

# **CSCE 448/748 – Computational Photography**

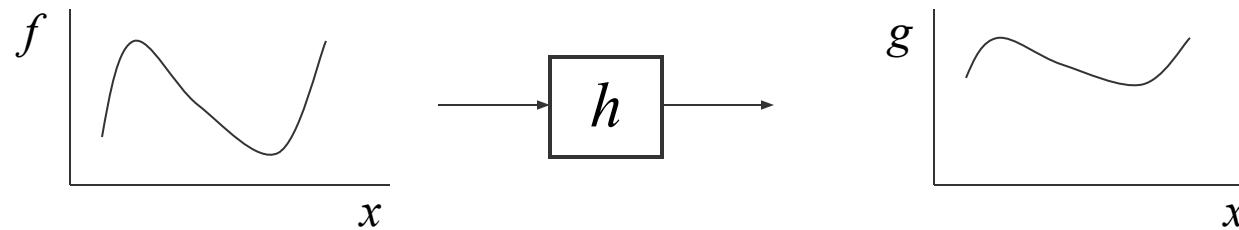
## **Point Processing and Image Warping**

**Nima Kalantari**

# Image Processing

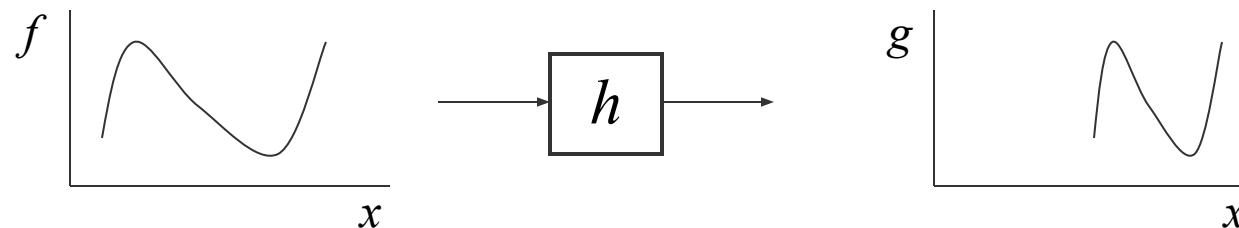
- Point processing: change *range* of image

$$g(x) = h(f(x))$$



- Image warping: change domain of image

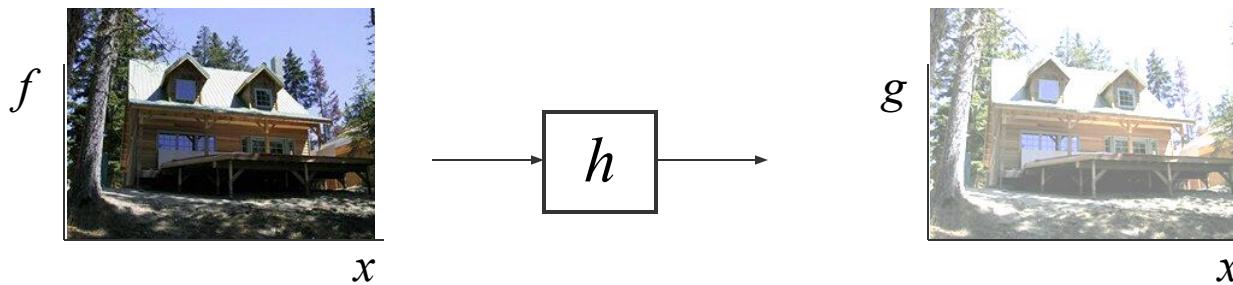
$$g(x) = f(h(x))$$



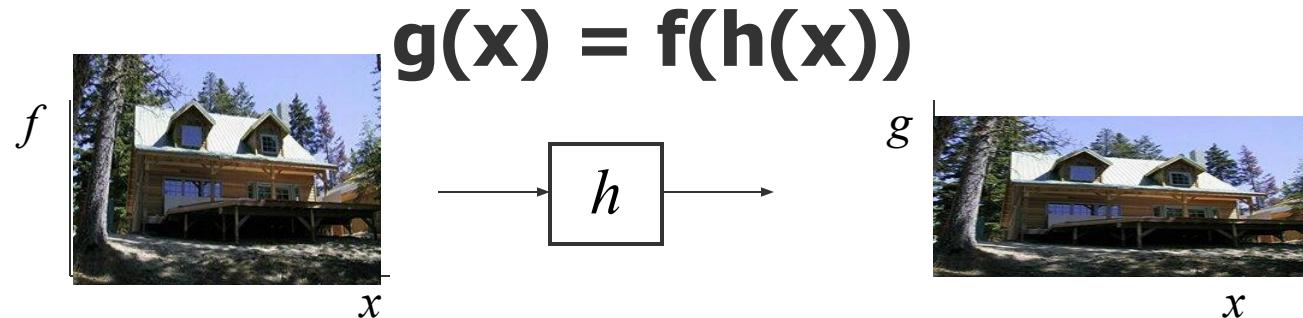
# Image Processing

## □ Point processing: change *range* of image

$$g(x) = h(f(x))$$



## □ Image warping: change domain of image



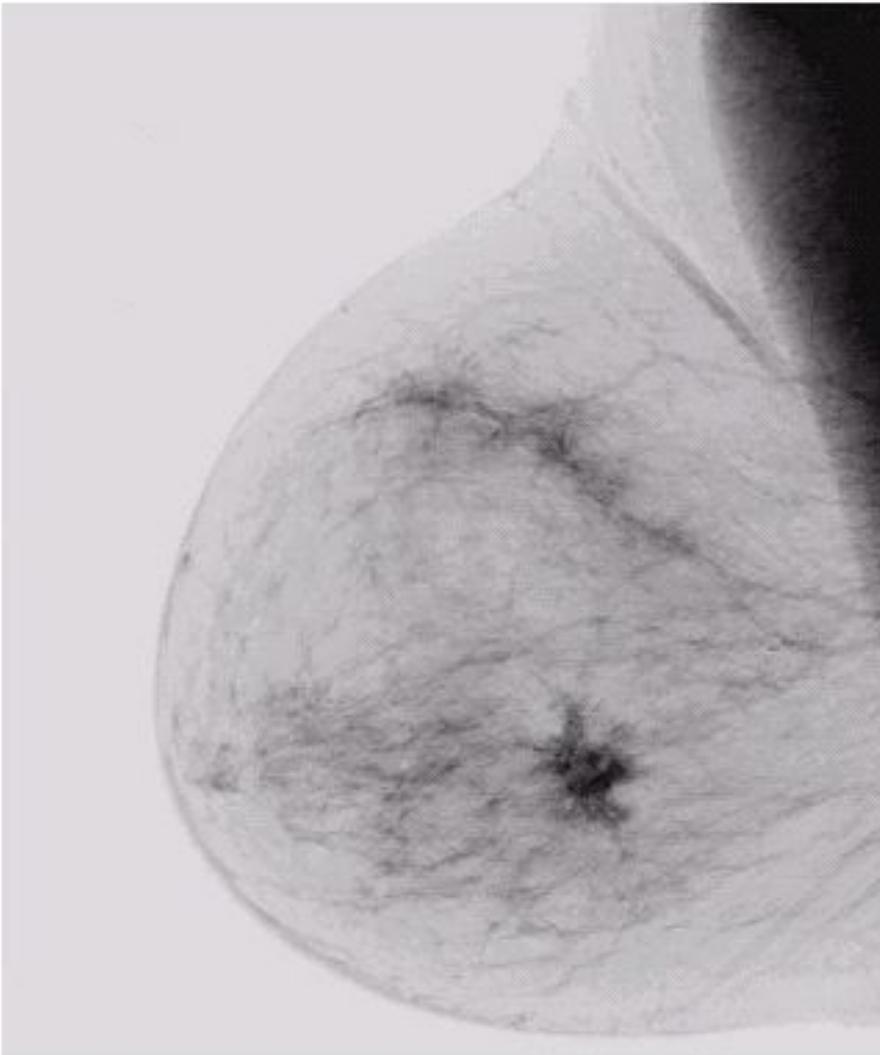
# Point Processing

- The simplest kind of range transformations are independent of position x,y:

$$g = t(f)$$

- This is called point processing.
- What can they do?
- Important: every pixel for itself

# Negative



a b

**FIGURE 3.4**

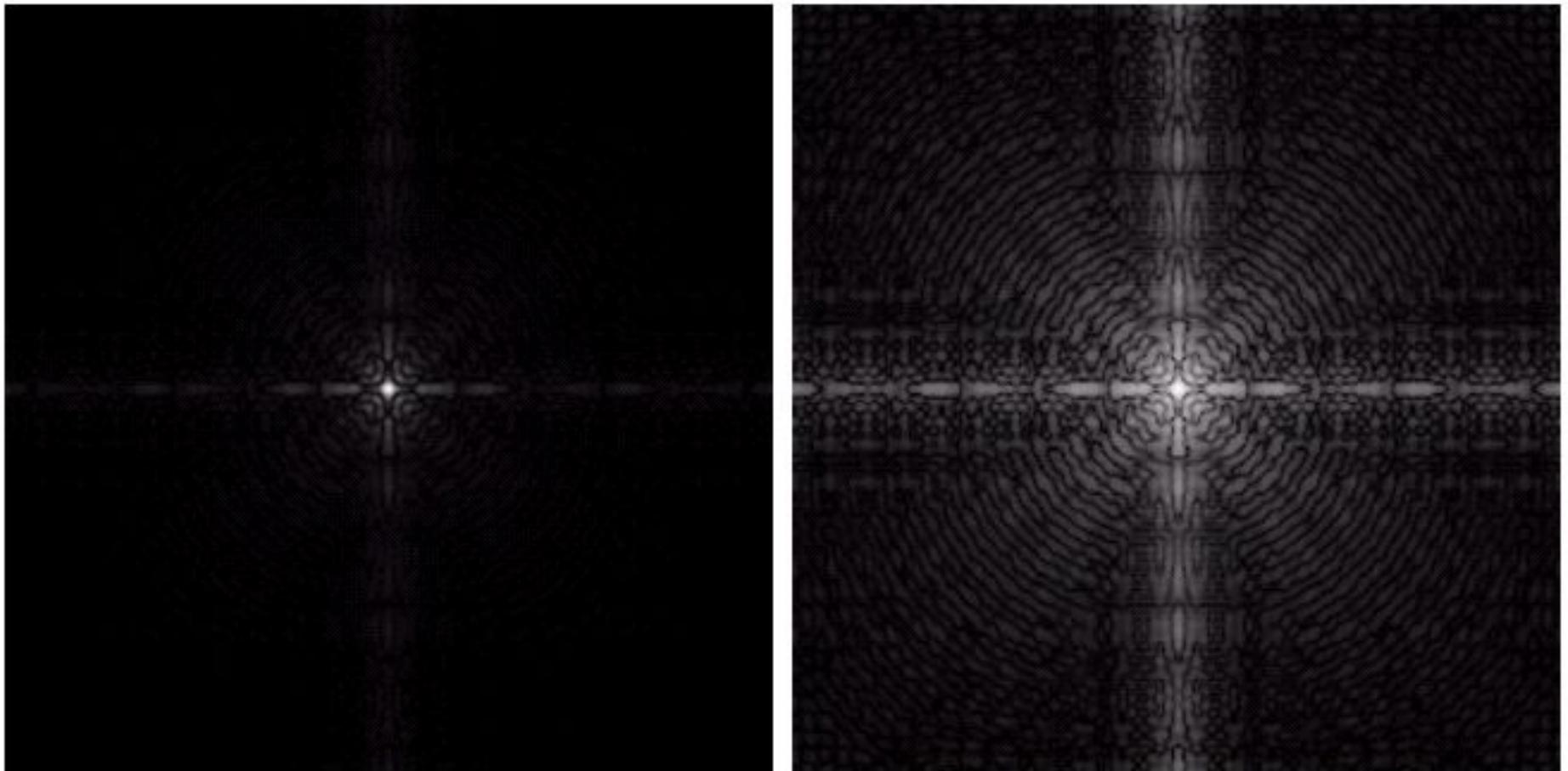
(a) Original digital mammogram.  
(b) Negative image obtained using the negative transformation in Eq. (3.2-1).  
(Courtesy of G.E. Medical Systems.)

# Log

a b

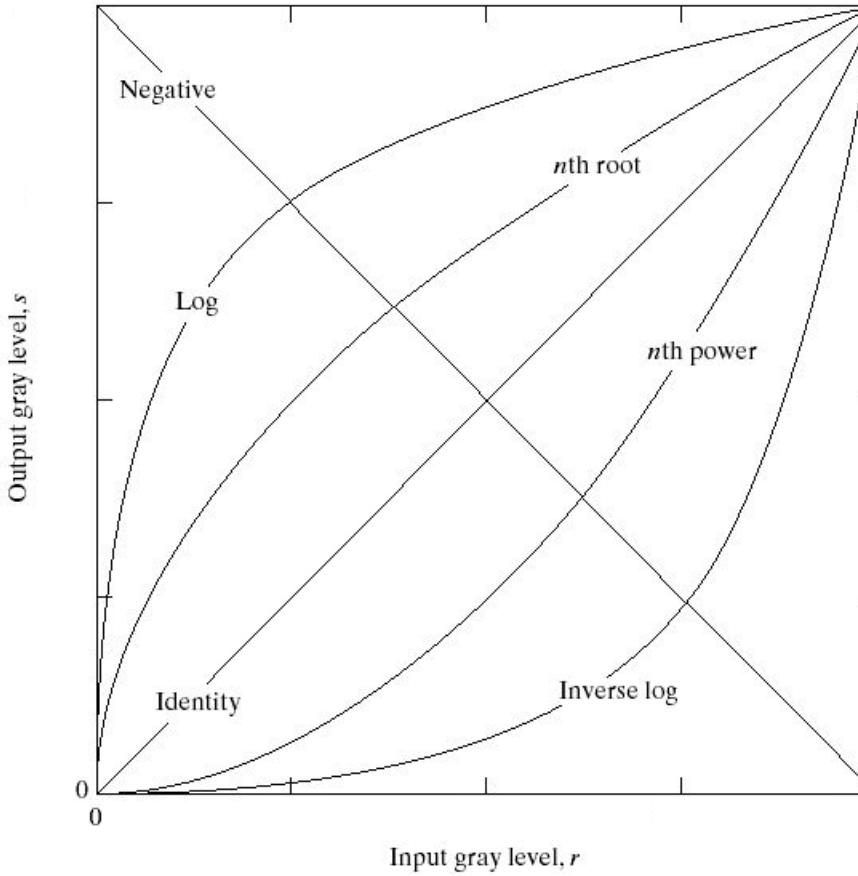
**FIGURE 3.5**

- (a) Fourier spectrum.  
(b) Result of applying the log transformation given in Eq. (3.2-2) with  $c = 1$ .
- 



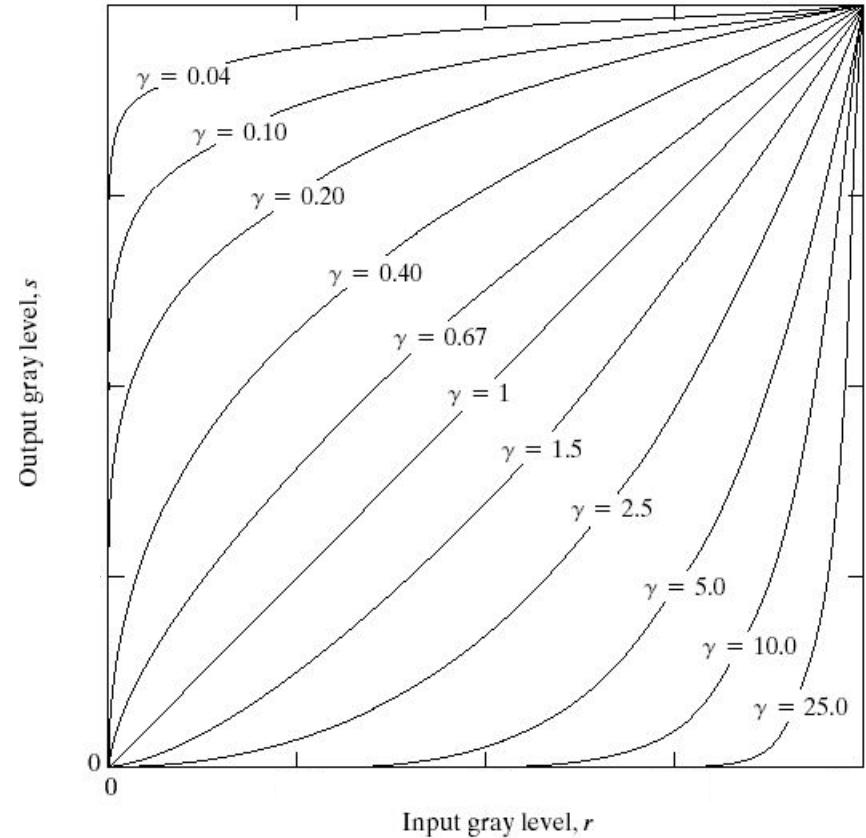
# Power law functions

**FIGURE 3.3** Some basic gray-level transformation functions used for image enhancement.



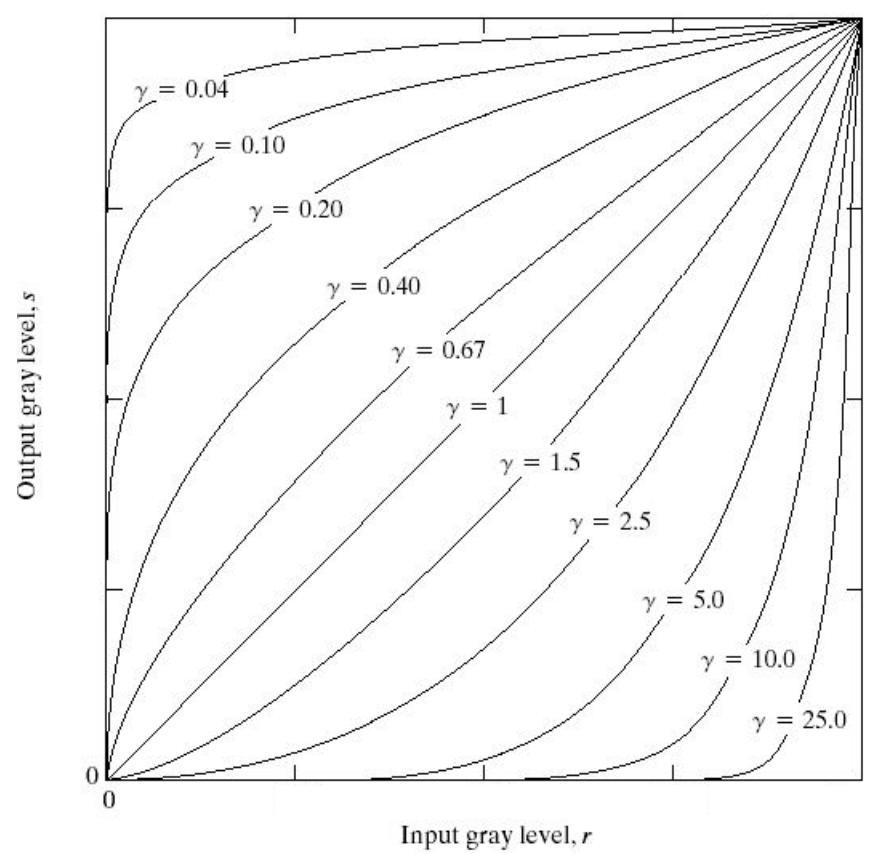
# Gamma “correction”

- Power law intensity transformations
- Typically, gamma = 2.2 for a display device, and 1/2.2 for encoding.



$$s = r^\gamma$$

# Gamma “correction”



$$s = r^\gamma$$



$\gamma = 1/2$

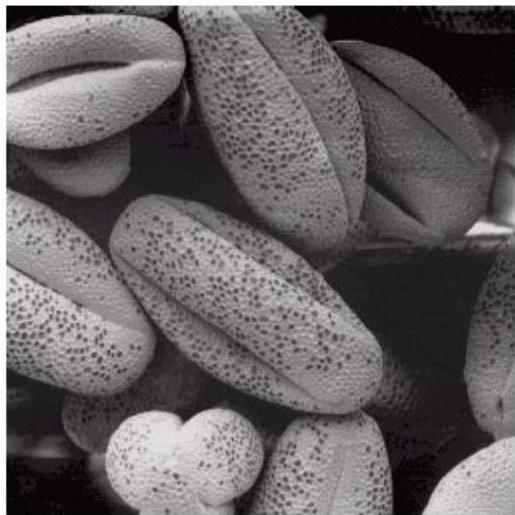
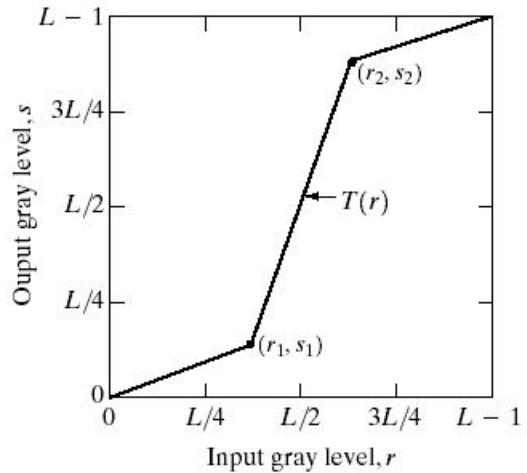


$\gamma = 1/3$



$\gamma = 1/4$

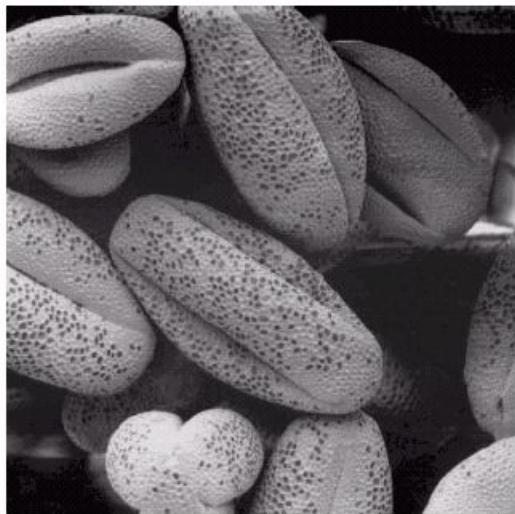
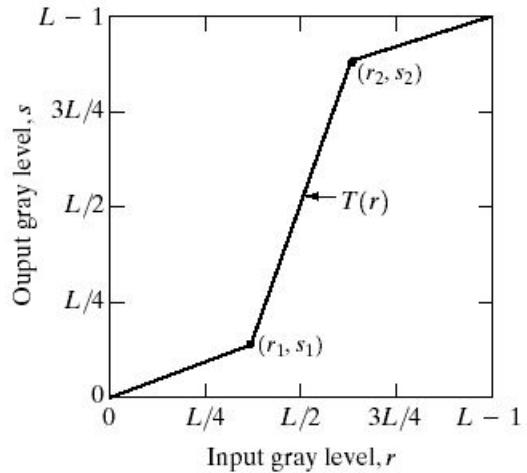
# Contrast Stretching



a | b  
c | d

**FIGURE 3.10**  
Contrast stretching.  
(a) Form of transformation function. (b) A low-contrast image. (c) Result of contrast stretching. (d) Result of thresholding. (Original image courtesy of Dr. Roger Heady, Research School of Biological Sciences, Australian National University, Canberra, Australia.)

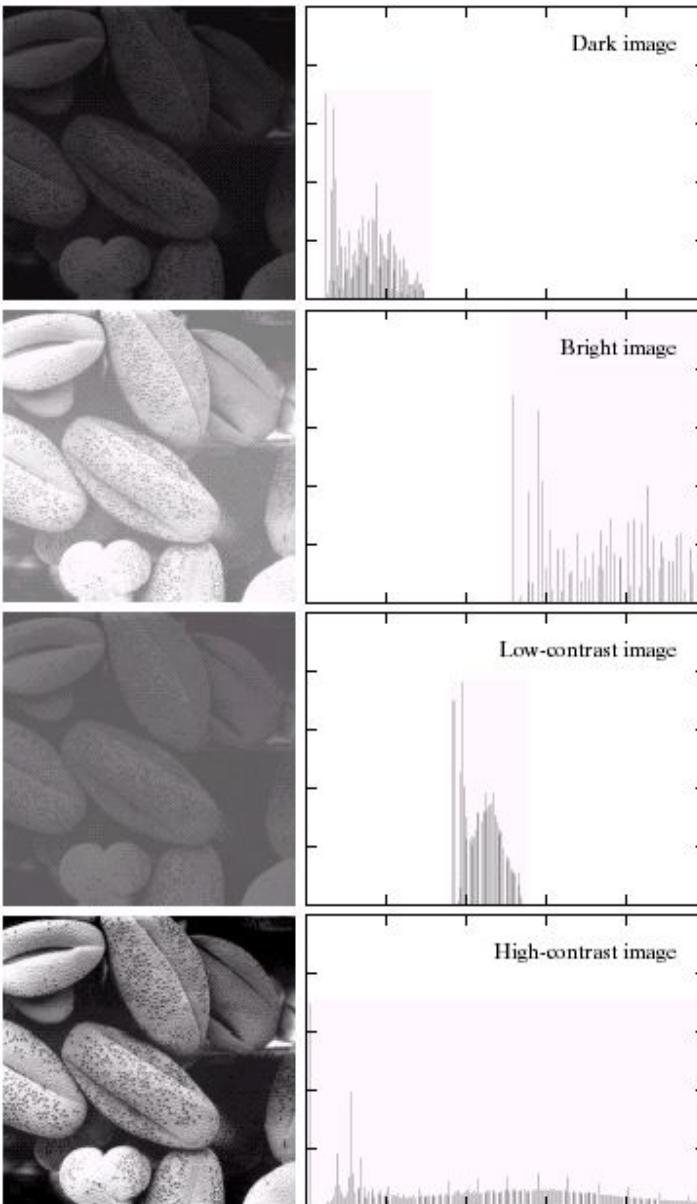
# Contrast Stretching



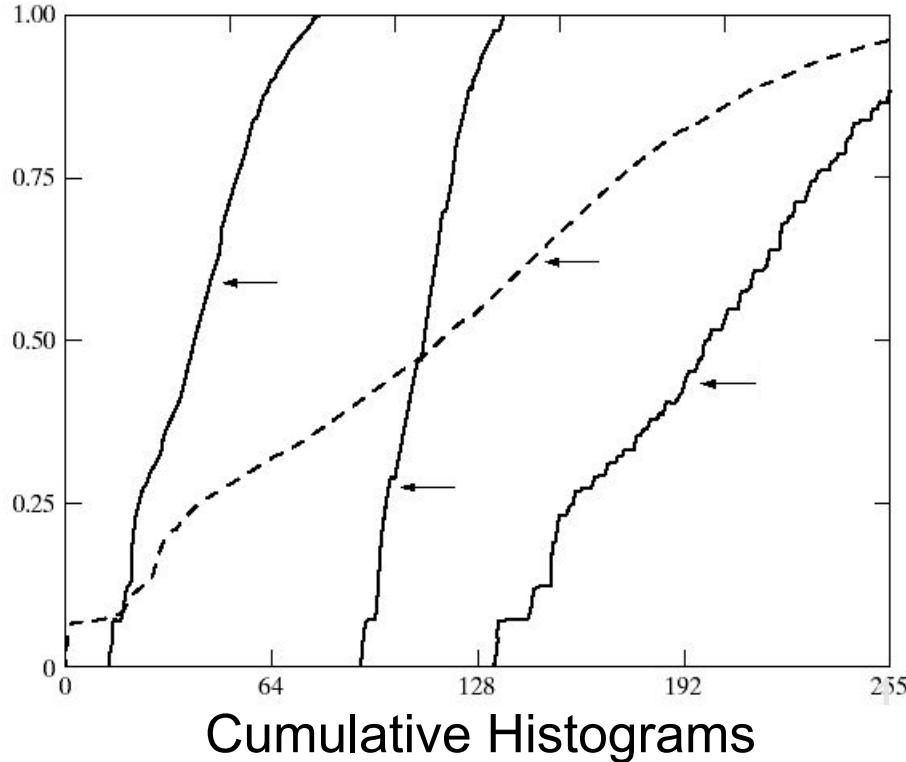
a | b  
c | d

**FIGURE 3.10**  
Contrast stretching.  
(a) Form of transformation function. (b) A low-contrast image. (c) Result of contrast stretching. (d) Result of thresholding. (Original image courtesy of Dr. Roger Heady, Research School of Biological Sciences, Australian National University, Canberra, Australia.)

# Image Histograms



1  
2  
3  
4



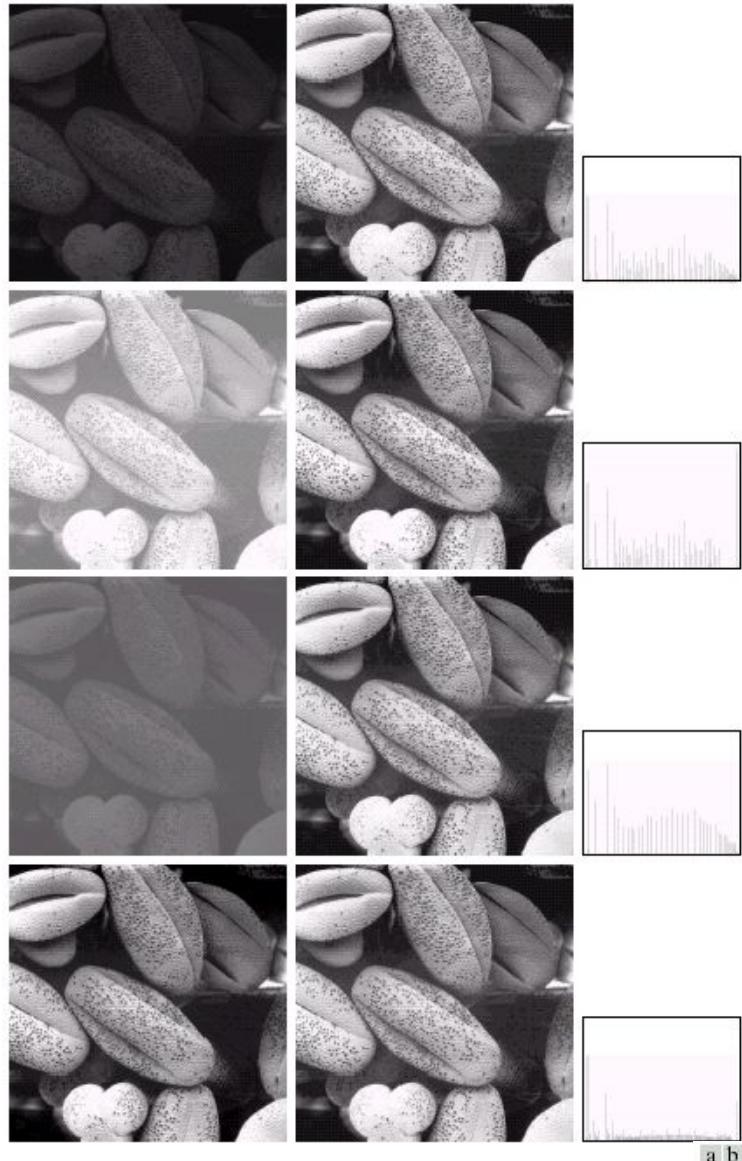
Cumulative Histograms

$$s = T(r)$$

a b

**FIGURE 3.15** Four basic image types: dark, light, low contrast, high contrast, and their corresponding histograms. (Original image courtesy of Dr. Roger Heady, Research School of Biological Sciences, Australian National University, Canberra, Australia.)

# Histogram Equalization

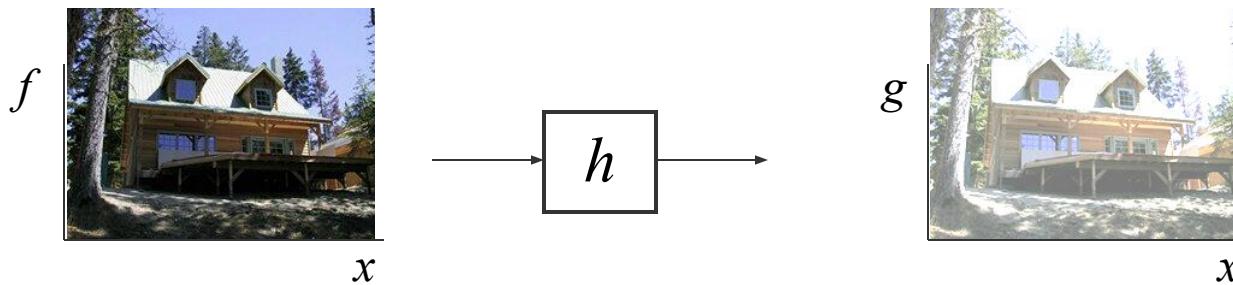


**FIGURE 3.17** (a) Images from Fig. 3.15. (b) Results of histogram equalization. (c) Corresponding histograms.

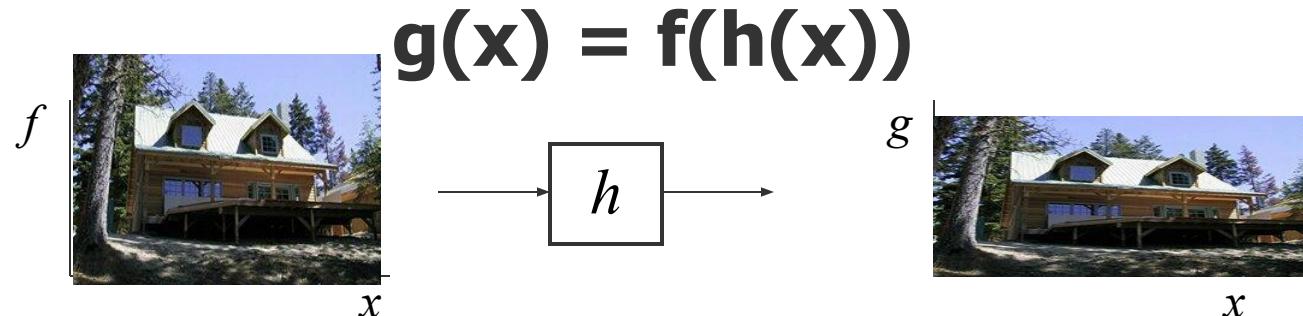
# Image Processing

## □ Point processing: change *range* of image

$$g(x) = h(f(x))$$



## □ Image warping: change domain of image



# Parametric (global) warping

## □ Examples of parametric warps:



translation



rotation



aspect



affine

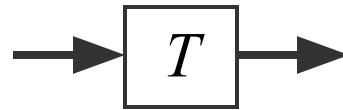


perspective



cylindrical

# Parametric (global) warping



$$p = (x, y)$$

$$p' = (x', y')$$

- Transformation  $T$  is a coordinate-changing machine:

$$p' = T(p)$$

- What does it mean that  $T$  is global and parametric?

- Is the same for any point  $p$
  - can be described by just a few numbers (parameters)

- Let's represent  $T$  as a matrix:

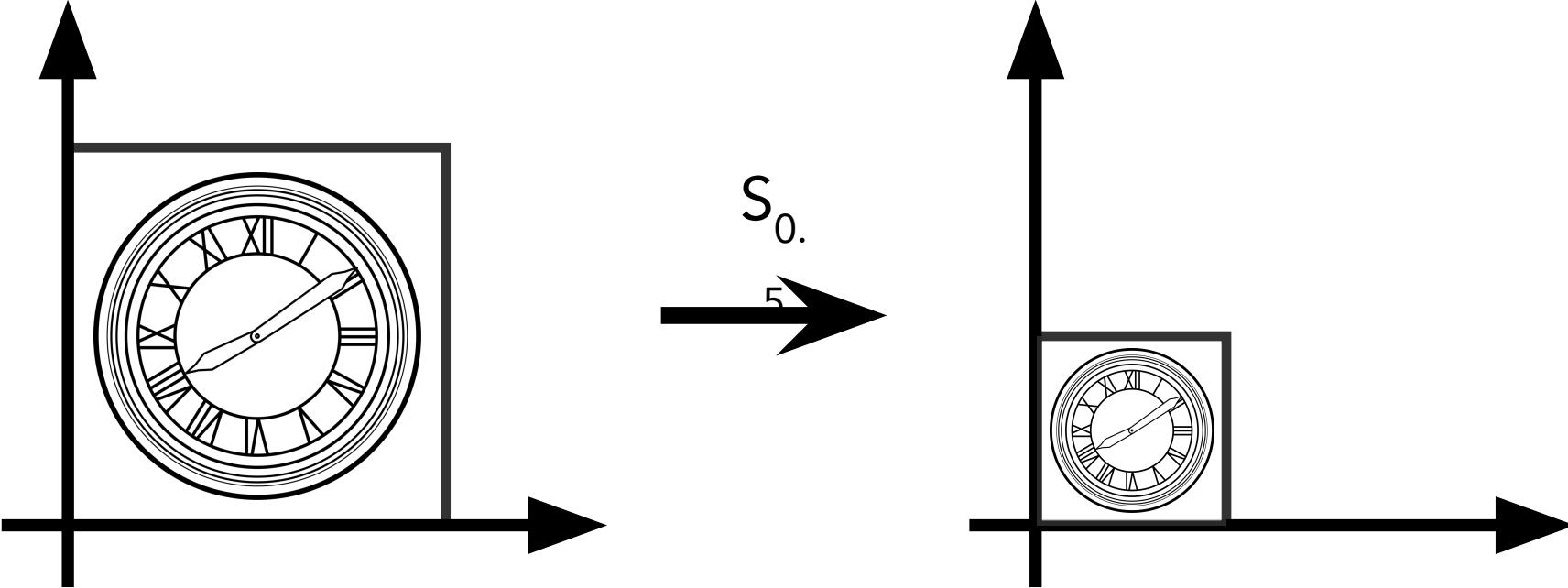
$$p' = Mp$$

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \mathbf{M} \begin{bmatrix} x \\ y \end{bmatrix}$$

# Linear transforms

$$T(ax + by) = aT(x) + bT(y)$$

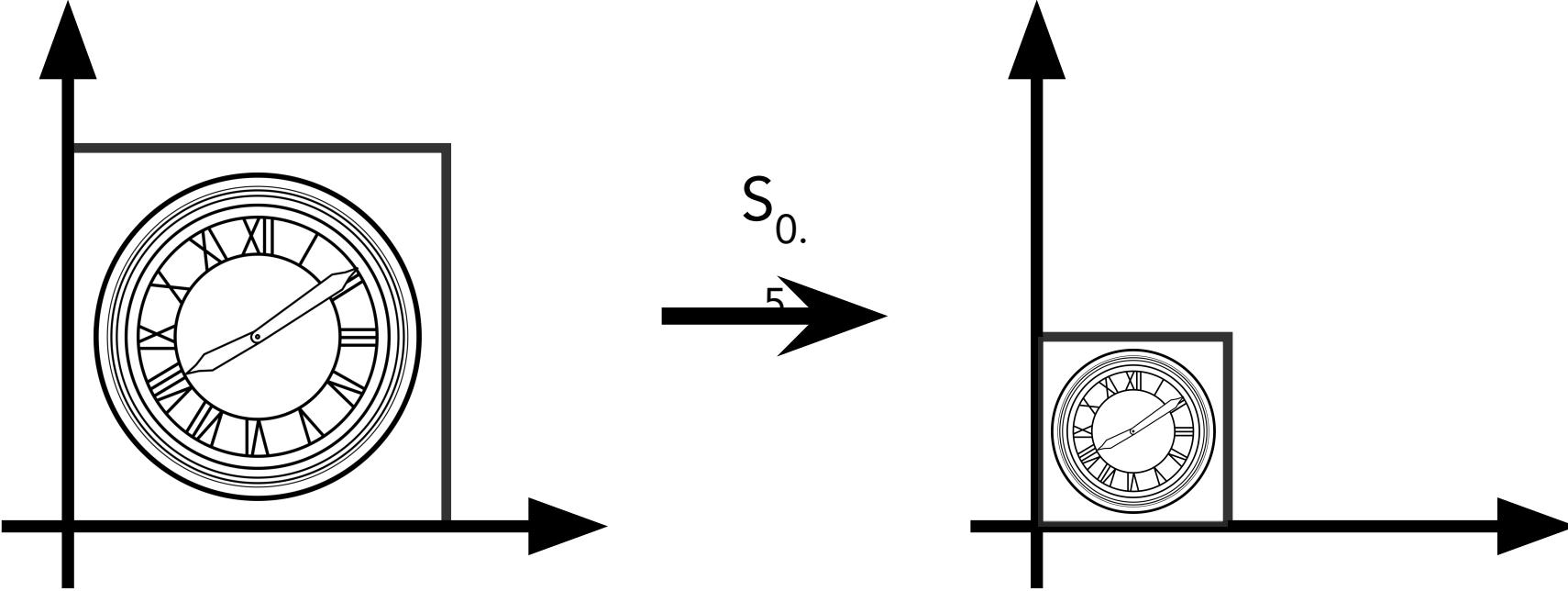
# Scale Transform (Uniform)



$$x' = sx$$

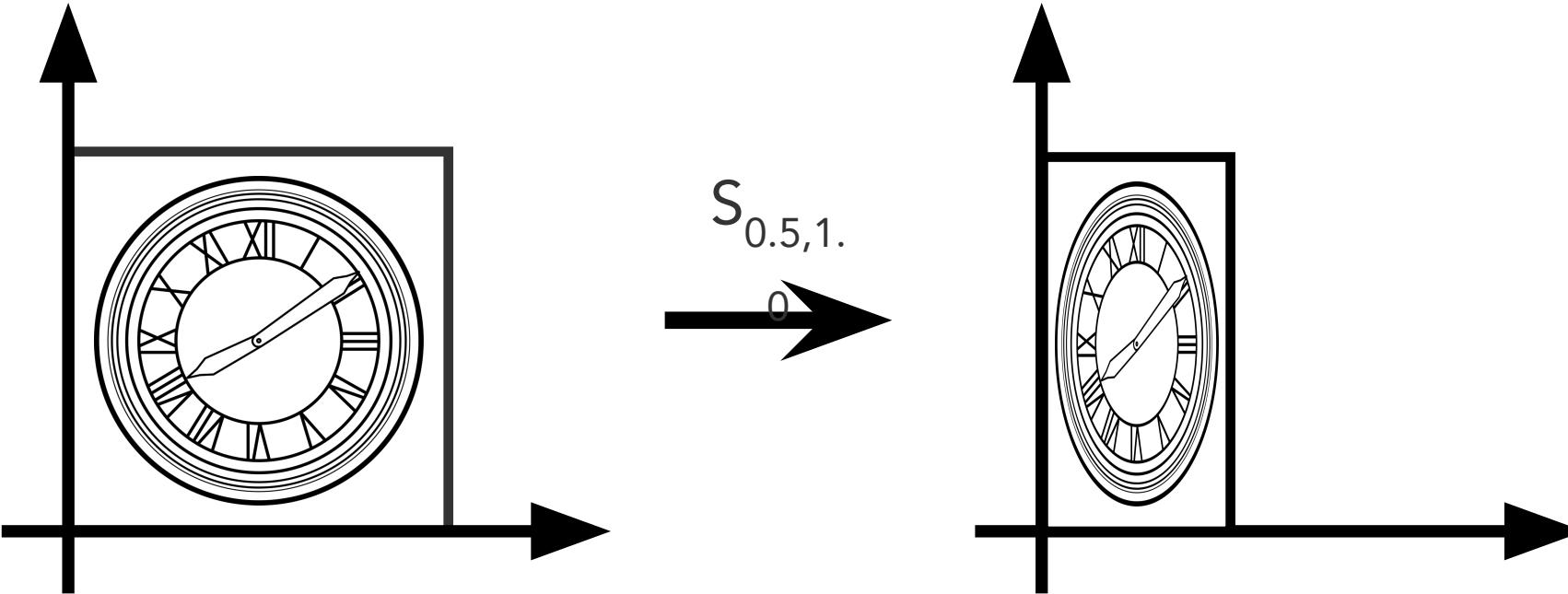
$$y' = sy$$

# Scale Matrix (Uniform)



$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

# Scale (Non-Uniform)



$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} s_x & 0 \\ 0 & s_y \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

# Scaling

## □ Operation

$$x' = s_x x$$

$$y' = s_y y$$

## □ In matrix form

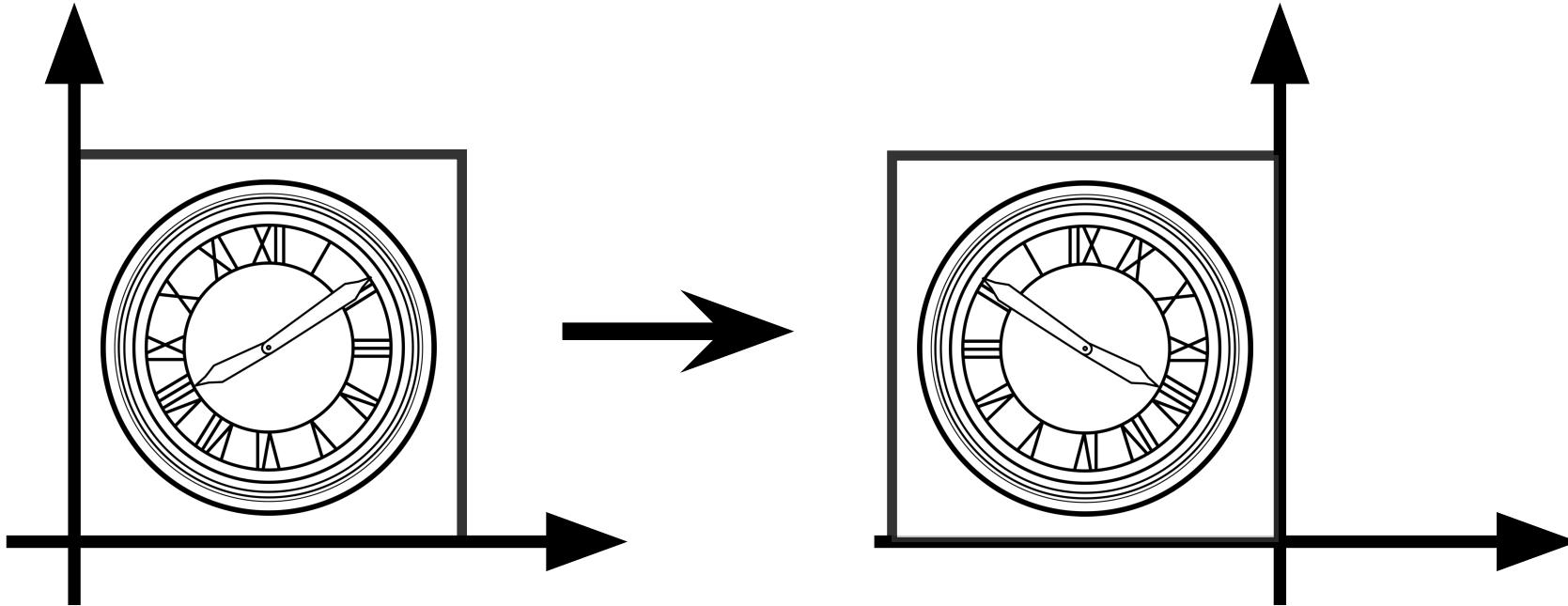
$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} s_x & 0 \\ 0 & s_y \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

## □ What's the inverse?

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} \frac{1}{s_x} & 0 \\ 0 & \frac{1}{s_y} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

## □ Is it linear? Yes!

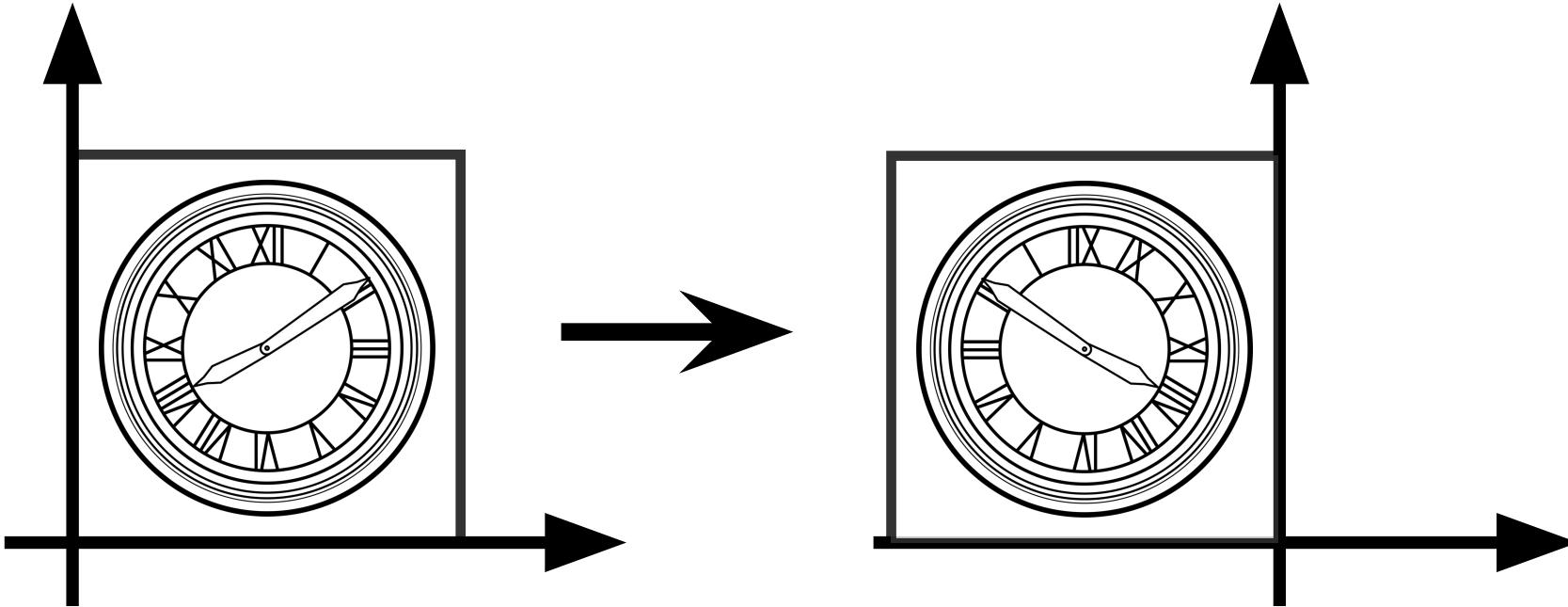
# Reflection



$$x' = -x$$

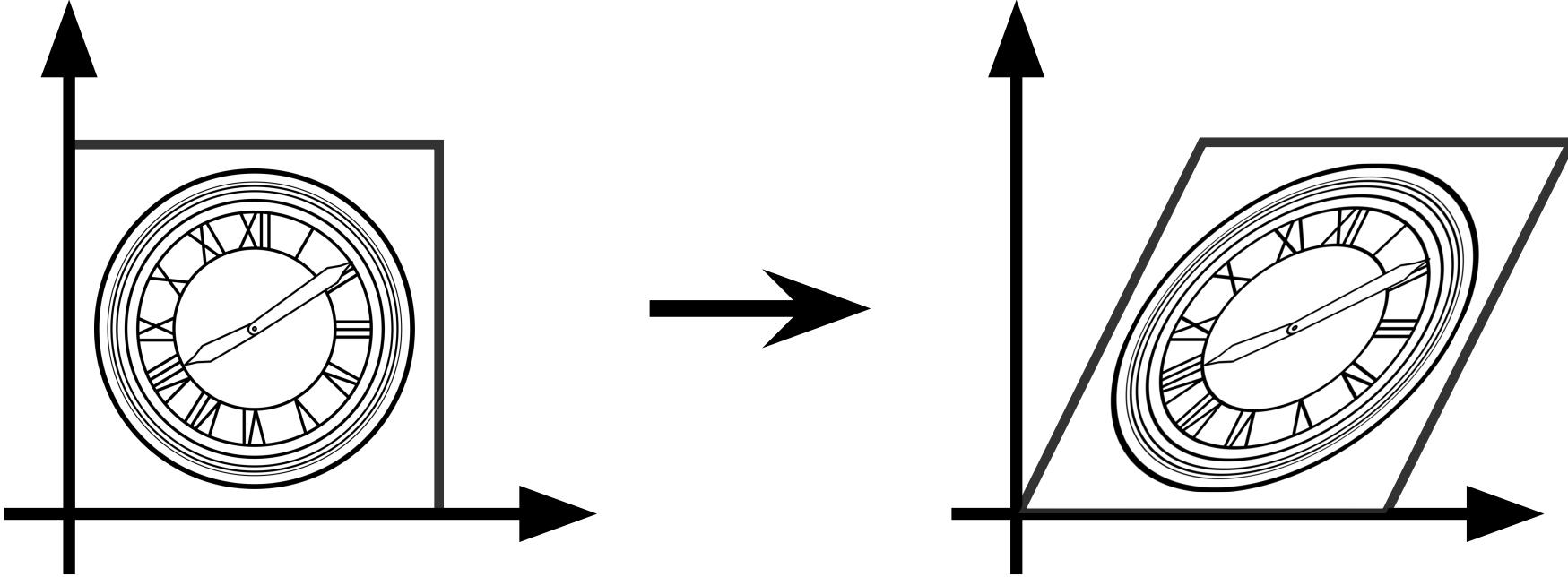
$$y' = y$$

# Reflection Matrix



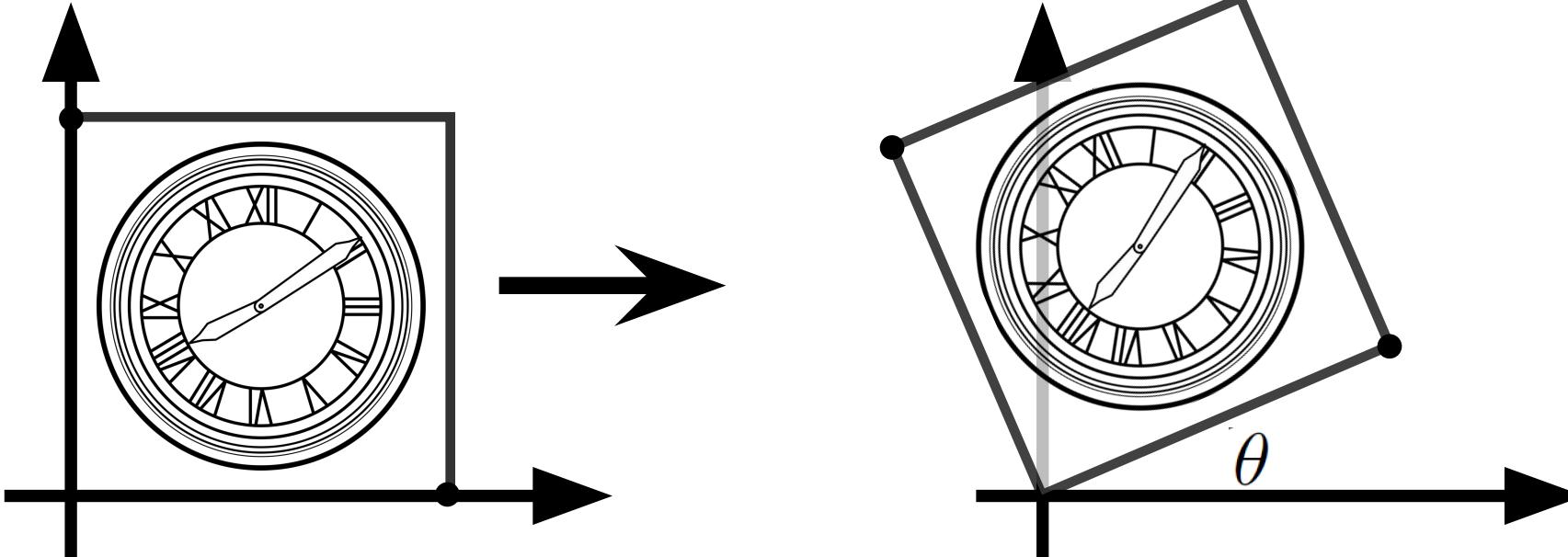
$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

# Shear Matrix

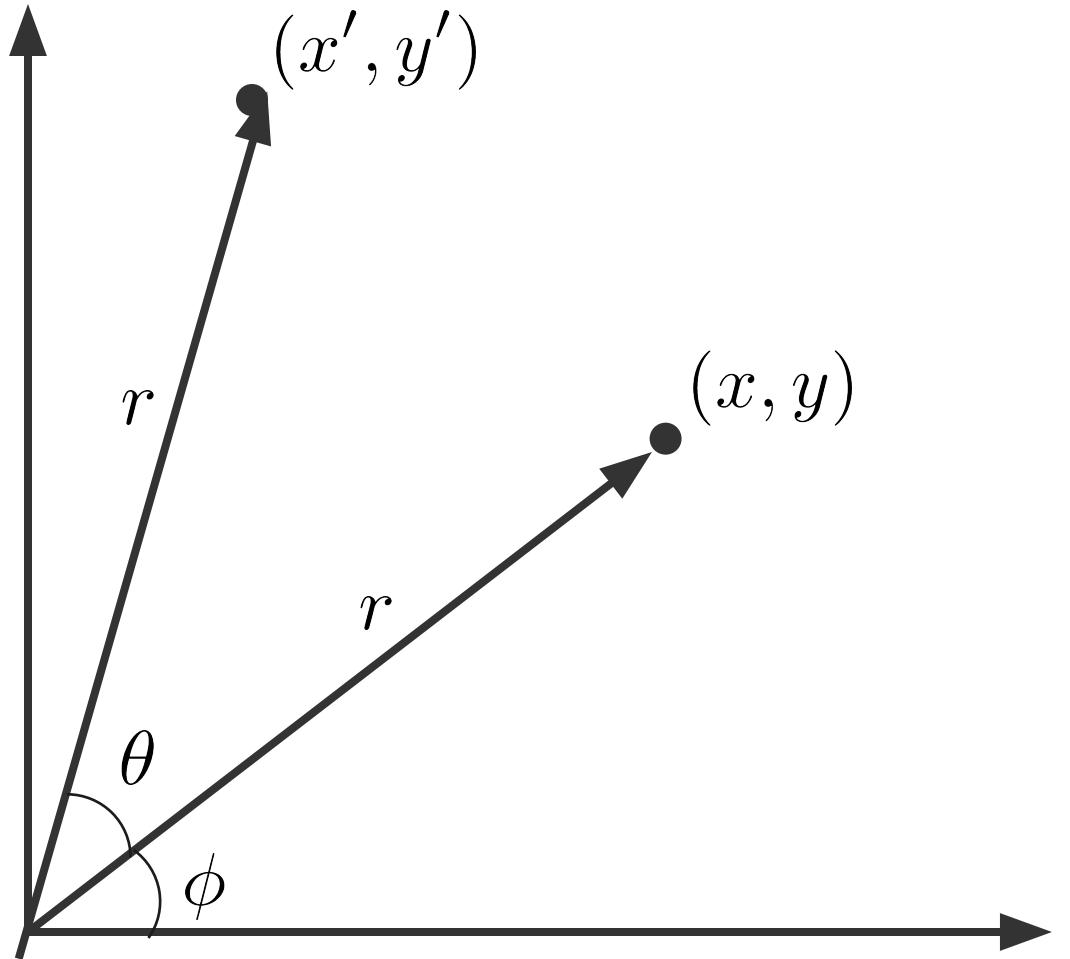


$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 1 & a \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

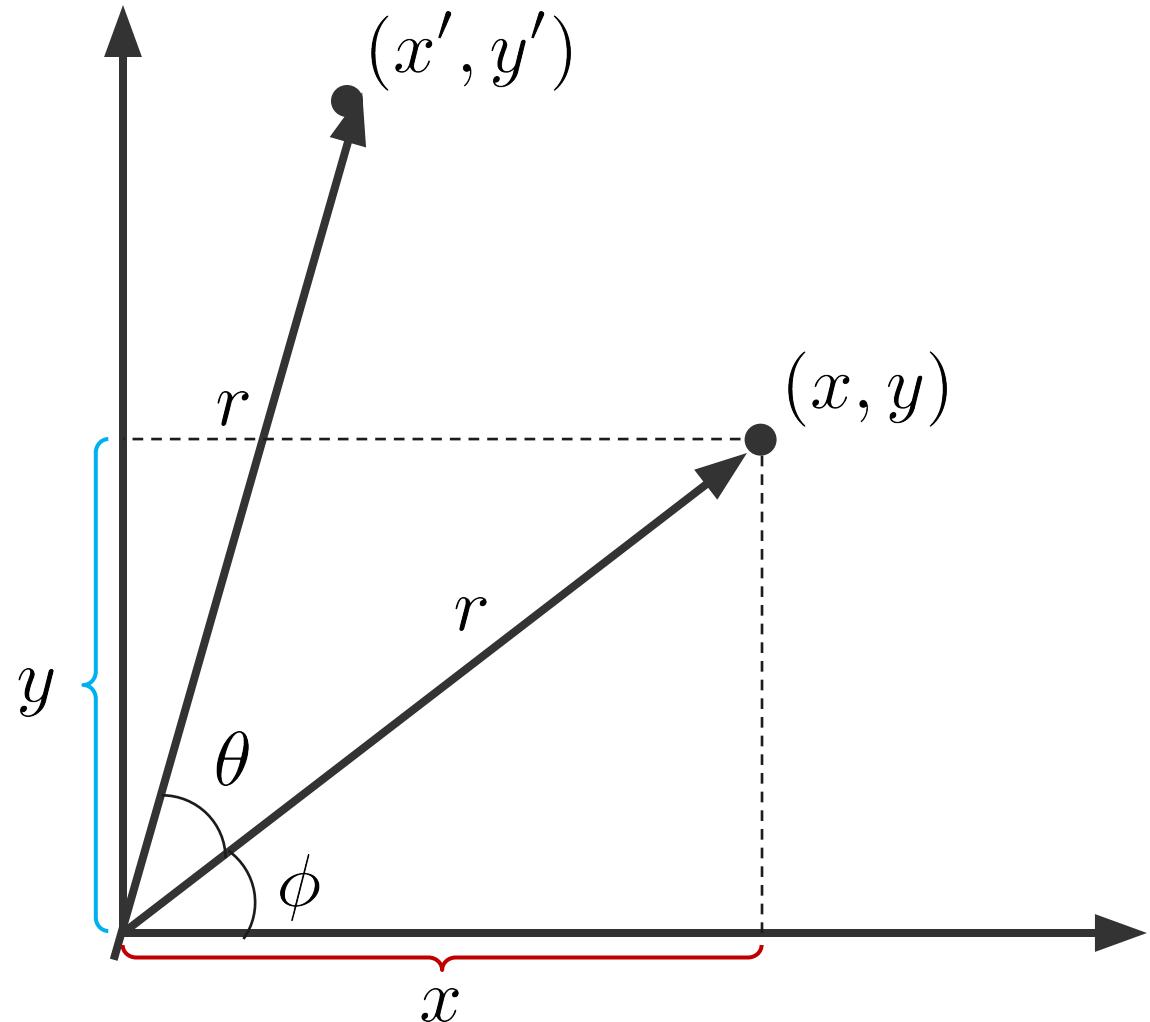
# Rotation Matrix



# 2-D Rotation



# 2-D Rotation



$$x = r \cos(\phi)$$

$$y = r \sin(\phi)$$

$$x' = r \cos(\phi + \theta)$$

$$y' = r \sin(\phi + \theta)$$

## Trigonometric Identity

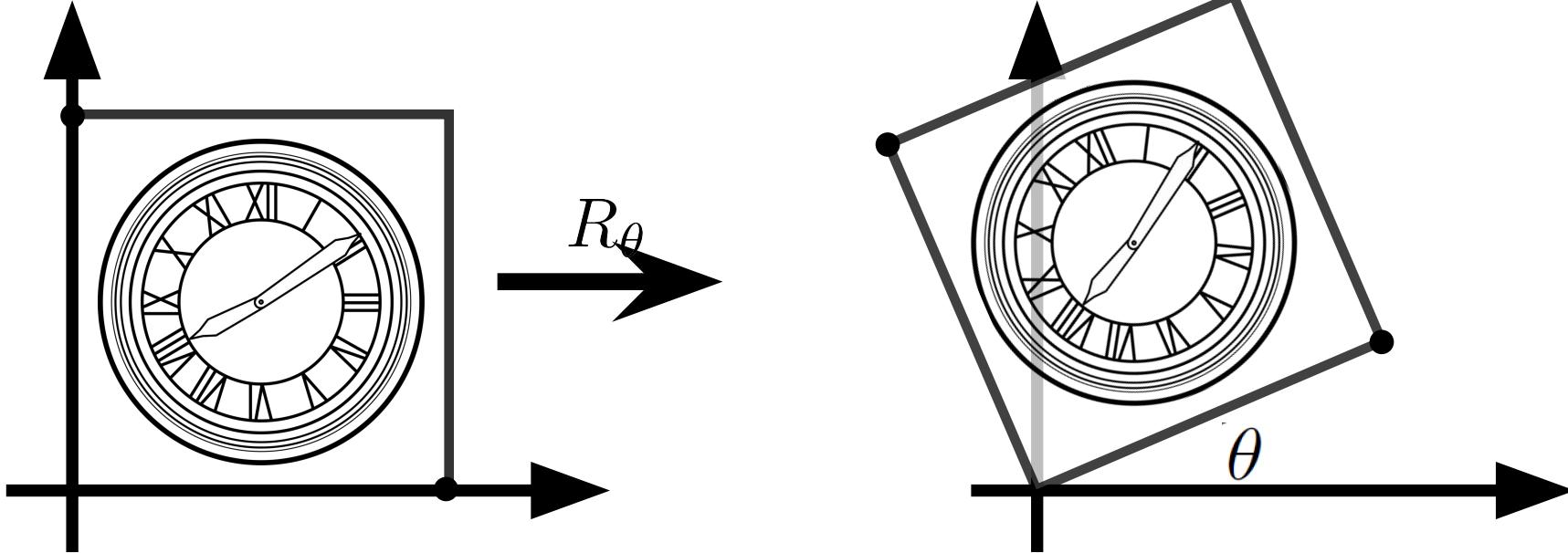
$$x' = \overbrace{r \cos(\phi)}^x \cos(\theta) - \overbrace{r \sin(\phi)}^y \sin(\theta)$$
$$y' = \underbrace{r \sin(\phi)}_y \cos(\theta) + \underbrace{r \cos(\phi)}_x \sin(\theta)$$

## Substitute

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

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

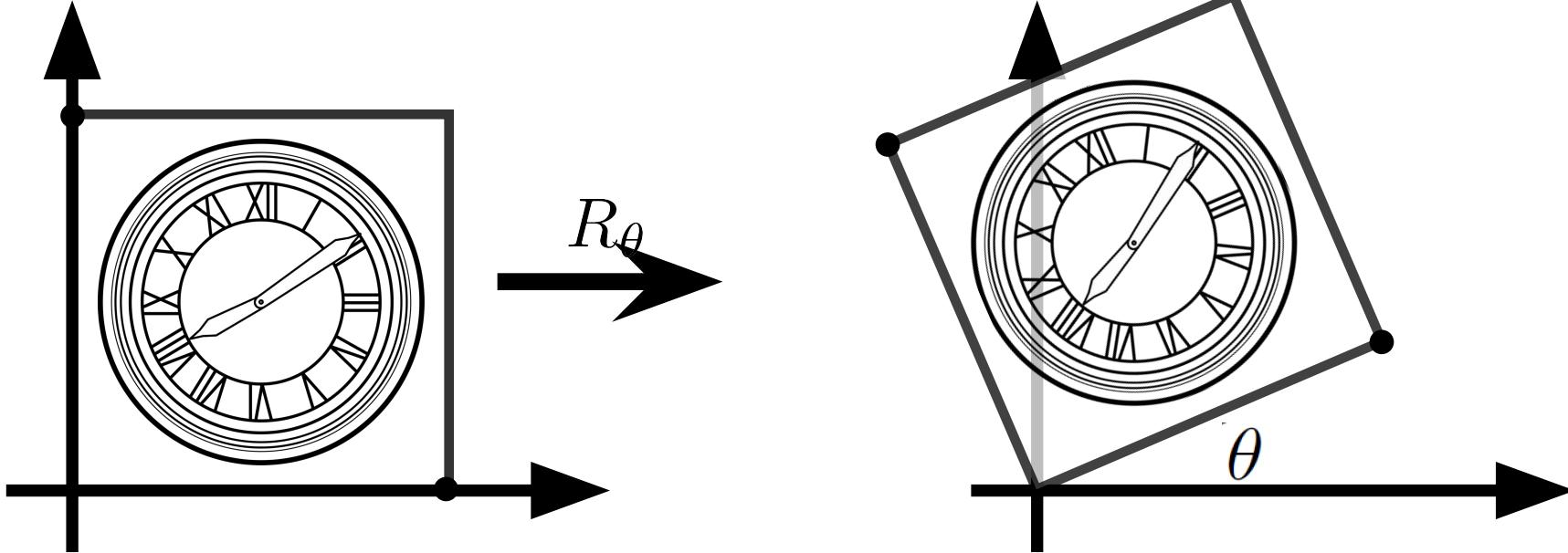
# Rotation Matrix



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

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} \cdot & \cdot \\ \cdot & \cdot \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

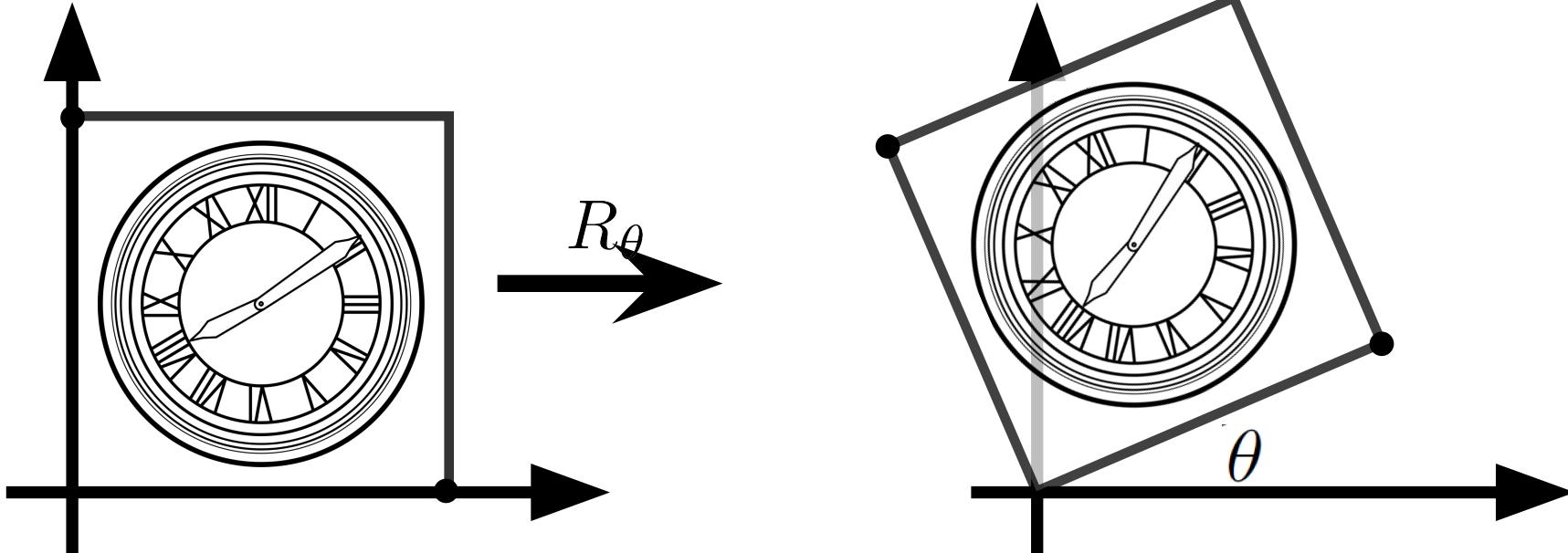
# Rotation Matrix



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

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} \cos(\theta) & . \\ . & . \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

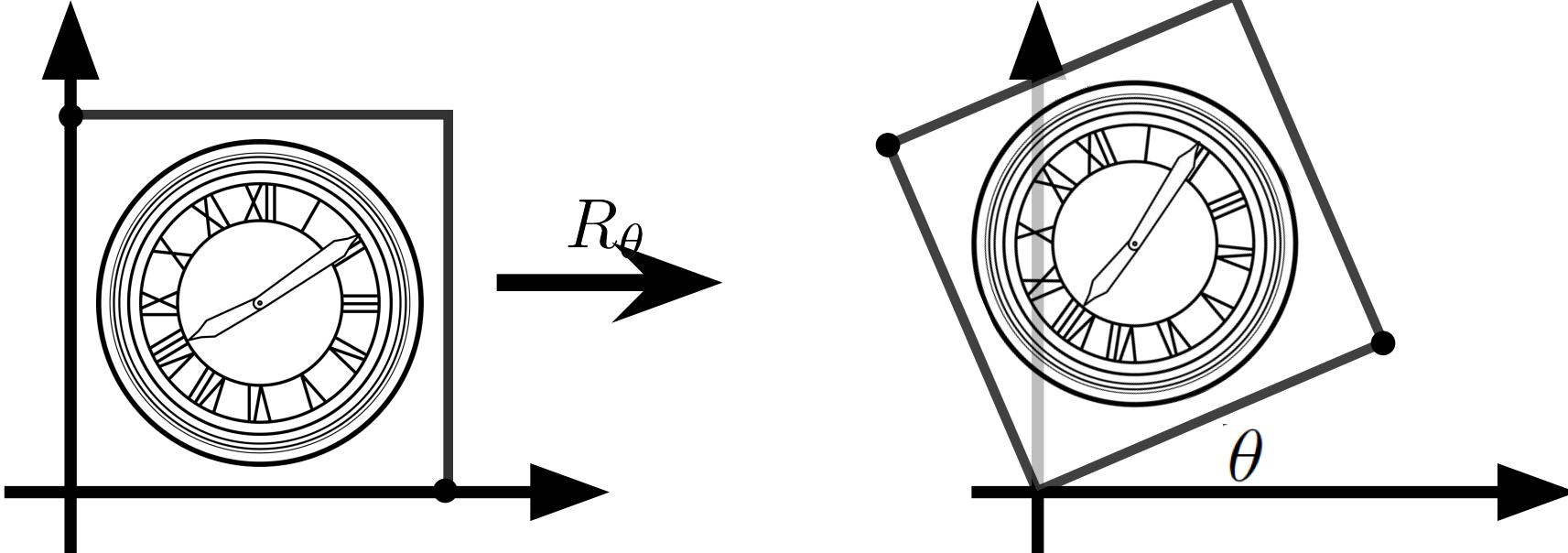
# Rotation Matrix



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

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

# Rotation Matrix



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

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \underbrace{\begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix}}_{R_\theta} \begin{pmatrix} x \\ y \end{pmatrix}$$

# Rotation

- Even though  $\sin(\theta)$  and  $\cos(\theta)$  are nonlinear functions of  $\theta$ 
  - $x'$  is a *linear combination* of  $x$  and  $y$
  - $y'$  is a *linear combination* of  $x$  and  $y$
- What's the inverse transformation?
  - Rotation by  $-\theta$

$$R_\theta^{-1} = R_{-\theta} = R^T$$

# Linear Transforms = Matrices

$$x' = a x + b y$$

$$y' = c x + d y$$

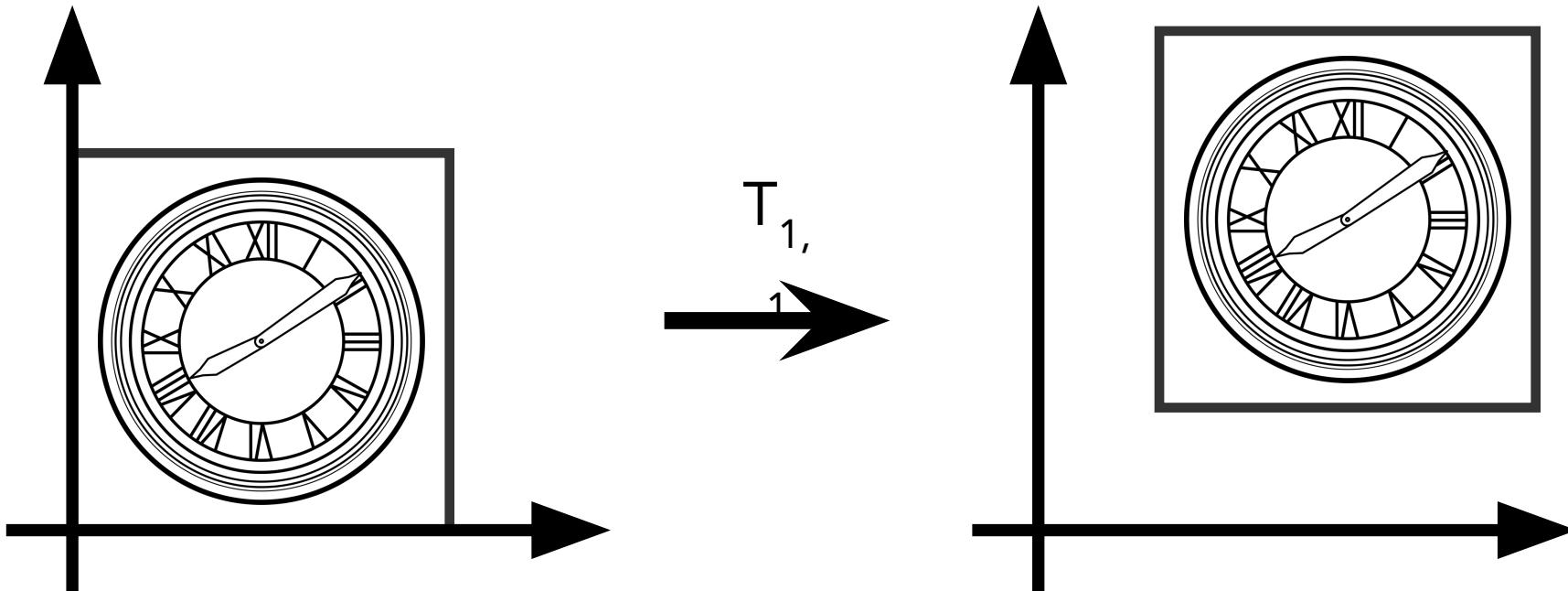
$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\mathbf{p}' = \mathbf{M}\mathbf{p}$$

# All 2D Linear Transformations

- **Linear transformations are combinations of ...**
  - **Scale,**
  - **Rotation,**
  - **Shear, and**
  - **Mirror**
- **Properties of linear transformations:**
  - **Origin maps to origin**
  - **Lines map to lines**
  - **Parallel lines remain parallel**
  - **Ratios of parallel line segments are preserved**
  - **Closed under composition**

# Translation?



$$x' = x + t_x$$

$$y' = y + t_y$$

# Translation

- **E.g., move x by +5 units, leave y**
- **We need appropriate matrix. What is it?**

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} ? & ? \\ ? & ? \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x + 5 \\ y \end{pmatrix}$$

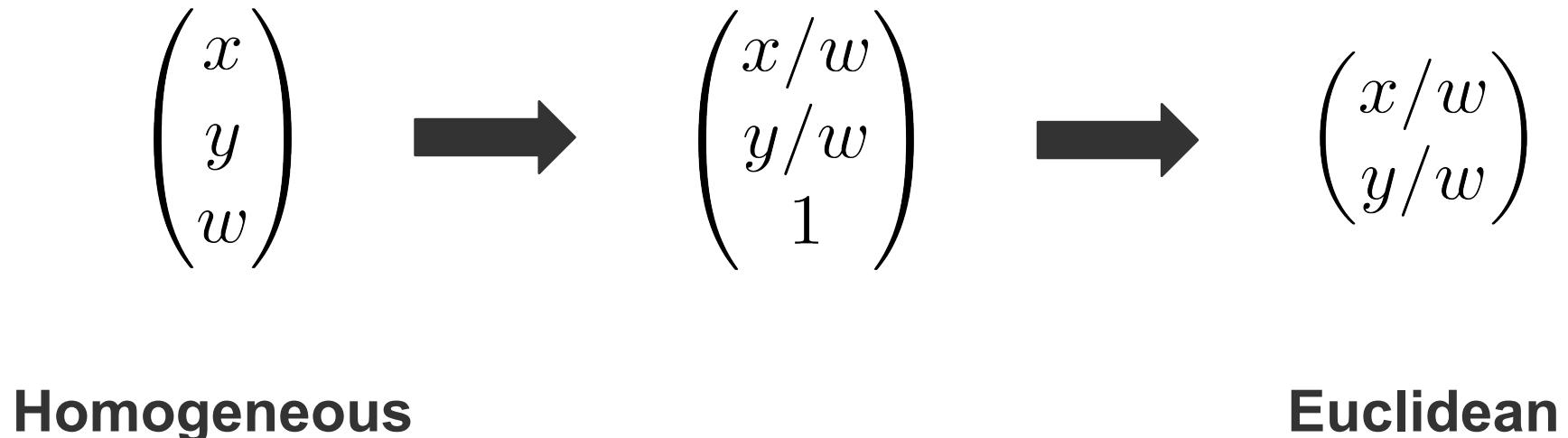
# Homogeneous Coordinates

- Represent an n-dimensional Euclidean point using an n+1 dimensional point
  - Set dimension n+1 (w) to 1
    - E.g., 2D Euclidean  $\rightarrow$  3D Homogeneous

$$\begin{pmatrix} x \\ y \end{pmatrix} \xrightarrow{\hspace{1cm}} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$$

# Homogeneous Coordinates

- To go from homogeneous coordinate to Euclidean
  - Divide by the last coordinate
  - Remove the normalized last coordinate



# Homogeneous Coordinates

- Multiplication by scalar has no effect

Homogeneous

$$\begin{pmatrix} x \\ y \\ w \end{pmatrix}$$

||

$$\begin{pmatrix} \alpha x \\ \alpha y \\ \alpha w \end{pmatrix}$$



$$\begin{pmatrix} x/w \\ y/w \\ 1 \end{pmatrix}$$



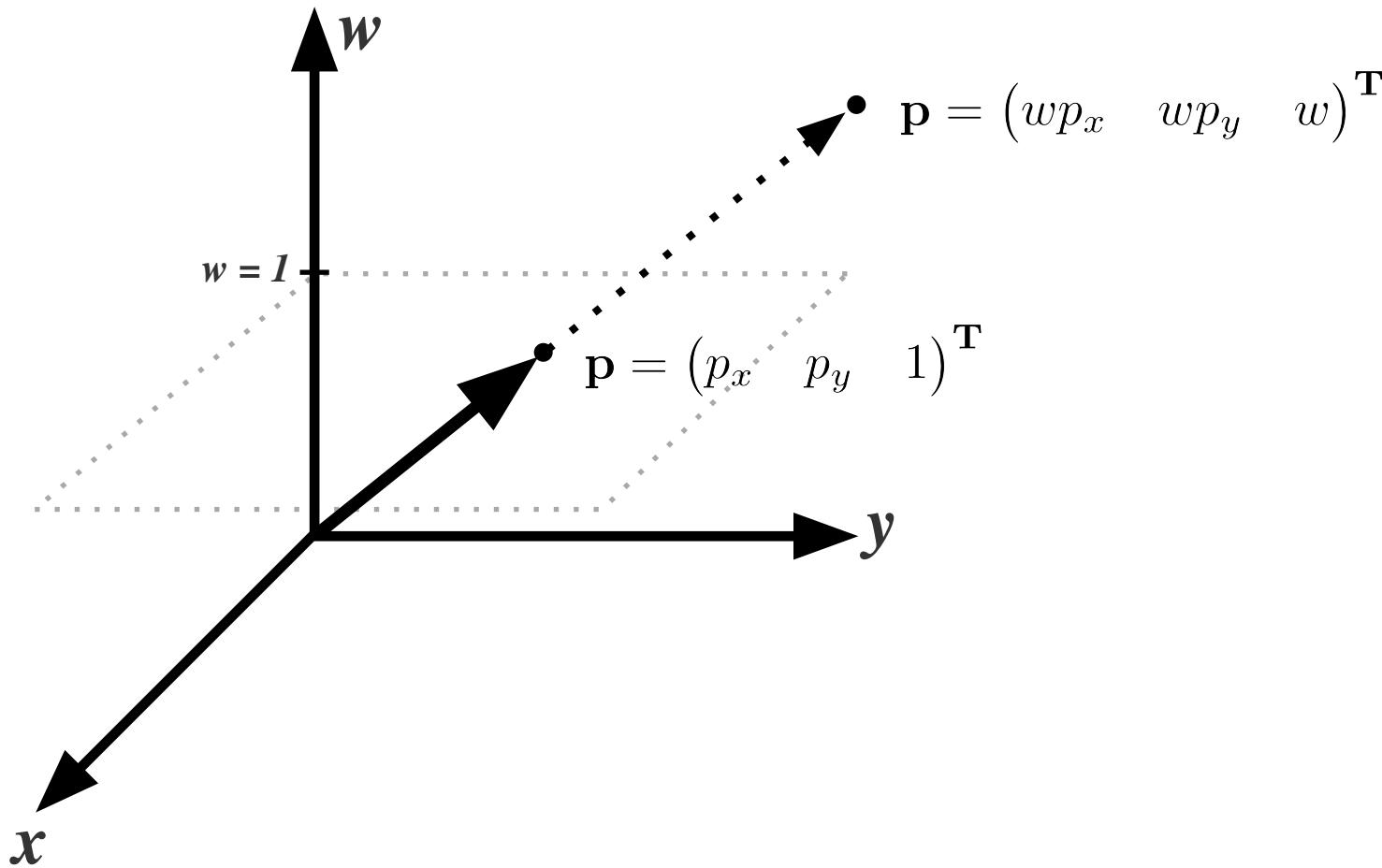
Euclidean

$$\begin{pmatrix} x/w \\ y/w \end{pmatrix}$$

||

$$\begin{pmatrix} x/w \\ y/w \end{pmatrix}$$

# Homogeneous coordinates: some intuition



# Translation in Homogeneous Coordinates

$$\begin{pmatrix} x' \\ y' \\ w' \end{pmatrix} = \begin{pmatrix} ? & ? & ? \\ ? & ? & ? \\ ? & ? & ? \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} = \begin{pmatrix} x + 5 \\ y \\ 1 \end{pmatrix}$$

# Translation in Homogeneous Coordinates

$$\begin{pmatrix} x' \\ y' \\ w' \end{pmatrix} = \begin{pmatrix} 1 & 0 & 5 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} = \begin{pmatrix} x + 5 \\ y \\ 1 \end{pmatrix}$$

# General Translation Matrix

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

# 2D Transformations as 3x3 matrices

## □ Scale

$$\mathbf{S}(s_x, s_y) = \begin{pmatrix} s_x & 0 & 0 \\ 0 & s_y & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

## □ Rotation

$$\mathbf{R}(\alpha) = \begin{pmatrix} \cos \alpha & -\sin \alpha & 0 \\ \sin \alpha & \cos \alpha & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

## □ Translation

$$\mathbf{T}(t_x, t_y) = \begin{pmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{pmatrix}$$

# Matrix Composition

## □ Combine transformations by matrix multiplication

$$\begin{bmatrix} x' \\ y' \\ w' \end{bmatrix} = \left( \begin{bmatrix} 1 & 0 & tx \\ 0 & 1 & ty \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} sx & 0 & 0 \\ 0 & sy & 0 \\ 0 & 0 & 1 \end{bmatrix} \right) \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

$$p' = T(t_x, t_y) R(\theta) S(s_x, s_y) p$$

1) Scale

2) Rotate

3) Translate

# Affine Transformations

- **Affine transformations are combinations of ...**

- **Linear transformations, and**
  - **Translations**

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

- **Properties of affine transformations:**

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

- **Will the last coordinate  $w$  always be 1?**

# Projective Transformations

- Projective transformations ...

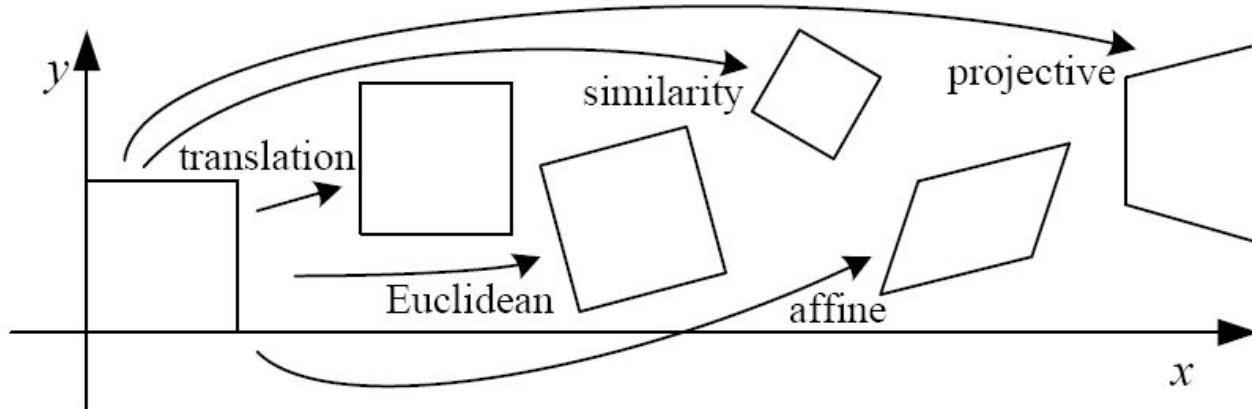
- Affine transformations, and
  - Projective warps

- Properties of projective transformations:

- Origin does not necessarily map to origin
  - Lines map to lines
  - Parallel lines do not necessarily remain parallel
  - Ratios are not preserved
  - Closed under composition

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

# 2D image transformations



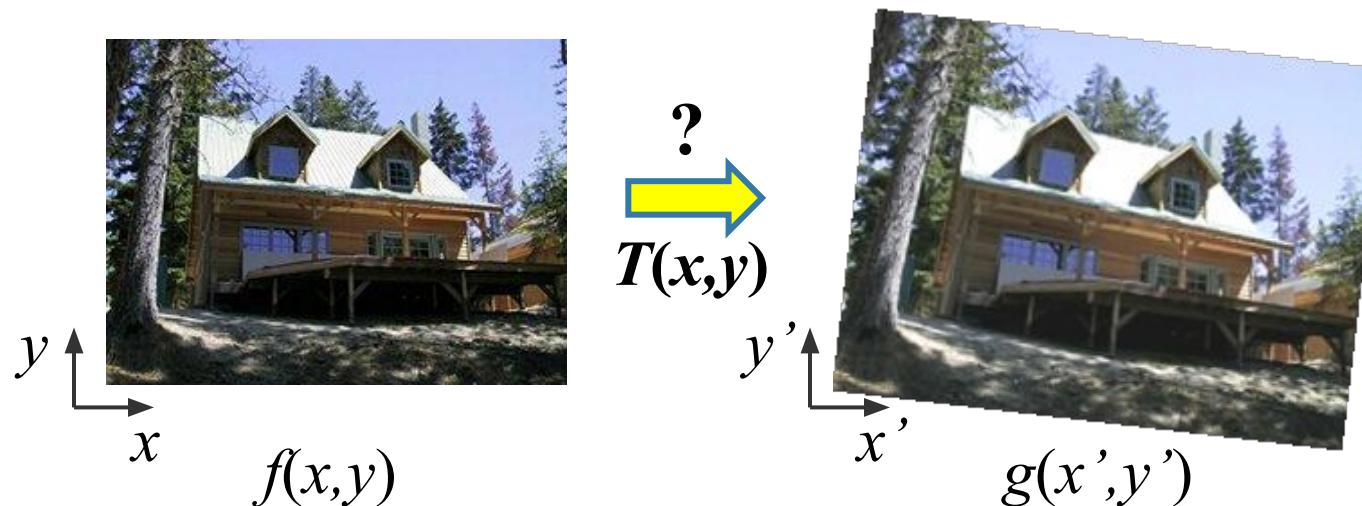
Name	# D.O.F.	Preserves:	Icon
translation	2	all	square
rigid (Euclidean)	3	angles	rotated square
similarity	4	angles, ratios	rotated and scaled square
affine	6	ratios	curved edges
projective	8	none	highly distorted edges

These transformations are a nested set of groups

- Closed under composition and inverse is a member

# Recovering Transformations

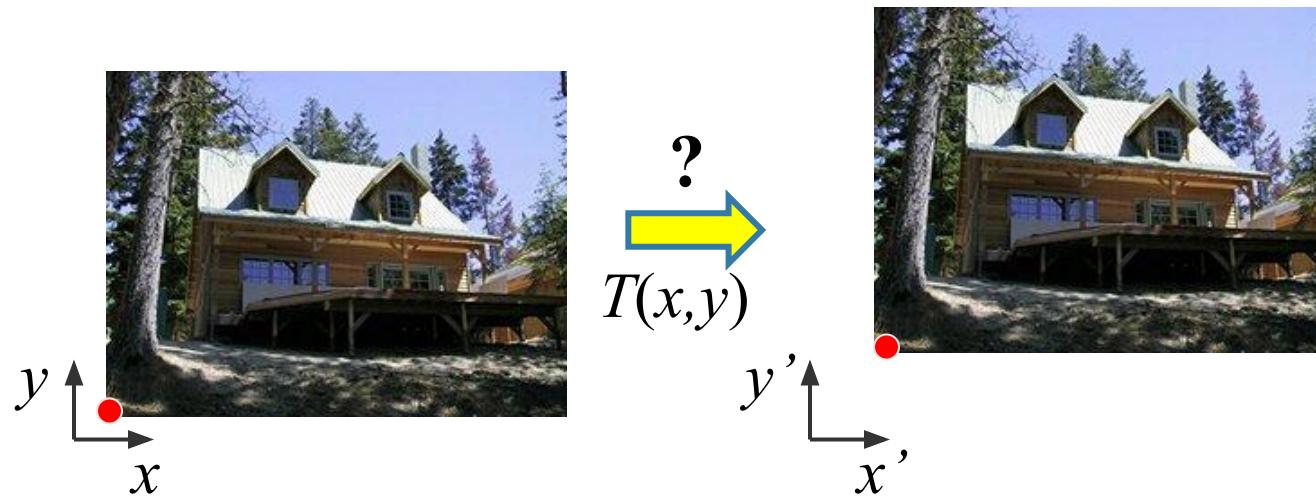
- What if we know  $f$  and  $g$  and want to recover the transform  $T$ ?
  - e.g., better align images from Project 2
  - Willing to let user provide correspondences
    - How many do we need?



# Translation: # correspondences?

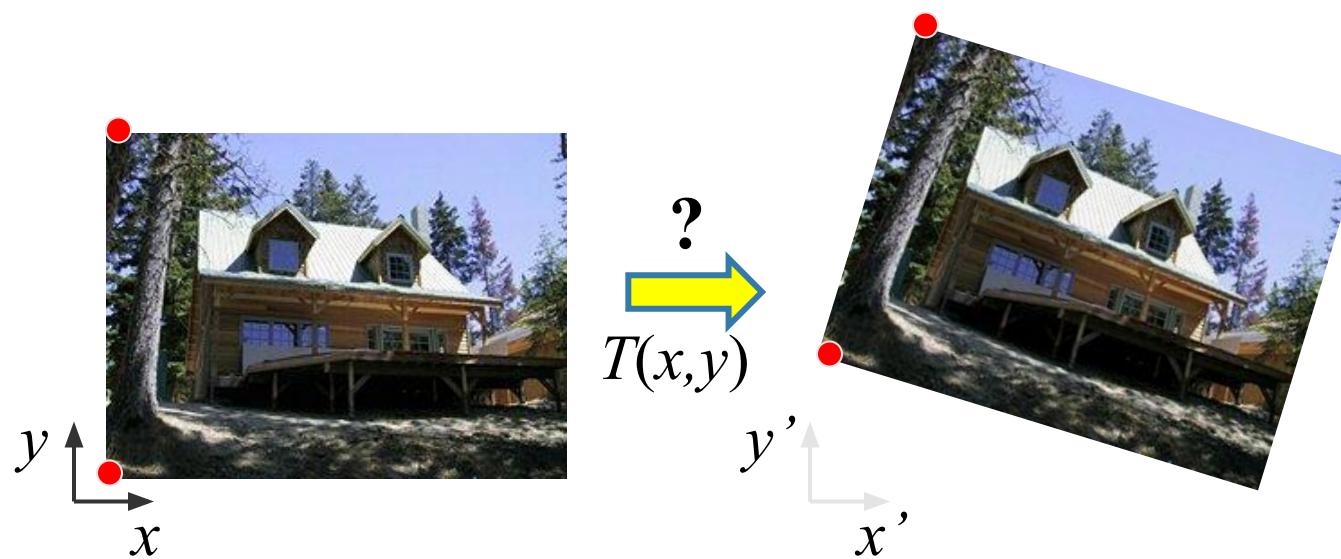
- How many correspondences needed for translation?
- How many Degrees of Freedom?
- What is the transformation matrix?

$$\mathbf{M} = \begin{bmatrix} 1 & 0 & p'_x - p_x \\ 0 & 1 & p'_y - p_y \\ 0 & 0 & 1 \end{bmatrix}$$



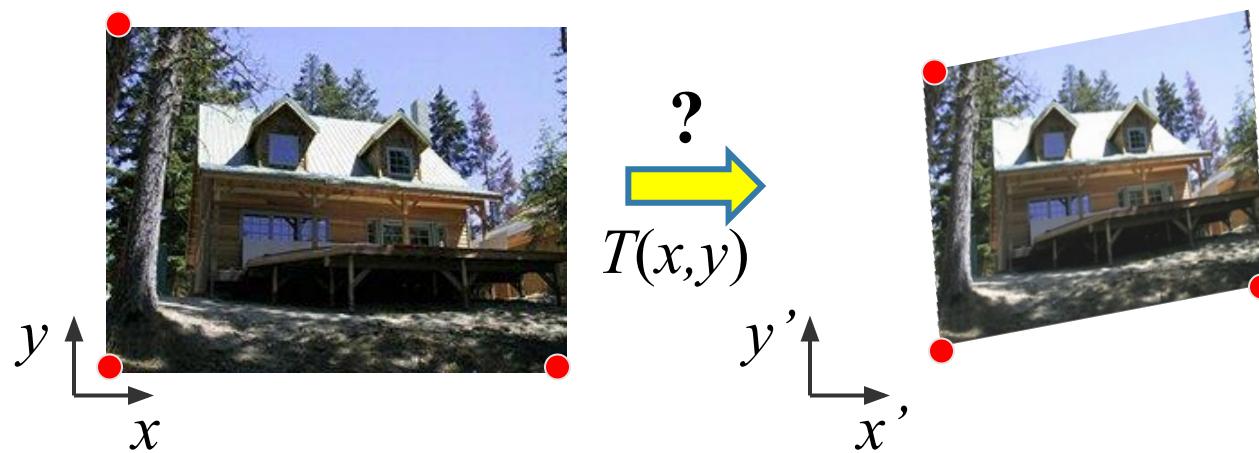
# Euclidian: # correspondences?

- How many correspondences needed for translation+rotation?
- How many DOF?



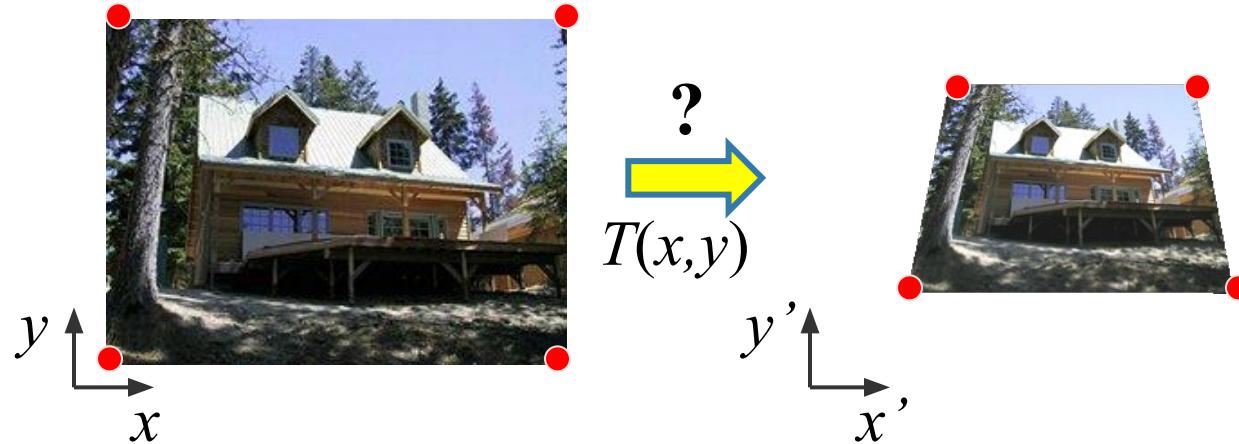
# Affine: # correspondences?

- How many correspondences needed for affine?
- How many DOF?



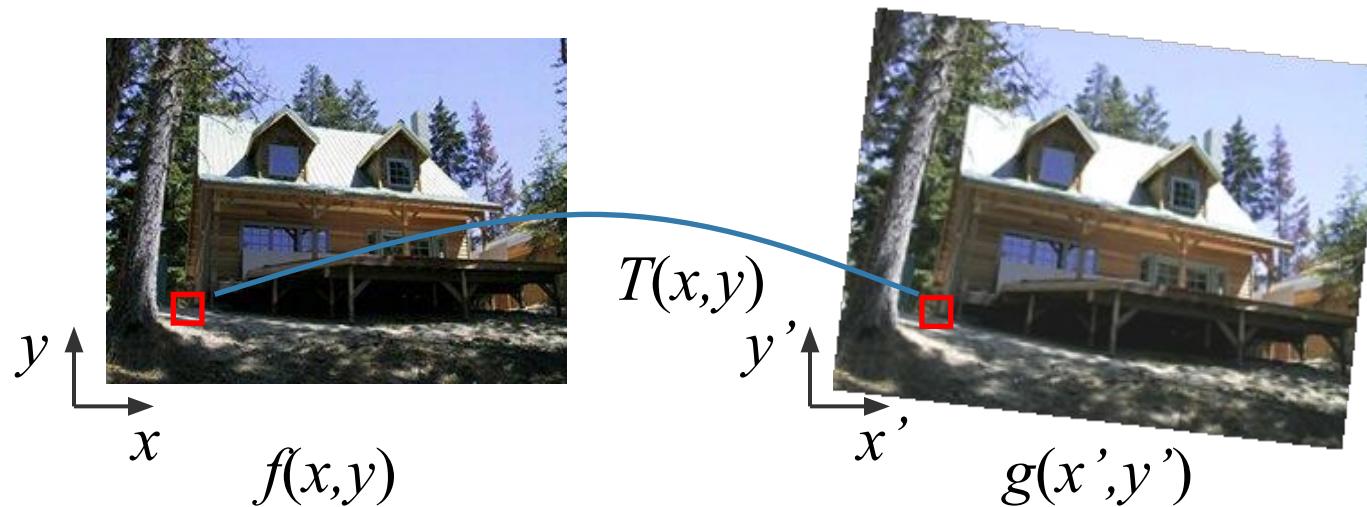
# Projective: # correspondences?

- How many correspondences needed for projective?
- How many DOF?



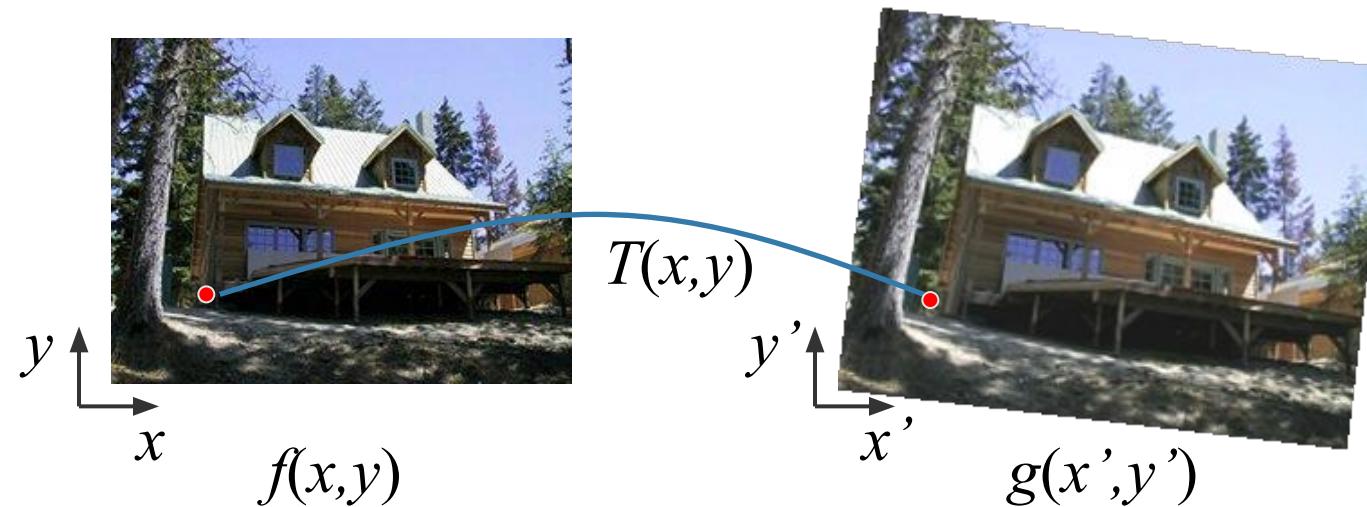
# Image warping

- Given a coordinate transform  $(x',y') = T(x,y)$  and a source image  $f(x,y)$ , how do we compute a transformed image?



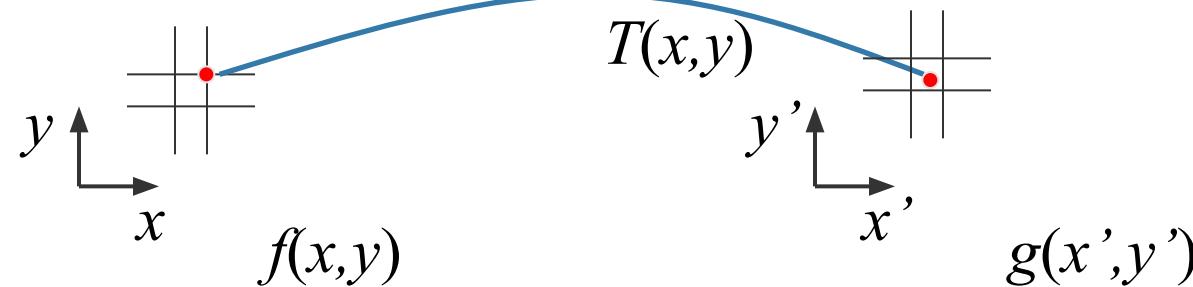
# Forward warping

- Send each pixel  $f(x,y)$  to its corresponding location
- $(x',y') = T(x,y)$  in the second image
- What if pixel lands “between” two pixels?



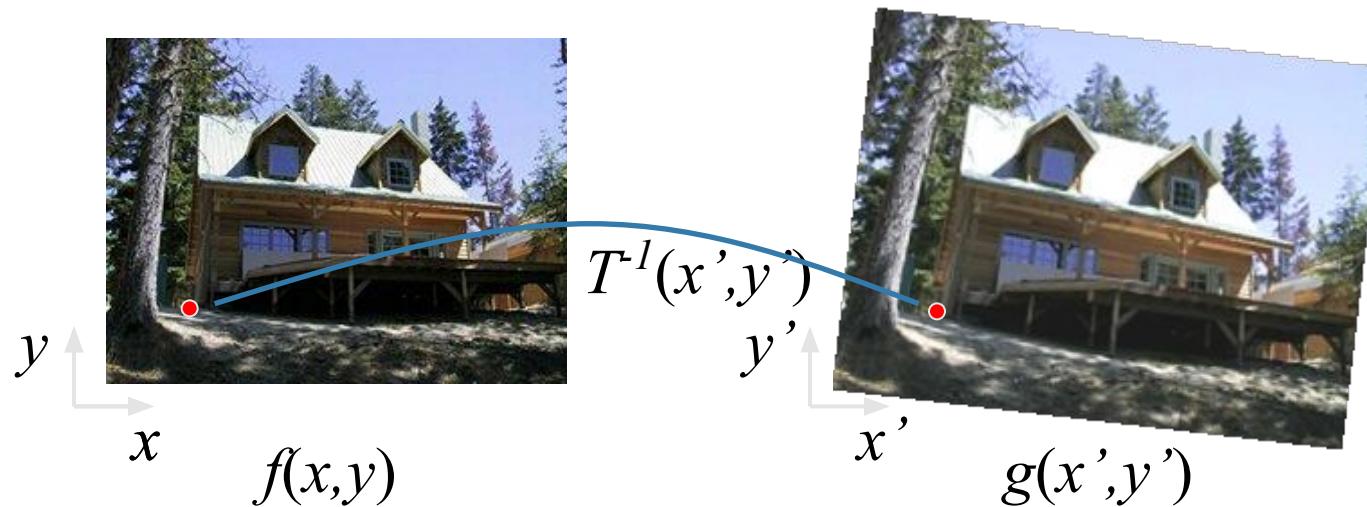
# Forward warping

- Send each pixel  $f(x,y)$  to its corresponding location
- $(x',y') = T(x,y)$  in the second image
- What if pixel lands “between” two pixels?
  - Distribute color among neighboring pixels  $(x',y')$ 
    - Known as “splatting”



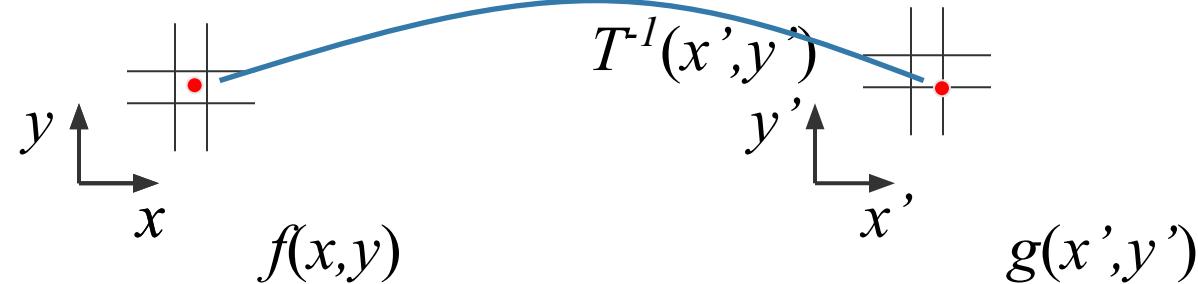
# Inverse warping

- Get each pixel  $g(x',y')$  from its corresponding location
- $(x,y) = T^{-1}(x',y')$  in the first image
- What if pixel comes from “between” two pixels?



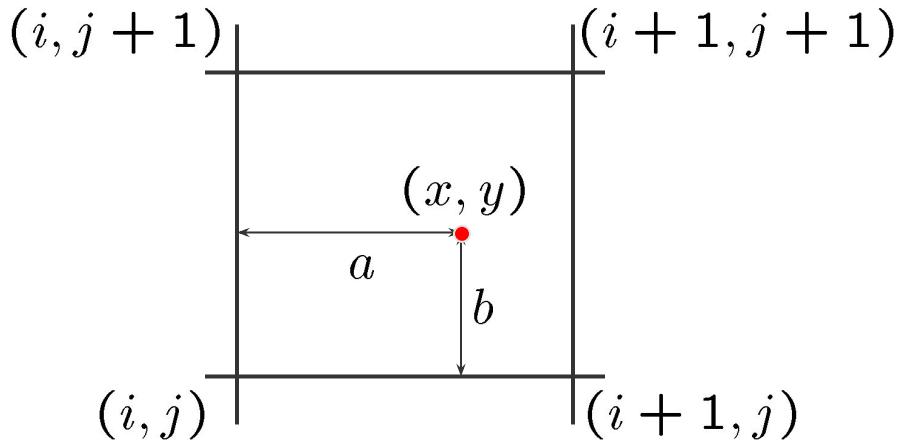
# Inverse warping

- Get each pixel  $g(x',y')$  from its corresponding location
- $(x,y) = T^{-1}(x',y')$  in the first image
- What if pixel comes from “between” two pixels?
- Interpolate color value from neighbors
  - nearest neighbor, bilinear, Gaussian, bicubic



# Bilinear interpolation

## □ A simple method for resampling images



$$\begin{aligned} f(x, y) = & (1 - a)(1 - b) f[i, j] \\ & + a(1 - b) f[i + 1, j] \\ & + ab f[i + 1, j + 1] \\ & + (1 - a)b f[i, j + 1] \end{aligned}$$

# **Forward vs. inverse warping**

- **Which is better?**
- **Usually inverse; eliminates holes**
  - **However, it requires an invertible warp function—not always possible...**