

Lecture 3 - Image Fundamentals

This lecture will cover:

- **Image operation (图像运算操作)**
- Image interpolation (图像插值)
- Image registration (图像配准)
- Image reconstruction (图像重建)

Image Operations

- Array and Matrix Operation
- Vector and Matrix Operation
- Linear and Nonlinear Operation
- Set and Logical Operation
- Arithmetic Operation
- Spatial Operation
- Image Transformation
- Probabilistic Methods

Array and Matrix Operation

Consider two 2 x 2 image

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \text{ and } \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

➤ Array product

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} = \begin{bmatrix} a_{11}b_{11} & a_{12}b_{11} \\ a_{21}b_{21} & a_{22}b_{22} \end{bmatrix}$$

➤ Matrix product

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} = \begin{bmatrix} a_{11}b_{11} + a_{12}b_{21} & a_{11}b_{12} + a_{12}b_{22} \\ a_{21}b_{11} + a_{22}b_{21} & a_{21}b_{12} + a_{22}b_{22} \end{bmatrix}$$

Vector and Matrix Operation

➤ Multispectral image processing

A pixel in a n -dimensional space can be expressed as a column vector

$Z = [z_1, z_2 \dots z_n]^T$, then a vector norm between two pixels Z and A

$$\begin{aligned}\|Z - A\| &= [(Z - A)^T (Z - A)]^{\frac{1}{2}} \\ &= [(z_1 - a_1)^2 + (z_2 - a_2)^2 + \dots + (z_n - a_n)^2]^{\frac{1}{2}}\end{aligned}$$

➤ Linear transformations

$$g = Hf + n$$

Linear and Nonlinear Operation

An operator

$$H[f(x, y)] = g(x, y)$$

is linear if

$$\begin{aligned} H[a_i f_i(x, y) + a_j f_j(x, y)] &= a_i H[f_i(x, y)] + a_j H[f_j(x, y)] \\ &= a_i g_i(x, y) + a_j g_j(x, y) \end{aligned}$$

- Additivity (相加性)
- Homogeneity (同质性)

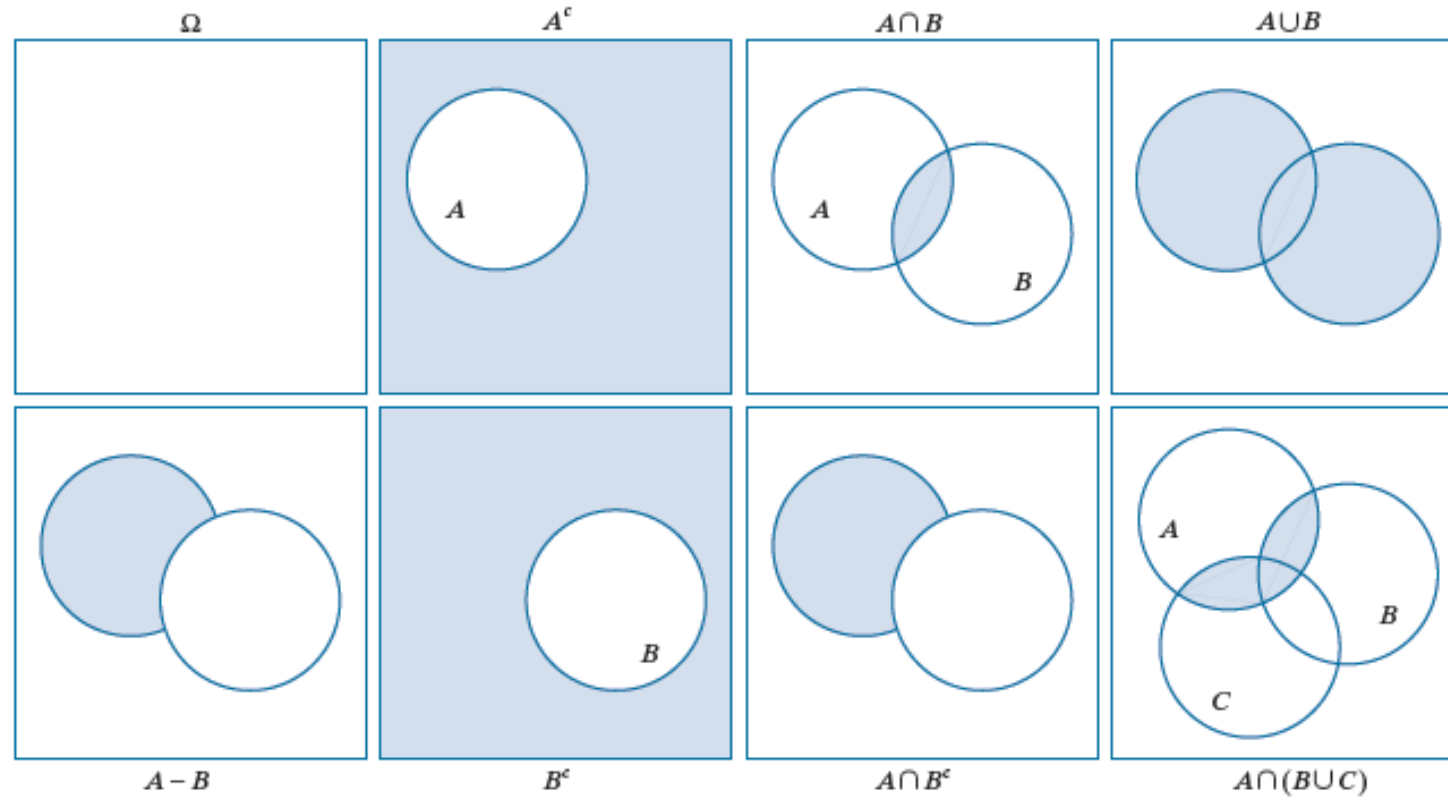
Set Operation

TABLE 2.1

Some important set operations and relationships.

Description	Expressions
Operations between the sample space and null sets	$\Omega^c = \emptyset; \emptyset^c = \Omega; \Omega \cup \emptyset = \Omega; \Omega \cap \emptyset = \emptyset$
Union and intersection with the null and sample space sets	$A \cup \emptyset = A; A \cap \emptyset = \emptyset; A \cup \Omega = \Omega; A \cap \Omega = A$
Union and intersection of a set with itself	$A \cup A = A; A \cap A = A$
Union and intersection of a set with its complement	$A \cup A^c = \Omega; A \cap A^c = \emptyset$
Commutative laws	$A \cup B = B \cup A$ $A \cap B = B \cap A$
Associative laws	$(A \cup B) \cup C = A \cup (B \cup C)$ $(A \cap B) \cap C = A \cap (B \cap C)$
Distributive laws	$(A \cup B) \cap C = (A \cap C) \cup (B \cap C)$ $(A \cap B) \cup C = (A \cup C) \cap (B \cup C)$
DeMorgan's laws	$(A \cup B)^c = A^c \cap B^c$ $(A \cap B)^c = A^c \cup B^c$

Set Operation (Coordinates)



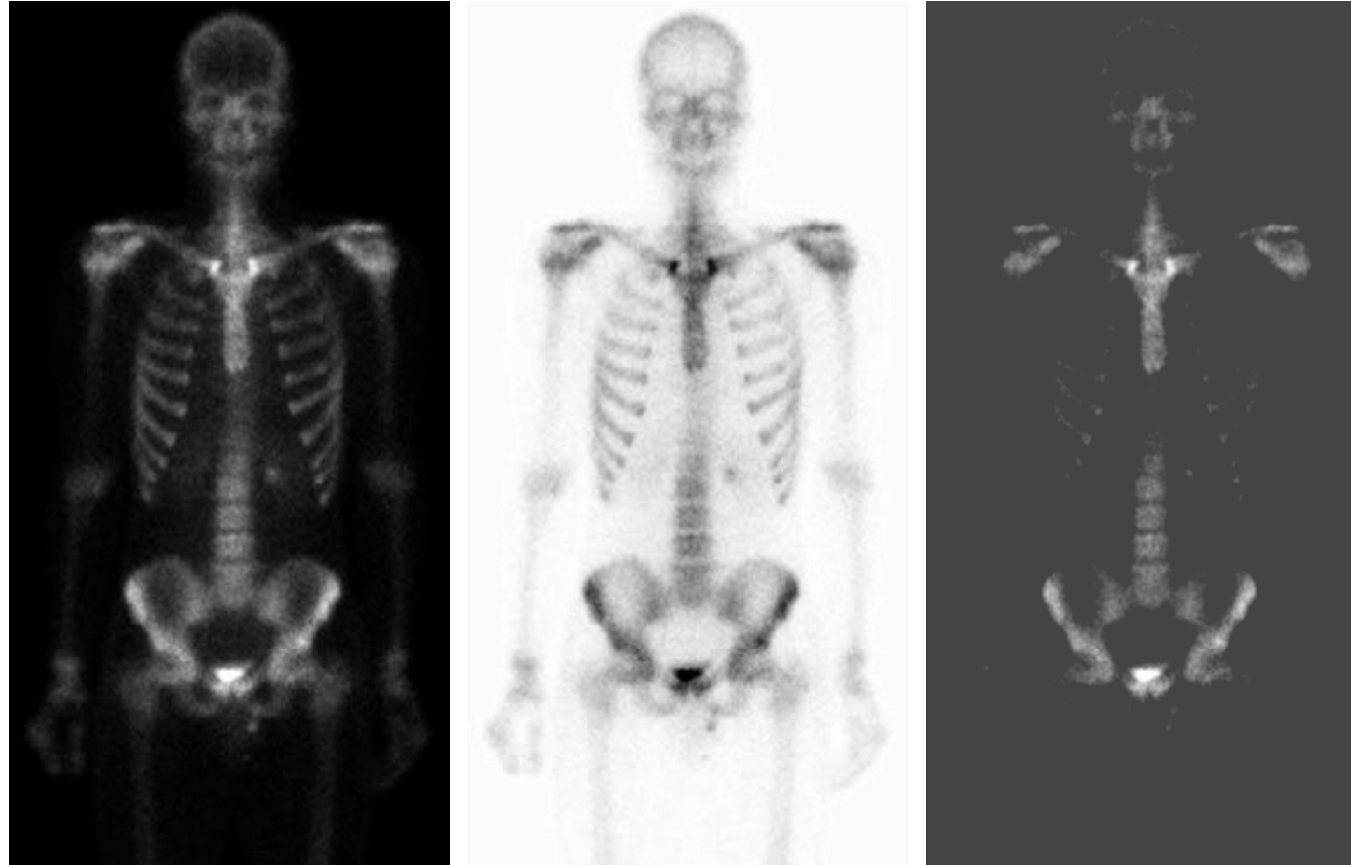
a b c d
e f g h

FIGURE 2.35 Venn diagrams corresponding to some of the set operations in Table 2.1. The results of the operations, such as A^c , are shown shaded. Figures (e) and (g) are the same, proving via Venn diagrams that $A - B = A \cap B^c$ [see Eq. (2-40)].

Set Operation (Intensity)

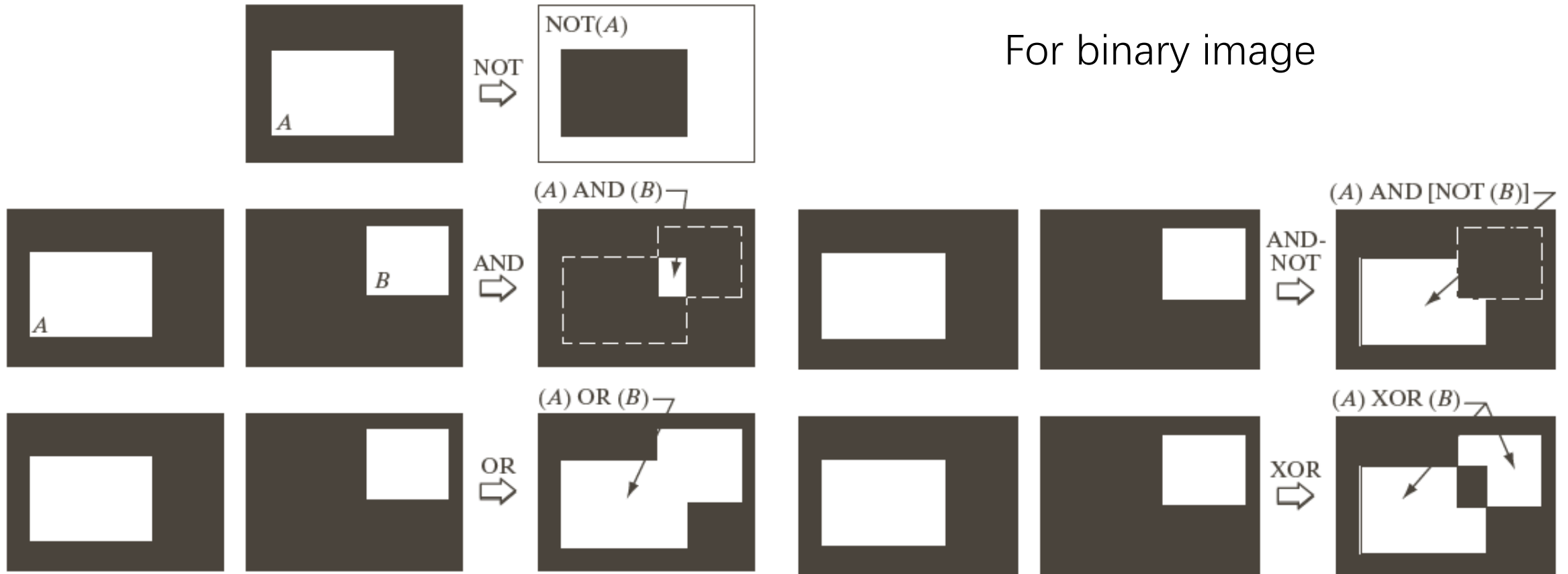
a b c

FIGURE 2.36
Set operations involving grayscale images. (a) Original image. (b) Image negative obtained using grayscale set complementation. (c) The union of image (a) and a constant image. (Original image courtesy of G.E. Medical Systems.)



Logical Operation

For binary image



Arithmetic Operation

➤ **Addition**

$$s(x, y) = f(x, y) + g(x, y)$$

➤ **Subtraction**

$$d(x, y) = f(x, y) - g(x, y)$$

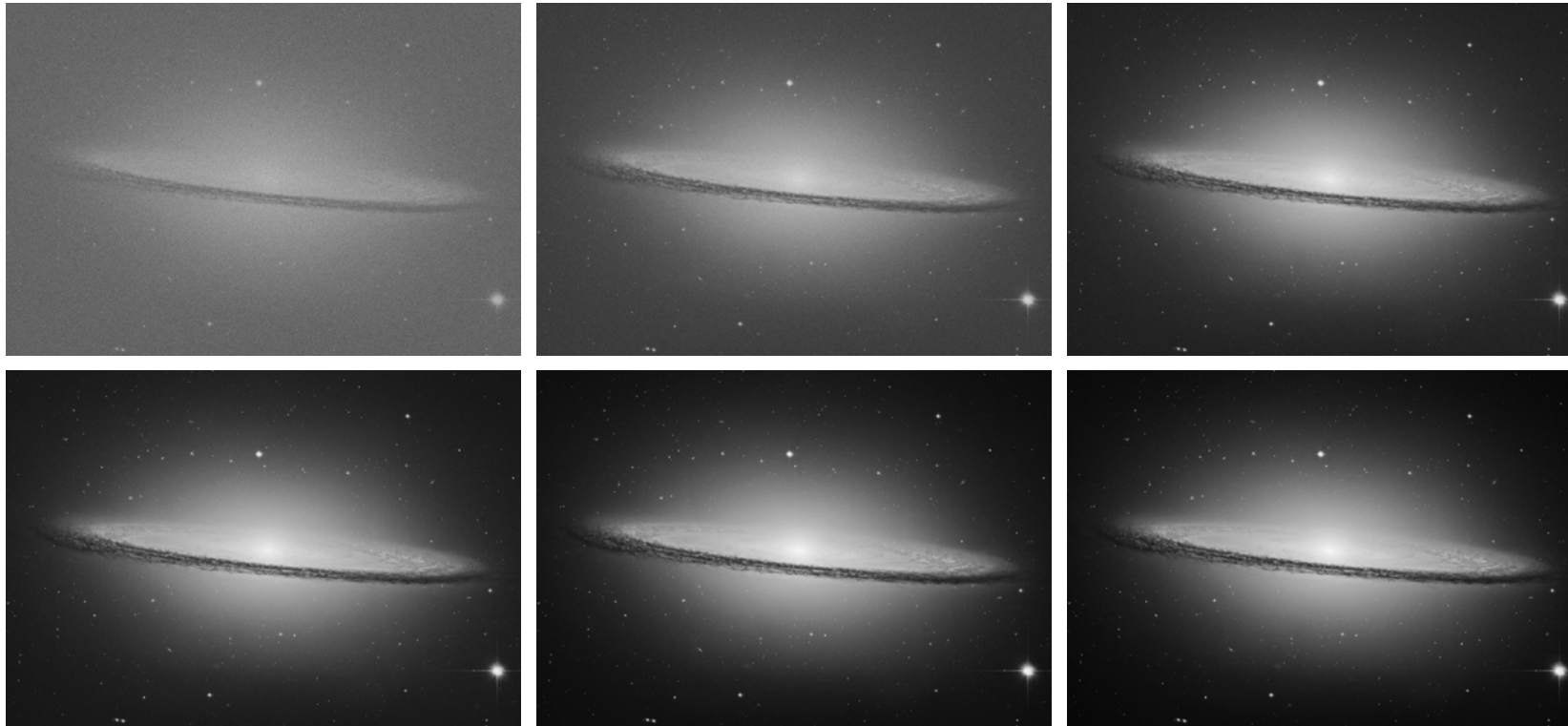
➤ **Multiplication**

$$p(x, y) = f(x, y) \times g(x, y)$$

➤ **Division**

$$v(x, y) = f(x, y) \div g(x, y)$$

Image Addition



a	b	c
d	e	f

FIGURE 2.29 (a) Sample noisy image of the Sombrero Galaxy. (b)-(f) Result of averaging 10, 50, 100, 500, and 1,000 noisy images, respectively. All images are of size 1548×2238 pixels, and all were scaled so that their intensities would span the full $[0, 255]$ intensity scale. (Discovered in 1767, the Sombrero Galaxy is 28 light years from Earth. Original image courtesy of NASA.)

Image Addition

If $f(x, y) + g(x, y) > L_{\max}$, $s(x, y)$ can be calculated as

➤ **Average**

$$s(x, y) = \frac{f(x, y) + g(x, y)}{2}$$

➤ **Scale**

$$\{\min[s(x, y)], \max[s(x, y)]\} = \{0, L_{\max}\}$$

➤ **Max intensity value**

$$\text{If } s(x, y) > L_{\max}, \quad s(x, y) = L_{\max}$$

Image Subtraction

a	b
c	d

FIGURE 2.32
Digital subtraction angiography.
(a) Mask image.
(b) A live image.
(c) Difference between (a) and (b). (d) Enhanced difference image.
(Figures (a) and (b) courtesy of the Image Sciences Institute, University Medical Center, Utrecht, The Netherlands.)

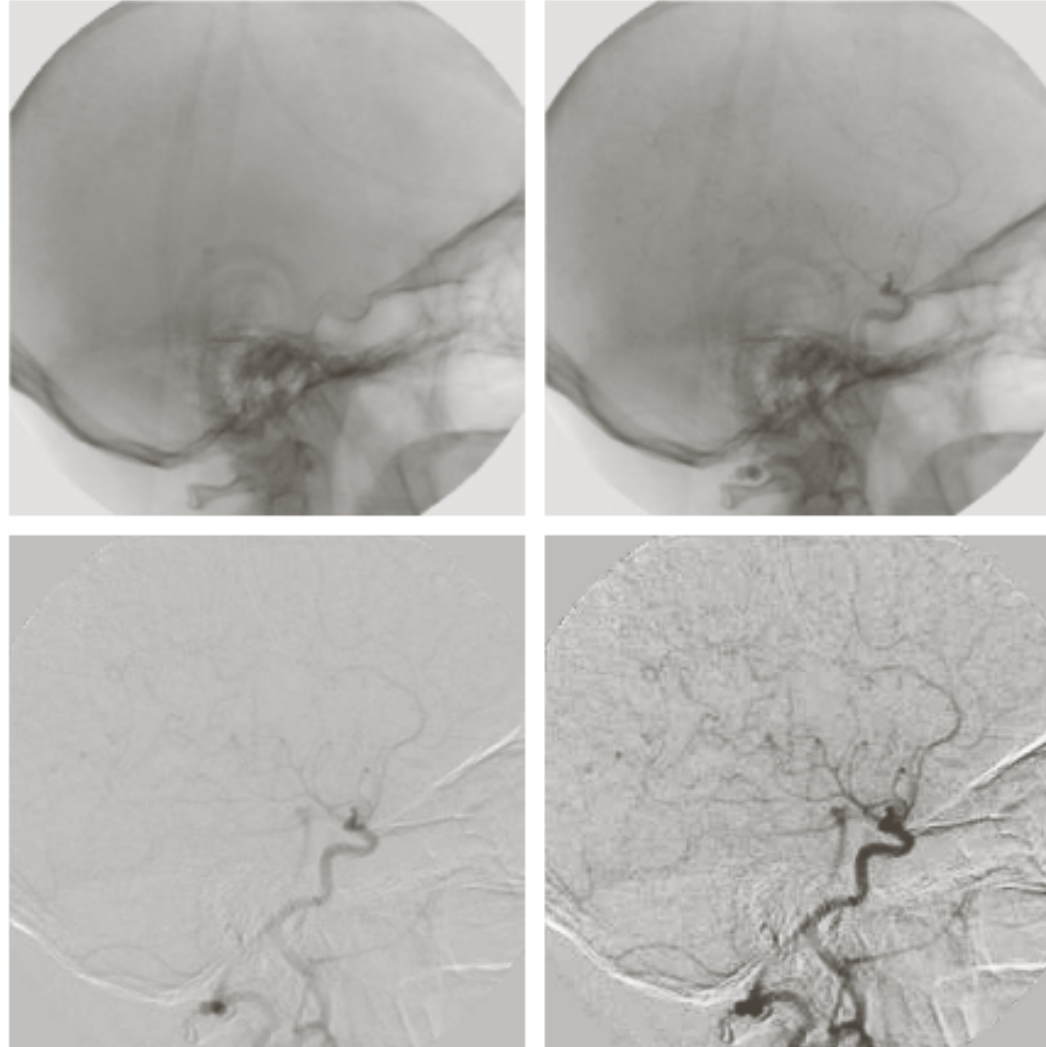
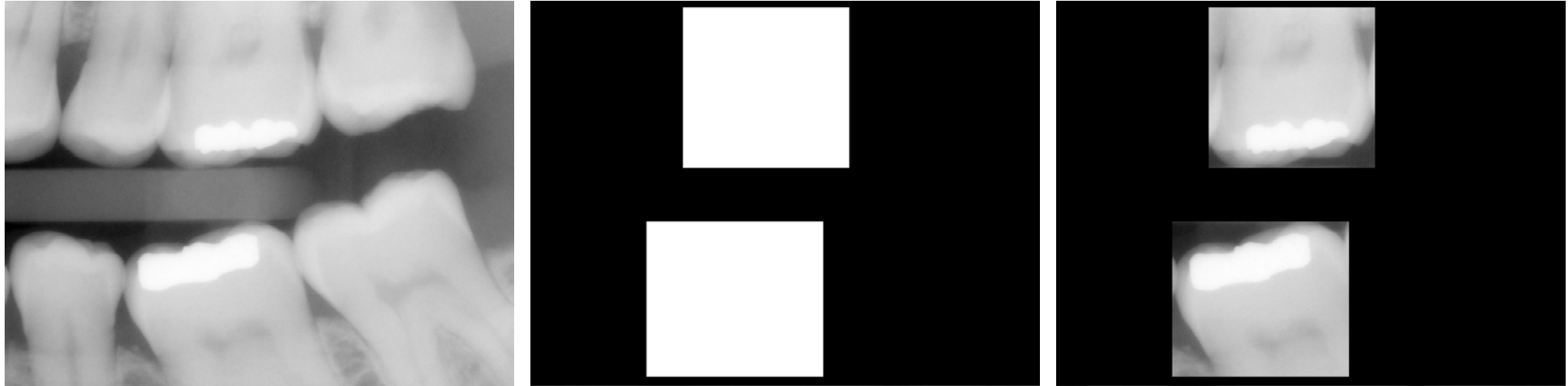


Image Multiplication



a b c

FIGURE 2.34 (a) Digital dental X-ray image. (b) ROI mask for isolating teeth with fillings (white corresponds to 1 and black corresponds to 0). (c) Product of (a) and (b).

Image Division



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

$$h(x, y)$$

$$f(x, y)$$

$$f(x, y) = g(x, y) / h(x, y)$$

Spatial Operation

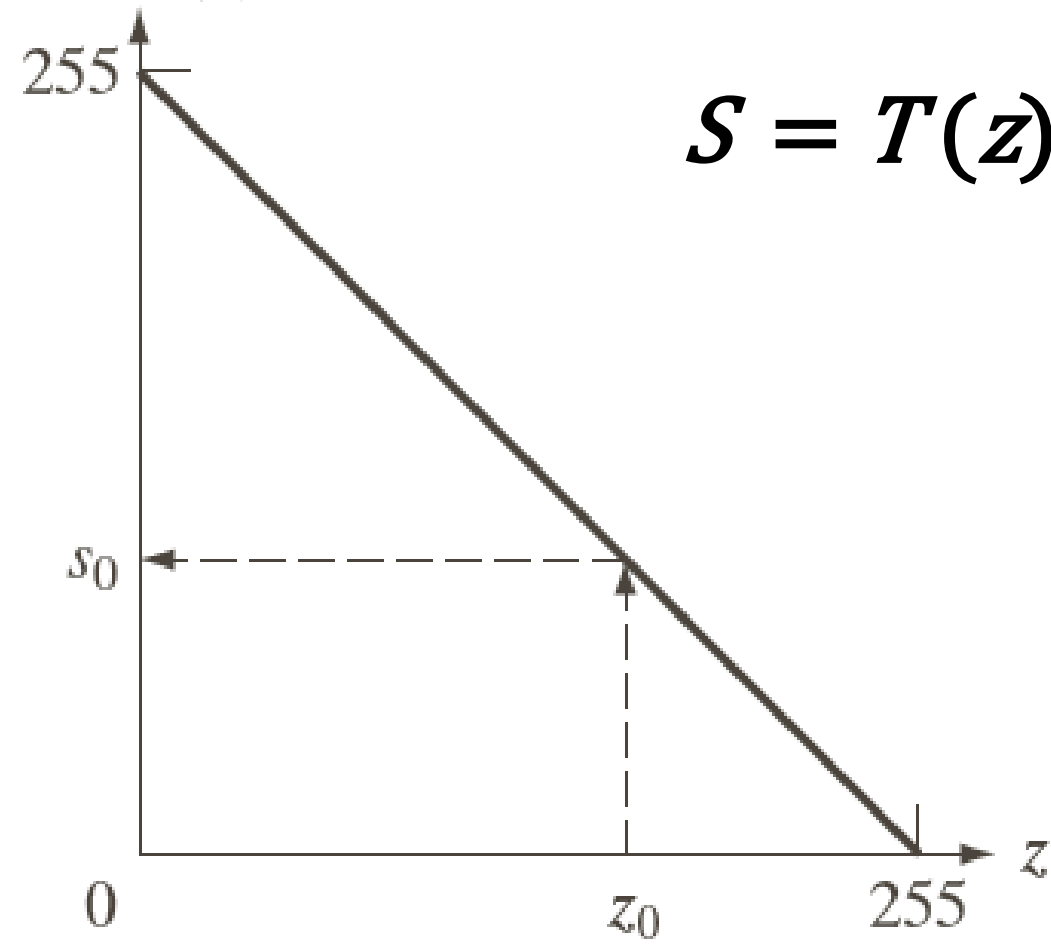
Performed directly on the pixels of the image

- Single-pixel operations
- Neighborhood operations
- Image geometry

Scale, Rotate, Translate, Mirror, Transpose, Shear, etc.

- Interpolation

Single-pixel Operation



Region operation

S_{xy} is a region with center (x, y) , $g(x, y) = \frac{1}{mn} \sum_{(r,c) \in S_{xy}} f(r, c)$

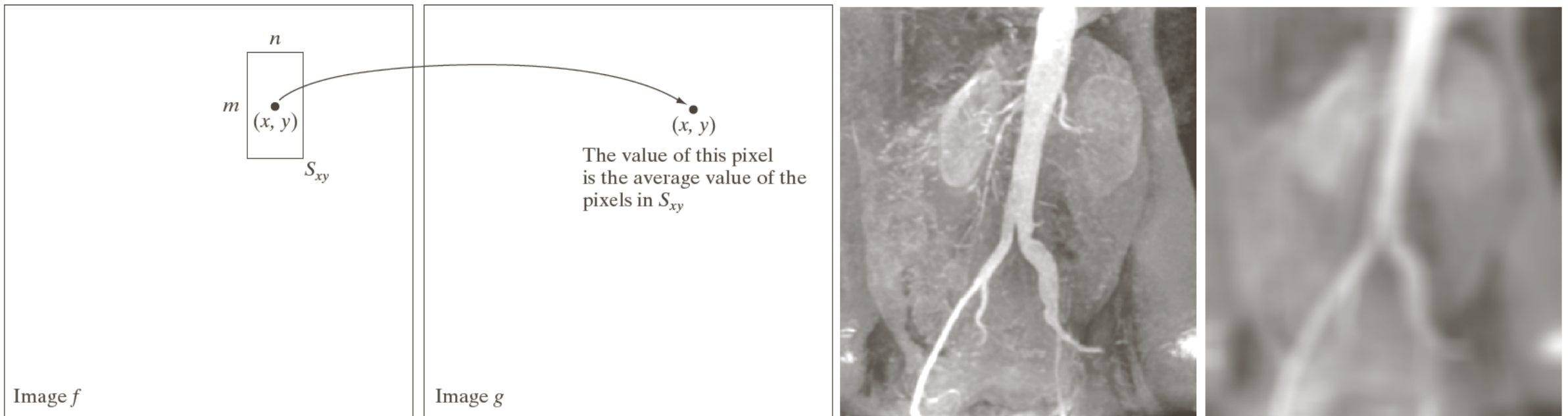


Image geometry

➤ Modify spatial relationship between pixels – *rubber-sheet*

- Forward mapping (前向映射): $(x \ y) = T(v \ w)$
- Inverse mapping (反向映射): $(v \ w) = T^{-1}(x \ y)$

➤ Affine transform (仿射变换)

$$[x \ y \ 1] = [v \ w \ 1]T = [v \ w \ 1] \begin{bmatrix} t_1 & t_4 & 0 \\ t_2 & t_5 & 0 \\ t_3 & t_6 & 1 \end{bmatrix}$$

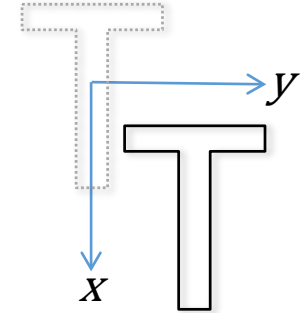
or

$$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = T \begin{bmatrix} v \\ w \\ 1 \end{bmatrix} = \begin{bmatrix} t_1 & t_2 & t_3 \\ t_4 & t_5 & t_6 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ w \\ 1 \end{bmatrix}$$

Affine Transform

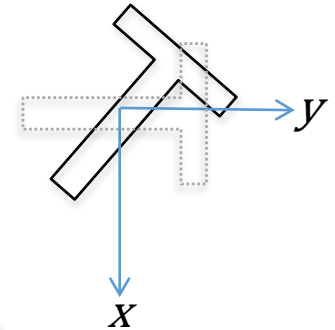
➤ Translation

$$\begin{cases} x = v + \Delta v \\ y = w + \Delta w \end{cases} \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & \Delta v \\ 0 & 1 & \Delta w \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ w \\ 1 \end{bmatrix}$$



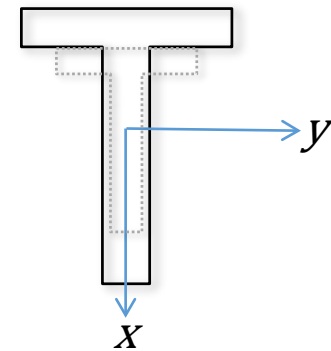
➤ Rotation

$$\begin{cases} x = v \cos \beta - w \sin \beta \\ y = v \sin \beta + w \cos \beta \end{cases} \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} \cos \beta & -\sin \beta & 0 \\ \sin \beta & \cos \beta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ w \\ 1 \end{bmatrix}$$



➤ Scaling

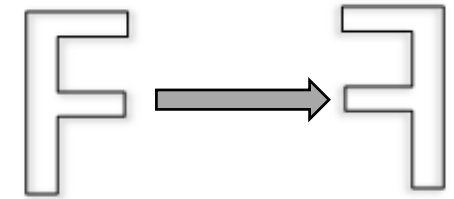
$$\begin{cases} x = c_x v \\ y = c_y w \end{cases} \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} c_x & 0 & 0 \\ 0 & c_y & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ w \\ 1 \end{bmatrix}$$



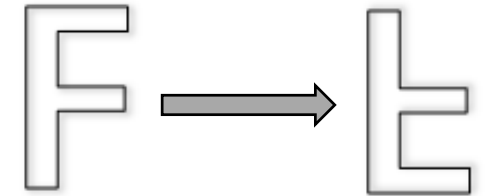
Affine Transform

➤ Mirror

$$\text{Horizontal: } \begin{cases} x = W - v \\ y = w \end{cases} \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} -1 & 0 & W \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ w \\ 1 \end{bmatrix}$$

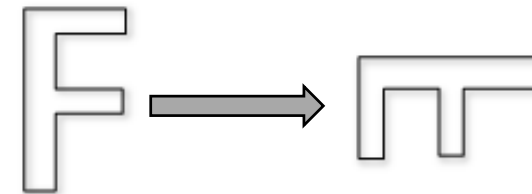


$$\text{Vertical: } \begin{cases} x = v \\ y = H - w \end{cases} \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & -1 & H \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ w \\ 1 \end{bmatrix}$$



➤ Transpose

$$\begin{cases} x = w \\ y = v \end{cases} \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ w \\ 1 \end{bmatrix}$$



Affine Transform

➤ Shear

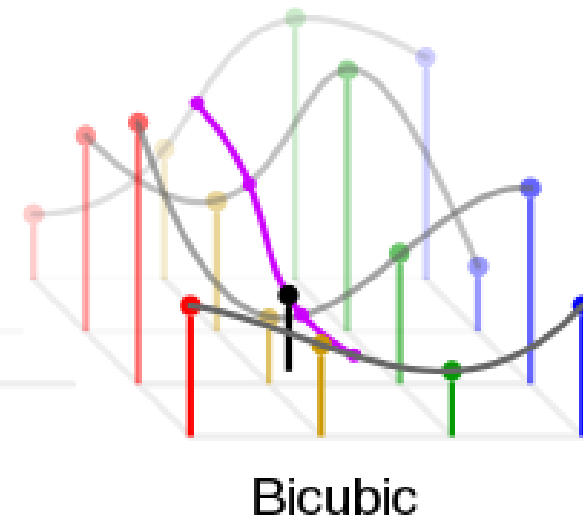
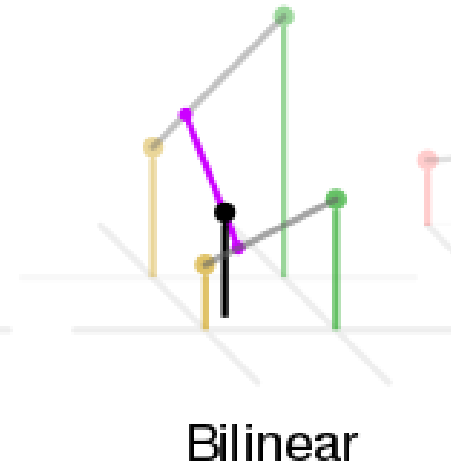
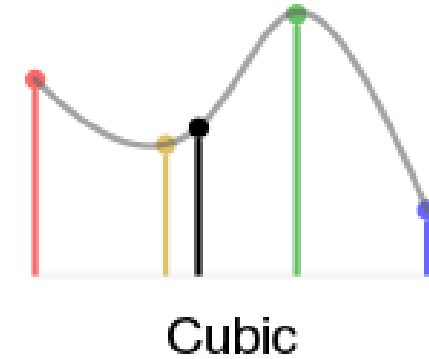
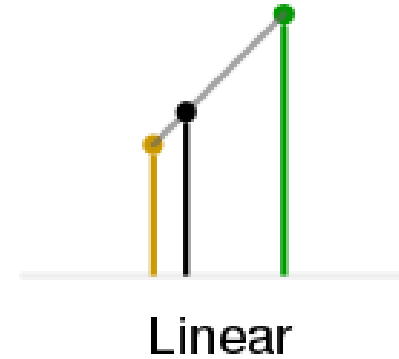
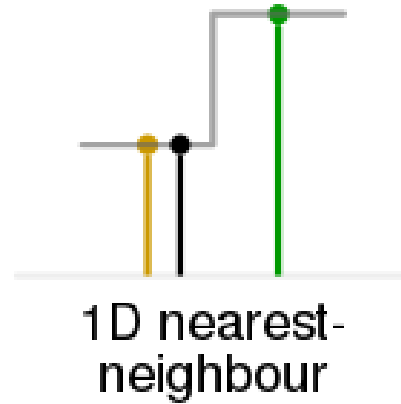
$$\text{Horizontal: } \begin{cases} x = v + c_y w \\ y = w \end{cases} \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & c_y & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ w \\ 1 \end{bmatrix}$$

$$\text{Vertical: } \begin{cases} x = v \\ y = c_x v + w \end{cases} \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ c_x & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ w \\ 1 \end{bmatrix}$$

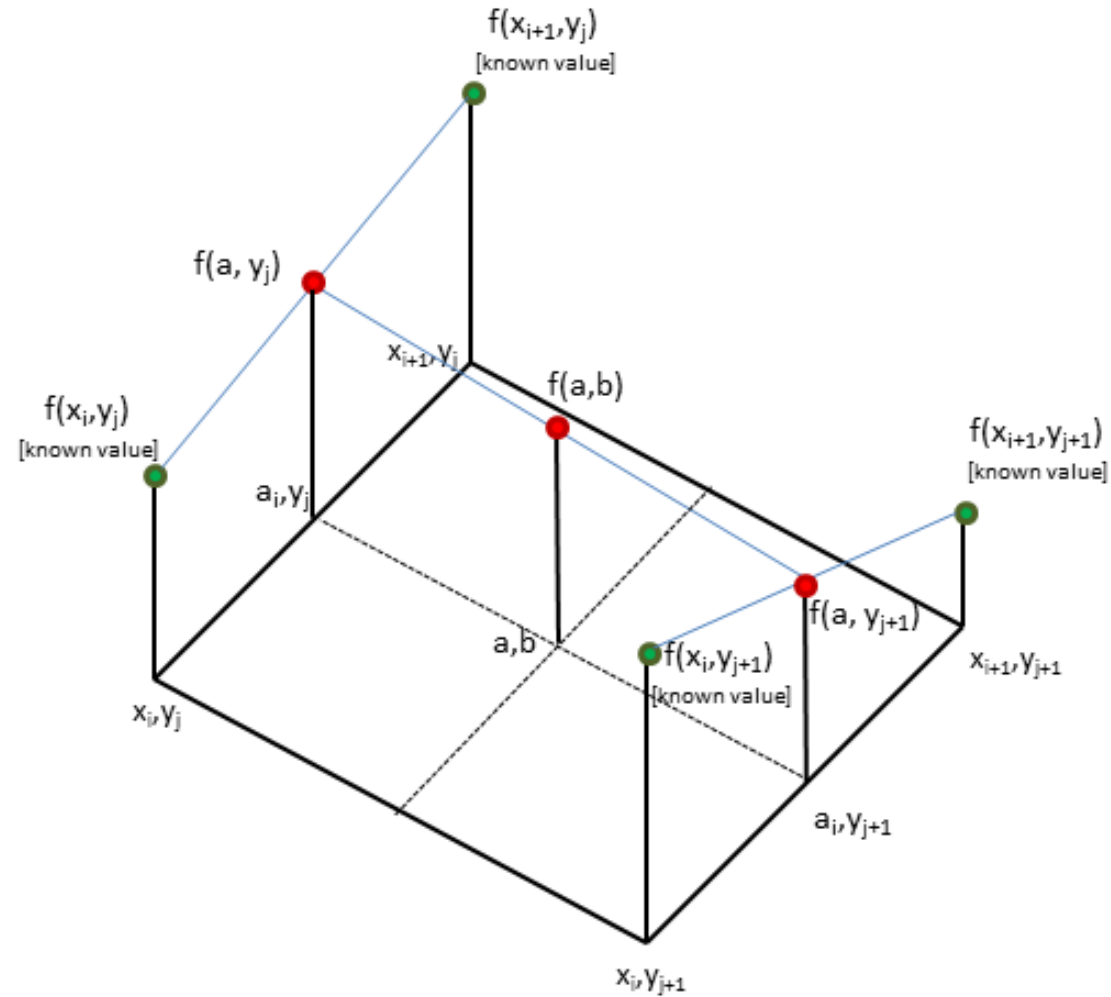


Image Interpolation (插值)

- Use known data to estimate values at unknown locations
- A resampling method
- Intensity interpolation



Bilinear interpolation



Interpolation

a b c
d e f

Image interpolation:

interpolate the image from 72dpi to 1250 dpi using (a)nearest neighbor method, (b) Bilinear method, (c) Bicubic method;

interpolate the image from 150dpi to 1250 dpi using (a)nearest neighbor method, (b) Bilinear method, (c) Bicubic method;



Image registration (图像配准)

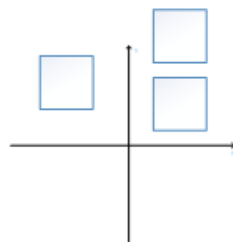
- To align two or more images of the same scene
- Given input and output images, to estimate the transformation functions and then use it to register the two images
- Tie point (约束点): $x = c_1v + c_2w + c_3vw + c_4$; $y = c_5v + c_6w + c_7vw + c_8$
- For large number of tie points
 - quadrilateral subimage formed by a group of 4 tie points
 - More complex model: polynomials fitted by least squares algorithms



Image reconstruction (图像重建)

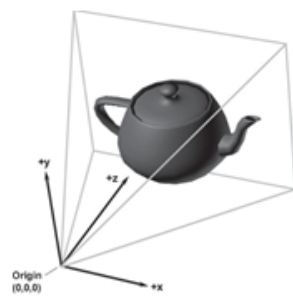
Some useful coordinate spaces

1. World Space (世界空间)



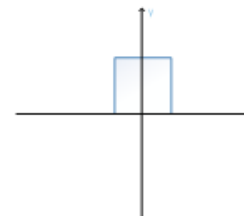
The *world coordinate system* is a special coordinate system that establishes the “global” reference frame for all other coordinate systems to be specified.

2. Camera/View Space (摄像机/观察空间)



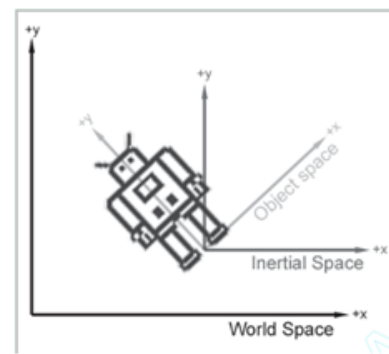
Camera space is the coordinate space associated with an observer.

3. Object Space (物体空间)



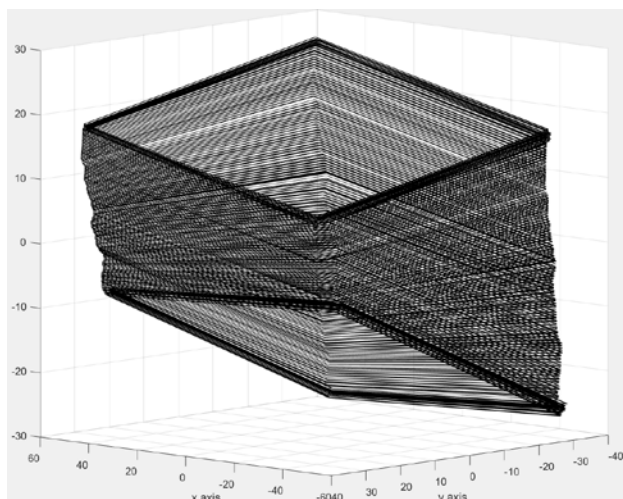
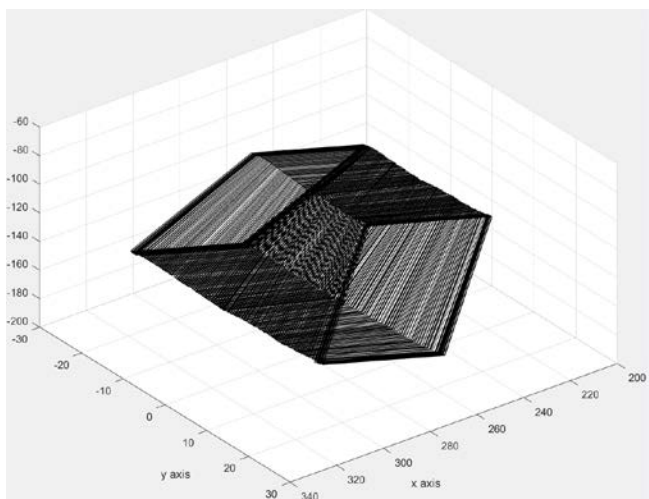
Object space is the coordinate space associated with a particular object. Every object has its own independent object space. For example, Fig.3 shows the image self-coordinate space

4. Inertial Space (惯性空间)



The origin of *inertial space* is the same as the origin of the *object space*, and the axes of *inertial space* are parallel with the axes of *world space*. Fig. 4 illustrates this principle in 2D.

Image reconstruction (图像重建)



$$\begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} t_1 & t_2 & t_3 & t_4 \\ t_5 & t_6 & t_7 & t_8 \\ t_9 & t_{10} & t_{11} & t_{12} \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \\ 1 \end{bmatrix}$$

