

Image Processing

Geometric Transform

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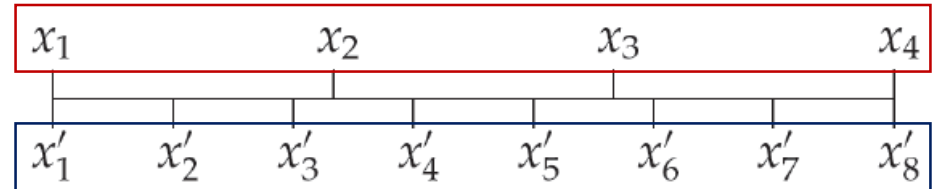
Chungnam National University

Objectives

- Interpolation
 - 1D signal interpolation
 - 2D image interpolation
- Relation between interpolation and spatial filtering
- Parametric transform
 - Scaling
 - Rotation
 - Translation
 - Affine

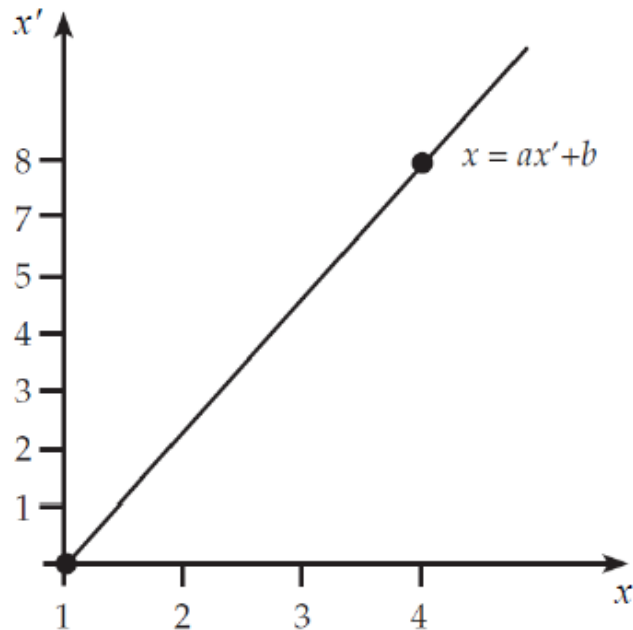
Interpolation

- A set of four values that we wish to enlarge to eight
 - In discrete signal, output x'_i points does not coincide with input x_i points, except for the first and last points
 - Input signal: 4 values
 - Output signal: 8 values



- Interpolation
 - Estimate $f(x'_i)$ based on the known values of nearby $f(x_i)$, i.e., neighboring pixels
 - 1. Compute the location x'_i
 - 2. Compute the function value (intensity) $f(x'_i)$

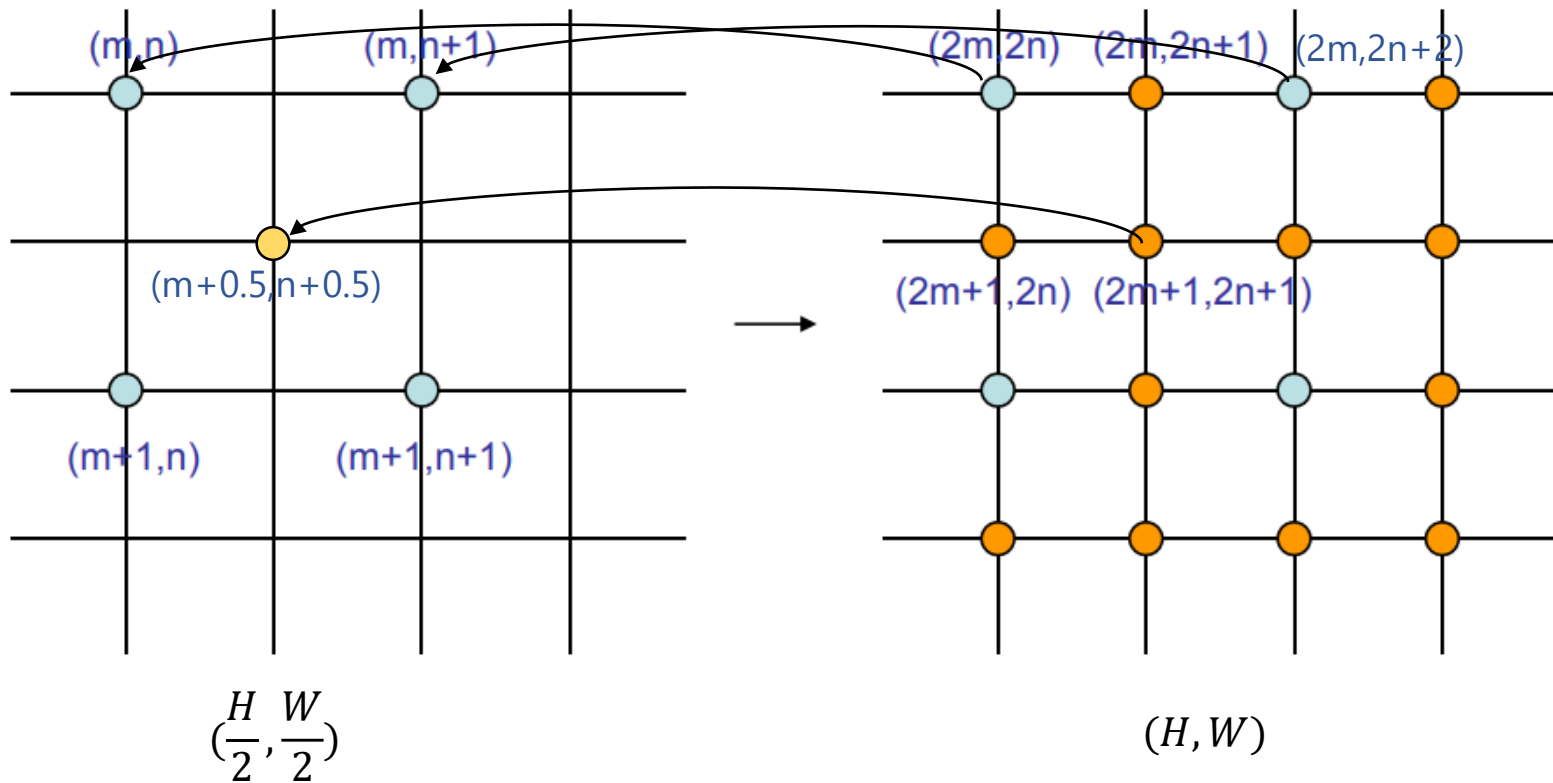
Compute the location x'_i -1D



$$\begin{cases} 1 = a + b \\ 4 = 8a + b \end{cases}$$

$$x = \frac{3}{7}x' + \frac{4}{7}$$

Compute the location x'_i -2D



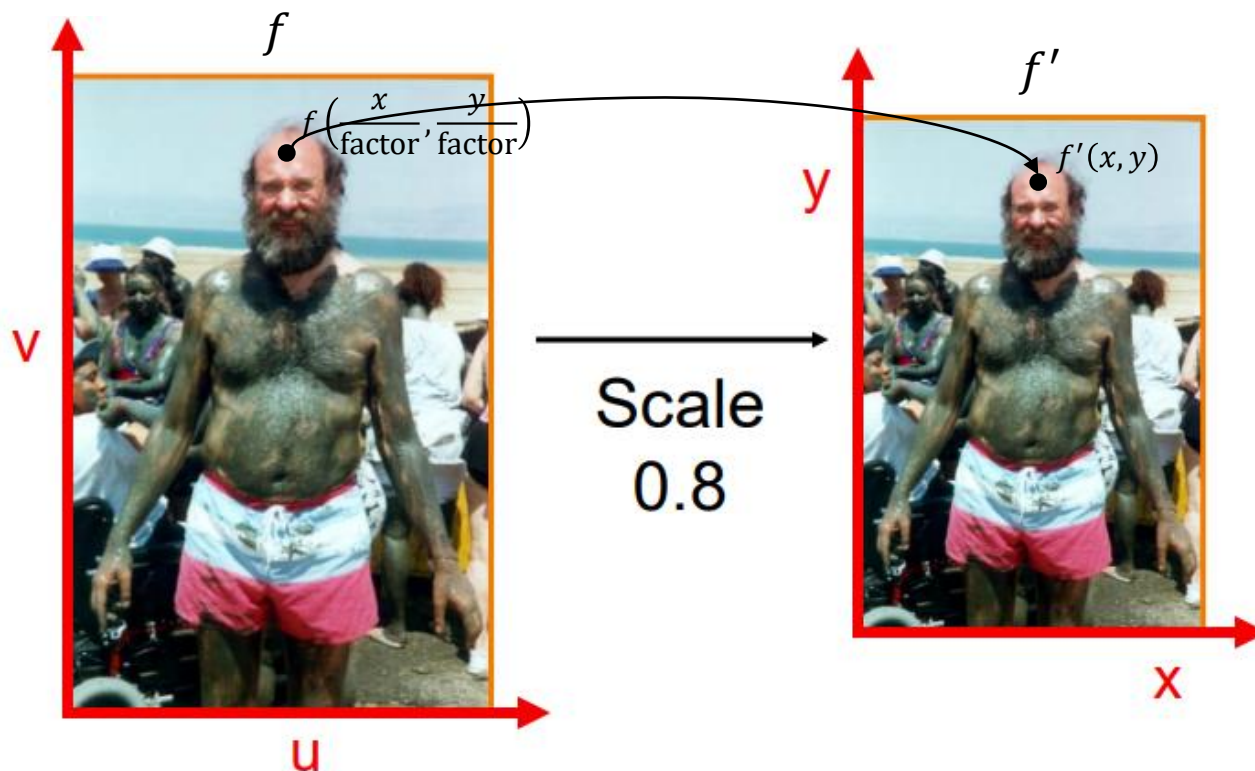
Factor of 2 Up-Sampling

Compute the location x'_i -2D

- Scale factor

- $x = \text{factor} \times u$
- $y = \text{factor} \times v$

$$f'(x, y) = f\left(\frac{x}{\text{factor}}, \frac{y}{\text{factor}}\right)$$

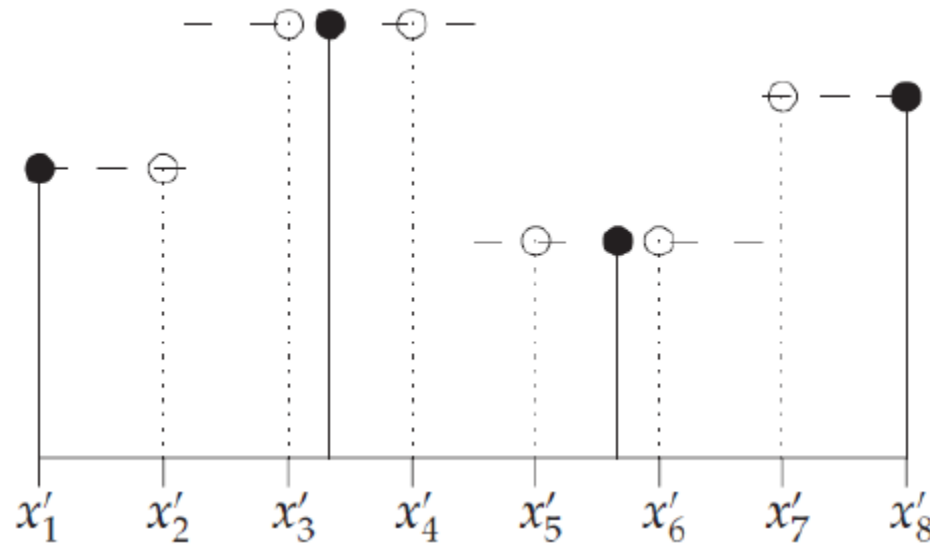


Compute the function value $f(x'_i)$

- Several interpolation methods in MATLAB
 - function `imresize()`
 - 'nearest': nearest-neighbor interpolation
 - 'bilinear': bilinear interpolation
 - 'bicubic': cubic interpolation (default in MATLAB)
 - 'lanczos2' and 'lanczos3': the best performance

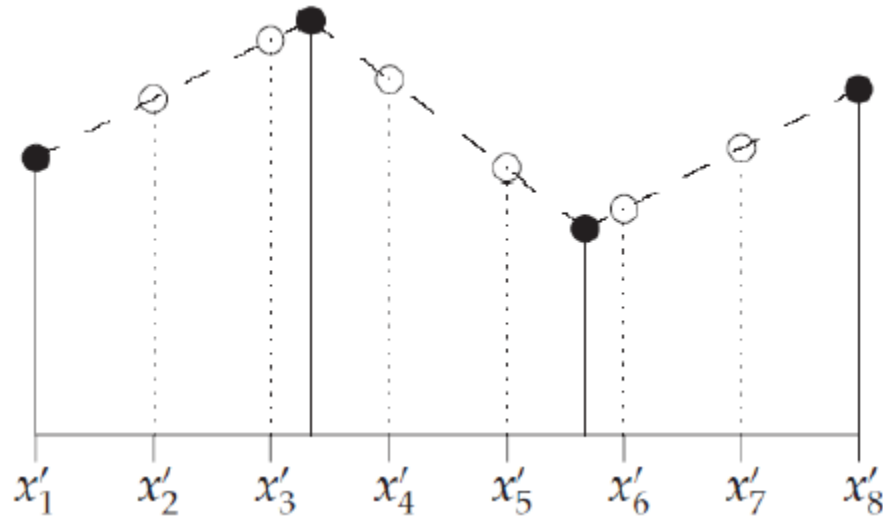
Compute the function value $f(x'_i)$

- Nearest-neighbor interpolation
 - Simply copy-and-paste the neighbor intensity value



Compute the function value $f(x'_i)$

- Linear interpolation
 - Using linear equation $f(m) = am + b$



Compute the function value $f(x'_i)$

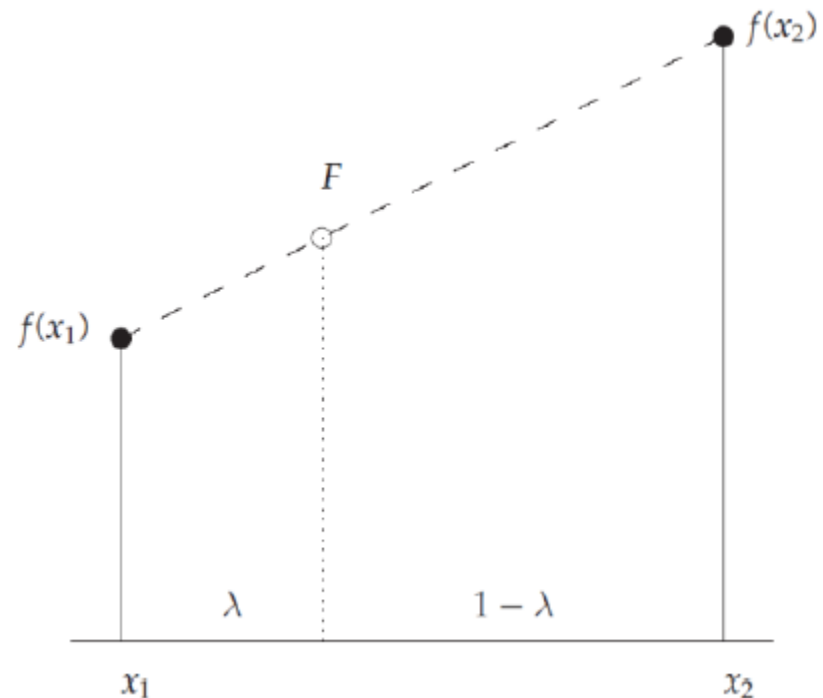
- Linear interpolation
 - Linear equation: using proportional formula

$$\frac{F - f(x_1)}{\lambda} = \frac{f(x_2) - f(x_1)}{1}$$

⇒ $F = \lambda f(x_2) + (1 - \lambda)f(x_1)$

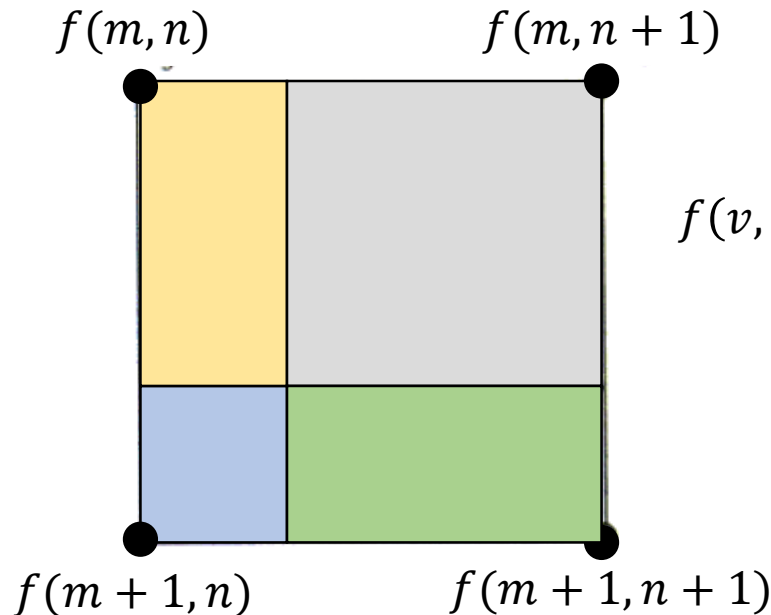
It is assumed that the distance between x_1 and x_2 is normalized (= 1).

Why?



Bilinear Interpolation (in 2D)

- Applying the linear interpolation to 2D data (image)

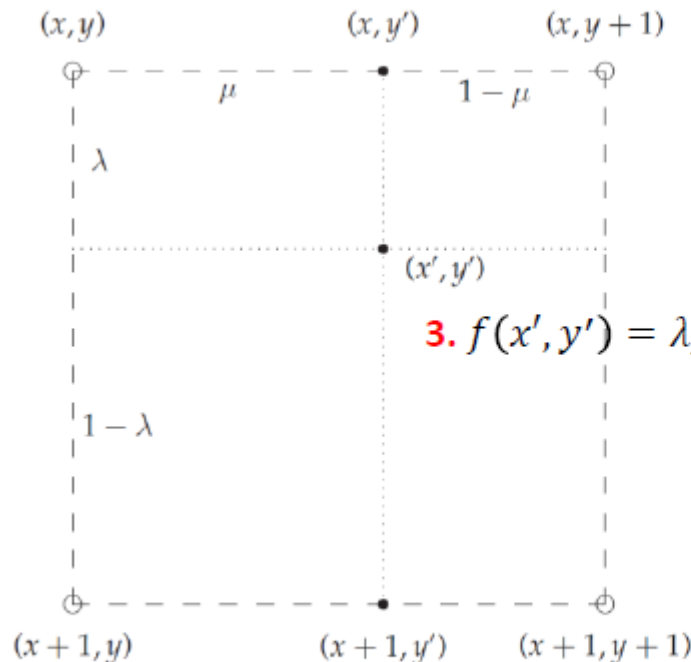


$$\begin{aligned} f(v, u) = & (1 - s)(1 - t) \cdot f(m, n) \\ & + s(1 - t) \cdot f(m, n + 1) \\ & + (1 - s)t \cdot f(m + 1, n) \\ & + st \cdot f(m + 1, n + 1) \end{aligned}$$

Bilinear Interpolation (in 2D)

- Applying the linear interpolation to 2D data (image)

1. $f(x, y') = \mu f(x, y + 1) + (1 - \mu)f(x, y)$



3. $f(x', y') = \lambda f(x + 1, y') + (1 - \lambda)f(x, y')$

FIGURE 6.8 Interpolation between four image points.

2. $f(x + 1, y') = \mu f(x + 1, y + 1) + (1 - \mu)f(x + 1, y)$

Bilinear Interpolation (in 2D)

- Interpolate $f\left(\frac{1}{2}, 0\right)$ using $f(0,0)$ and $f(1,0)$
- Interpolate $f\left(\frac{1}{2}, 1\right)$ using $f(0,1)$ and $f(1,1)$
- Interpolate $f\left(\frac{1}{2}, \frac{1}{2}\right)$ using $f\left(\frac{1}{2}, 0\right)$ and $f\left(\frac{1}{2}, 1\right)$

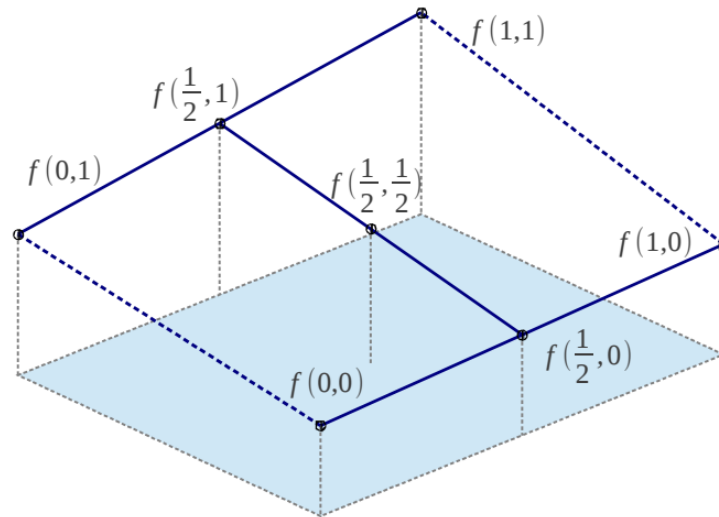


Image Interpolation

- Linear interpolation

Nearest neighbor vs. Bilinear interpolation



FIGURE 6.9 *The head.*



(a)

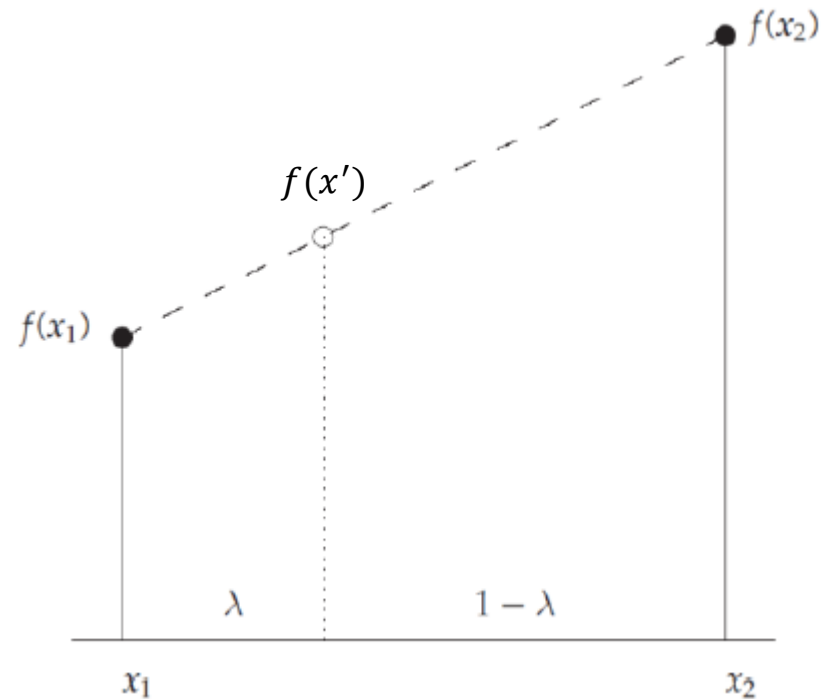
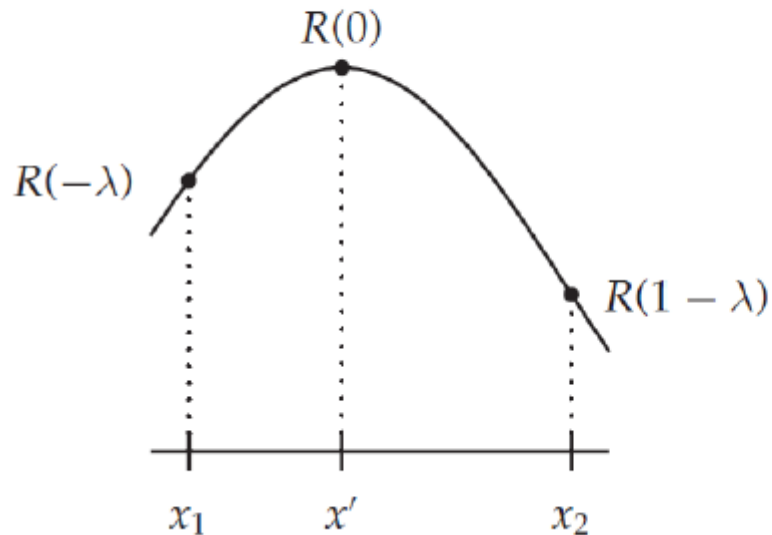


(b)

General Interpolation

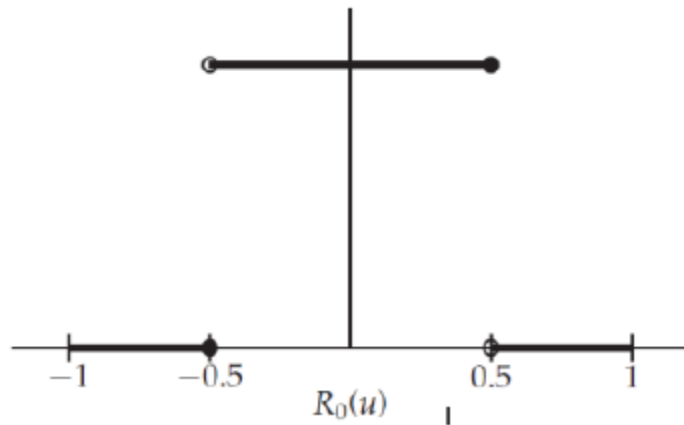
- More general form for interpolation

$$f(x') = R(-\lambda) f(x_1) + R(1 - \lambda) f(x_2)$$

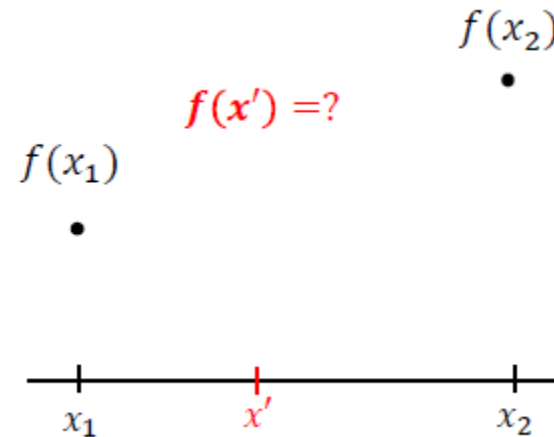


Linear interpolation

Nearest Neighbor Interpolation



$$R_0(u) = \begin{cases} 1 & \text{if } -0.5 < u \leq 0.5 \\ 0 & \text{otherwise} \end{cases}$$



$$\begin{aligned} x' - x_1 &= \lambda \\ x_2 - x' &= 1 - \lambda \end{aligned}$$

$$f(x') = R_0(-\lambda) f(x_1) + R_0(1 - \lambda) f(x_2)$$

$$\Rightarrow f(x') = \begin{cases} f(x_1) & \text{if } \lambda < 0.5 \\ f(x_2) & \text{otherwise} \end{cases}$$

Linear Interpolation

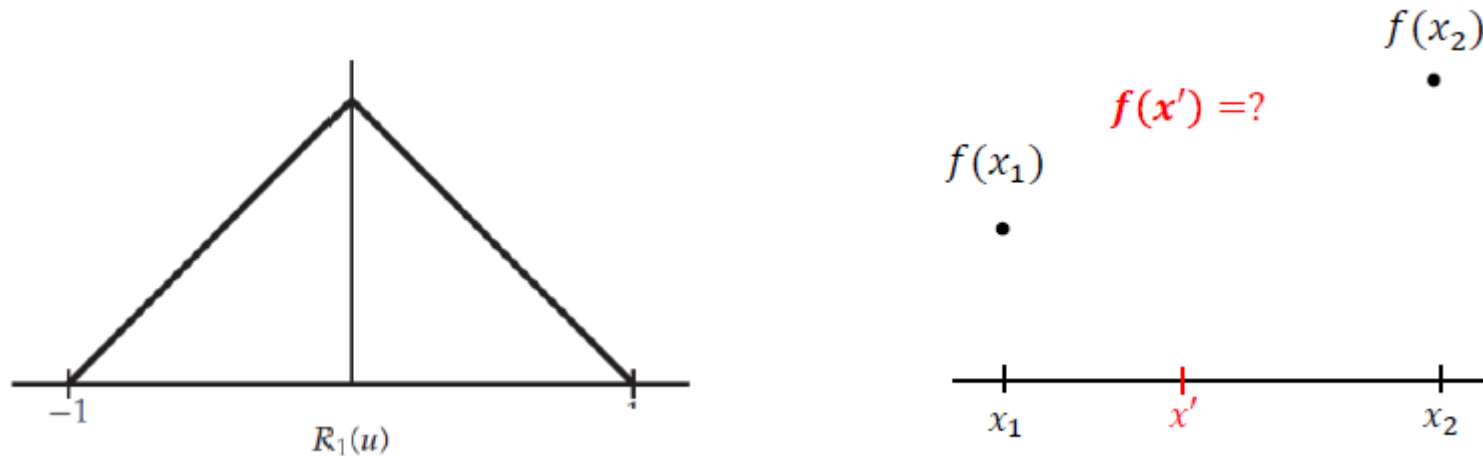


FIGURE 6.12 Two interpolation functions.

$$R_1(u) = \begin{cases} 1 + u & \text{if } u \leq 0 \\ 1 - u & \text{otherwise} \end{cases}$$

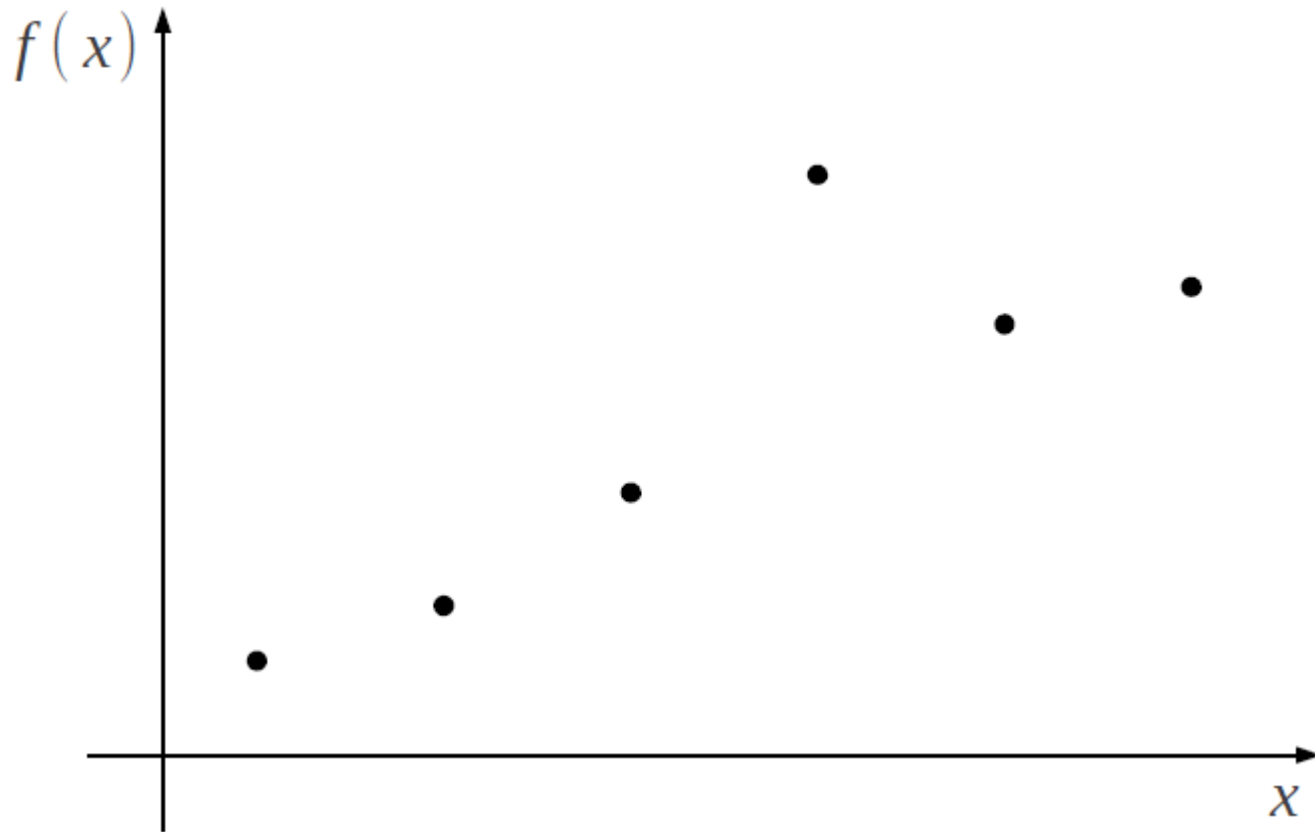
$$\begin{aligned} x' - x_1 &= \lambda \\ x_2 - x' &= 1 - \lambda \end{aligned}$$

$$f(x') = R_1(-\lambda) f(x_1) + R_1(1 - \lambda) f(x_2)$$

$$\Rightarrow f(x') = (1 - \lambda) f(x_1) + \lambda f(x_2)$$

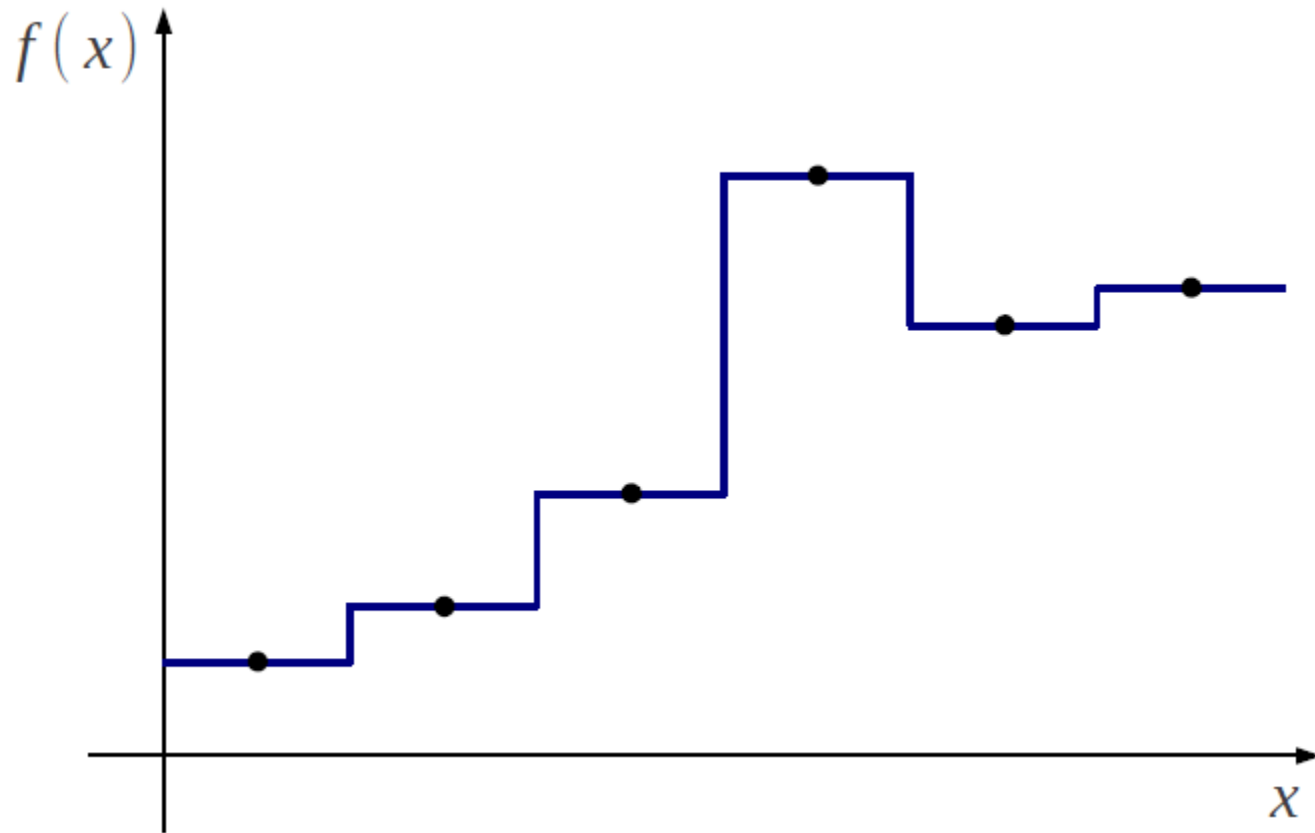
Cubic Interpolation

- Interpolation



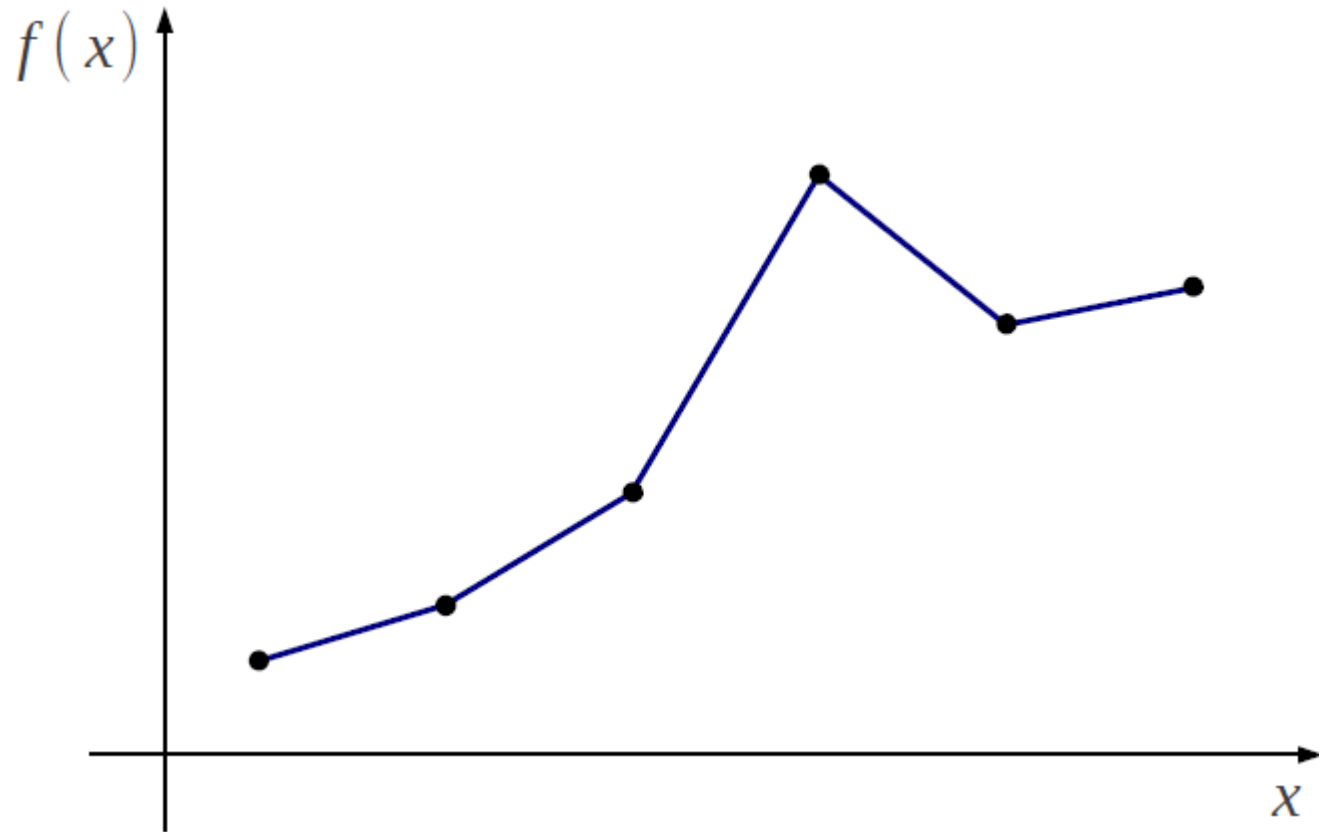
Cubic Interpolation

- Nearest neighbor interpolation



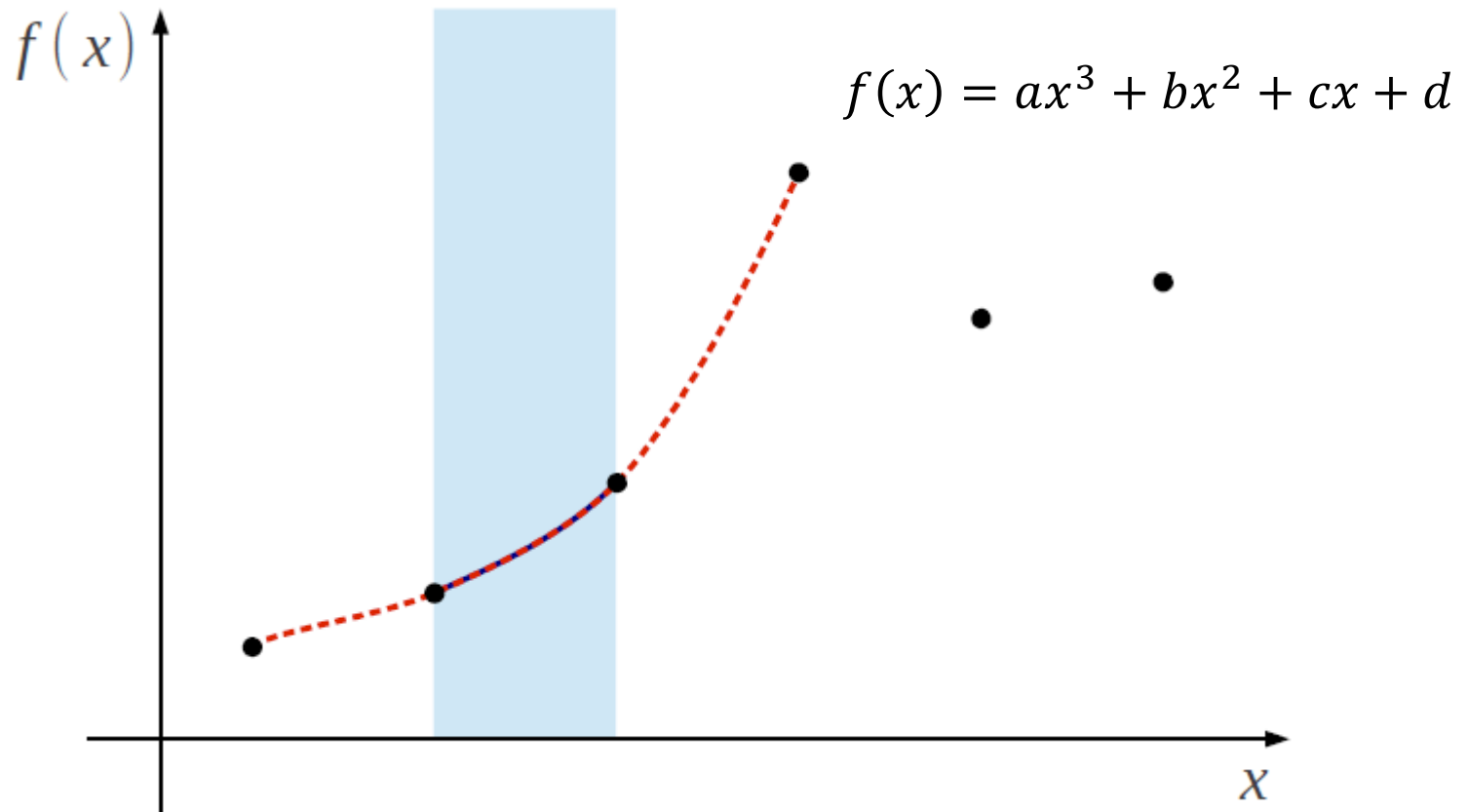
Cubic Interpolation

- Linear interpolation



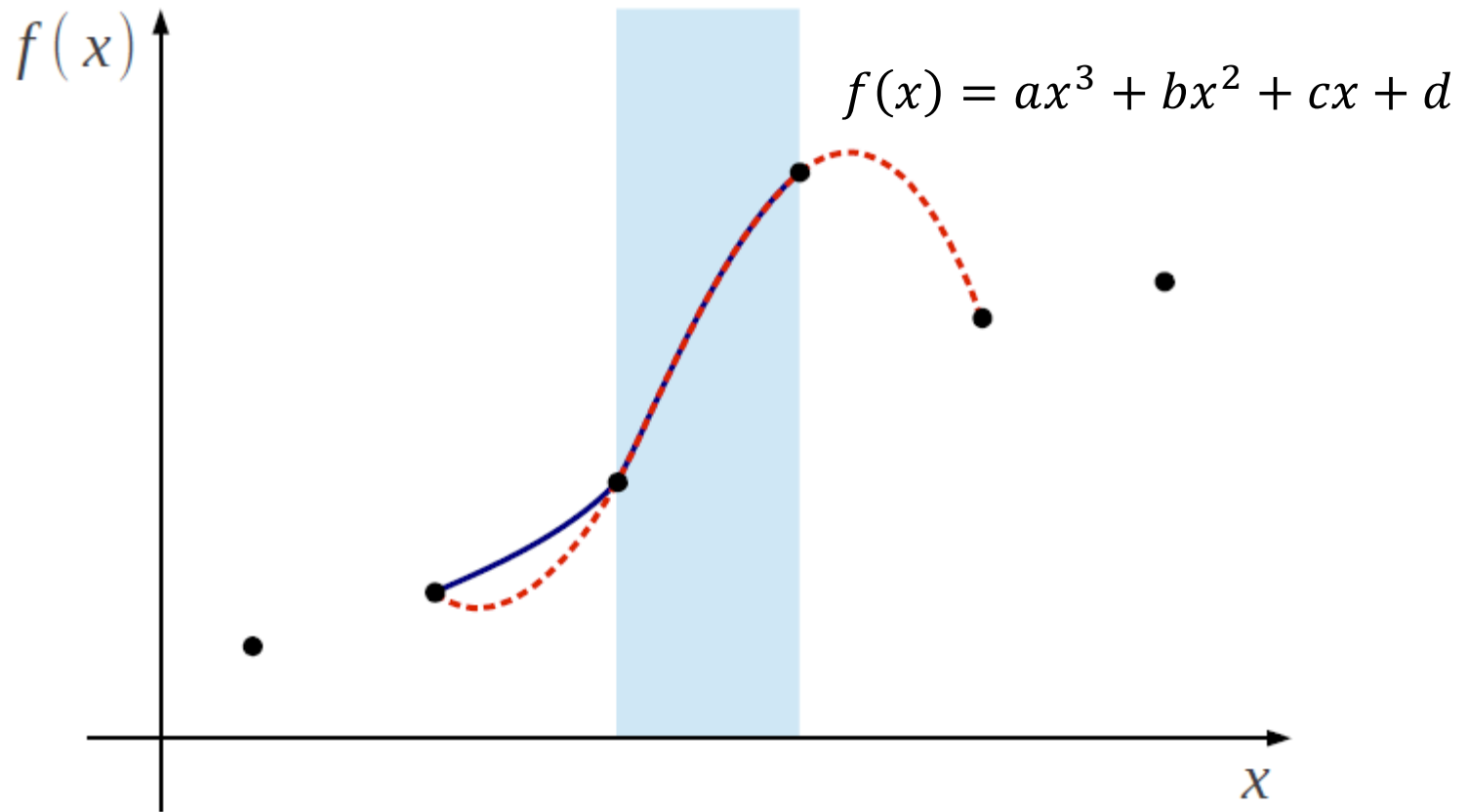
Cubic Interpolation

- Cubic interpolation



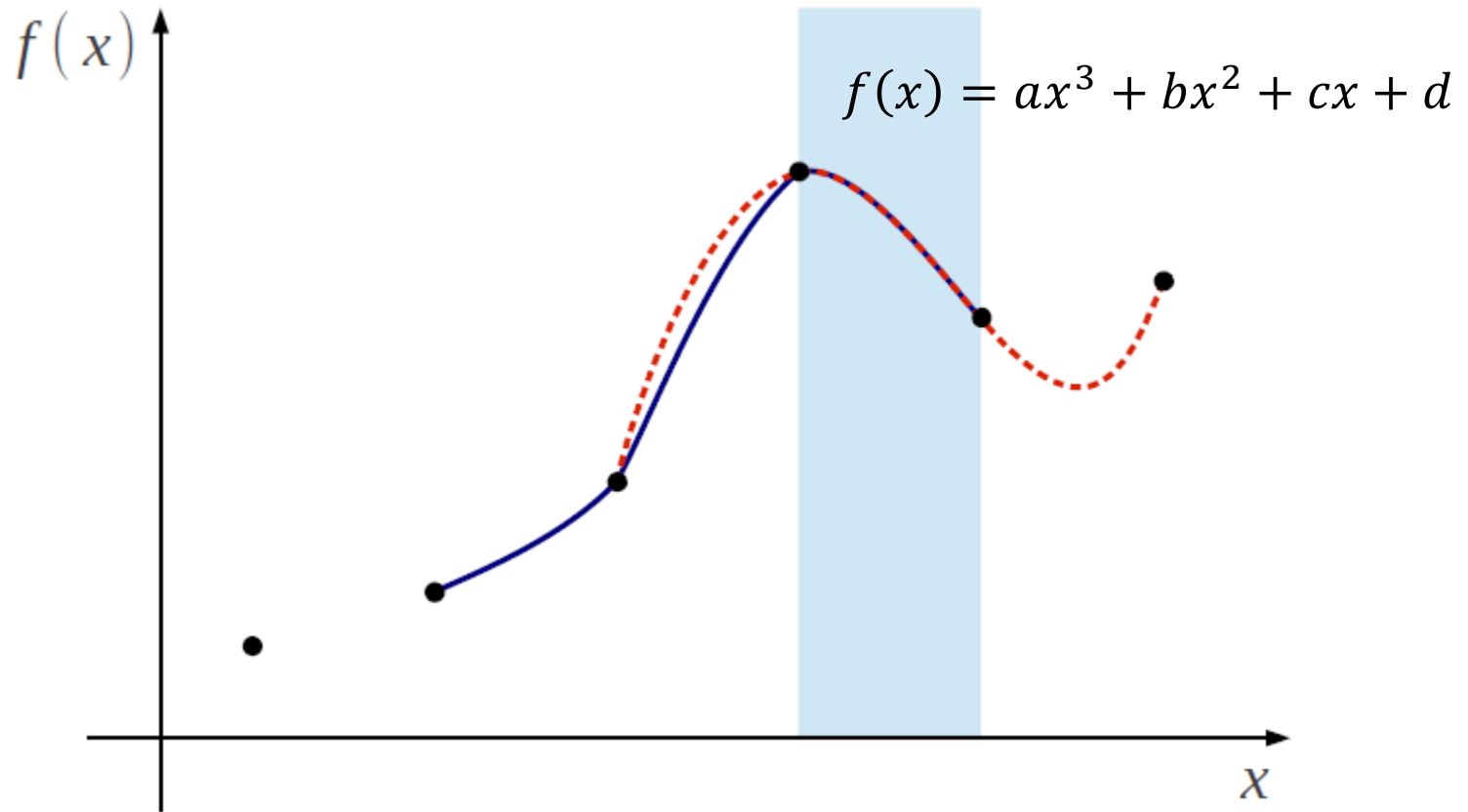
Cubic Interpolation

- Cubic interpolation



Cubic Interpolation

- Cubic interpolation



Cubic Interpolation

$$R_3(u) = \begin{cases} 1.5|u|^3 - 2.5|u|^2 + 1 & \text{if } |u| \leq 1 \\ -0.5|u|^3 + 2.5|u|^2 - 4|u| + 2 & \text{if } 1 < |u| \leq 2 \end{cases}$$

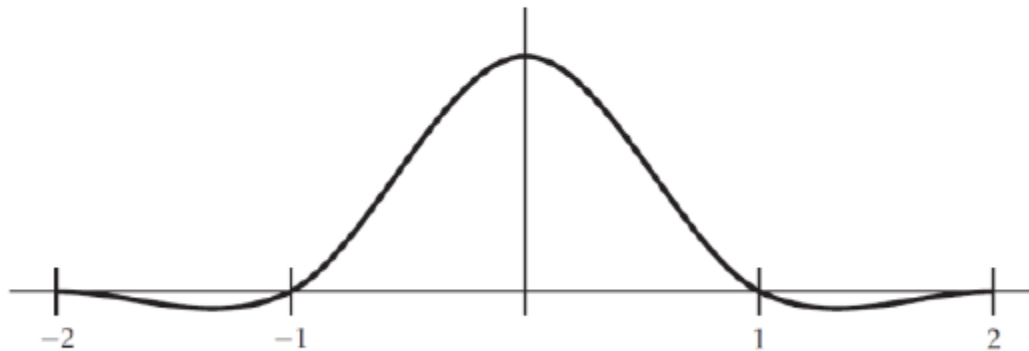
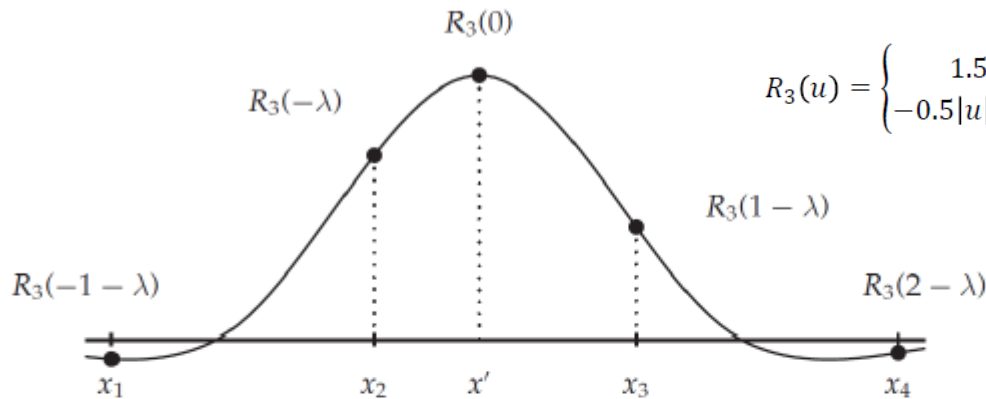


FIGURE 6.13 The cubic interpolation function $R_3(u)$.

Q: What is the difference from R_0 and R_1 ?

Cubic Interpolation

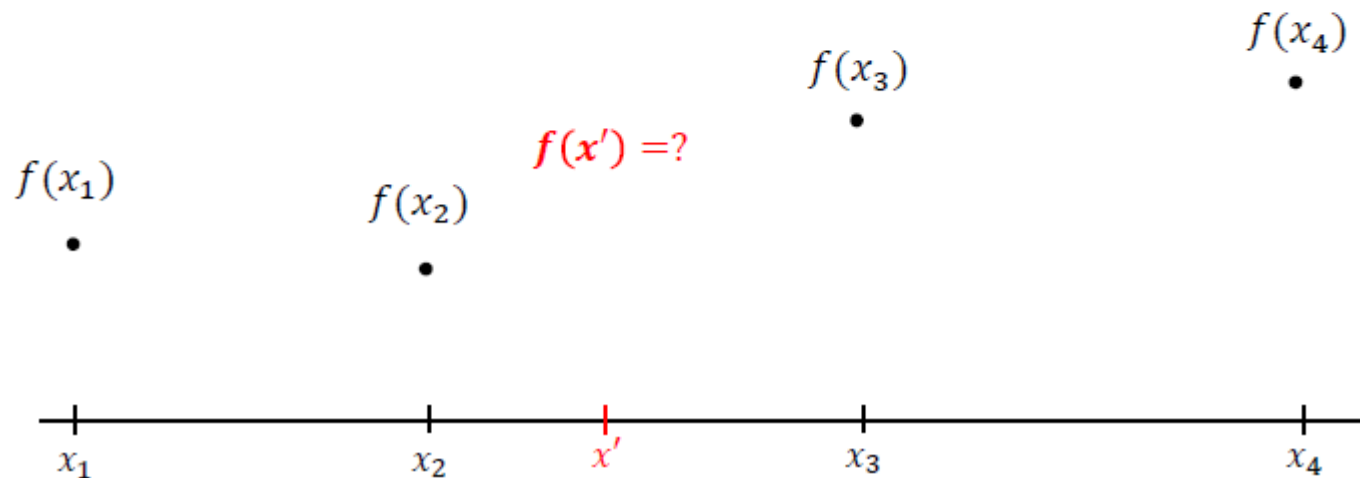
$$f(x') = R_3(-1 - \lambda)f(x_1) + R_3(-\lambda)f(x_2) + R_3(1 - \lambda)f(x_3) + R_4(2 - \lambda)f(x_4),$$



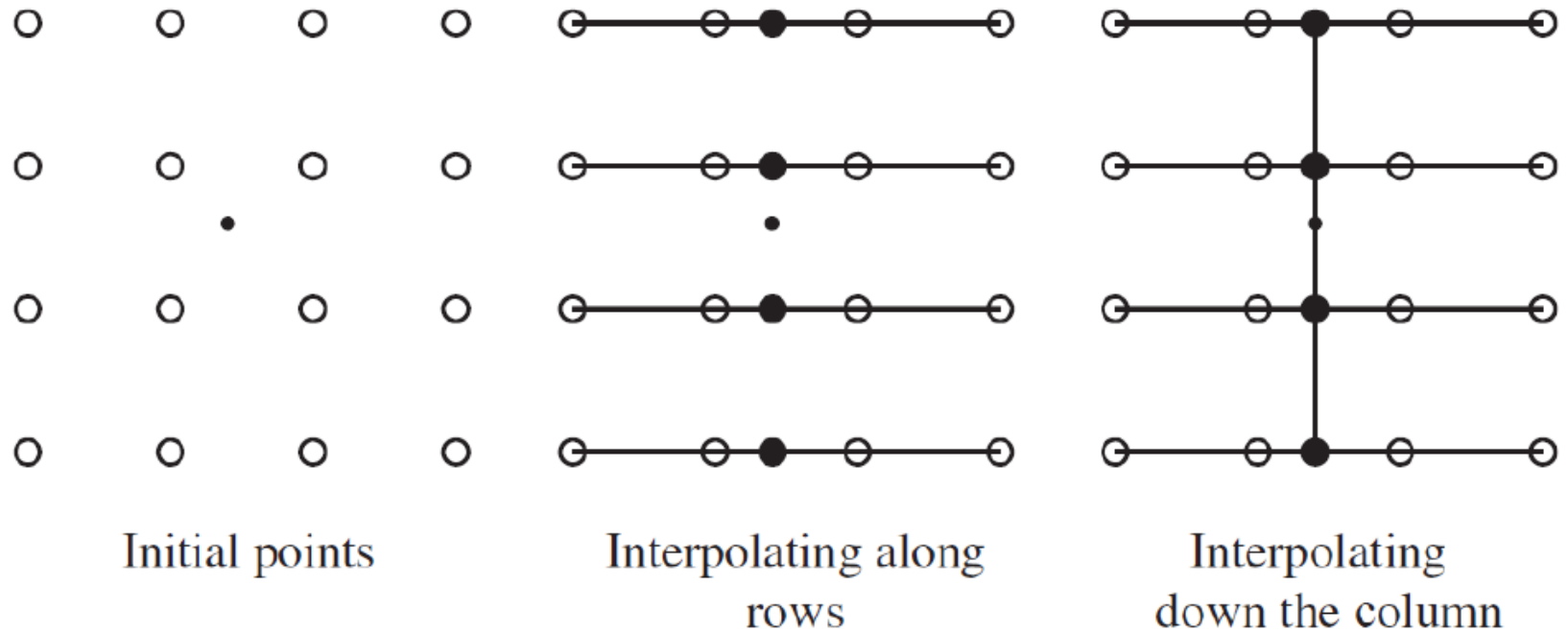
$$R_3(u) = \begin{cases} 1.5|u|^3 - 2.5|u|^2 + 1 & \text{if } |u| \leq 1 \\ -0.5|u|^3 + 2.5|u|^2 - 4|u| + 2 & \text{if } 1 < |u| \leq 2 \end{cases}$$

FIGURE 6.14 Using $R_3(u)$ for interpolation.

$$\begin{aligned} x' - x_2 &= \lambda \\ x_3 - x' &= 1 - \lambda \end{aligned}$$



Bicubic Interpolation



Nearest Neighbor

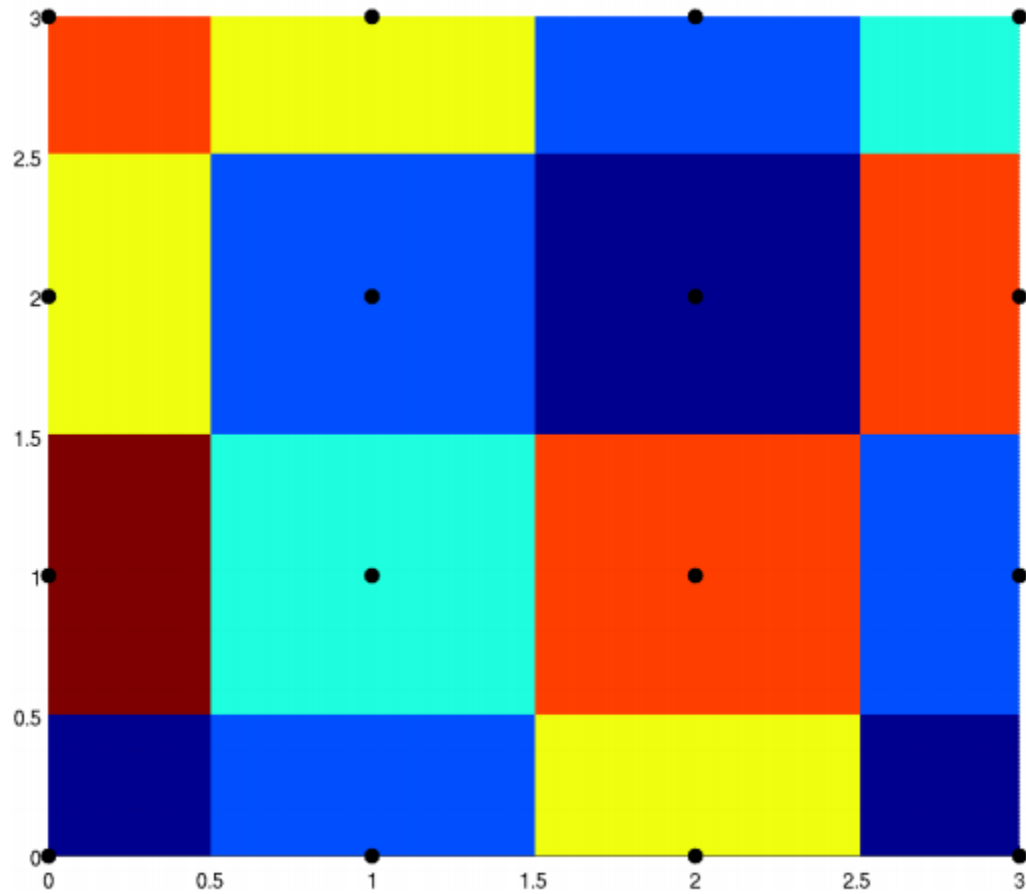


Figure : Nearest Neighbour Interpolation

Bilinear vs. Bicubic

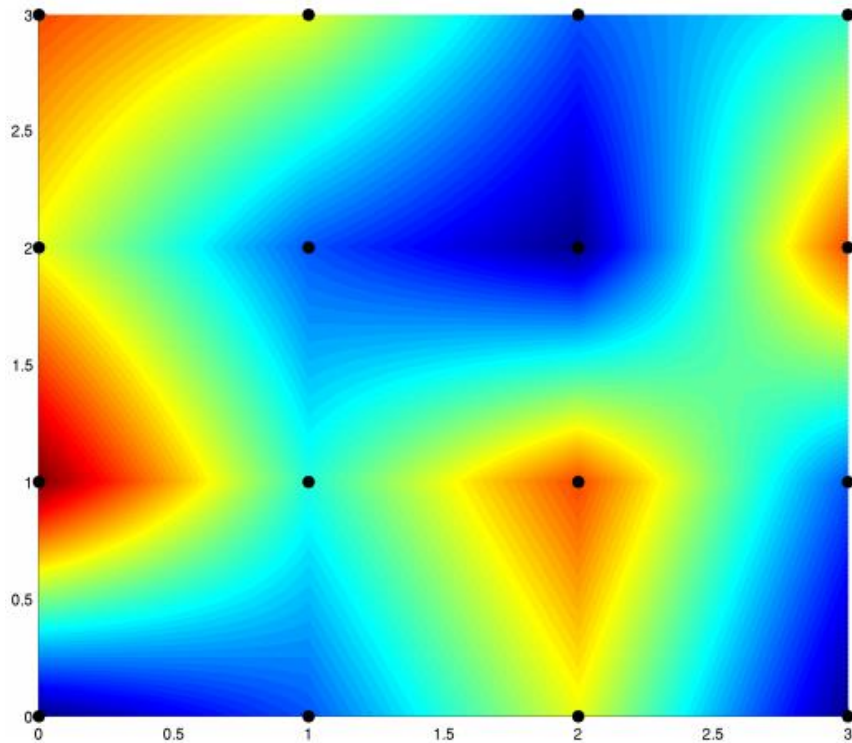


Figure : Bilinear Interpolation

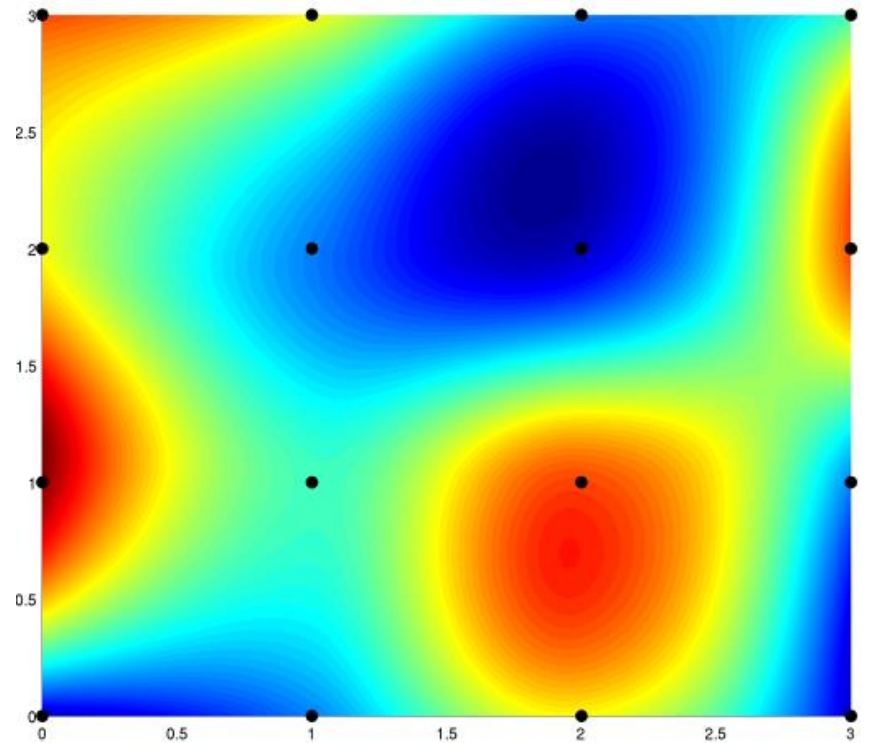


Figure : Bicubic Spline Interpolation

Interpolation



Nearest Neighbor vs. Bilinear vs. Bicubic interpolation

Interpolation with Filtering

- Interpolation \approx Filtering
- 2 times interpolation

– $H \times W$ image $\rightarrow 2H \times 2W$ image

1) Generate zero-interleaved image

```
>> m=magic(4)

m =

    16     2     3    13
     5    11    10     8
     9     7     6    12
     4    14    15     1
```



```
>> m2=zerosint(m)

m2 =

    16     0     2     0     3     0    13     0
     0     0     0     0     0     0     0     0
     5     0    11     0    10     0     8     0
     0     0     0     0     0     0     0     0
     9     0     7     0     6     0    12     0
     0     0     0     0     0     0     0     0
     4     0    14     0    15     0     1     0
     0     0     0     0     0     0     0     0
```

2) Replace the zeros by applying a spatial filter

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

$$\frac{1}{4} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

$$\frac{1}{64} \begin{bmatrix} 1 & 4 & 6 & 4 & 1 \\ 4 & 16 & 24 & 16 & 4 \\ 6 & 24 & 36 & 24 & 6 \\ 4 & 16 & 24 & 16 & 4 \\ 1 & 4 & 6 & 4 & 1 \end{bmatrix}$$

Interpolation with Filtering

- Interpolation \approx Filtering
- 2 times interpolation

– $H \times W$ image $\rightarrow 2H \times 2W$ image

1) Generate zero-interleaved image

```
>> m=magic(4)

m =

    16     2     3    13
     5    11    10     8
     9     7     6    12
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```



```
>> m2=zerosint(m)

m2 =

    16     0     2     0     3     0    13     0
     0     0     0     0     0     0     0     0
     5     0    11     0    10     0     8     0
     0     0     0     0     0     0     0     0
     9     0     7     0     6     0    12     0
     0     0     0     0     0     0     0     0
     4     0    14     0    15     0     1     0
     0     0     0     0     0     0     0     0
```

2) Replace the zeros by applying a spatial filter

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

$$\frac{1}{4} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

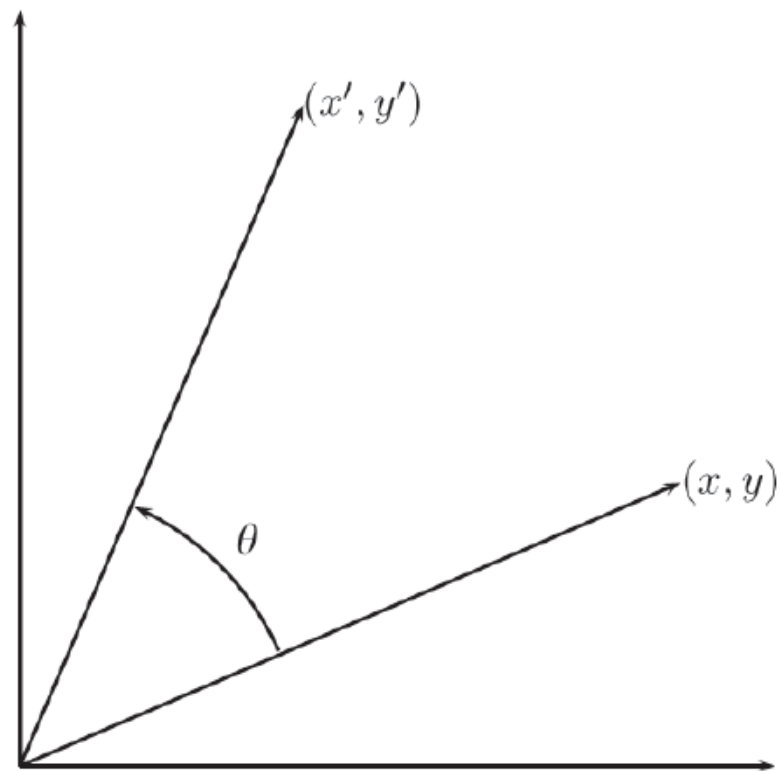
$$\frac{1}{64} \begin{bmatrix} 1 & 4 & 6 & 4 & 1 \\ 4 & 16 & 24 & 16 & 4 \\ 6 & 24 & 36 & 24 & 6 \\ 4 & 16 & 24 & 16 & 4 \\ 1 & 4 & 6 & 4 & 1 \end{bmatrix}$$

Rotation

- In a continuous signal, the rotation can be done using the following equation

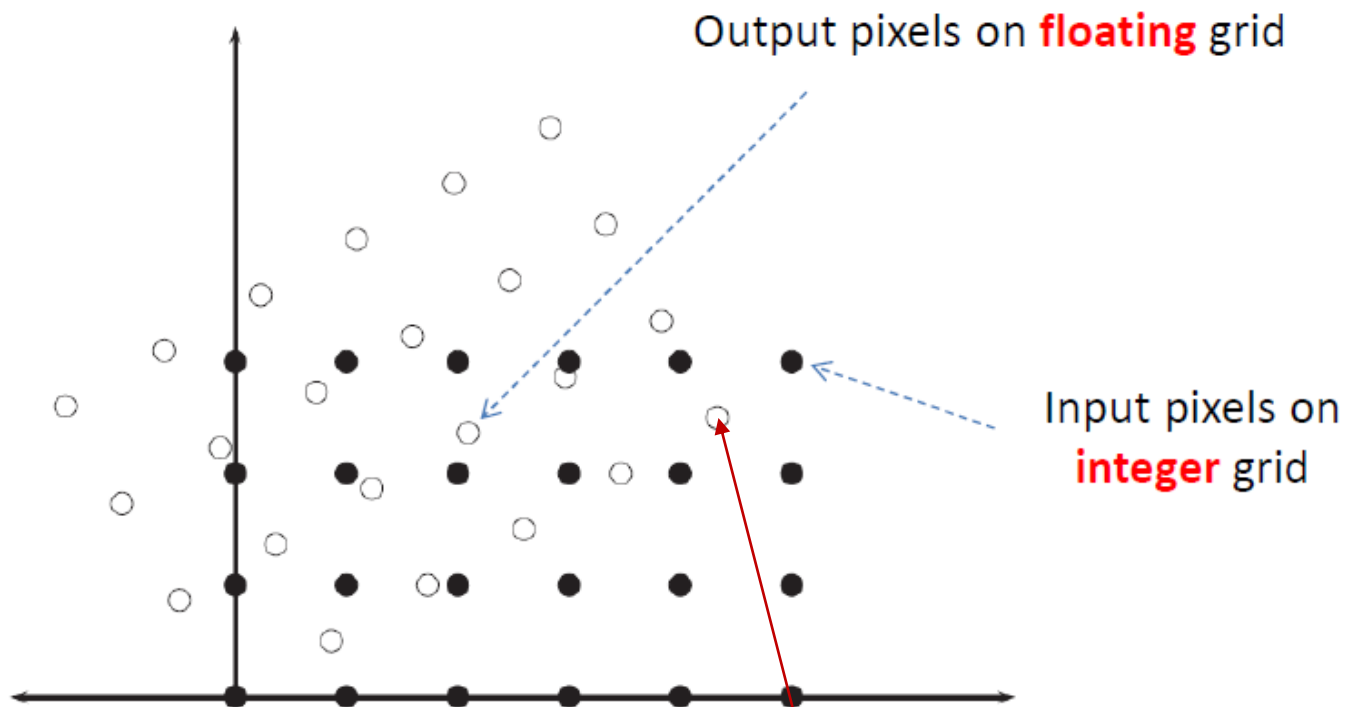
$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x' \\ y' \end{bmatrix}$$



Rotation

- Discrete signal?
 - Rotation does not always produce an integer pixel location
 - So, the interpolation (or re-sampling) is needed

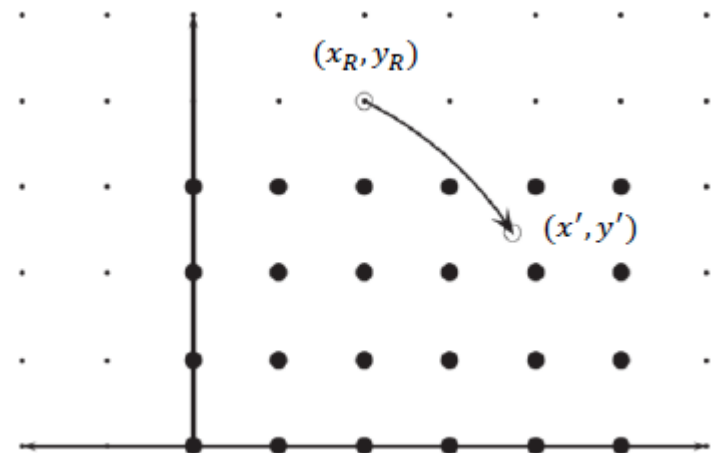
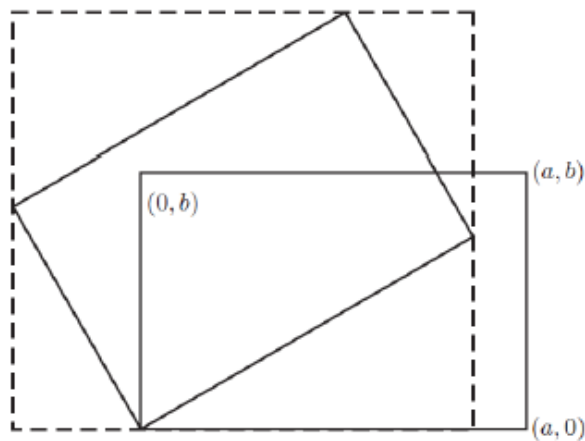


Rotation

- Define an enlarged, rotated image I_R whose size is larger than that of an original image
- Assume $(x_R, y_R) \in I_R$. Then, compute the transform $(x_R, y_R) \rightarrow (x', y')$
Note that (x', y') is a floating pixel

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x_R \\ y_R \end{bmatrix}$$

- Compute $f(x', y')$ using the interpolation technique
- Then $f(x_R, y_R) = f(x', y')$



Rotation in MATLAB

```
>> cr=imrotate(c,60);  
>> imshow(cr)  
>> crc=imrotate(c,60,'bicubic');  
>> imshow(crc)
```



(a)



(b)

Rotation

- In a continuous signal, the rotation can be done using the following equation

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x' \\ y' \end{bmatrix}$$

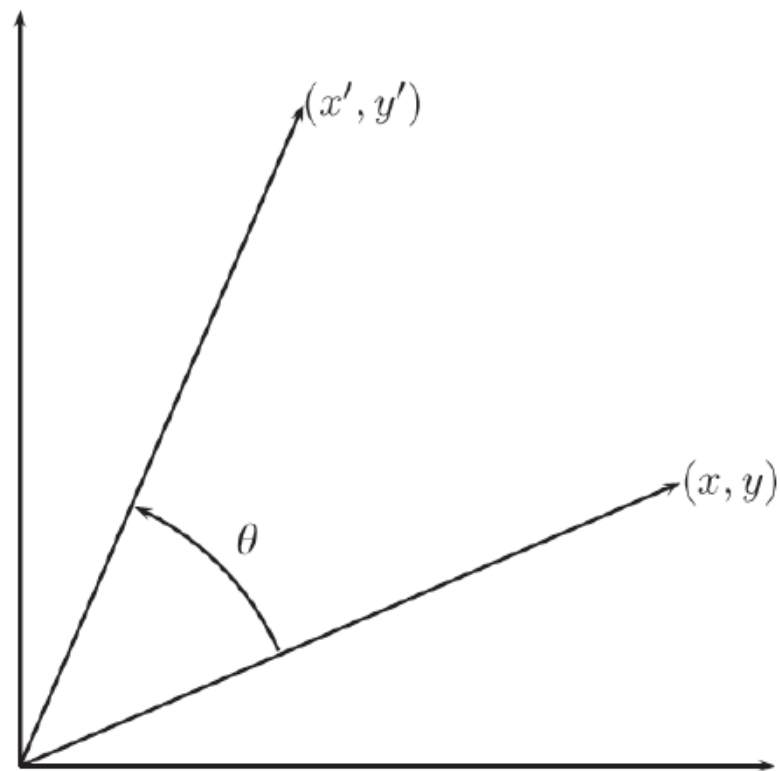


Image Warping

- Parametric (global) warping



translation



rotation



aspect



affine



perspective



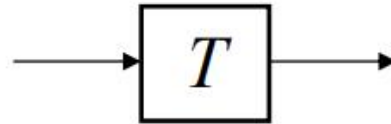
cylindrical

Image Warping

- Parametric (global) warping



$$\mathbf{p} = (x, y)$$



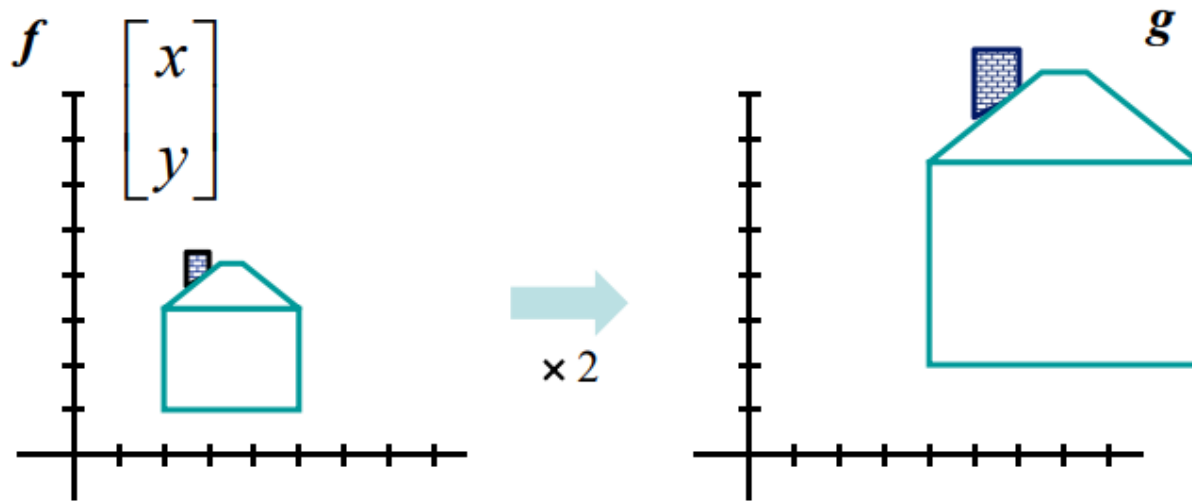
$$\mathbf{p}' = (x', y')$$

- Transform T is a coordinate-changing machine: $\mathbf{p}' = T(\mathbf{p})$
- What does it mean that T is global?
 - Is the same for any point \mathbf{p}
- Represent T as a matrix: $\mathbf{p}' = \mathbf{M}\mathbf{p}$

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

Parametric Warping

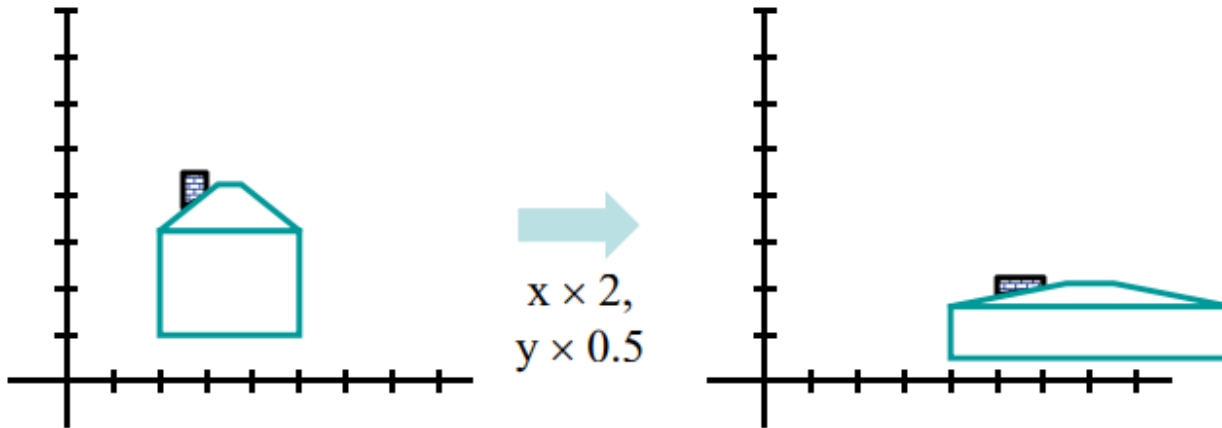
- Scaling
 - Uniform scaling



$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 2x \\ 2y \end{bmatrix}$$

Parametric Warping

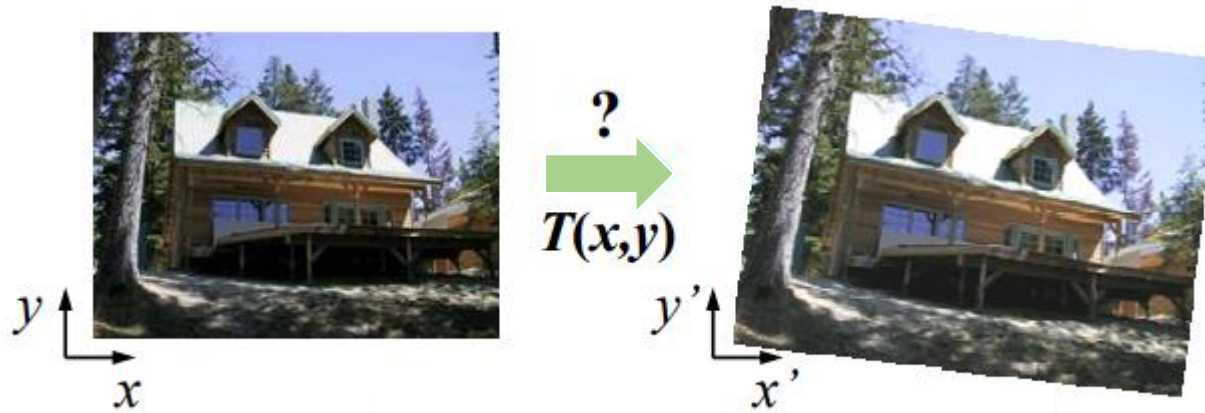
- Scaling
 - Non-uniform scaling: different scalars per component



$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 0.5 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 2x \\ 0.5y \end{bmatrix} \quad \rightarrow \quad \begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} a & 0 \\ 0 & b \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} ax \\ by \end{bmatrix}$$

Parametric Warping

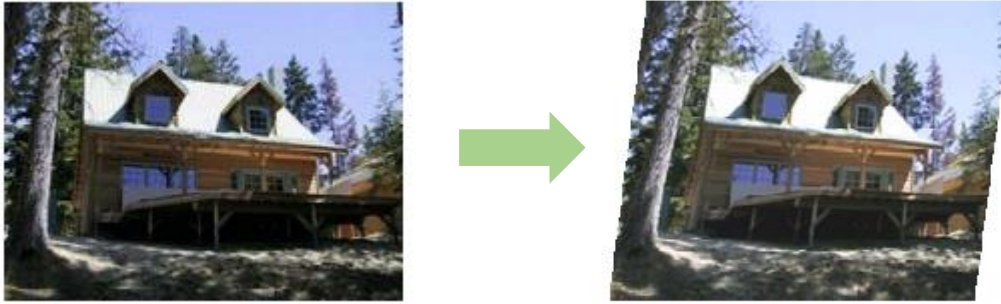
- Rotation



$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

Parametric Warping

- Shear



$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 1 & h_x \\ h_y & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

Parametric Warping

- 2×2 Matrices
 - What types of transformation can be represented with a 2×2 matrix?

Scale matrix: $\begin{bmatrix} x' \\ y' \end{bmatrix} = \mathbf{S}\mathbf{x} = \begin{bmatrix} a & 0 \\ 0 & b \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} ax \\ by \end{bmatrix}$

Rotation matrix: $\begin{bmatrix} x' \\ y' \end{bmatrix} = \mathbf{R}\mathbf{x} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \cos \theta x - \sin \theta y \\ \sin \theta x + \cos \theta y \end{bmatrix}$

Shear matrix: $\begin{bmatrix} x' \\ y' \end{bmatrix} = \mathbf{H}\mathbf{x} = \begin{bmatrix} 1 & h_x \\ h_y & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x + h_x y \\ x h_y + y \end{bmatrix}$



x' is a linear combination of x and y
 y' is a linear combination of x and y


Parametric Warping

- 2×2 Matrices
 - What types of transformation can be represented with a 2×2 matrix?

Scale matrix: $\begin{bmatrix} x' \\ y' \end{bmatrix} = \mathbf{S}\mathbf{x} = \begin{bmatrix} a & 0 \\ 0 & b \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} ax \\ by \end{bmatrix}$

Rotation matrix: $\begin{bmatrix} x' \\ y' \end{bmatrix} = \mathbf{R}\mathbf{x} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \cos \theta x - \sin \theta y \\ \sin \theta x + \cos \theta y \end{bmatrix}$

Shear matrix: $\begin{bmatrix} x' \\ y' \end{bmatrix} = \mathbf{H}\mathbf{x} = \begin{bmatrix} 1 & h_x \\ h_y & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x + h_x y \\ x h_y + y \end{bmatrix}$

 $\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} ax + by \\ dx + ey \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \mathbf{M}\mathbf{x}$

Parametric Warping

- 2×2 Matrices
 - Scaling + Shear + Rotate

$$\mathbf{x}' = \mathbf{R} \left(\mathbf{H}((\mathbf{S}\mathbf{x})) \right) = \mathbf{M}\mathbf{x}$$

If we have to transform many points, it is easier to do one matrix multiplication

Parametric Warping

- What types of transformation can **not** be represented with a 2×2 matrix?

Translation

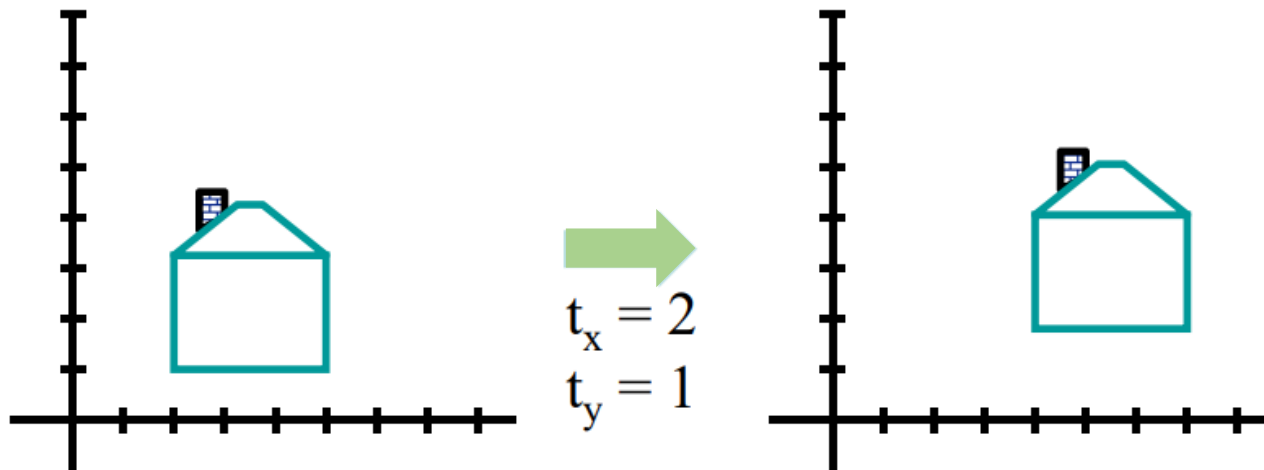
$$x' = x + t_x$$

$$y' = y + t_y$$

Parametric Warping

- Example of translation

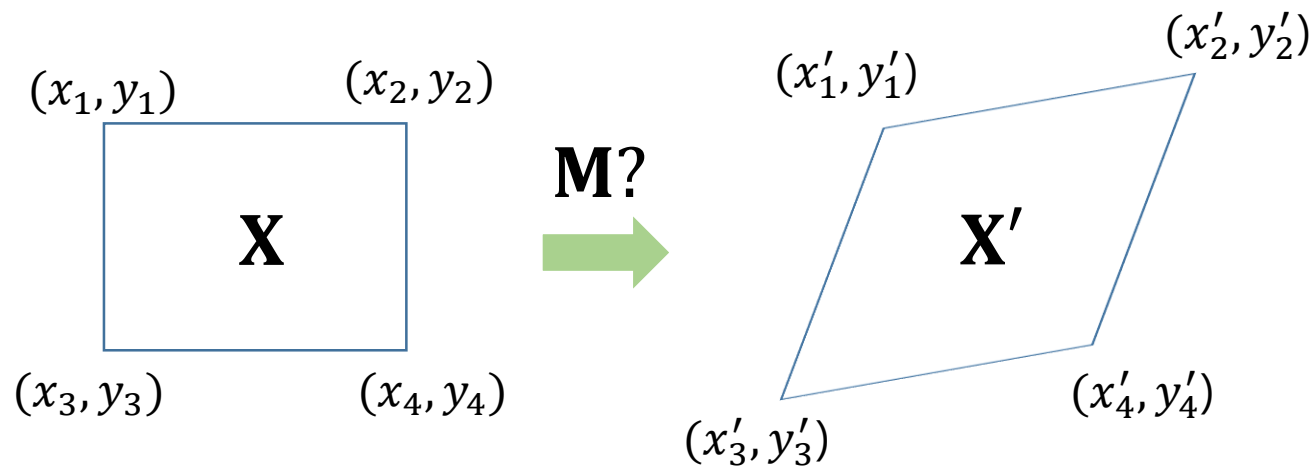
$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} x + t_x \\ y + t_y \\ 1 \end{bmatrix}$$



Parametric Warping

- Affine transform
 - Linear transform + translation

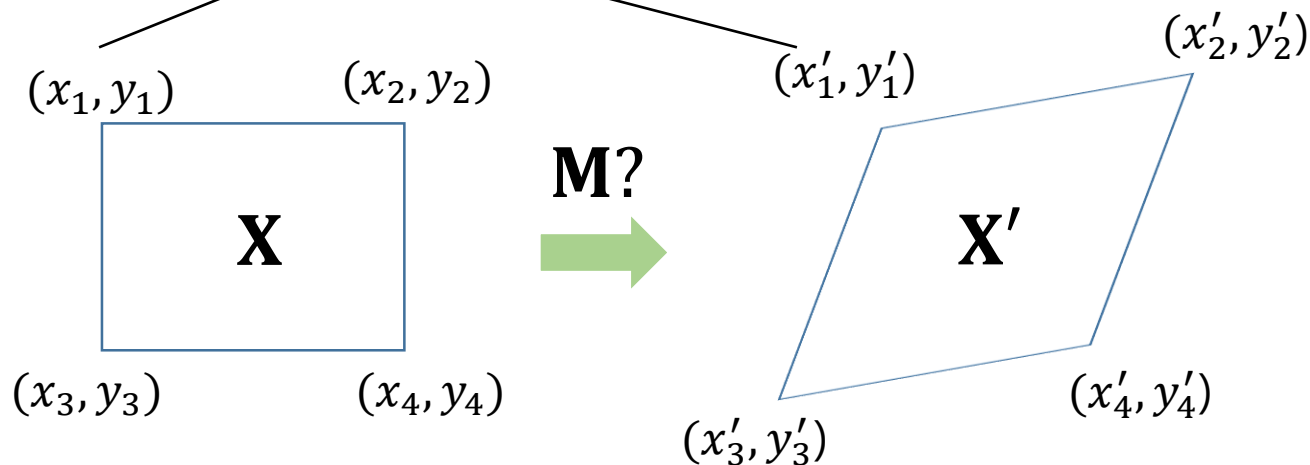
$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} a & b & c \\ d & e & f \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} ax + by + c \\ dx + ey + f \\ 1 \end{bmatrix}$$



Parametric Warping

- Affine transform
 - Linear transform + translation

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



Parametric Warping

- Affine transform

$$\begin{array}{ll} x'_1 = ax_1 + by_1 + c & x'_2 = ax_2 + by_2 + c \\ y'_1 = dx_1 + ey_1 + f & y'_2 = dx_2 + ey_2 + f \end{array}$$

$$\begin{array}{ll} x'_3 = ax_3 + by_3 + c & x'_4 = ax_4 + by_4 + c \\ y'_3 = dx_3 + ey_3 + f & y'_4 = dx_4 + ey_4 + f \end{array}$$

Parametric Warping

- Affine transform

$$x'_1 = ax_1 + by_1 + c$$

$$y'_1 = dx_1 + ey_1 + f$$

$$x'_2 = ax_2 + by_2 + c$$

$$y'_2 = dx_2 + ey_2 + f$$

$$x'_3 = ax_3 + by_3 + c$$

$$y'_3 = dx_3 + ey_3 + f$$

$$x'_4 = ax_4 + by_4 + c$$

$$y'_4 = dx_4 + ey_4 + f$$

$$\begin{bmatrix} x_1 & y_1 & c & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \\ e \\ f \end{bmatrix} = \begin{bmatrix} x'_1 \end{bmatrix}$$

Parametric Warping

- Affine transform

$$x'_1 = ax_1 + by_1 + c$$

$$y'_1 = dx_1 + ey_1 + f$$

$$x'_2 = ax_2 + by_2 + c$$

$$y'_2 = dx_2 + ey_2 + f$$

$$x'_3 = ax_3 + by_3 + c$$

$$y'_3 = dx_3 + ey_3 + f$$

$$x'_4 = ax_4 + by_4 + c$$

$$y'_4 = dx_4 + ey_4 + f$$

$$\begin{bmatrix} x_1 & y_1 & c & 0 & 0 & 0 \\ 0 & 0 & 0 & x_1 & y_1 & f \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \\ e \\ f \end{bmatrix} = \begin{bmatrix} x'_1 \\ y'_1 \end{bmatrix}$$

Parametric Warping

- Affine transform

$$\begin{aligned} x'_1 &= ax_1 + by_1 + c & x'_2 &= ax_2 + by_2 + c \\ y'_1 &= dx_1 + ey_1 + f & y'_2 &= dx_2 + ey_2 + f \end{aligned}$$

$$\begin{aligned} x'_3 &= ax_3 + by_3 + c & x'_4 &= ax_4 + by_4 + c \\ y'_3 &= dx_3 + ey_3 + f & y'_4 &= dx_4 + ey_4 + f \end{aligned}$$

$$\begin{array}{c} \updownarrow \\ 8 \end{array} \begin{bmatrix} x_1 & y_1 & c & 0 & 0 & 0 \\ 0 & 0 & 0 & x_1 & y_1 & f \\ & & & \vdots & & \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \\ e \\ f \end{bmatrix} = \begin{bmatrix} x'_1 \\ y'_1 \\ \vdots \end{bmatrix} \rightarrow \mathbf{Ax} = \mathbf{b}$$

Parametric Warping

- Affine transform
 - Least-squares (최소자승법)

$$\mathbf{Ax} - \mathbf{b} = \mathbf{0}$$

$$\mathbf{x}^* = \operatorname{argmin}_{\mathbf{x}} \|\mathbf{Ax} - \mathbf{b}\|^2 = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b}$$

Parametric Warping

- Affine transform

$$\begin{aligned} E(x) &= (Ax - b)^T (Ax - b) = (x^T A^T - b^T)(Ax - b) \\ &= x^T A^T A x - x^T A^T b - b^T A x + b^T b \quad (x^T A^T b = b^T A x) \\ &= x^T A^T A x - 2b^T A x + b^T b \end{aligned}$$

$$\nabla E(x) = \frac{dE}{dx} = x^T (A^T A + A^T A) - 2b^T A = 2A^T A x - 2A^T b$$

$$A^T A x = A^T b \quad (\text{normal equations})$$

$$\begin{aligned} \Rightarrow x^+ &= (A^T A)^+ A^T b \\ &= (A^T A)^{-1} A^T b \quad (\text{if the inverse } (A^T A)^{-1} \text{ exists}) \end{aligned}$$

$$(Ax)^T = x^T A^T$$

$$\frac{d}{dx} (x^T A x) = x^T (A + A^T)$$

Parametric Warping

- Affine transform

