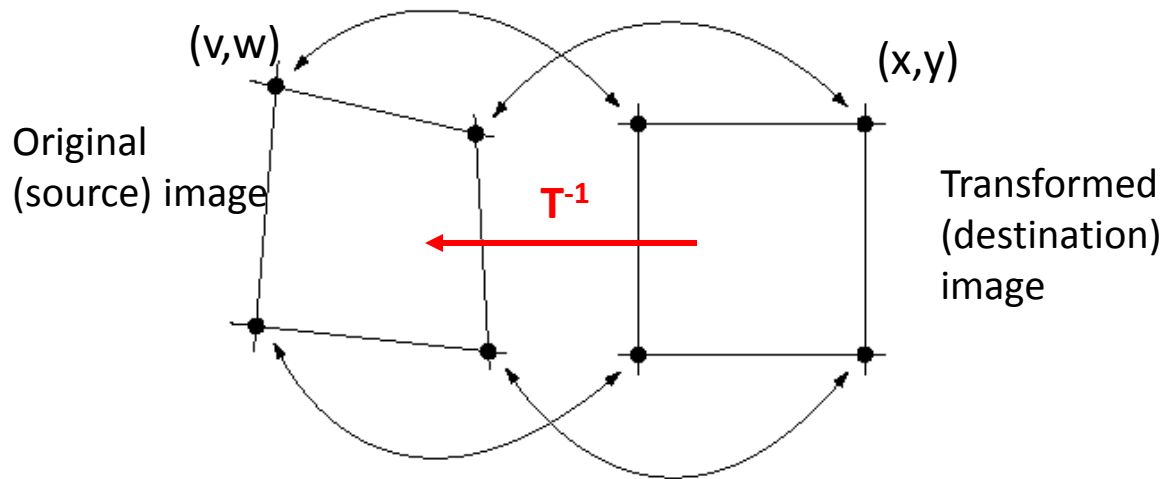


FIGURE 5.32
Corresponding tiepoints in two image segments.

- We need to do gray level interpolation to find the values in the original image

Estimating parameters of the transformation

- If we have three tie points

$$\begin{bmatrix} x_1 & y_1 & 1 \\ x_2 & y_2 & 1 \\ x_3 & y_3 & 1 \end{bmatrix} = \begin{bmatrix} u_1 & v_1 & 1 \\ u_2 & v_2 & 1 \\ u_3 & v_3 & 1 \end{bmatrix} \mathbf{T} \quad \text{or} \quad \mathbf{X} = \mathbf{UT}. \quad \text{Then } \mathbf{T} = \mathbf{U}^{-1}\mathbf{X}$$

- If we have more than three tie points (more is better for accuracy)

$$\begin{bmatrix} x_1 & y_1 & 1 \\ x_2 & y_2 & 1 \\ x_3 & y_3 & 1 \\ & \vdots & \\ x_n & y_n & 1 \end{bmatrix} = \begin{bmatrix} u_1 & v_1 & 1 \\ u_2 & v_2 & 1 \\ u_3 & v_3 & 1 \\ & \vdots & \\ u_n & v_n & 1 \end{bmatrix} \mathbf{T}$$
$$\begin{aligned} \mathbf{X} &= \mathbf{UT} \\ \mathbf{U}^T \mathbf{X} &= \mathbf{U}^T \mathbf{UT} \\ (\mathbf{U}^T \mathbf{U})^{-1} \mathbf{U}^T \mathbf{X} &= \mathbf{T} \end{aligned}$$

Example from Matlab

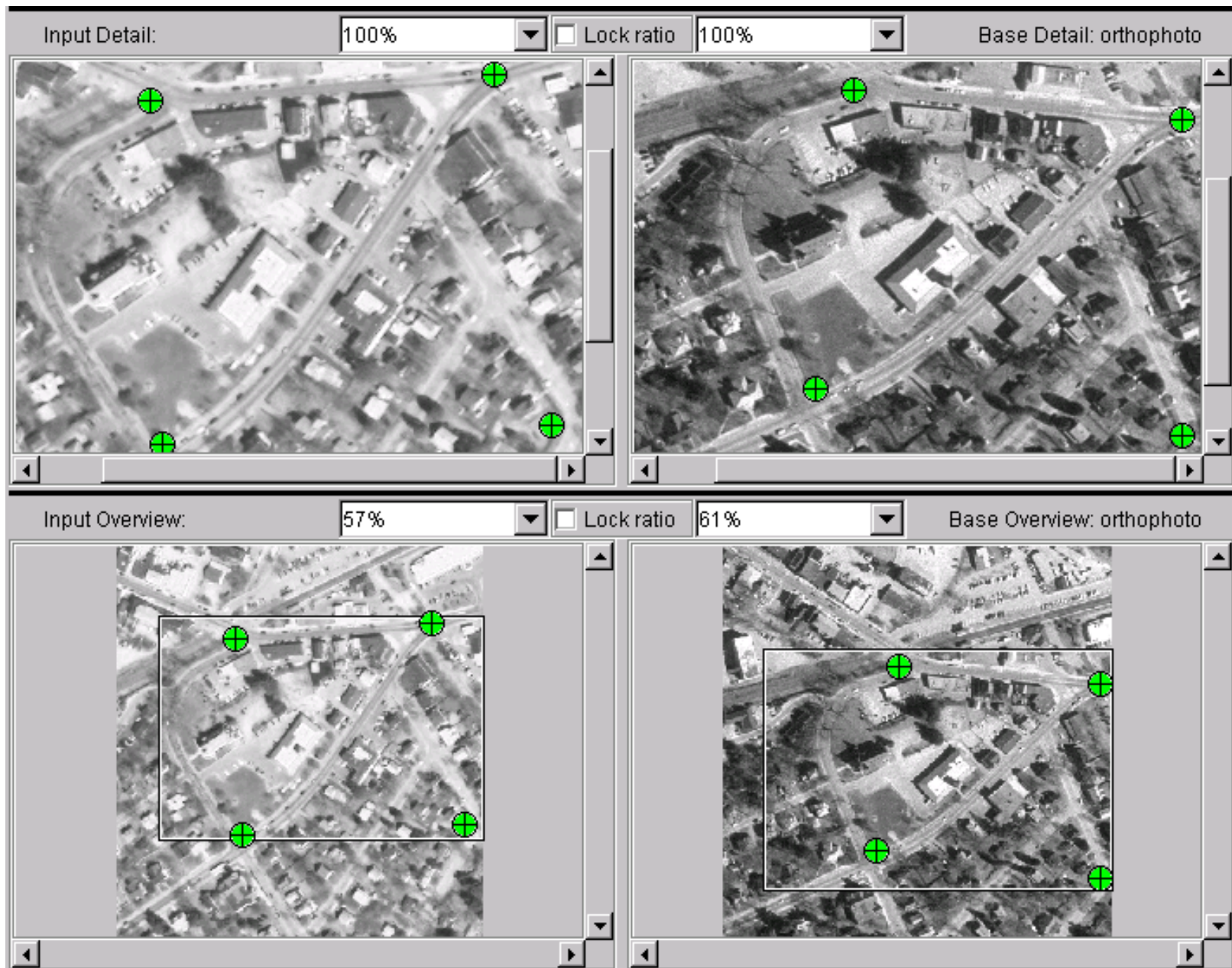
- We want to transform the new image on the left to register (align) it to the old image on the right (to compare them for changes, for example)
- See Image Processing Toolbox>Examples>Image registration



westconcordaerial.png



westconcordorthophoto.png



See Matlab functions `cpselect`, `cp2tform`, `imtransform`



Registered Image



Orthophoto Image



```
diff = imabsdiff(registeredOriginal,  
registered(:, :, 1));
```

Can look at the 3x3 affine
transformation matrix

Summary / Questions

- To do a spatial transformation, we need to (a) specify a function that maps points from one image to the other, and (b) interpolate intensity values at non-integer positions.
- Affine transformations can perform rotation, translation, scaling, and shear.
- When performing the transformation to generate the new image, why is the inverse mapping useful?