

Affine Linear Transformations

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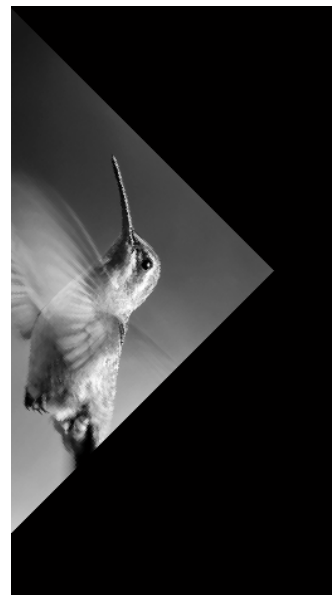
(notes modified from Noah Snavely, Spring 2009)



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More image transformations

- Recall from last time ... the default rotation is around the point $(0, 0)$ – the upper-left corner of the image



- This probably isn't really what we want...



Translation

- We really want to rotate around the *center* of the image
- Nice trick: move the center of the image to the origin, apply the default rotation, then move the center back. If T is the translation, R the desired rotation, and D the default rotation, then $R = T^{-1}DT$. We often call this *conjugation* of D by T ... it's a very useful trick in a great many different contexts!
- But technically, translation isn't a linear function - check the two definition properties for $\mathbf{f}(\mathbf{v}) = \mathbf{v} + \mathbf{w}$ (for some constant vector \mathbf{w})...
 - $\mathbf{f}(\mathbf{v} + \mathbf{u}) \stackrel{?}{=} \mathbf{f}(\mathbf{v}) + \mathbf{f}(\mathbf{u})$ $LHS = \mathbf{v} + \mathbf{u} + \mathbf{w}$ $RHS = \mathbf{v} + \mathbf{u} + 2\mathbf{w}$
 - $\mathbf{f}(a\mathbf{v}) \stackrel{?}{=} a \mathbf{f}(\mathbf{v})$ $LHS = a\mathbf{v} + \mathbf{w}$ $RHS = a\mathbf{v} + 2a\mathbf{w}$
- Formally we call this an *affine* linear function



Reminder: shear from lecture 15

- Shear



$$H = \begin{bmatrix} 1 & a \\ 0 & 1 \end{bmatrix}$$

*Remember again that for images, the positive y direction is **downwards**!!*



Detour: Homogeneous coordinates

- We want a way to use MATLAB's matrix routines to compute translations, but matrices only do linear functions...
- Ans: add a dimension! Add a 1 to the end of our 2D point $(x, y) \rightarrow (x, y, 1)$ to give a point on the plane parallel to the 2D coordinate plane at height 1 = "homogeneous" 2D points
- Essentially each line through the origin in 3D 'corresponds' to a point in 2D (what about the lines actually in the xy-plane?)
- We can represent transformations on 2D homogeneous coordinates as 3D matrices
- 'Shear' in 3D is linear, and 'is' translation in (homog) 2D !!!!



Translation via homogeneous coordinates

$$T = \begin{bmatrix} 1 & 0 & s \\ 0 & 1 & t \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} s \\ t \\ 1 \end{bmatrix}$$

Shear parallel to the xy-plane by adding s to the x direction and t to the y direction

- Other transformations just add an extra row and column with $[0 \ 0 \ 1]$

$$S = \begin{bmatrix} s & 0 & 0 \\ 0 & s & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad R = \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

scale *rotation*



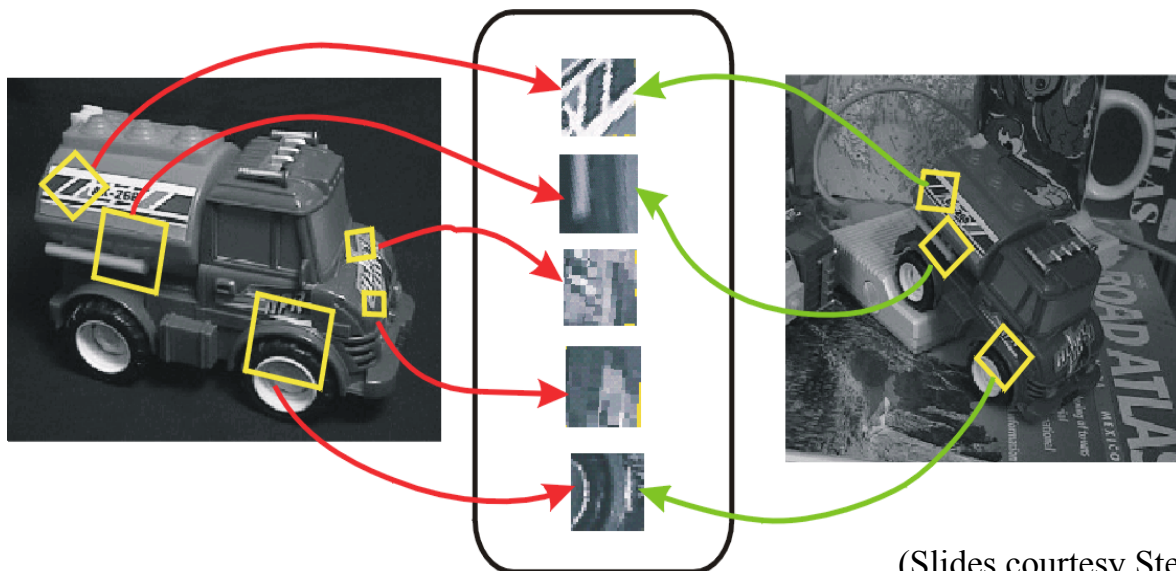
Rotation about image centre

- Translate center to origin $T = \begin{bmatrix} 1 & 0 & -w/2 \\ 0 & 1 & -h/2 \\ 0 & 0 & 1 \end{bmatrix}$
- Rotate $R = \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$
- Translate back to center $T^{-1} = \begin{bmatrix} 1 & 0 & w/2 \\ 0 & 1 & h/2 \\ 0 & 0 & 1 \end{bmatrix}$
- Combining these gives $T^{-1} R T$



Invariant local features

- Find features that are invariant to transformations
 - geometric invariance: translation, rotation, scale
 - photometric invariance: brightness, exposure, ...



(Slides courtesy Steve Seitz)



Why local features?

- Locality
 - features are local, so robust to occlusion and clutter
- Distinctiveness:
 - can differentiate a large database of objects
- Quantity
 - hundreds or thousands in a single image
- Efficiency
 - real-time performance achievable



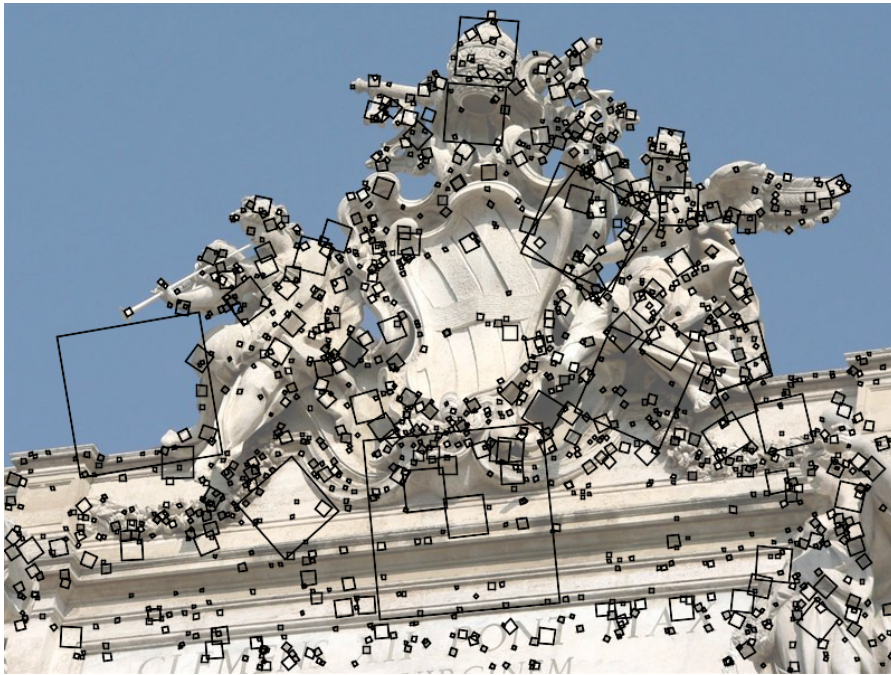
More motivation...

- Feature points are used for:
 - Image alignment (e.g., mosaics)
 - 3D reconstruction
 - Motion tracking
 - Object recognition
 - Robot navigation
 - ...



SIFT Features

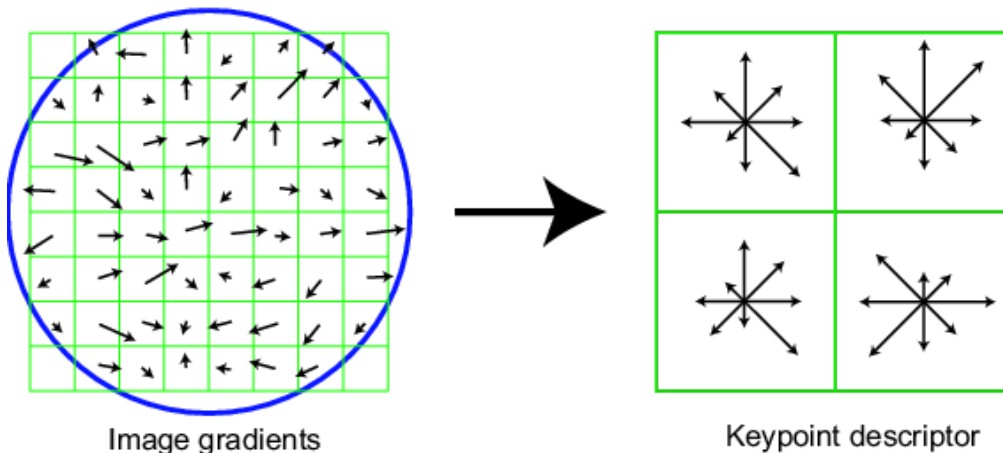
- **Scale-Invariant Feature Transform**



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SIFT descriptor

- Very complicated, but very powerful
- (The details aren't all that important for this class.)
- 128 dimensional descriptor



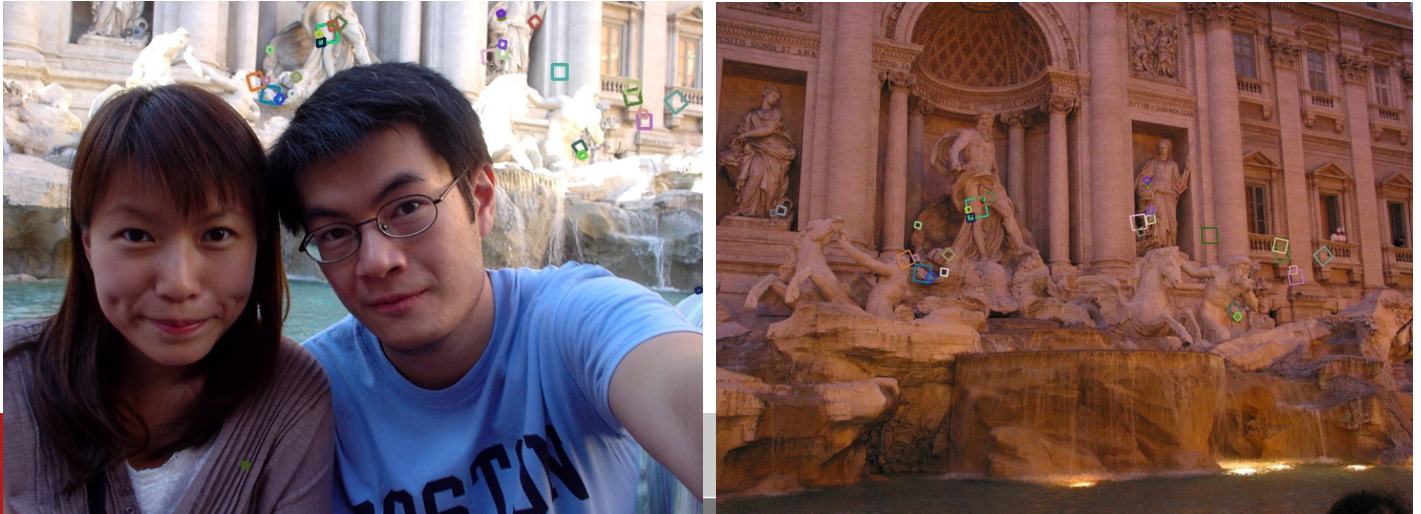
Adapted from a slide by David Lowe



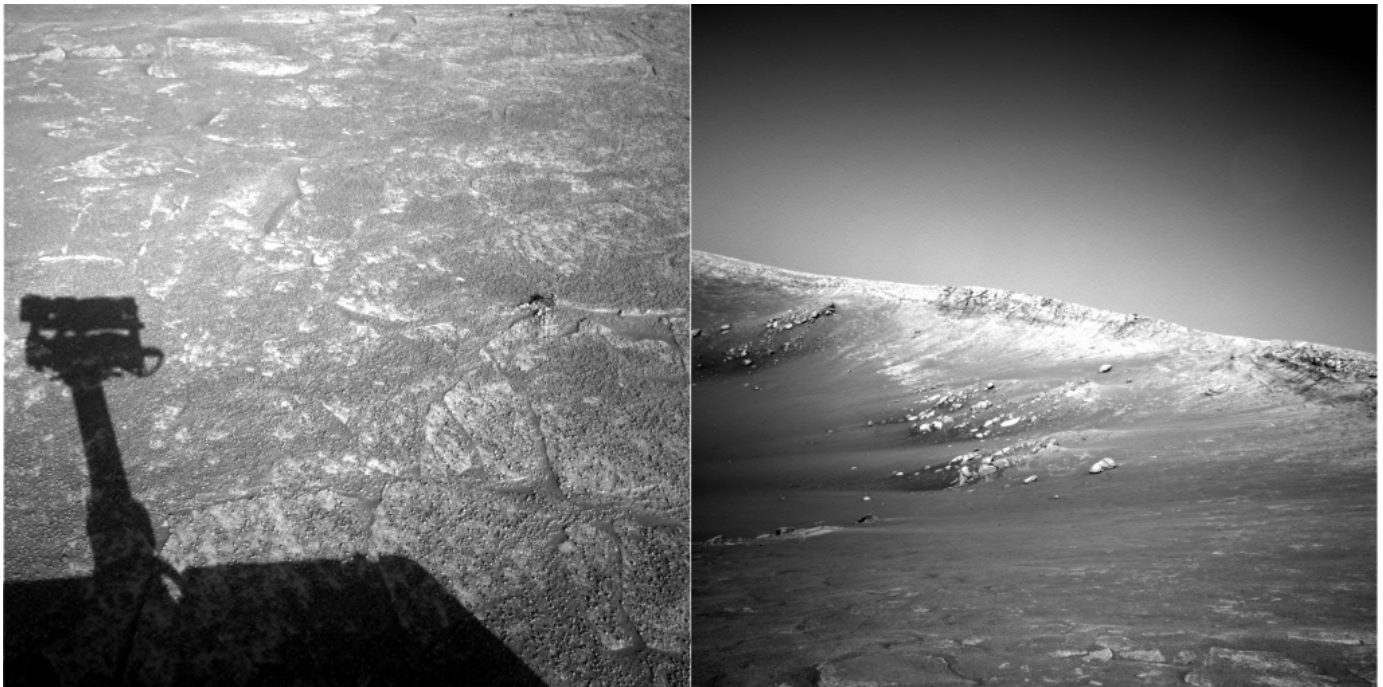
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Properties of SIFT

- Extraordinarily robust matching technique
 - Can handle significant changes in illumination
 - Sometimes even day vs. night (below)
 - Fast and efficient—can run in real time
 - Lots of code available
 - http://people.csail.mit.edu/albert/ladypack/wiki/index.php/Known_implementations_of_SIFT



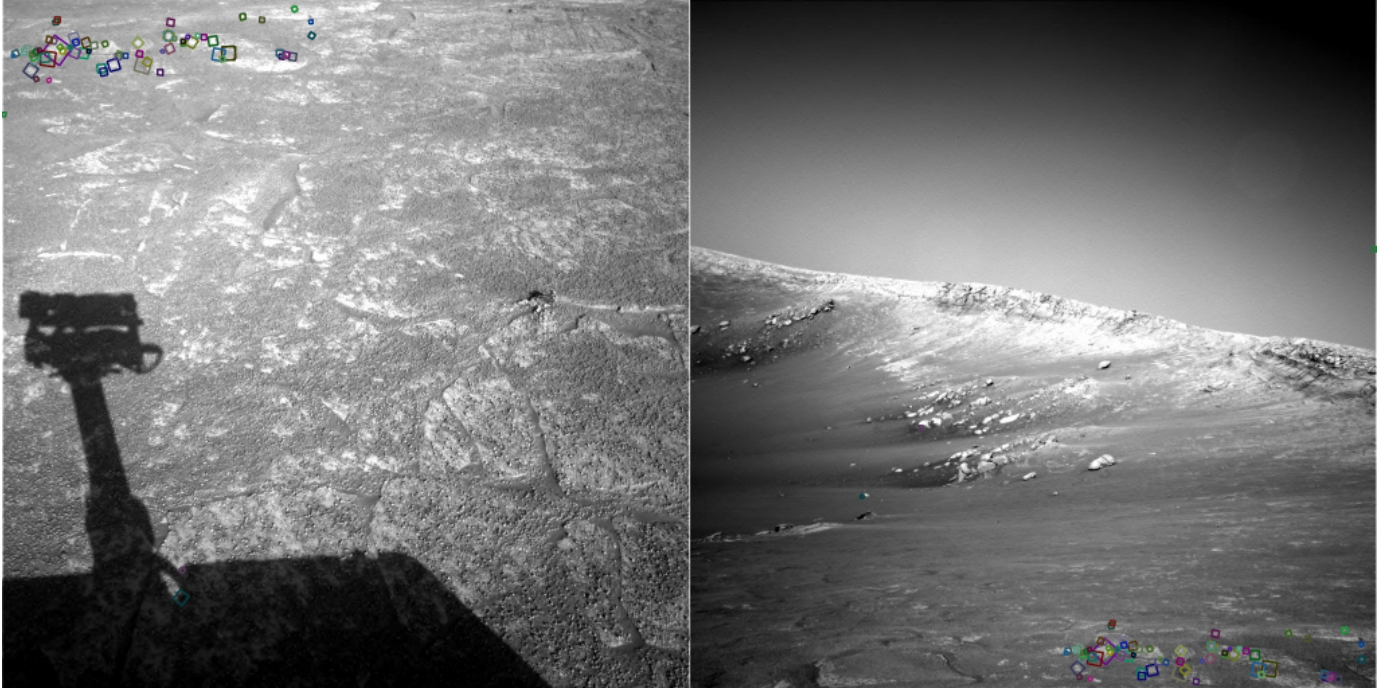
Do these two images overlap?



NASA Mars Rover images



Answer below



NASA Mars Rover images



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Image transformations

- What about a general homogeneous transformation?

$$T = \begin{bmatrix} a & b & c \\ d & e & f \\ 0 & 0 & 1 \end{bmatrix}$$

$$\begin{bmatrix} a & b & c \\ d & e & f \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} ax + by + c \\ dx + ey + f \\ 1 \end{bmatrix}$$

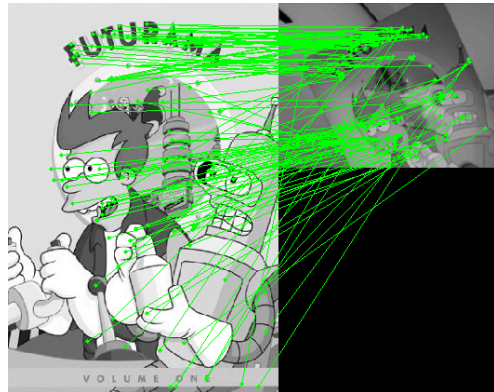
- Called a 2D *affine* transformation



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Solving for image transformations

- Given a set of matching points between image 1 and image 2...



... can we solve for an affine transformation T mapping 1 to 2?



