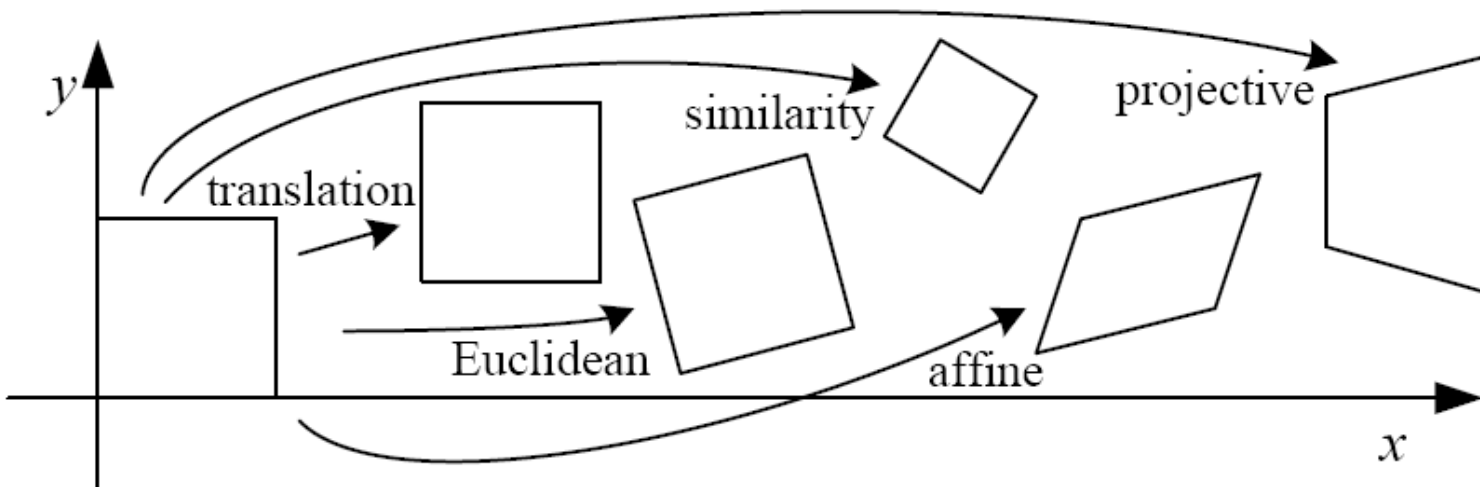


CSCI 497P/597P: Computer Vision



Lecture 14

Homogeneous Points Intuition

Fitting Transformations

Announcements

Goals

- Know the definition of a **projective (homography) transformation**, and gain some geometric intuition for what it represents in 2D.
 - Understand the meaning of homogeneous **points at infinity**.
 - Know why a homography has 8 degrees of freedom, not 9 [hw2]
- Know how to find a least-squares best-fit transformation for the given models:
translation, affine, homography

Transformations: Linear

$$\begin{bmatrix} x' \\ y' \\ w \end{bmatrix} = \begin{bmatrix} a & b & c \\ d & e & f \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ w \end{bmatrix}$$

- Properties:

- Lines map to lines
- Parallel lines remain parallel
- Ratios of lengths along lines are preserved
- Closed under composition
- Origin maps to origin

linear

Transformations: Affine

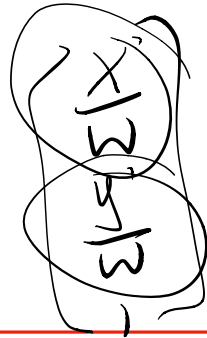
$$\begin{bmatrix} x' \\ y' \\ w \end{bmatrix} = \begin{bmatrix} a & b & c \\ d & e & f \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ w \end{bmatrix}$$

- Properties:

- Lines map to lines
- Parallel lines remain parallel
- Ratios of lengths along lines are preserved
- Closed under composition
- Origin **does not** necessarily map to origin

affine

Transformations: Projective (Homography)



$$\begin{bmatrix} x' \\ y' \\ w' \end{bmatrix} = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ w \end{bmatrix}$$

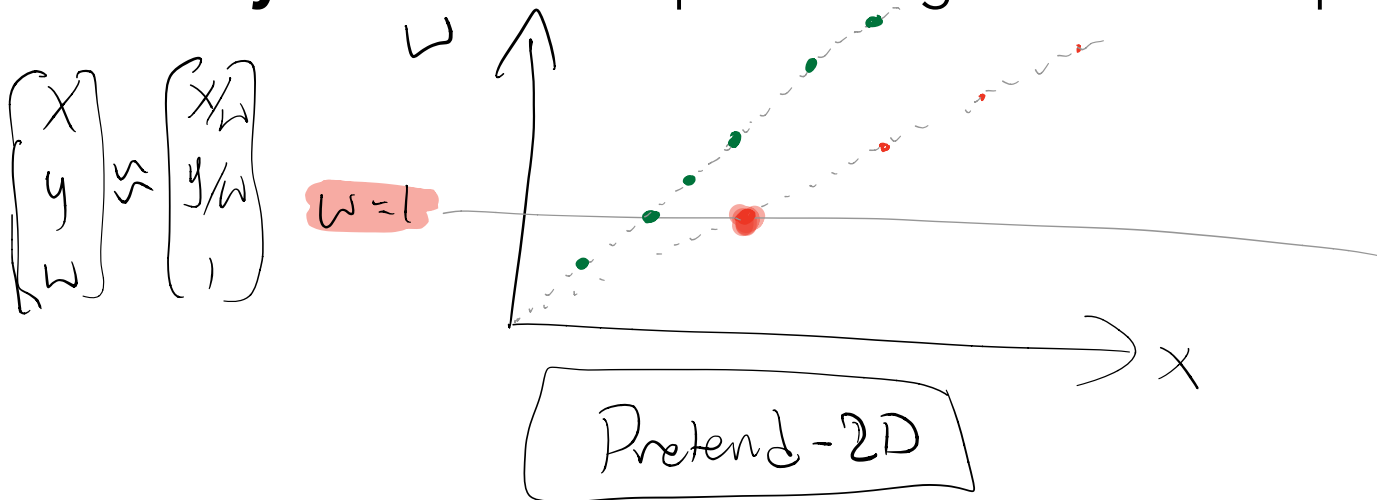
- Properties:

- Lines map to lines
- Parallel lines **do not** remain parallel
- Ratios of lengths along lines are **not** preserved.
- Closed under composition
- Origin **does not** necessarily map to origin **projective**

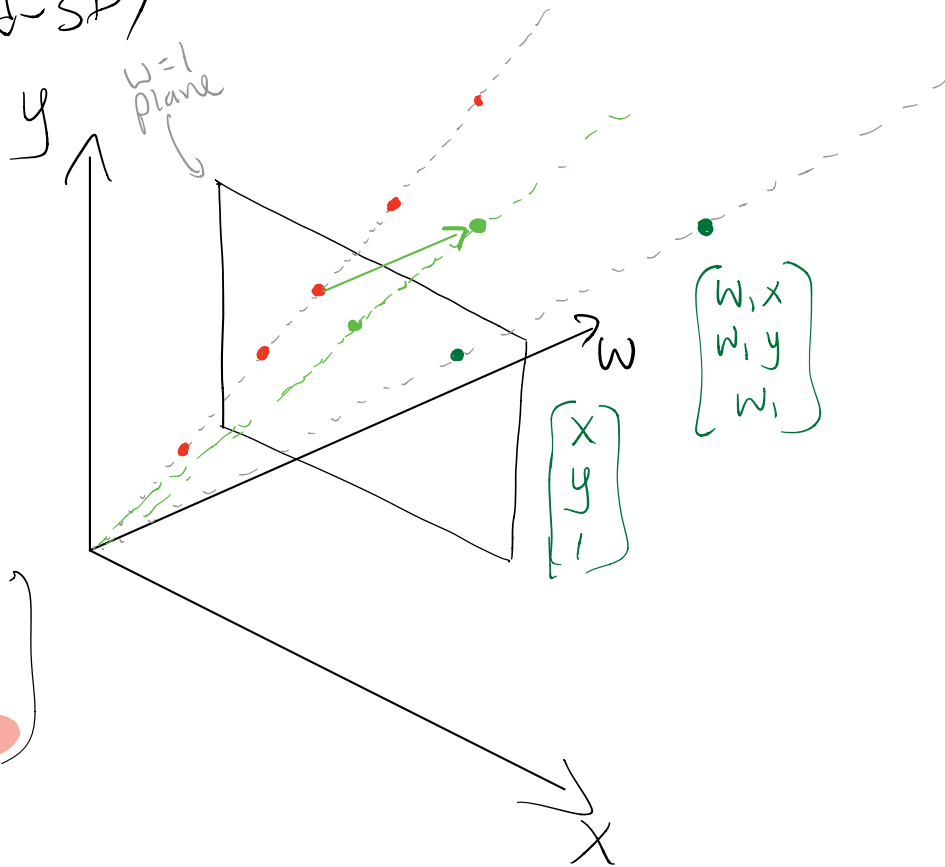
Homogeneous Coordinates: Intuition for our math hack

A 3-vector belongs to a **family** of 3-vectors representing the same 2D point.

1D example, for intuition: A 2-vector belongs to a **family** of 2-vectors representing the same 1D point.



In 2D: (Pretend-3D)



Affine:

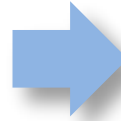
$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} a & b & c \\ d & e & f \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

Homography:

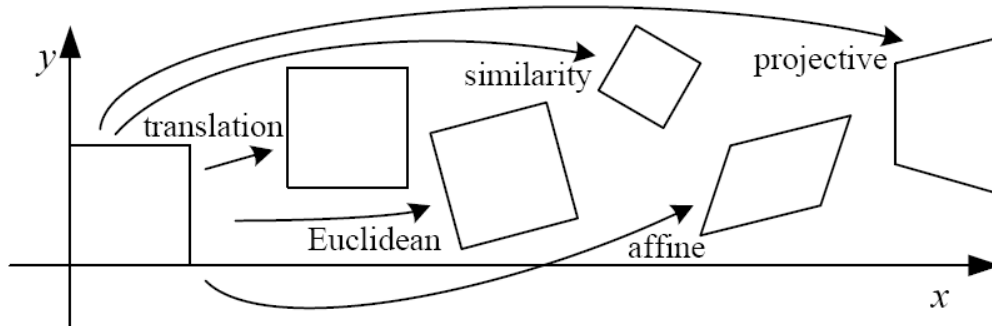
$$\begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$






What are we looking at?

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & 1 \end{bmatrix}$$



Transformations: A Hierarchy



Name	Matrix	# D.O.F.	Preserves:	Icon
translation	$\begin{bmatrix} \mathbf{I} & \mathbf{t} \end{bmatrix}_{2 \times 3}$	2	orientation + ...	
rigid (Euclidean)	$\begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix}_{2 \times 3}$	3	lengths + ...	
similarity	$\begin{bmatrix} s\mathbf{R} & \mathbf{t} \end{bmatrix}_{2 \times 3}$	4	angles + ...	
affine	$\begin{bmatrix} \mathbf{A} \end{bmatrix}_{2 \times 3}$	6	parallelism + ...	
projective	$\begin{bmatrix} \tilde{\mathbf{H}} \end{bmatrix}_{3 \times 3}$	8	straight lines	

Homographies for image alignment



Why the 1 in the bottom right?

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & 1 \end{bmatrix}$$

Prove on HW2: For any 3x3 matrix H , there exists another 3x3 matrix H' which:

- Has a 1 in the bottom-right corner
- Has the same effect on homogeneous points.

Why the 1 in the bottom right?

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & 1 \end{bmatrix}$$

Prove on HW2: For any 3x3 matrix H , there exists another 3x3 matrix H' which:

- Has a 1 in the bottom-right corner
- Has the same effect on homogeneous points.

Fixing the 1 is merely a convention.

Another convention: values in H as a vector have magnitude 1.

Why the 1 in the bottom right?

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & 1 \end{bmatrix}$$

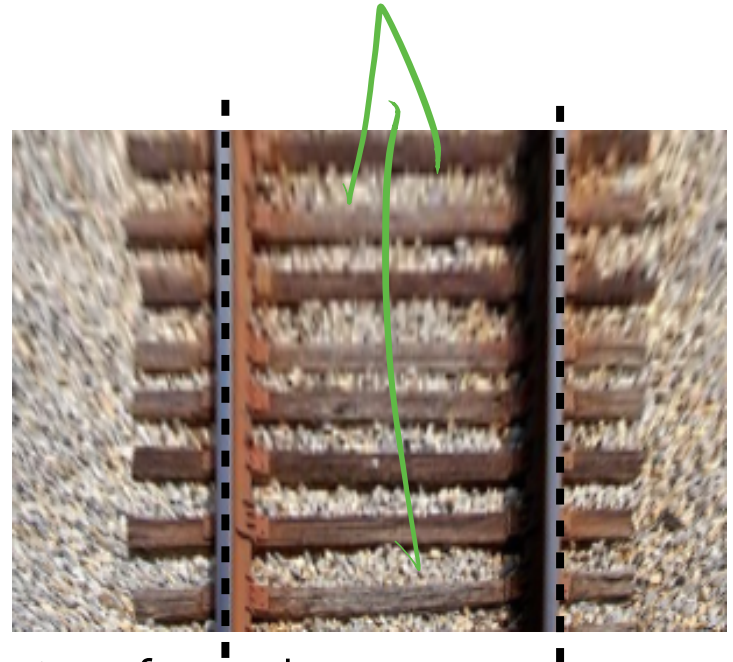
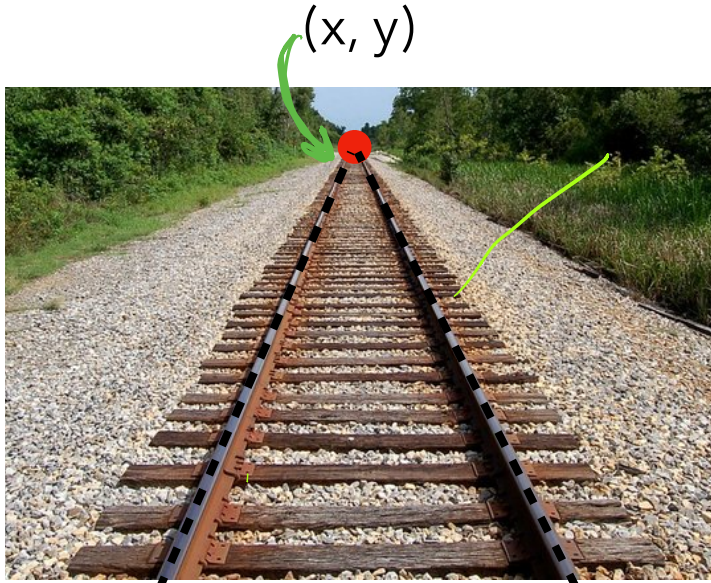
Prove on HW2: For any 3x3 matrix H , there exists another 3x3 matrix H' which:

- Has a 1 in the bottom-right corner
- Has the same effect on homogeneous points.

Consequently:

- homographies have 8 **degrees of freedom (DOF)**.
- affine transformations have 6 DOF
- translations have 2 DOF

About those parallel lines...



The homography H transforms points from the left image to the right image.

Not apart \rightarrow
 point at infinity \rightarrow

$$\begin{bmatrix} x \\ y \\ 0 \end{bmatrix}$$

What is H $\begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$?

$$\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \leftarrow \begin{bmatrix} 010 - 010 \end{bmatrix}$$

font a, ... (0)

Next up:

(imperfect)

Given a set of feature matches, how do I **find** the **transformation** that relates the two images?

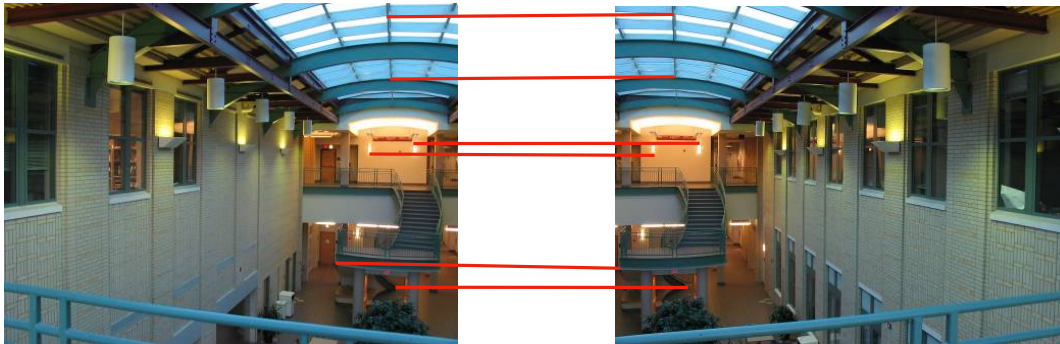
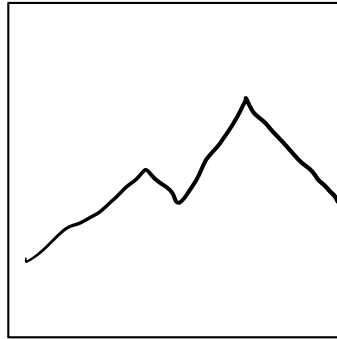
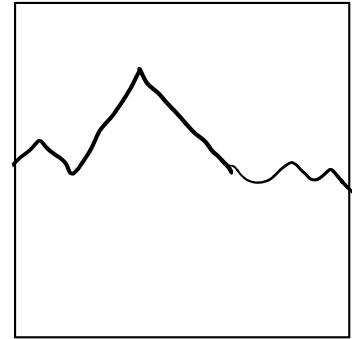


Image Alignment: Translation

I_1



I_2

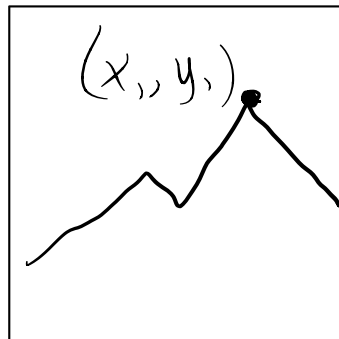


$$t_x = x'_1 - x_1$$

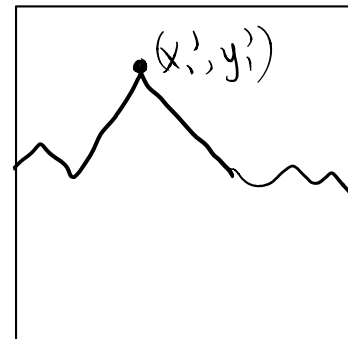
$$t_y = y'_1 - y_1$$

$$\begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix}$$

I_1



I_2



Multiple Matches?

$$t_x = \frac{1}{n} \sum_{i=0}^n (x_i' - x_i)$$

$$t_y = \frac{1}{n} \sum_{i=0}^n (y_i' - y_i)$$

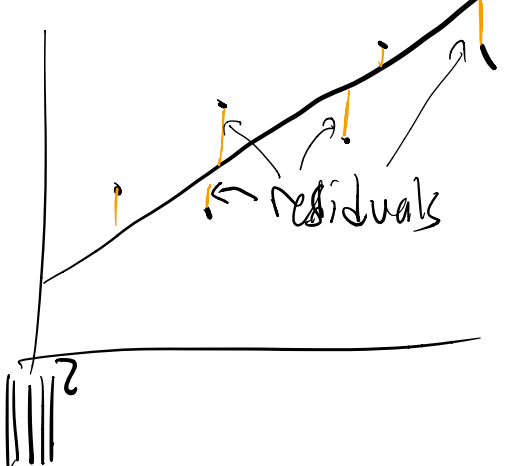
Multiple Matches: Linear Algebra Edition!

Goal: Minimize difference between

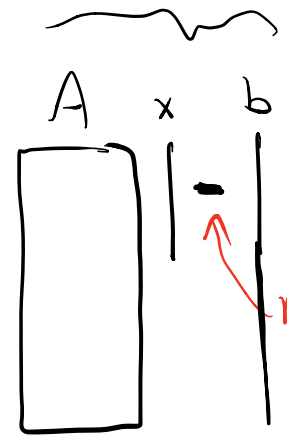
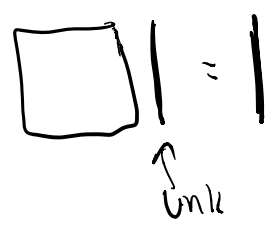
Observed landing pt and model's prediction:

$$\sum_{i=0}^n \left(\begin{bmatrix} x_i' \\ y_i' \end{bmatrix} - \begin{bmatrix} x_i + t_x \\ y_i + t_y \end{bmatrix} \right)^2$$

$$\min_x \|Ax - b\|^2$$



$$Ax = b$$



not possible to have =

For Translation:

$$\begin{aligned} - r_i(t_x)^2 &= (x_i^y - (x_i + t_x))^2 \\ - r_i(t_y)^2 &= (y_i^x - (y_i + t_y))^2 \end{aligned} \left. \vphantom{\begin{aligned} - r_i(t_x)^2 \\ - r_i(t_y)^2 \end{aligned}} \right\} z_n$$

$$\min_x \underbrace{Ax - b}_{\substack{2n \times 2 \quad 2 \times 1 \quad 2n \times 1}}$$

unkennungs

$$(t_x - (x_i' - x_i))^2$$

$$\left. \begin{matrix} 1 \\ 0 \\ 1 \\ 0 \\ \vdots \end{matrix} \right\} 2n$$

$$\left. \begin{matrix} 0 \\ 1 \\ 0 \\ 1 \\ \vdots \end{matrix} \right\}$$

$$\begin{bmatrix} t_x \\ t_y \end{bmatrix}$$

$$\begin{bmatrix} x_1' - x_1 \\ y_1' - y_1 \\ x_2' - x_2 \\ y_2' - y_2 \\ \vdots \end{bmatrix}$$

$$\begin{matrix} t_x - (x_i' - x_i) \\ t_y - (y_i' - y_i) \end{matrix}$$

Solving $\min_x \|Ax - b\|^2$

math class: normal equations

$$(A^T A)^{-1} (A^T b)$$

$$x = (A^T A)^{-1} A^T b$$

On a computer:

`np.linalg.lstsq(A, b)`

↖ use this!