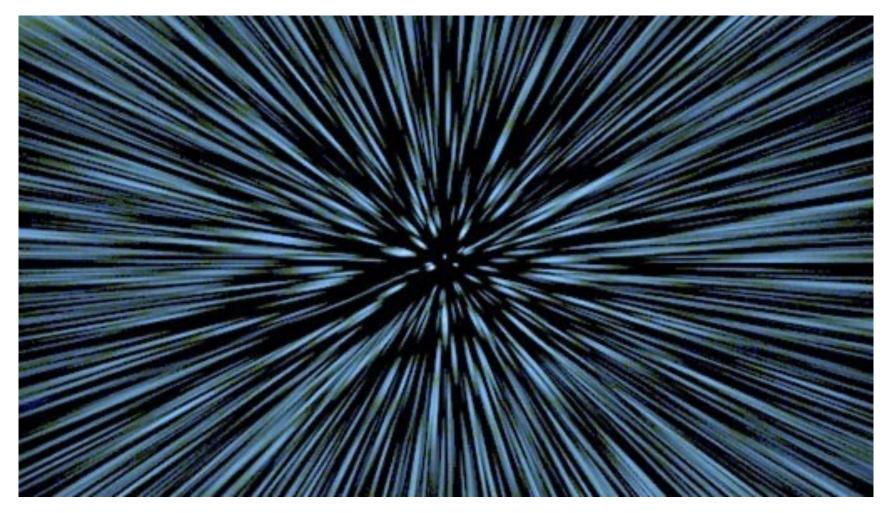
2D transformations (a.k.a. warping)



16-385 Computer Vision Fall 2023, Lecture 7

http://16385.courses.cs.cmu.edu/

Overview of today's lecture

- Reminder: image transformations.
- 2D transformations.
- Projective geometry 101.
- Transformations in projective geometry.
- Classification of 2D transformations.
- Determining unknown 2D transformations.
- Determining unknown image warps.

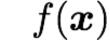
Slide credits

Most of these slides were adapted from:

• Kris Kitani (16-385, Spring 2017).

Reminder: image transformations

What is an image?

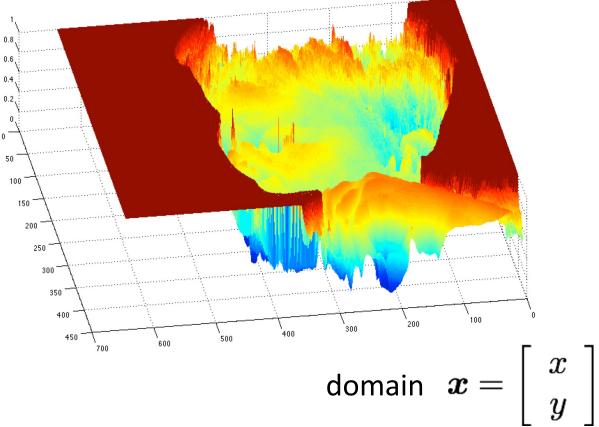


0.8



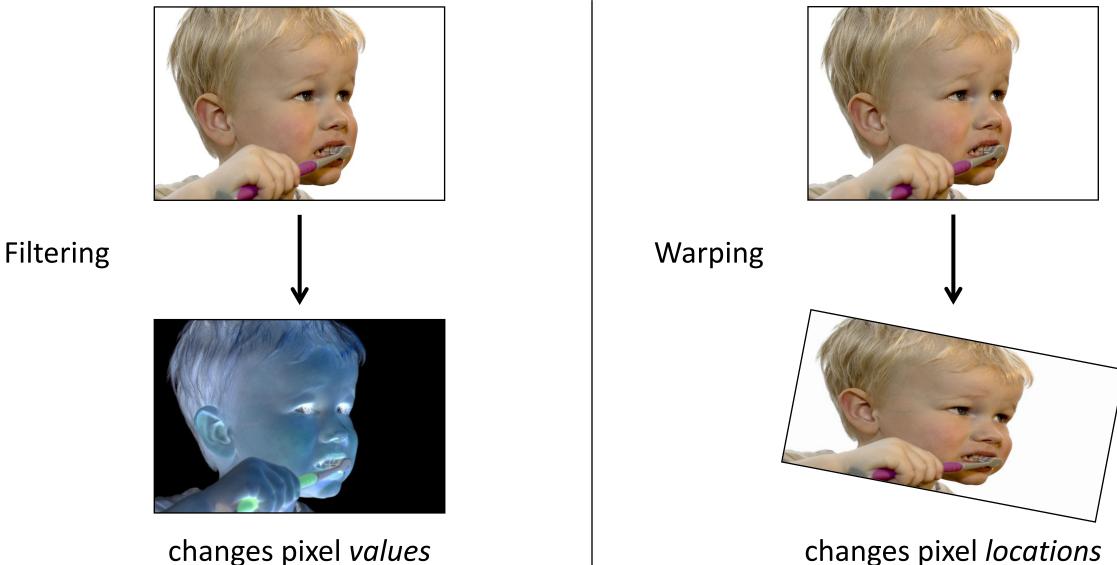
grayscale image

What is the range of the image function f?



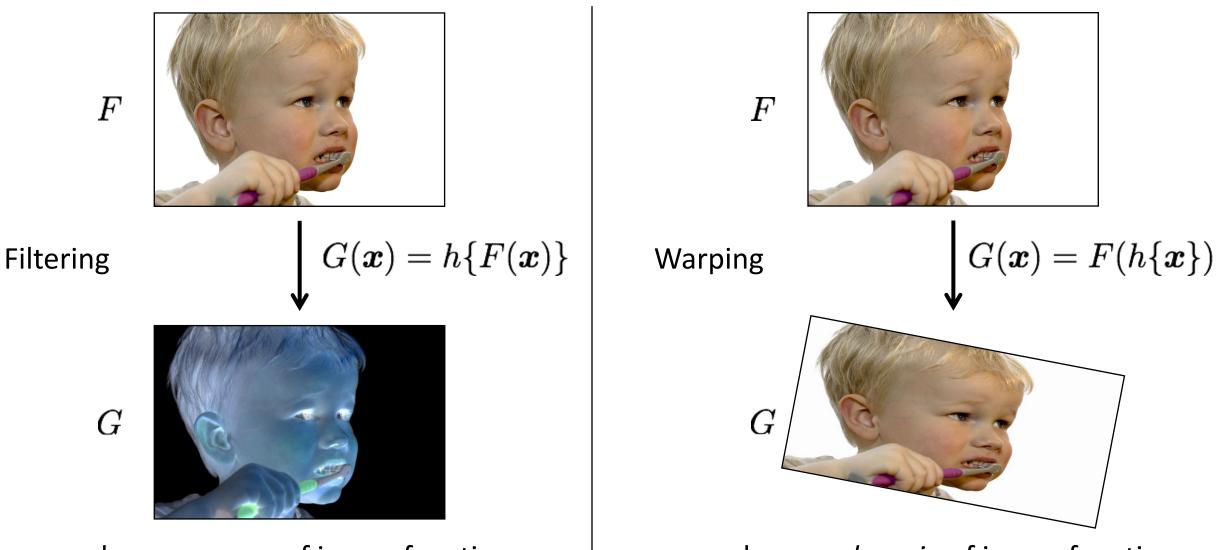
A (grayscale) image is a 2D function.

What types of image transformations can we do?



changes pixel *locations*

What types of image transformations can we do?

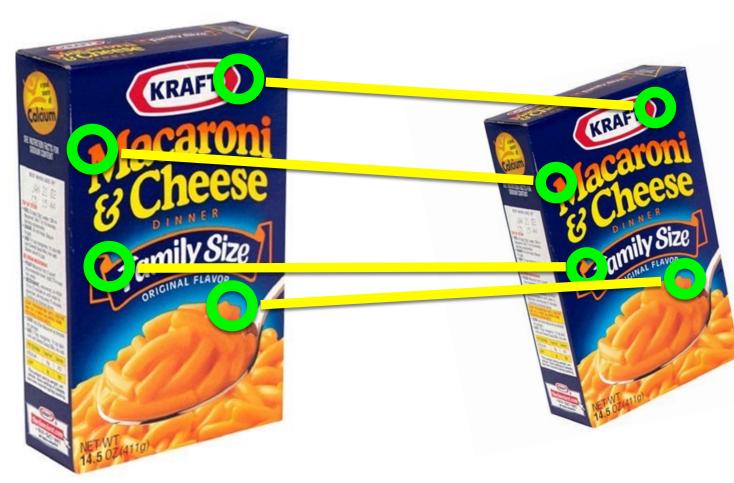


changes *range* of image function

changes domain of image function



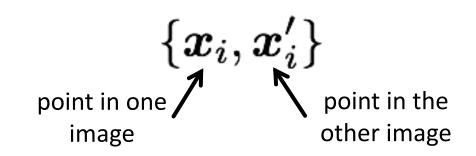




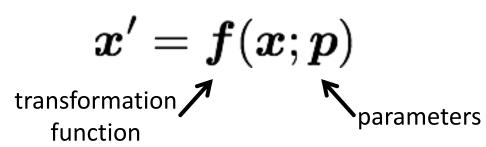
- object recognition
- 3D reconstruction
- augmented reality
- image stitching

How do you compute the transformation?

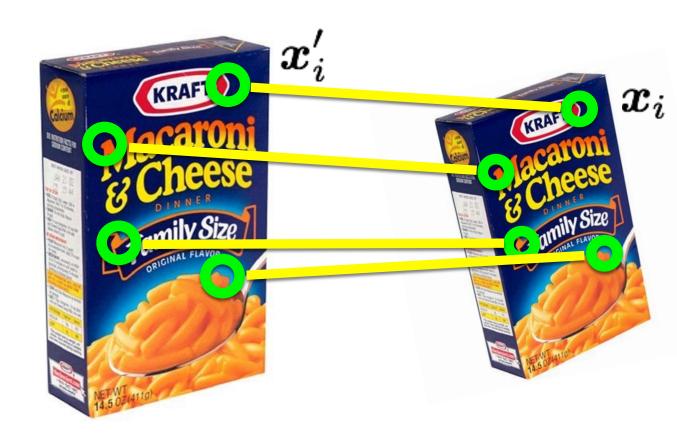
Given a set of matched feature points:



and a transformation:



find the best estimate of the parameters



What kind of transformation functions f are there?

2D transformations

2D transformations







translation

rotation

aspect

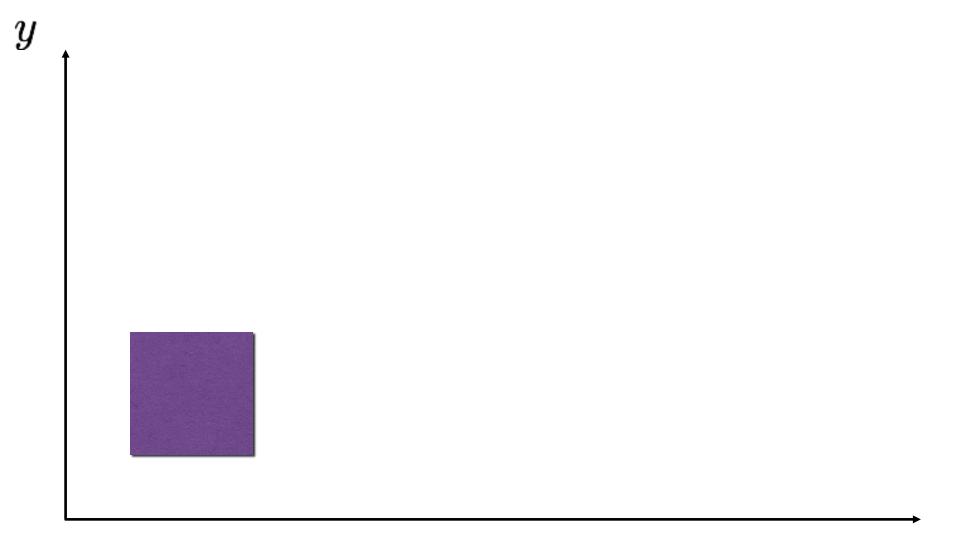


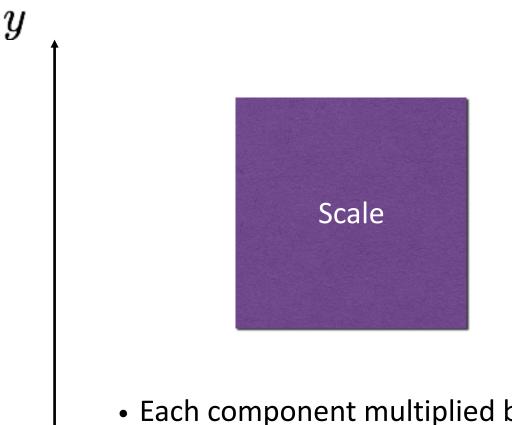




cylindrical

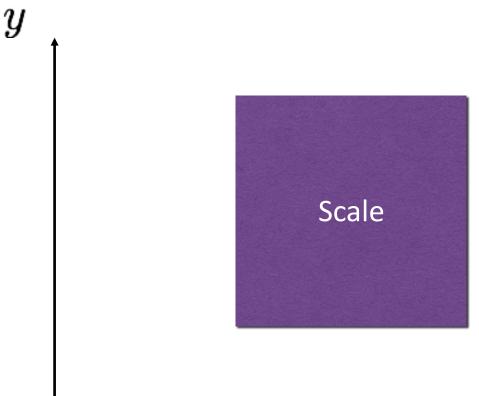
perspective





How would you implement scaling?

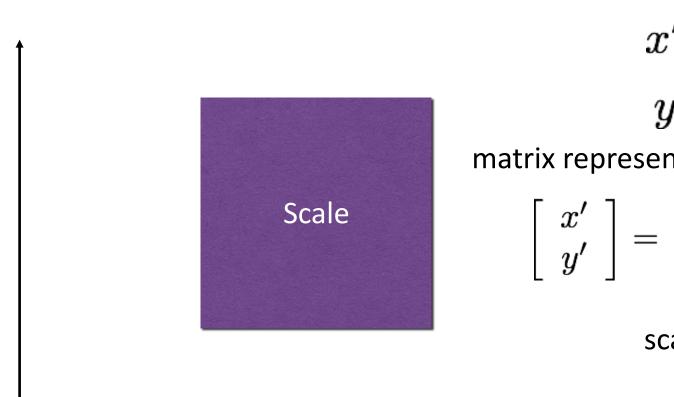
- Each component multiplied by a scalar
- Uniform scaling same scalar for each component



$$\begin{aligned} x' &= ax\\ y' &= by \end{aligned}$$

What's the effect of using different scale factors?

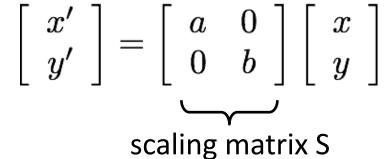
- Each component multiplied by a scalar
- Uniform scaling same scalar for each component



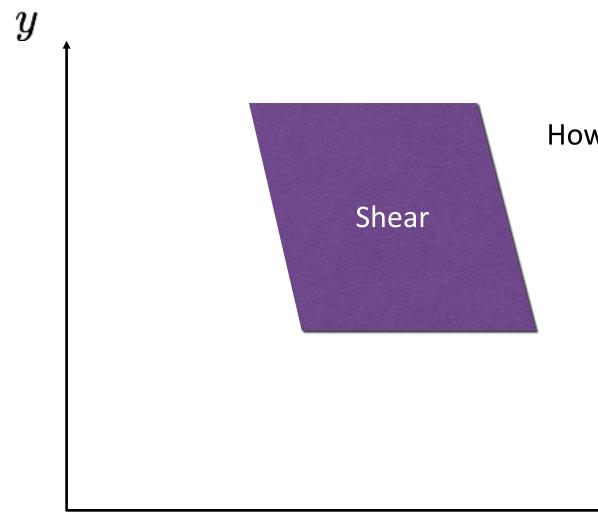
y

 $\begin{aligned} x' &= ax \\ y' &= by \end{aligned}$

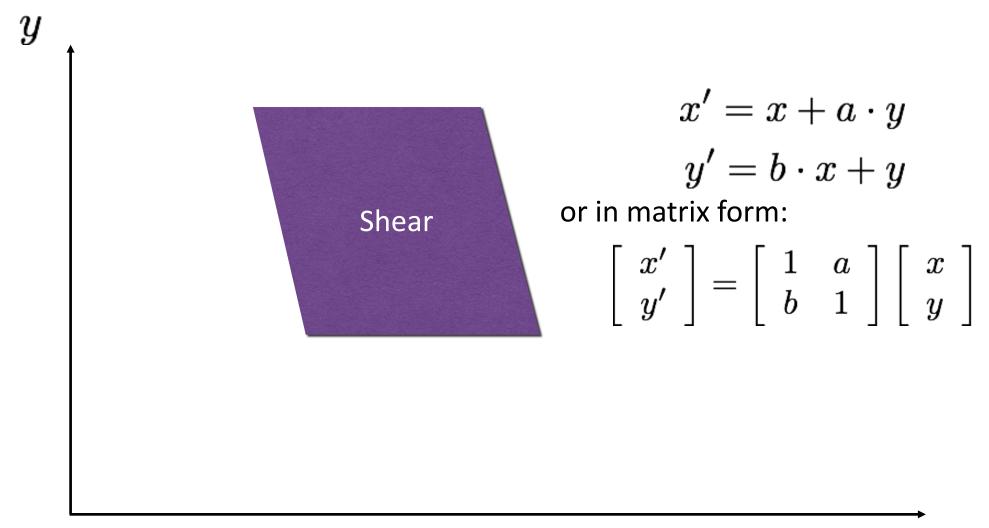
matrix representation of scaling:

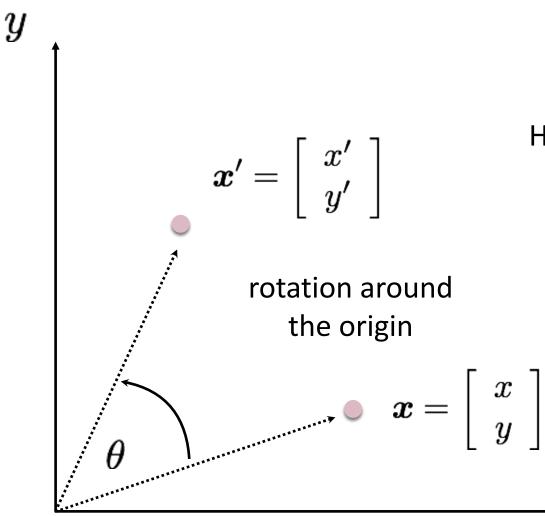


- Each component multiplied by a scalar
- Uniform scaling same scalar for each component

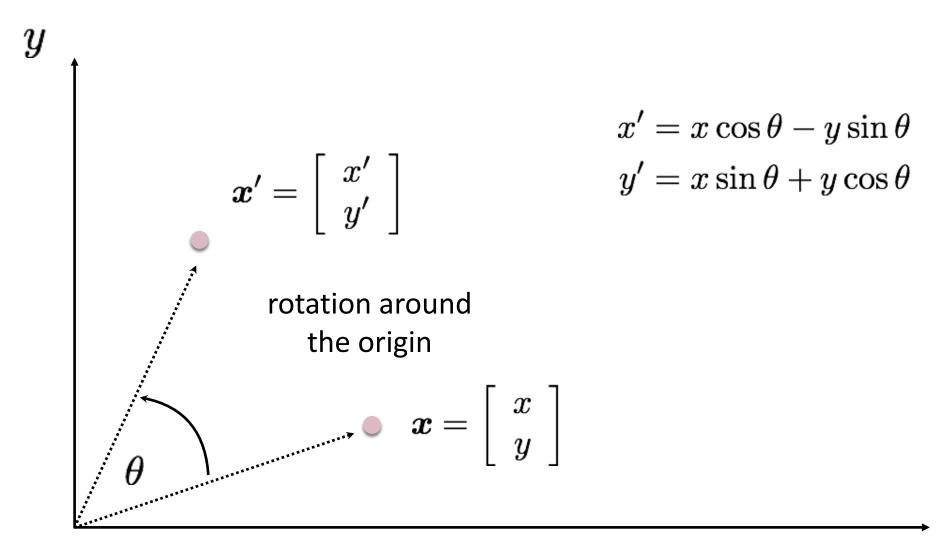


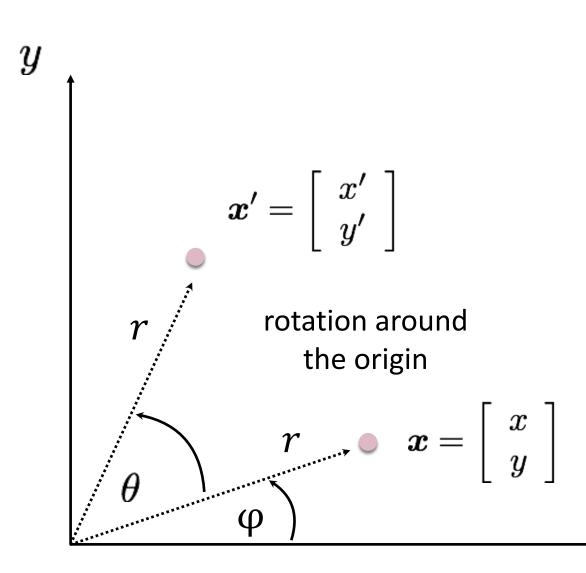
How would you implement shearing?





How would you implement rotation?



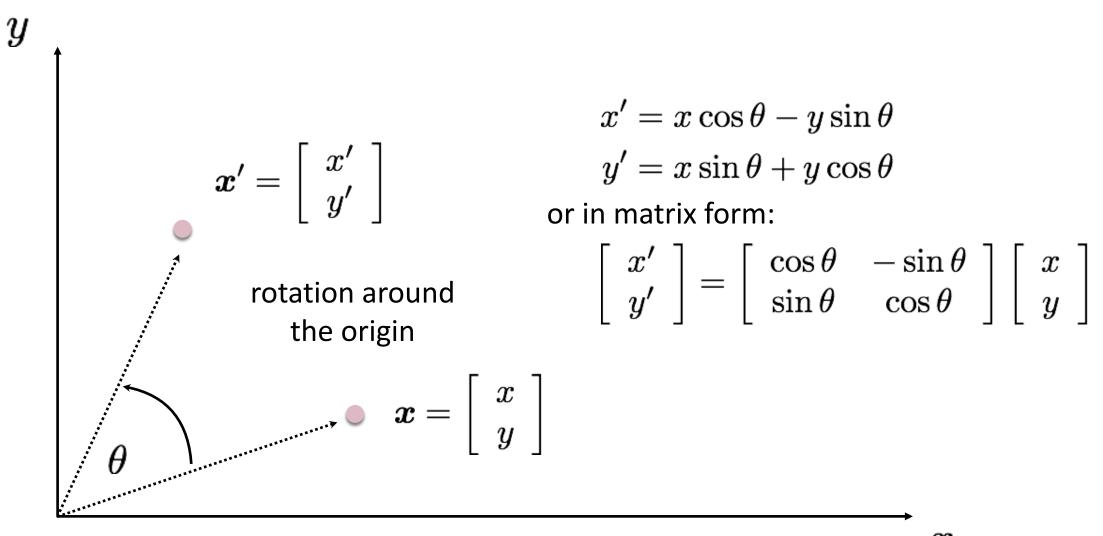


Polar coordinates... $x = r \cos (\phi)$ $y = r \sin (\phi)$ $x' = r \cos (\phi + \theta)$ $y' = r \sin (\phi + \theta)$

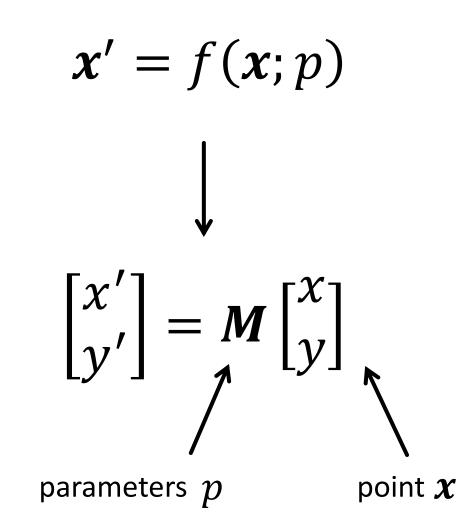
Trigonometric Identity...

 $\begin{aligned} x' &= r \cos(\phi) \cos(\theta) - r \sin(\phi) \sin(\theta) \\ y' &= r \sin(\phi) \cos(\theta) + r \cos(\phi) \sin(\theta) \end{aligned}$

Substitute... $x' = x \cos(\theta) - y \sin(\theta)$ $y' = x \sin(\theta) + y \cos(\theta)$

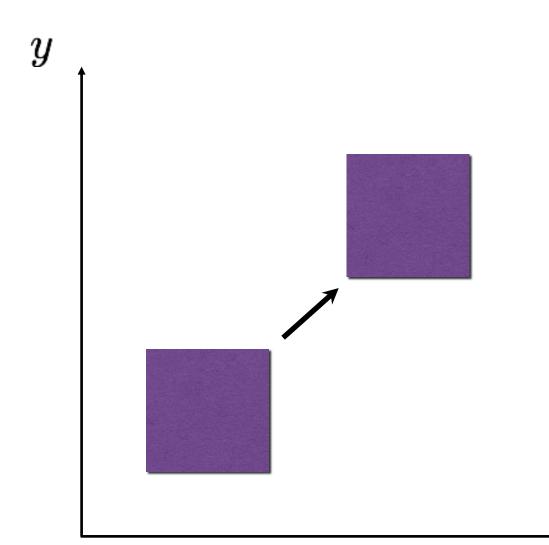


2D planar and linear transformations

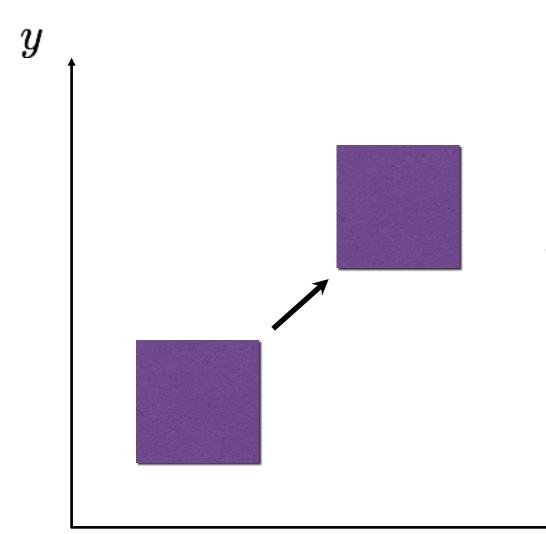


2D planar and linear transformations

Flip across y Scale $\mathbf{M} = \begin{bmatrix} s_x & 0\\ 0 & s_y \end{bmatrix} \qquad \qquad \mathbf{M} = \begin{bmatrix} -1 & 0\\ 0 & 1 \end{bmatrix}$ Rotate Flip across origin $\mathbf{M} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \qquad \mathbf{M} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}$ Shear Identity $\mathbf{M} = \begin{bmatrix} 1 & s_x \\ s_y & 1 \end{bmatrix} \qquad \qquad \mathbf{M} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$



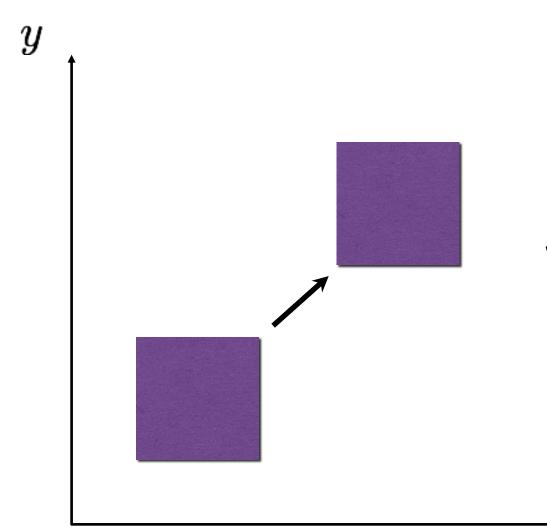
How would you implement translation?



$$x' = x + t_x$$
$$y' = y + t_x$$

What about matrix representation?

$$\mathbf{M} = \left[\begin{array}{cc} ? & ? \\ ? & ? \end{array} \right]$$



$$x' = x + t_x$$
$$y' = y + t_x$$

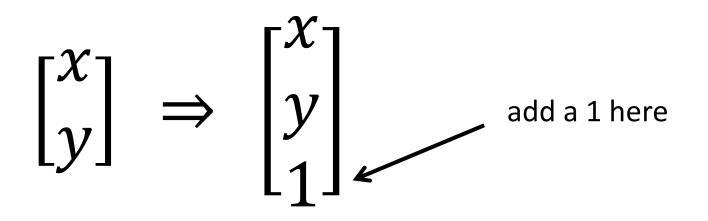
What about matrix representation?

Not possible.

Projective geometry 101

Homogeneous coordinates

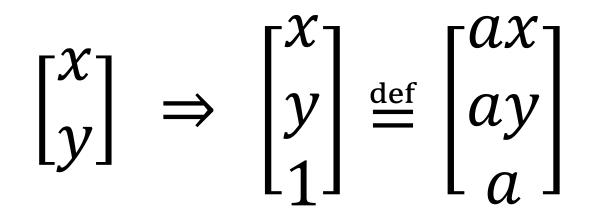
heterogeneous homogeneous coordinates coordinates



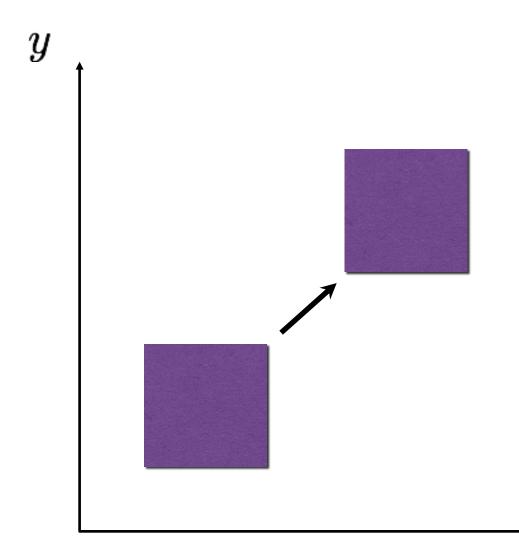
• Represent 2D point with a 3D vector

Homogeneous coordinates

heterogeneous homogeneous coordinates coordinates

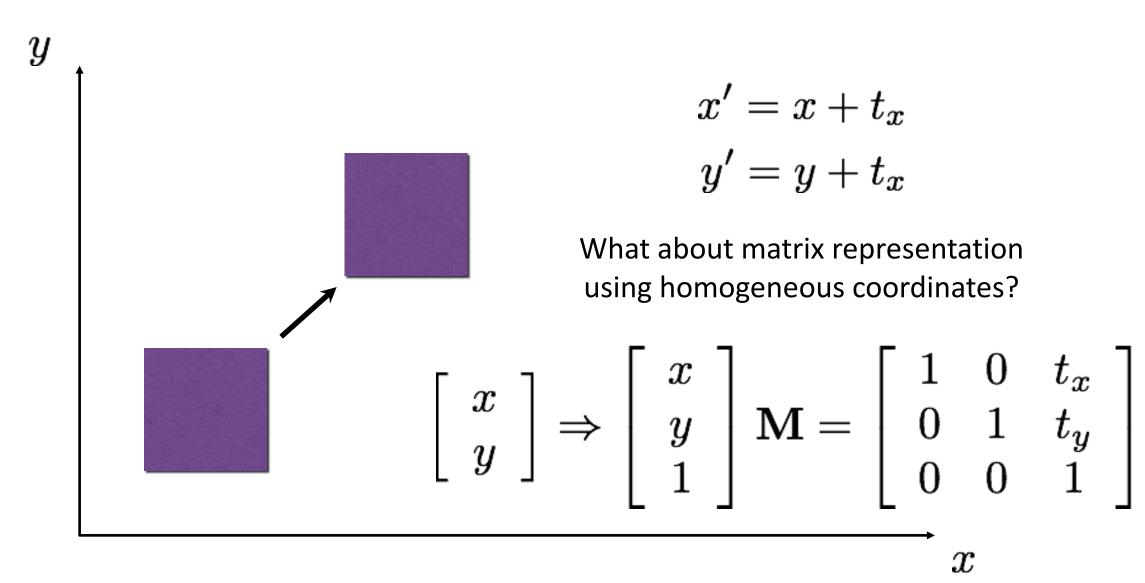


- Represent 2D point with a 3D vector
- 3D vectors are only defined up to scale



$$x' = x + t_x$$
$$y' = y + t_x$$

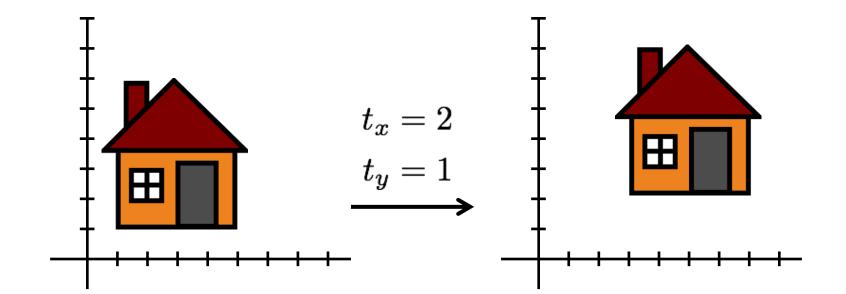
What about matrix representation using homogeneous coordinates?



x

2D translation using homogeneous coordinates

$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} x + t_x \\ y + t_y \\ 1 \end{bmatrix}$$



Homogeneous coordinates

Conversion:

heterogeneous → homogeneous

 $\left[\begin{array}{c} x\\ y\end{array}\right] \Rightarrow \left[\begin{array}{c} x\\ y\\ 1\end{array}\right]$

- homogeneous \rightarrow heterogeneous

$$\left[\begin{array}{c} x\\ y\\ w \end{array}\right] \Rightarrow \left[\begin{array}{c} x/w\\ y/w \end{array}\right]$$

• scale invariance

$$\begin{bmatrix} x & y & w \end{bmatrix}^{\top} = \lambda \begin{bmatrix} x & y & w \end{bmatrix}^{\top}$$

Special points:

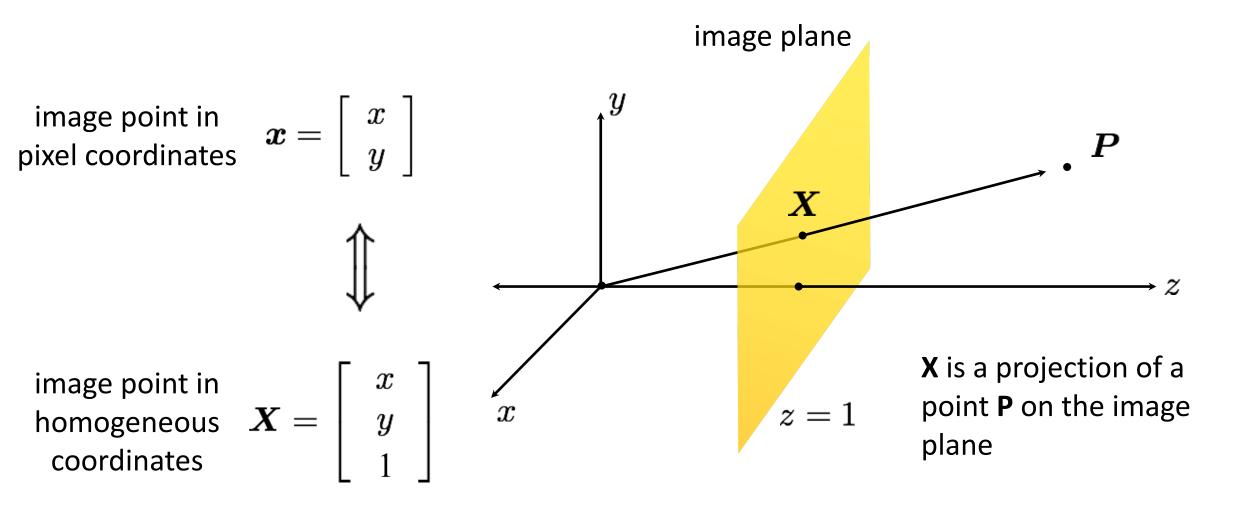
• point at infinity

$$\left[egin{array}{ccc} x & y & 0 \end{array}
ight]$$

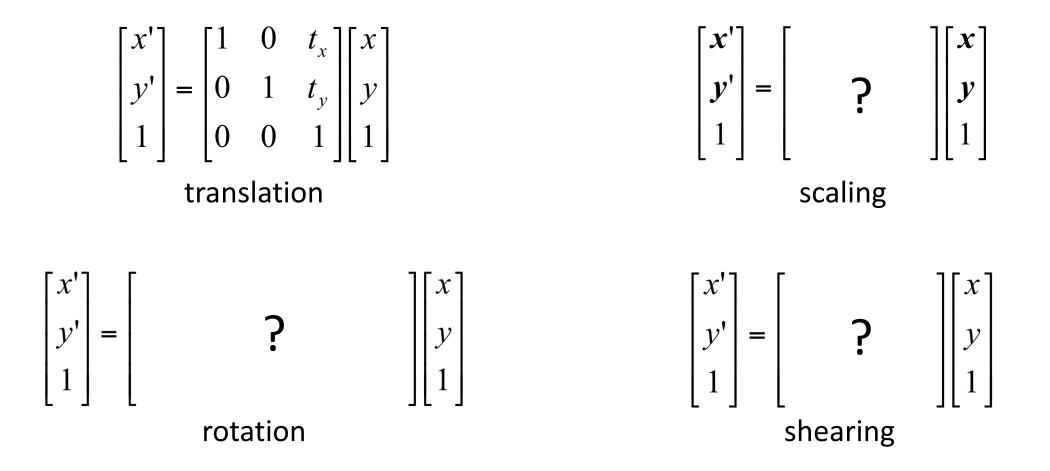
undefined

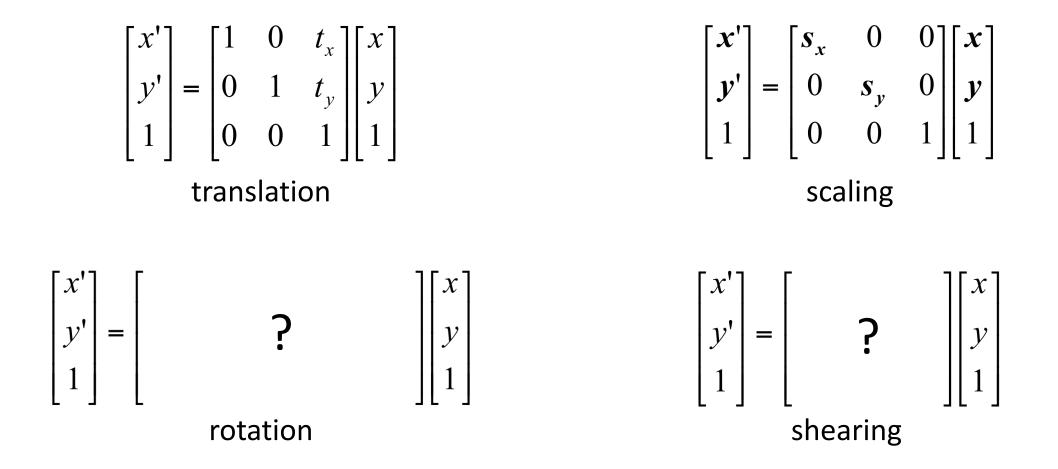
$$\left[\begin{array}{ccc} 0 & 0 & 0 \end{array}\right]$$

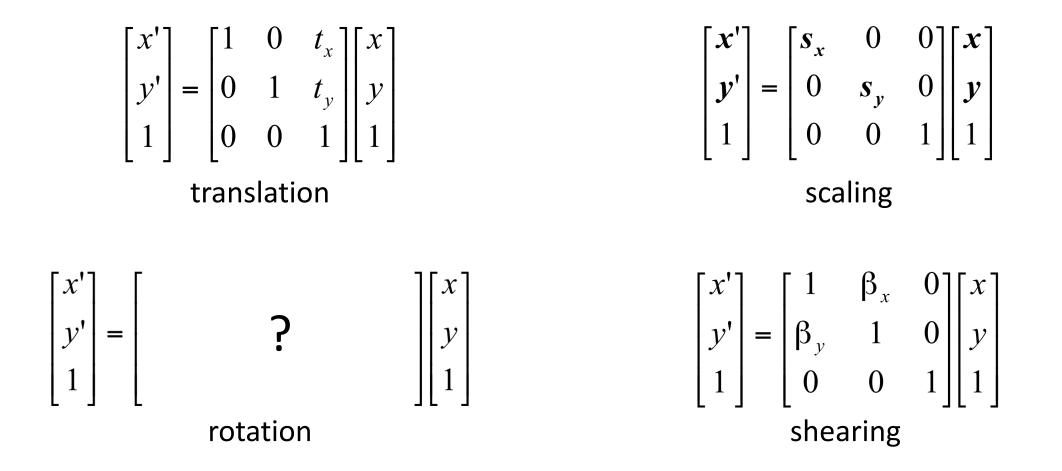
Projective geometry

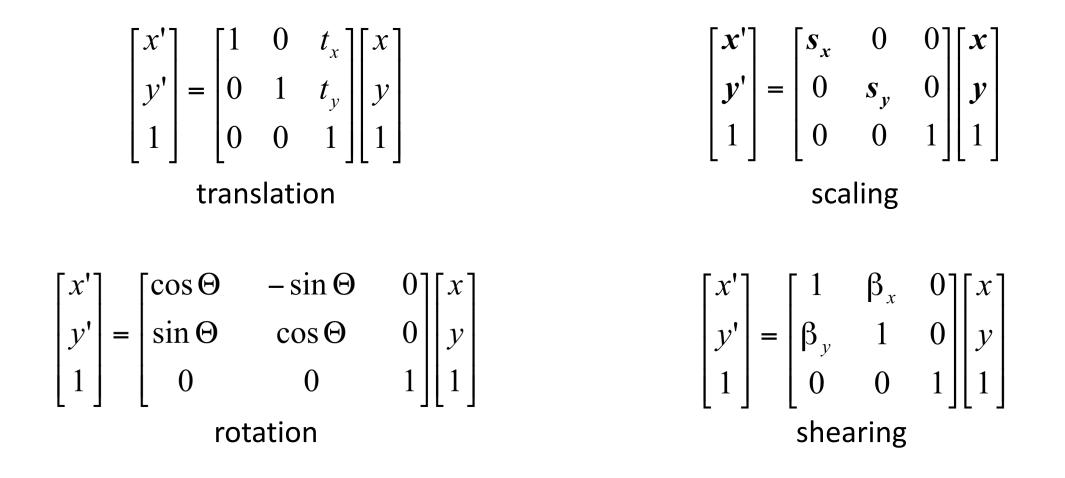


Transformations in projective geometry









Matrix composition

Transformations can be combined by matrix multiplication:

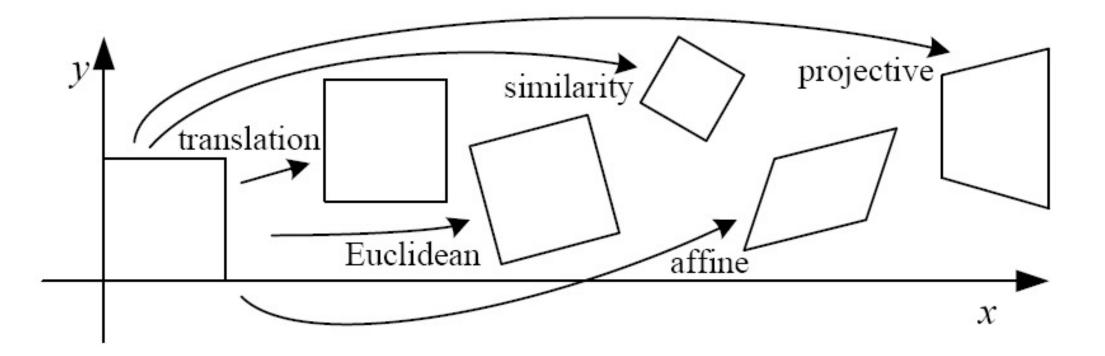
$$\begin{bmatrix} x'\\y'\\w' \end{bmatrix} = \begin{pmatrix} \begin{bmatrix} 1 & 0 & tx\\0 & 1 & ty\\0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos\Theta & -\sin\Theta & 0\\\sin\Theta & \cos\Theta & 0\\0 & 0 & 1 \end{bmatrix} \begin{bmatrix} sx & 0 & 0\\0 & sy & 0\\0 & 0 & 1 \end{bmatrix} \begin{pmatrix} x\\y\\w \end{bmatrix}$$
$$\mathbf{p}' = \mathbf{P} + \mathbf{P} +$$

Matrix composition

Transformations can be combined by matrix multiplication:

$$\begin{bmatrix} x'\\y'\\w' \end{bmatrix} = \begin{pmatrix} \begin{bmatrix} 1 & 0 & tx\\0 & 1 & ty\\0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos\Theta & -\sin\Theta & 0\\\sin\Theta & \cos\Theta & 0\\0 & 0 & 1 \end{bmatrix} \begin{bmatrix} sx & 0 & 0\\0 & sy & 0\\0 & 0 & 1 \end{bmatrix} \begin{pmatrix} x\\y\\w \end{bmatrix}$$
$$\mathbf{p}' = \text{translation}(\mathbf{t}_{x},\mathbf{t}_{y}) \qquad \text{rotation}(\theta) \qquad \text{scale}(s,s) \quad \mathbf{p}$$

Does the multiplication order matter?

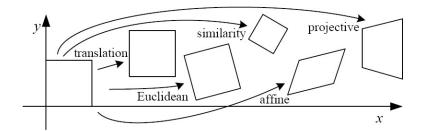


Name	Matrix	# D.O.F.
translation	$\left[egin{array}{c c} I & t \end{array} ight]$?
rigid (Euclidean)	$\left[egin{array}{c c} R & t \end{array} ight]$?
similarity	$\left[\left. s \boldsymbol{R} \right \boldsymbol{t} \right]$?
affine	$\begin{bmatrix} A \end{bmatrix}$?
projective	$\left[egin{array}{c} ilde{H} \end{array} ight]$?

Translation:

$$\left[\begin{array}{rrrr} 1 & 0 & t_1 \\ 0 & 1 & t_2 \\ 0 & 0 & 1 \end{array} \right]$$

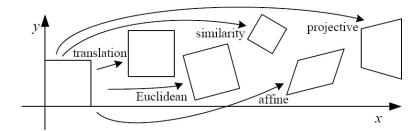
How many degrees of freedom?



Euclidean (rigid): rotation + translation

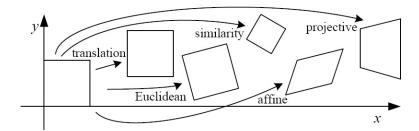
$$\left[\begin{array}{rrrr} r_1 & r_2 & r_3 \\ r_4 & r_5 & r_6 \\ 0 & 0 & 1 \end{array}\right]$$

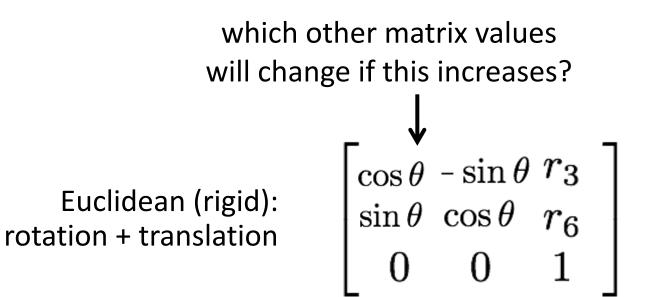
Are there any values that are related?

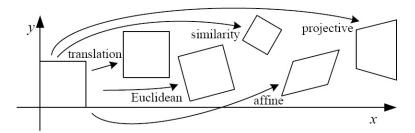


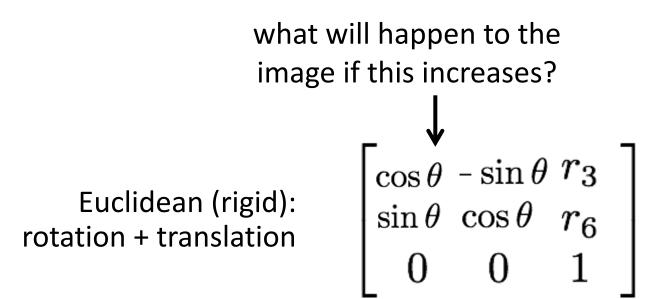
Euclidean (rigid): rotation + translation $\begin{bmatrix} \cos\theta & -\sin\theta & r_3 \\ \sin\theta & \cos\theta & r_6 \\ 0 & 0 & 1 \end{bmatrix}$

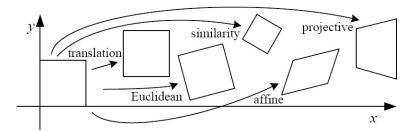
How many degrees of freedom?











 $\begin{bmatrix} \cos\theta & -\sin\theta & r_3 \\ \sin\theta & \cos\theta & r_6 \\ 0 & 0 & 1 \end{bmatrix}$

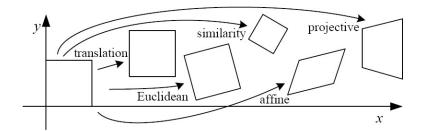
what will happen to the image if this increases?

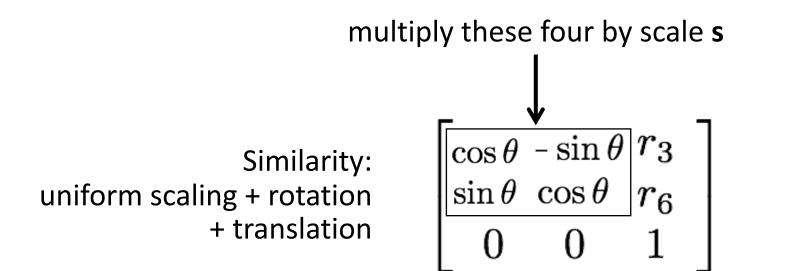
Euclidean (rigid): rotation + translation

y:
$$\begin{bmatrix} r_1 & r_2 & r_3 \\ r_4 & r_5 & r_6 \\ 0 & 0 & 1 \end{bmatrix}$$

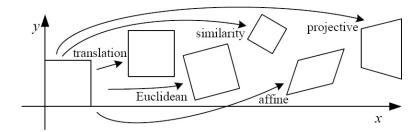
Similarity uniform scaling + rotatior + translatior

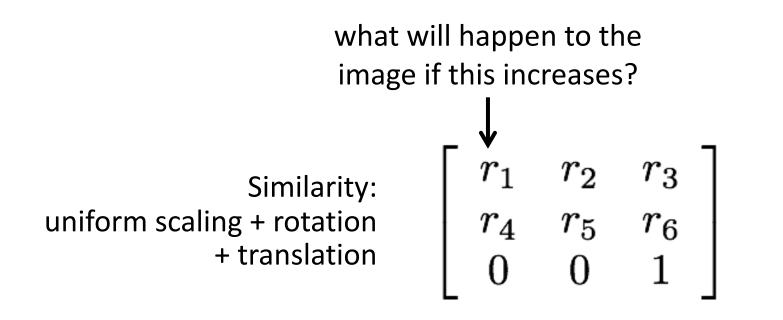
Are there any values that are related?

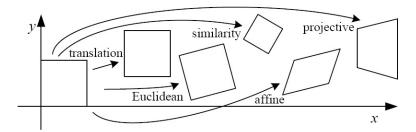




How many degrees of freedom?



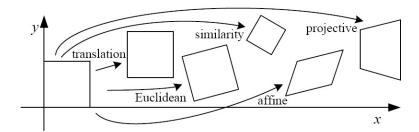




Affine transform: uniform scaling + shearing + rotation + translation

 $\left[\begin{array}{rrrrr}a_1 & a_2 & a_3\\ a_4 & a_5 & a_6\\ 0 & 0 & 1\end{array}\right]$

Are there any values that are related?



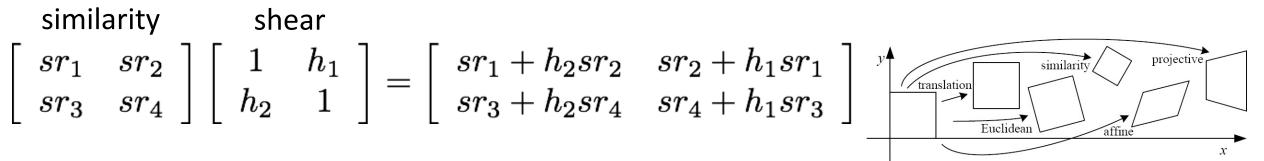
Affine transform:
uniform scaling + shearing
+ rotation + translation
$$\begin{bmatrix} a_1 & a_2 & a_3 \\ a_4 & a_5 & a_6 \\ 0 & 0 & 1 \end{bmatrix}$$

Are there any values that are related?

$$\begin{bmatrix} similarity & shear \\ sr_1 & sr_2 \\ sr_3 & sr_4 \end{bmatrix} \begin{bmatrix} 1 & h_1 \\ h_2 & 1 \end{bmatrix} = \begin{bmatrix} sr_1 + h_2 sr_2 & sr_2 + h_1 sr_1 \\ sr_3 + h_2 sr_4 & sr_4 + h_1 sr_3 \end{bmatrix} \xrightarrow{y^4}_{\text{translation}} \xrightarrow{\text{similarity}}_{\text{Euclidean}} \xrightarrow{\text{projective}}_{\text{fine}} \xrightarrow{x}_{\text{translation}}$$

Affine transform:
uniform scaling + shearing
+ rotation + translation
$$\begin{bmatrix} a_1 & a_2 & a_3 \\ a_4 & a_5 & a_6 \\ 0 & 0 & 1 \end{bmatrix}$$

How many degrees of freedom?



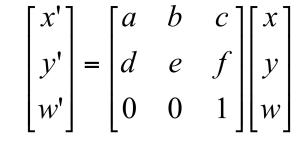
Affine transformations

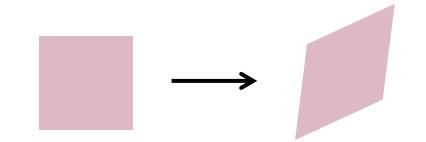
Affine transformations are combinations of

- arbitrary (4-DOF) linear transformations; and
- translations

Properties of affine transformations:

- origin does not necessarily map to origin
- lines map to lines
- parallel lines map to parallel lines
- ratios are preserved
- compositions of affine transforms are also affine transforms





Does the last coordinate w ever change?

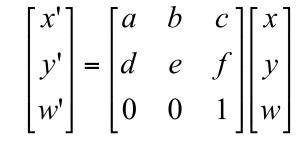
Affine transformations

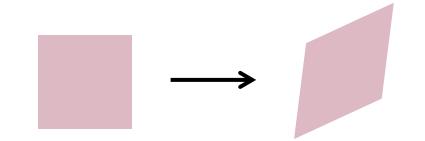
Affine transformations are combinations of

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Properties of affine transformations:

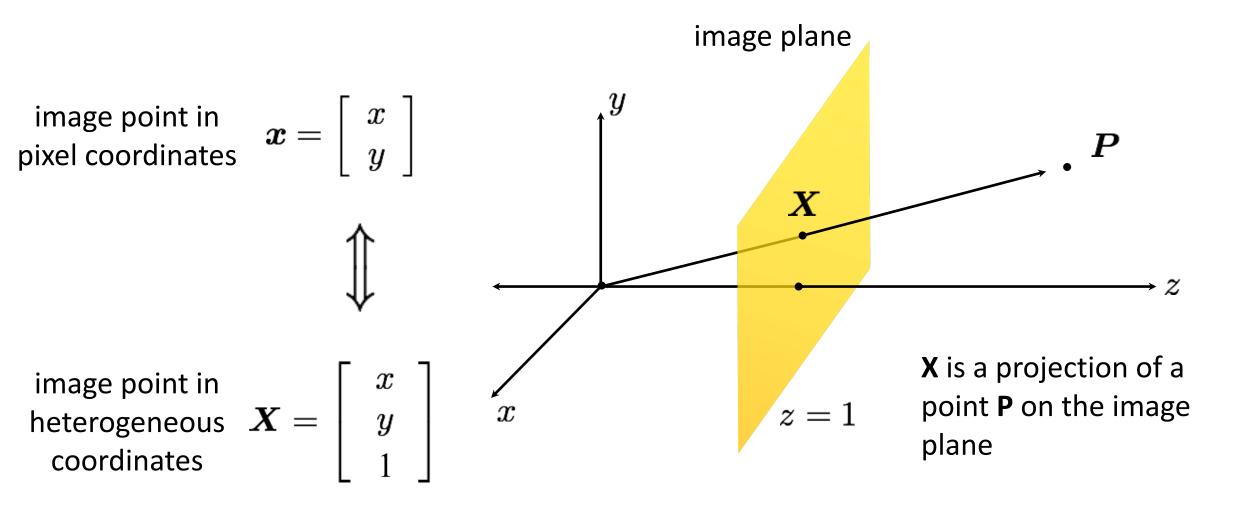
- origin does not necessarily map to origin
- lines map to lines
- parallel lines map to parallel lines
- ratios are preserved
- compositions of affine transforms are also affine transforms





Nope! But what does that mean?

How to interpret affine transformations here?



Projective transformations (aka homographies)

Projective transformations are combinations of

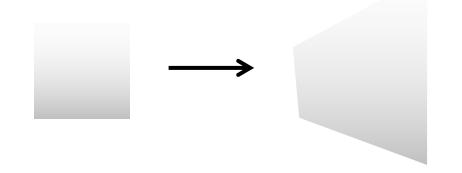
- affine transformations; and
- projective wraps

Properties of projective transformations:

- origin does not necessarily map to origin
- lines map to lines
- parallel lines do not necessarily map to parallel lines
- ratios are not necessarily preserved
- compositions of projective transforms are also projective transforms

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

How many degrees of freedom?



Projective transformations (aka homographies)

Projective transformations are combinations of

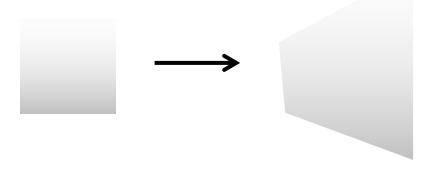
- affine transformations; and
- projective wraps

Properties of projective transformations:

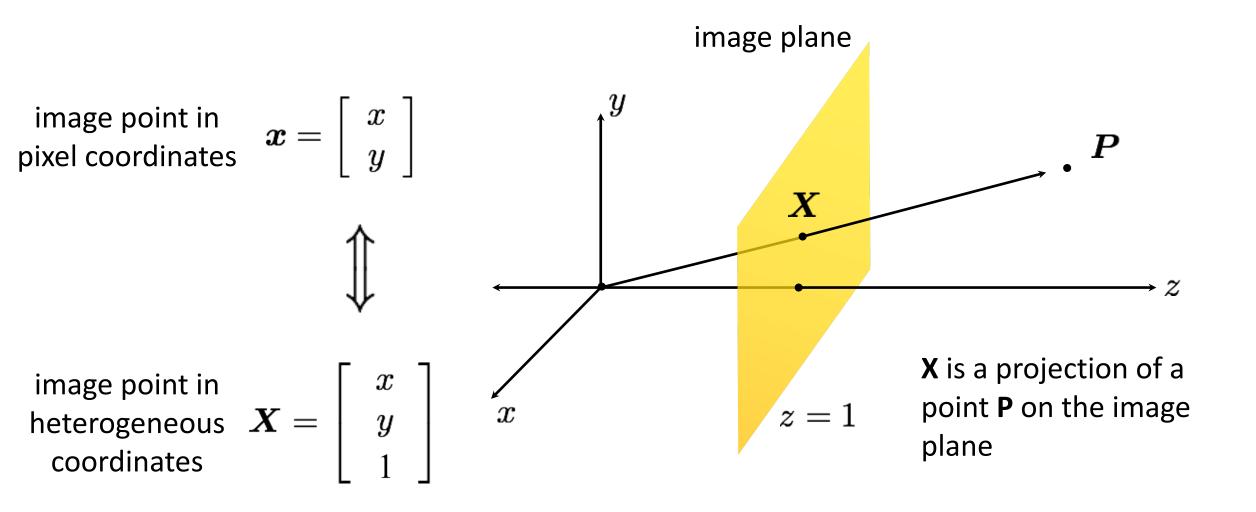
- origin does not necessarily map to origin
- lines map to lines
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- ratios are not necessarily preserved
- compositions of projective transforms are also projective transforms

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

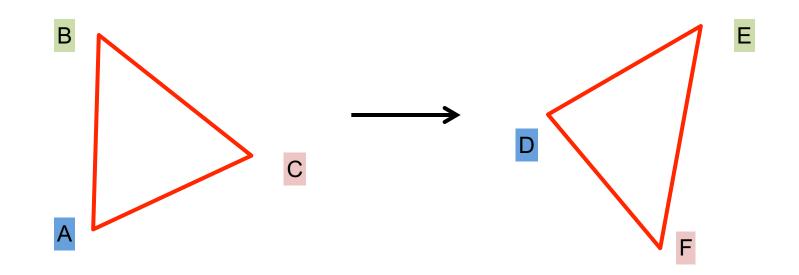
8 DOF: vectors (and therefore matrices) are defined up to scale)



How to interpret projective transformations here?

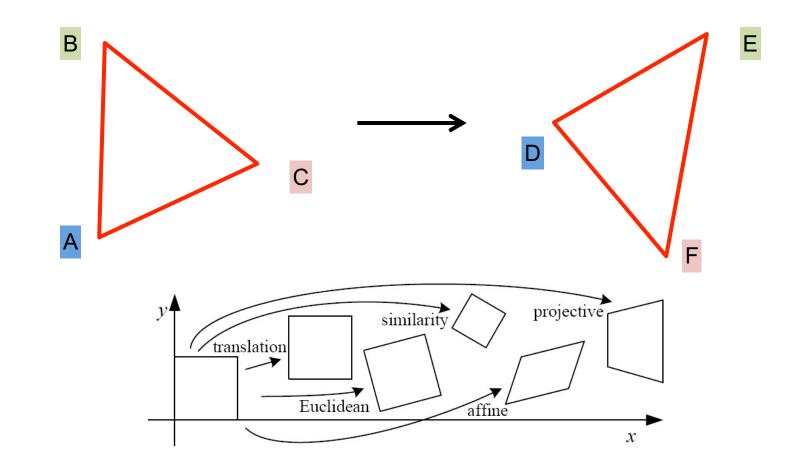


Determining unknown (affine) 2D transformations

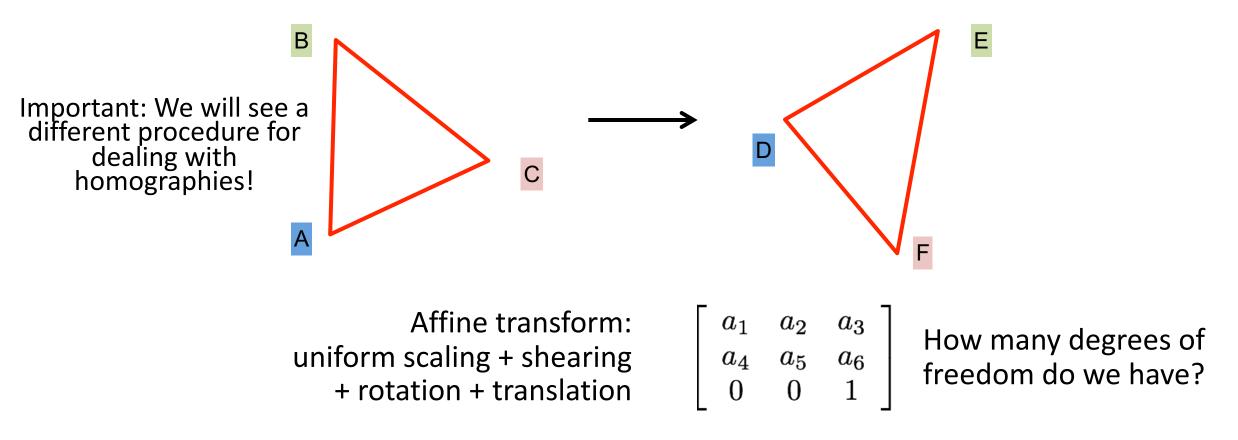


Suppose we have two triangles: ABC and DEF.

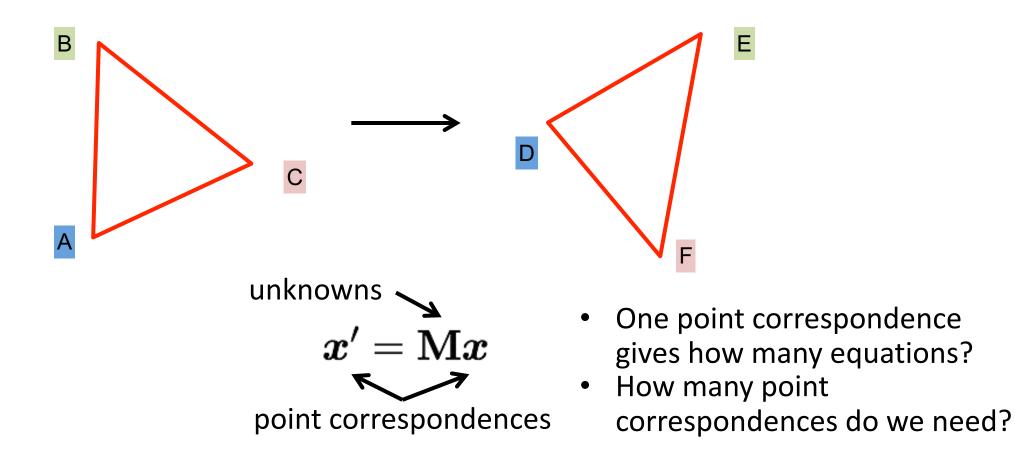
• What type of transformation will map A to D, B to E, and C to F?



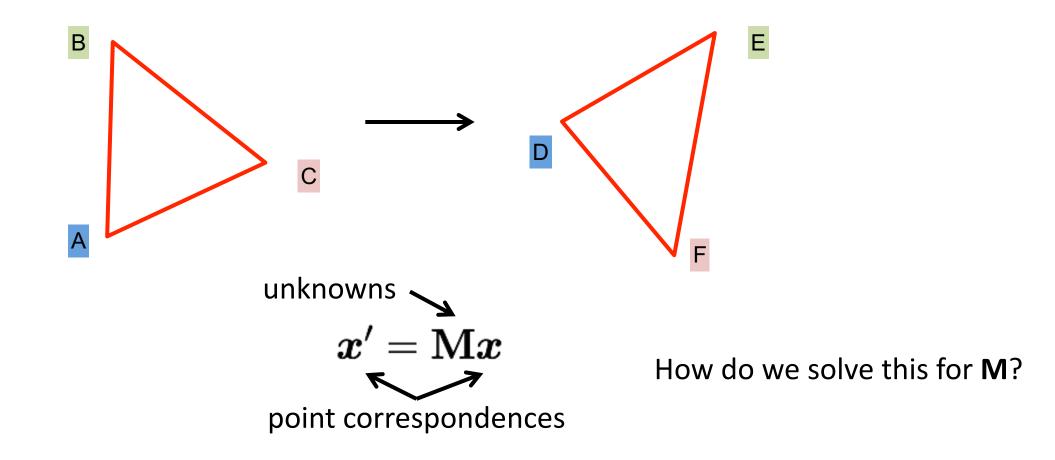
- What type of transformation will map A to D, B to E, and C to F?
- How do we determine the unknown parameters?

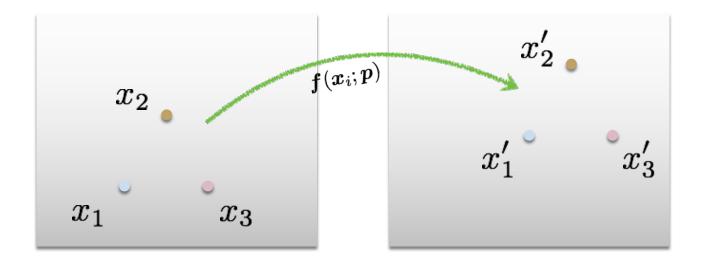


- What type of transformation will map A to D, B to E, and C to F?
- How do we determine the unknown parameters?



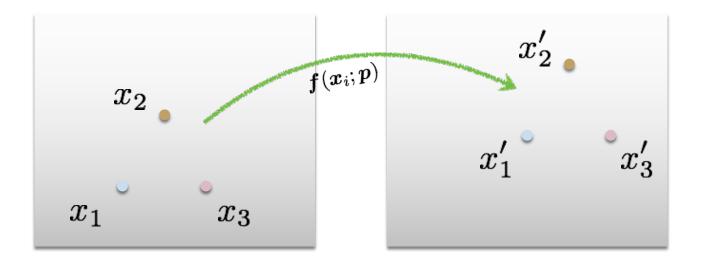
- What type of transformation will map A to D, B to E, and C to F?
- How do we determine the unknown parameters?

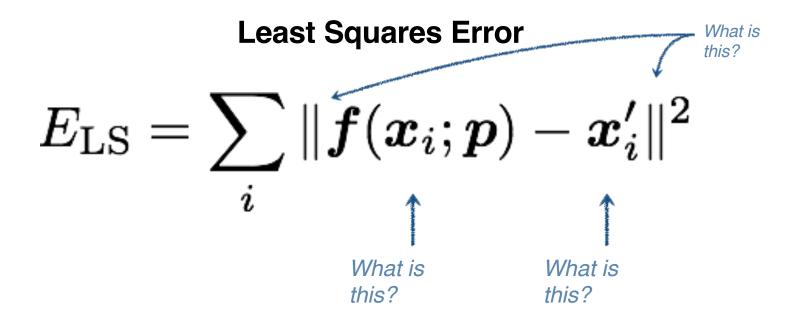


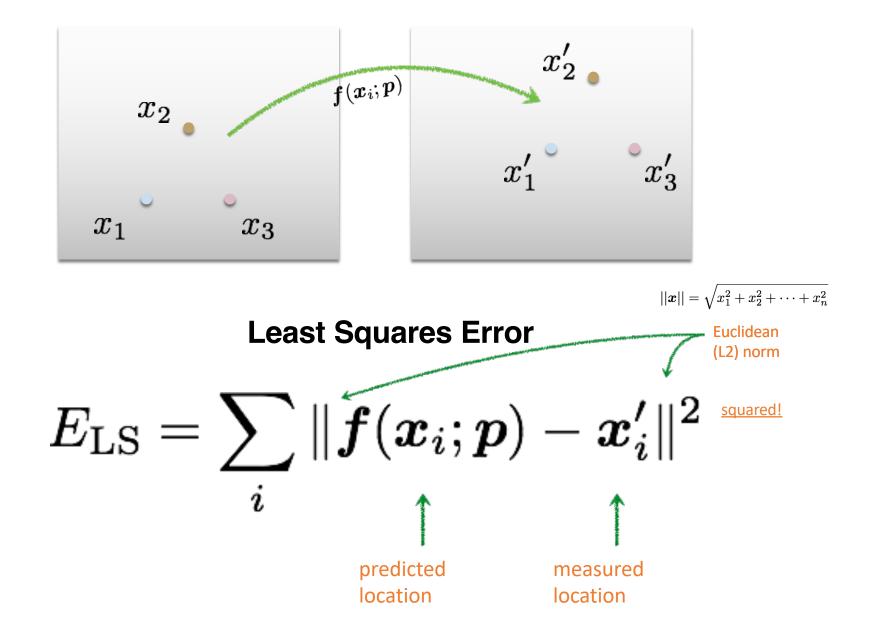


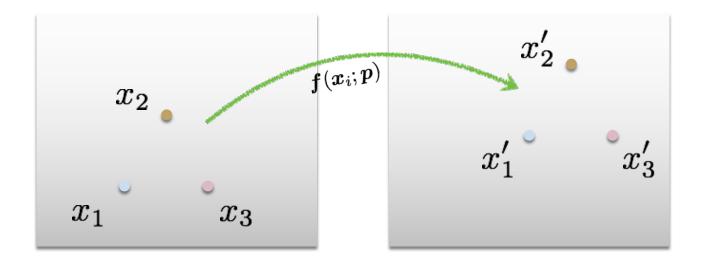
Least Squares Error

$$E_{\rm LS} = \sum_i \|\boldsymbol{f}(\boldsymbol{x}_i; \boldsymbol{p}) - \boldsymbol{x}'_i\|^2$$

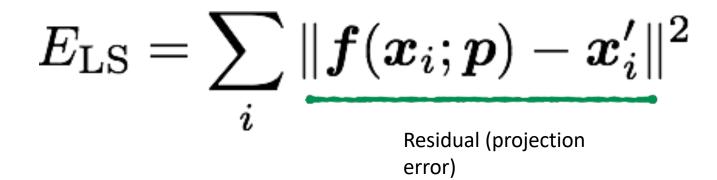


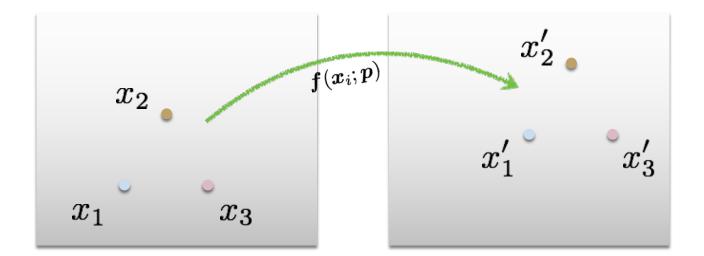






Least Squares Error

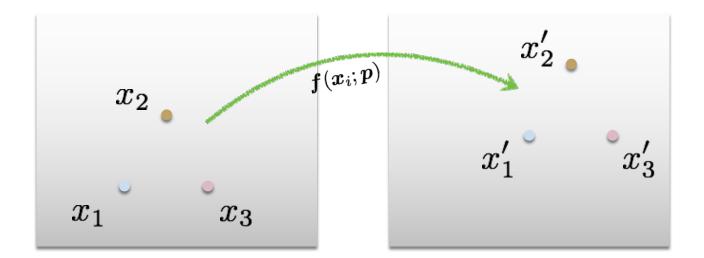




Least Squares Error

$$E_{\rm LS} = \sum_i \|\boldsymbol{f}(\boldsymbol{x}_i; \boldsymbol{p}) - \boldsymbol{x}'_i\|^2$$

What do we want to optimize?



Find parameters that minimize squared error

$$\hat{oldsymbol{p}} = rgmin_{oldsymbol{p}} \sum_{oldsymbol{i}} \|oldsymbol{f}(oldsymbol{x}_i;oldsymbol{p}) - oldsymbol{x}_i'\|^2$$

General form of linear least squares

(**Warning:** change of notation. x is a vector of parameters!)

$$E_{ ext{LLS}} = \sum_i |oldsymbol{a}_i oldsymbol{x} - oldsymbol{b}_i|^2 \ = \|oldsymbol{A}oldsymbol{x} - oldsymbol{b}\|^2 \quad ext{(matrix form}$$

Affine transformation:

Vectorize transformation parameters:

Stack equations from point correspondences:

Notation in system form:

$$\left[\begin{array}{cc} x' \\ y' \end{array}
ight] = \left[\begin{array}{cc} p_1 & p_2 & p_3 \\ p_4 & p_5 & p_6 \end{array}
ight] \left[\begin{array}{c} x \\ y \\ 1 \end{array}
ight]$$
 Why can we drop the last line?

 $\begin{bmatrix} x' \\ y' \\ x' \\ y' \end{bmatrix} = \begin{bmatrix} x & y & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & x & y & 1 \\ x & y & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & x & y & 1 \end{bmatrix} \begin{bmatrix} p_1 \\ p_2 \\ p_3 \\ p_4 \end{bmatrix}$ p_5 b \boldsymbol{x}

General form of linear least squares

(**Warning:** change of notation. x is a vector of parameters!)

$$E_{ ext{LLS}} = \sum_i |oldsymbol{a}_i oldsymbol{x} - oldsymbol{b}_i|^2 \ = \|oldsymbol{A}oldsymbol{x} - oldsymbol{b}\|^2 \quad ext{(matrix form)}$$

This function is quadratic. *How do you find the root of a quadratic?*

Solving the linear system

Convert the system to a linear least-squares problem:

$$E_{\text{LLS}} = \|\mathbf{A}\boldsymbol{x} - \boldsymbol{b}\|^2$$

Expand the error:

$$E_{\text{LLS}} = \boldsymbol{x}^{\top} (\mathbf{A}^{\top} \mathbf{A}) \boldsymbol{x} - 2 \boldsymbol{x}^{\top} (\mathbf{A}^{\top} \boldsymbol{b}) + \| \boldsymbol{b} \|^2$$

Solve for x $\ oldsymbol{x} = (\mathbf{A}^{ op} \mathbf{A})^{-1} \mathbf{A}^{ op} \mathbf{b}$ $\boldsymbol{\leftarrow}$

Minimize the error:

Set derivative to 0
$$\,\,(\mathbf{A}^{ op}\mathbf{A})oldsymbol{x} = \mathbf{A}^{ op}oldsymbol{b}$$

In Python:

Note: You almost <u>never</u> want to compute the inverse of a matrix.

Linear least squares estimation only works when the transform function is ?

Linear least squares estimation only works when the transform function is linear! (duh)

Also doesn't deal well with outliers