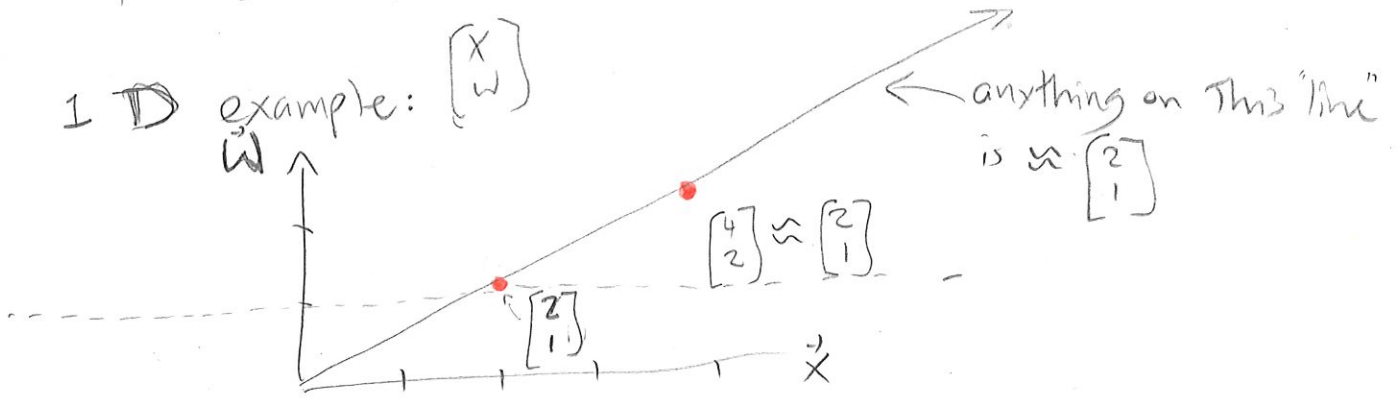
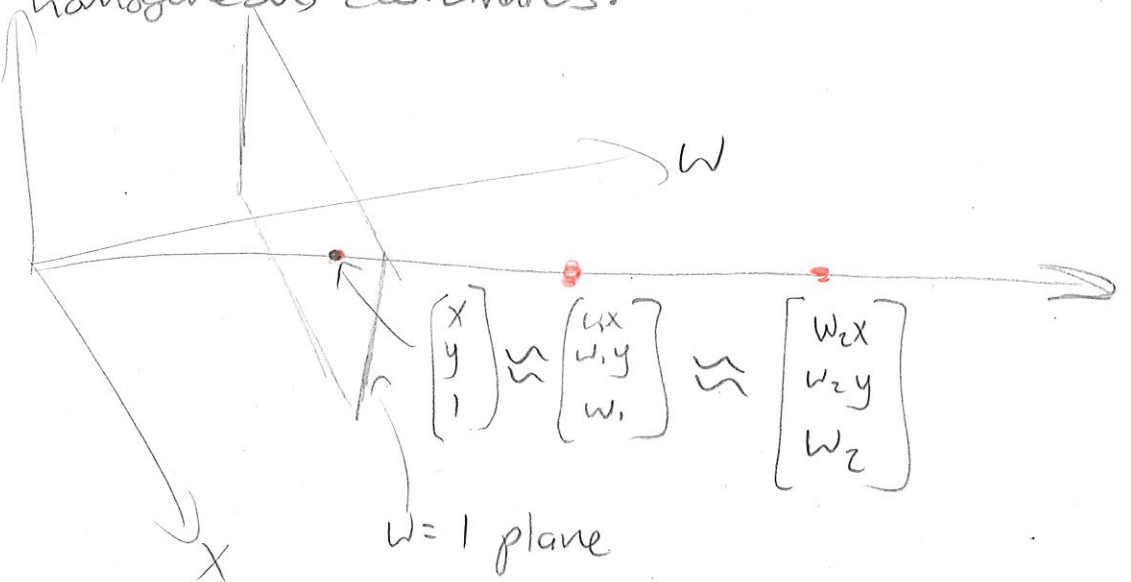


Homogeneous Coordinates: Not 3D, but let's pretend.



In 2D homogeneous coordinates:



Affine:

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

points can't leave projective plane

"projective plane"

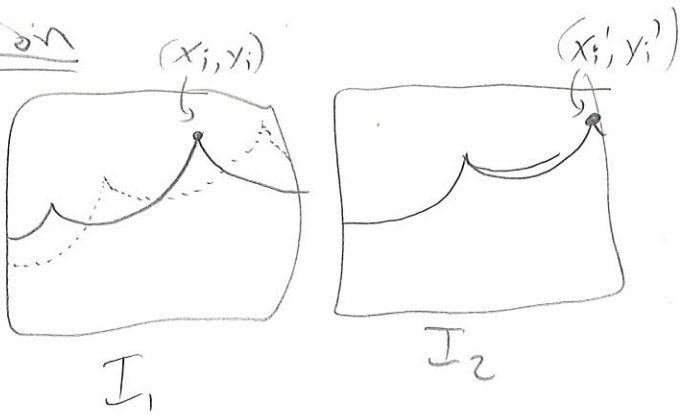
Homography

$\begin{bmatrix} g & h & i \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$

can get

Translation

Single Match



matches

unknowns



$$x_i' = x_i + x_t$$

$$y_i' = y_i + y_t$$

Transformation Fitting

- Tx
- Affine
- Homography
- ↳ SVD
- ||Ah=0||

$$x_t = x_i' - x_i$$

$$y_t = y_i' - y_i$$

More Matches: Average them I guess? same unknown

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

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

$$x_t = x_1' - x_1$$

$$x_t = x_2' - x_2$$

⋮

More Matches: Linear Algebra Edition

$$x_t = x_i' - x_i, \quad y_t = y_i' - y_i$$

⋮

2n equations
2 unknowns.

Least Squares!

$$\text{Min } \|Ax = b\|^2$$

↑ residuals

unknowns

$$r_{x_i}(x_t) = x_i' - (x_i + x_t)$$

$$r_{y_i}(y_t) = y_i' - (y_i + y_t)$$

$$\begin{matrix}
 & A & & t & & b \\
 \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \\ 0 & 1 \\ \vdots & \vdots \\ 1 & 0 \\ 0 & 1 \end{pmatrix} & & \begin{bmatrix} x_t \\ y_t \end{bmatrix} & - & \begin{pmatrix} x_1' - x_1 \\ y_1' - y_1 \\ x_2' - x_2 \\ \vdots \\ x_n' - x_n \\ y_n' - y_n \end{pmatrix}
 \end{matrix}$$

Affine

②

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

Unknowns: a, b, c, d, e, f (6)

Eqs: 2 per match

$$x'_i = ax_i + by_i + c$$

$$y'_i = dx_i + ey_i + f$$

Residuals:

$$(ax_i + by_i + c) - x'_i$$

$$(dx_i + ey_i + f) - y'_i$$

$$\text{Min} \|A t - b\|^2$$

\uparrow rank

$$\begin{bmatrix} x_1 & y_1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & x_1 & y_1 & 1 \\ x_2 & y_2 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & x_1 & y_1 & 1 \\ & & & \vdots & & \\ x_n & y_n & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & x_n & y_n & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \\ e \\ f \end{bmatrix} = \begin{bmatrix} x'_1 \\ y'_1 \\ x'_2 \\ y'_2 \\ \vdots \\ x'_n \\ y'_n \end{bmatrix}$$

Homography

(3)

$$\begin{bmatrix} X_h \\ y_h \\ w_h \end{bmatrix} = \begin{bmatrix} h_{00} & h_{01} & h_{02} \\ h_{10} & h_{11} & h_{12} \\ h_{20} & h_{21} & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix}$$

make this h_{22}
then rewrite

$$X' = \frac{h_{00}X + h_{01}Y + h_{02}}{h_{20}X + h_{21}Y + 1} \left. \begin{array}{l} \} X_h \\ \} w_h \end{array} \right\}$$

$$Y' = \frac{h_{10}X + h_{11}Y + h_{12}}{w_h}$$

Not linear!



$$\begin{bmatrix} X_h \\ y_h \\ w_h \end{bmatrix} = \begin{bmatrix} h_{00} & h_{01} & h_{02} \\ h_{10} & h_{11} & h_{12} \\ h_{20} & h_{21} & h_{22} \end{bmatrix} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix}$$

$$\frac{h_{00}X + h_{01}Y + h_{02}}{h_{20}X + h_{21}Y + h_{22}} = X'$$

$$h_{00}X + h_{01}Y + h_{02} = X'(h_{20}X + h_{21}Y + h_{22})$$

* Residual: $(h_{00}X + h_{01}Y + h_{02}) - X'(h_{20}X + h_{21}Y + h_{22})$
 $= (h_{00}X + h_{01}Y + h_{02}) - (h_{20}XX' + h_{21}YX' + h_{22}X')$

$$\frac{h_{10}X + h_{11}Y + h_{12}}{h_{20}X + h_{21}Y + h_{22}} = Y'$$

$y_{\text{residual}}: h_{10}X + h_{11}Y + h_{12} - (h_{20}XY' + h_{21}YY' + h_{22}Y')$

We had: $X' = \frac{X_h}{w_h}$ now $X_h = X'w_h$

$y' = \frac{y_h}{w_h}$ now $y_h = y'w_h$

Residuals: X_n $X'W_n$

$$X: (h_{00}X + h_{01}Y + h_{02} - h_{20}XX' - h_{21}X'Y - h_{22}X')$$

$$Y: h_{10}X + h_{11}Y + h_{12} - h_{20}XY' - h_{21}YY' - h_{22}Y'$$

$$\min \|Ah - b\|^2$$

$$\begin{bmatrix}
 x_1, y_1, 1 & 0 & 0 & 0 & x_1^2 & x_1 y_1 & x_1' \\
 0 & 0 & 0 & x_1, y_1, 1 & x_1 y_1' & y_1 y_1' & y_1' \\
 \vdots & & & & & & \\
 X_n Y_n & 1 & 0 & 0 & 0 & X_n X_n' & X_n' Y_n & X_n' \\
 0 & 0 & 0 & X_n Y_n & 1 & X_n Y_n' & Y_n Y_n' & Y_n'
 \end{bmatrix}
 -
 \begin{bmatrix}
 h_{00} \\
 h_{01} \\
 h_{02} \\
 h_{10} \\
 h_{11} \\
 h_{12} \\
 h_{20} \\
 h_{21} \\
 h_{22} \\
 0 \\
 0
 \end{bmatrix}$$

Minimizing $\|Ah - 0\|^2$

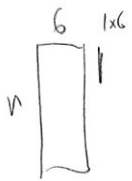
5

Trivial: $h = \vec{0}$

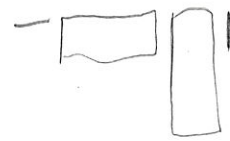
Solution: constrain $\|h\| = 1$

Min $\|Ah - 0\|^2$ s.t. $\|h\| = 1$

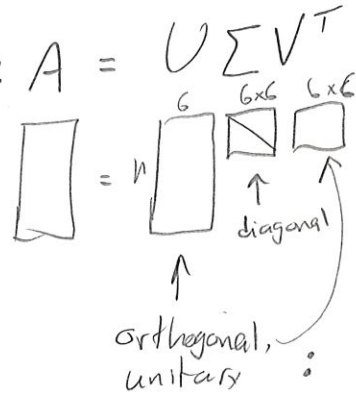
min $\|Ah\|^2$ s.t. $\|h\| = 1$



$\|Ah\|^2 = (Ah)^T(Ah) = h^T A^T A h$



Singular value decomposition: $A = U \Sigma V^T$

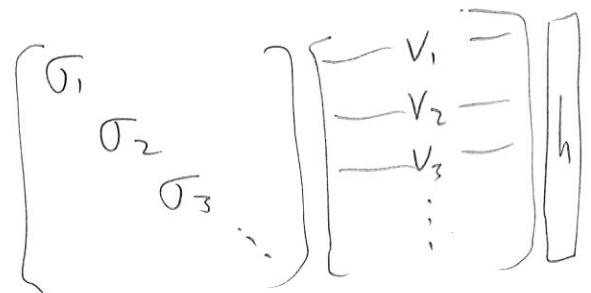


$h^T A^T A h = h^T U \Sigma U^T U \Sigma V^T h$
 $= h^T U \Sigma \Sigma V^T h$

$U_i^T \cdot U_j = 0$
 $U_i^T U_i = 1$



$\Sigma V^T h$



$\|h\| = 1$

$\|v_i\| = 1$

$v_i^T \cdot v_j = 1$

$h = v_3 \rightarrow$

$\sigma_1, v_1, v_3 \rightarrow 0$
 $\sigma_2, v_2, v_3 \rightarrow 0$
 $\sigma_3, v_3, v_3 \rightarrow \sigma_3$
 \vdots
 0

$= \begin{bmatrix} \sigma_1 & & & \\ & \sigma_2 & & \\ & & \sigma_3 & \\ & & & \ddots \end{bmatrix} \begin{bmatrix} v_1 h \\ v_2 h \\ v_3 h \\ \vdots \end{bmatrix} = \begin{bmatrix} \sigma_1 v_1 h \\ \sigma_2 v_2 h \\ \sigma_3 v_3 h \\ \vdots \end{bmatrix}$

In Practice:

6

1. Compute SVD of A .
2. Find index of smallest σ_i in Σ
3. Take the i th column of V (i th row of V^T) as solution h .