



上海交通大学  
SHANGHAI JIAO TONG UNIVERSITY



**IVM**

*Image, Video, and Multimedia Communications Laboratory*

# Digital Image Processing

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# This lecture will cover

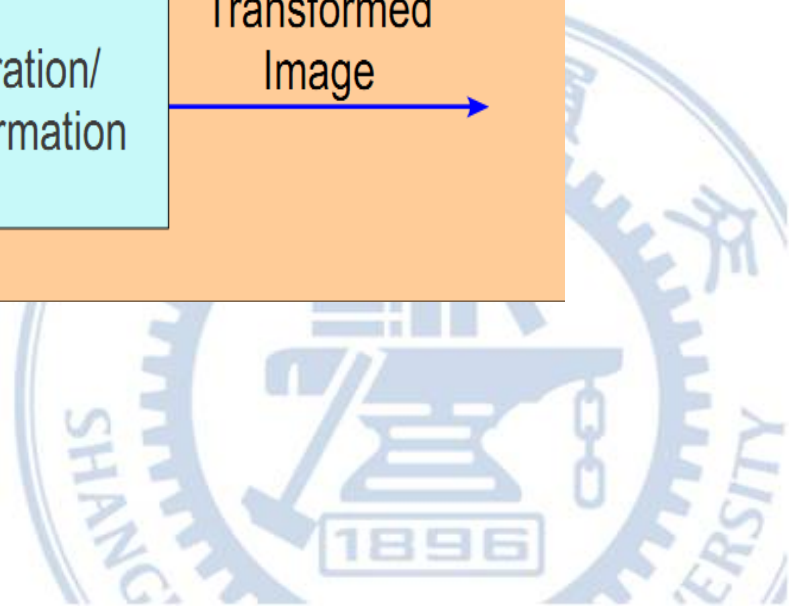
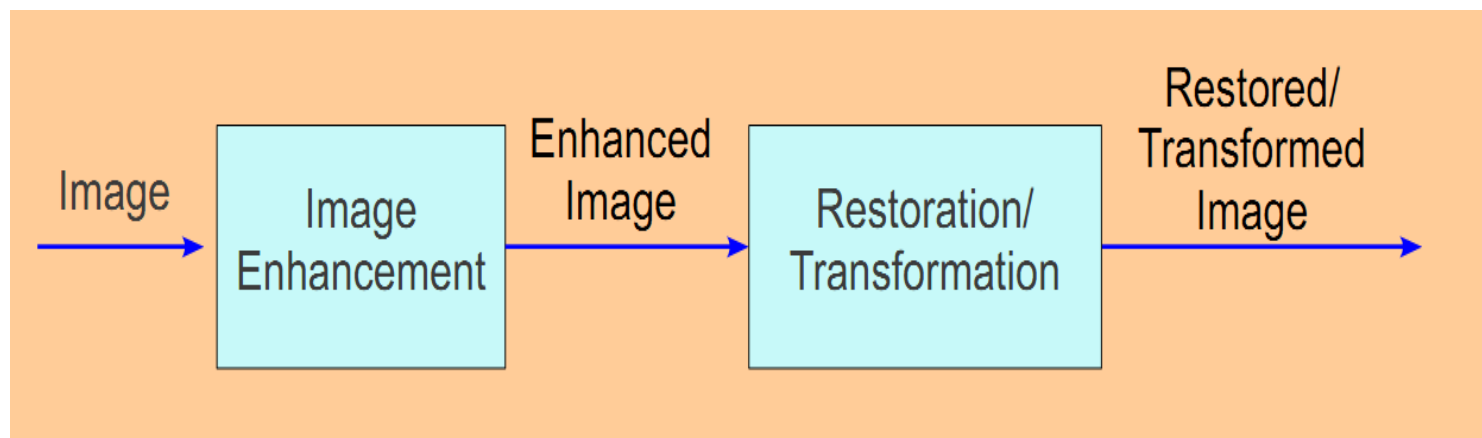
- **Representation and Descriptor**
  - **SIFT descriptor**
  - **Boundary descriptor**
  - **Region descriptor**





# Low-level image processing

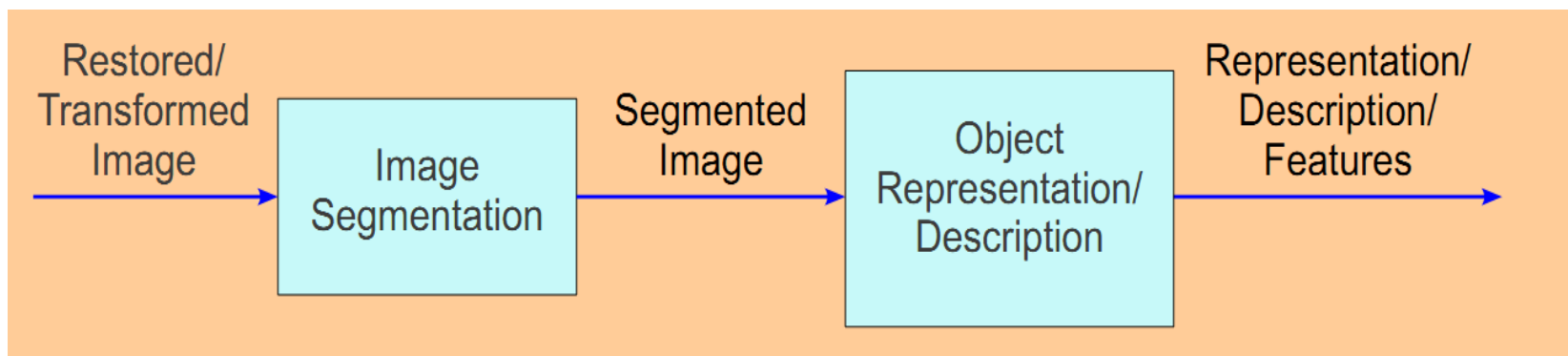
- Image enhancement, restoration, transformation...





# Mid-level image processing (image understanding)

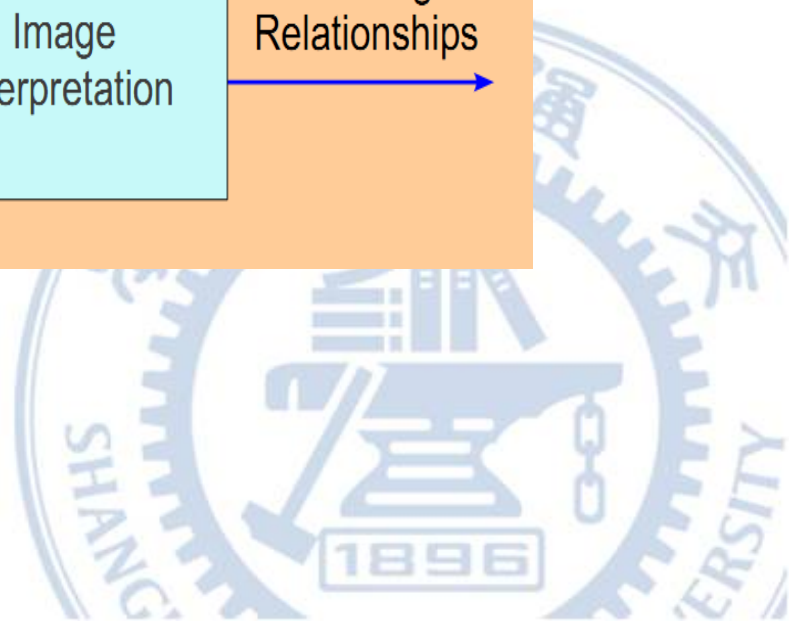
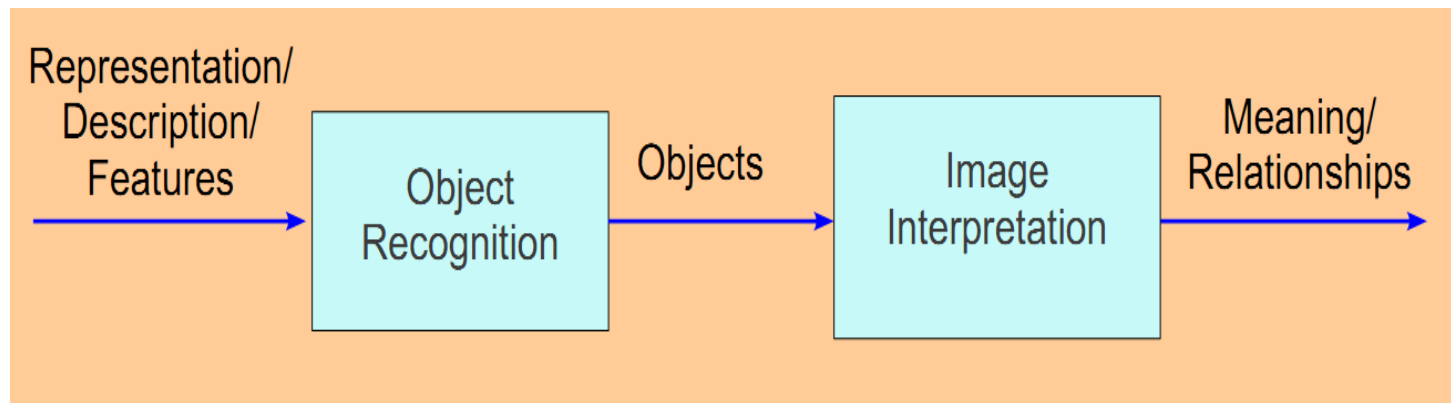
- Object representation, description





# High-level image processing (recognition and interpretation)

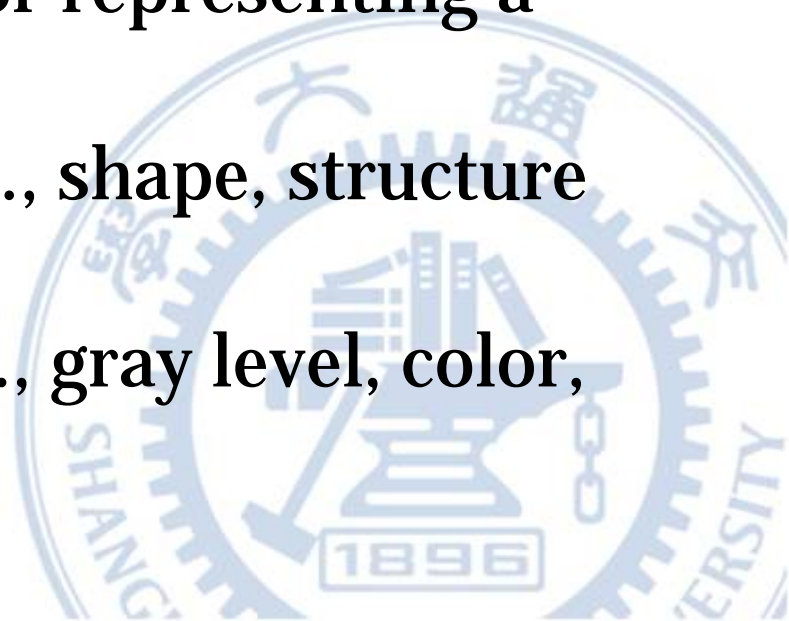
- Object recognition, interpretation of object relationships





# Goal

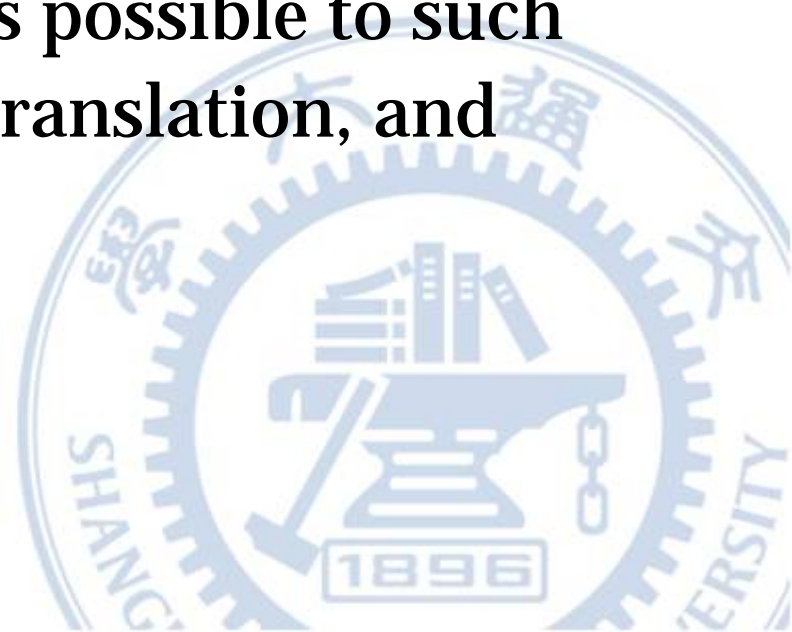
- To represent and describe the resulting aggregate of segmented pixels in a form suitable for further computer processing.
- There are two information for representing a region:
  - I. external characteristic (e.g., shape, structure relation)
  - II. internal characteristic (e.g., gray level, color, texture)





## Continued

- In general, description is usually based on the representation.
- It is important that the features selected as descriptors be as insensitive as possible to such variations as changes in size, translation, and rotation

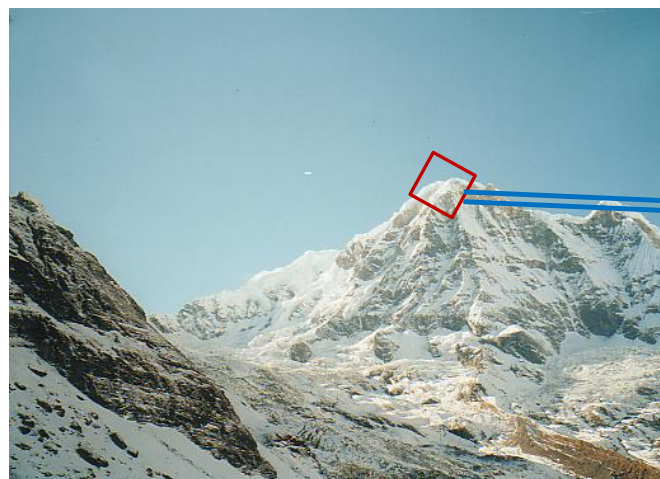




# Descriptors - SIFT (A brief introduction)

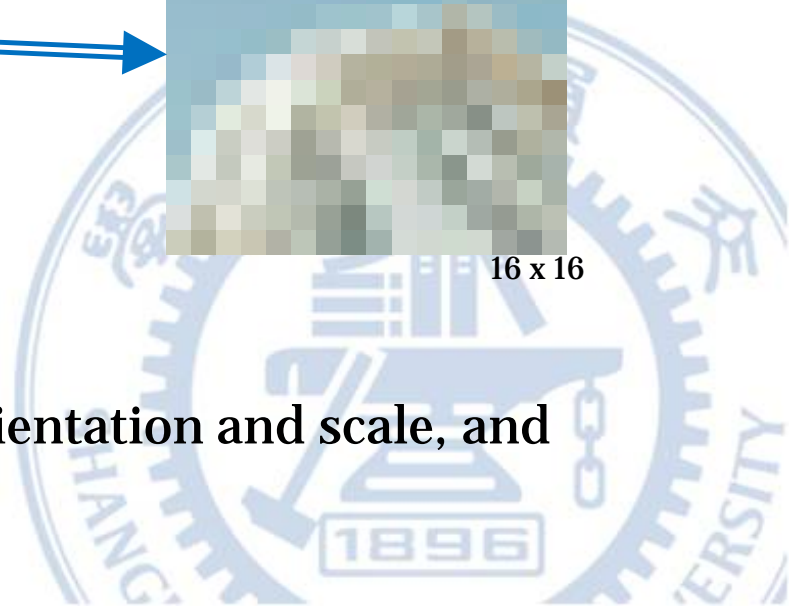
## How to compute SIFT descriptor?

Input: an image and a location to compute the descriptor



16 x 16

Step 1: Warp the image to the correct orientation and scale, and then extract the feature as 16×16 pixels

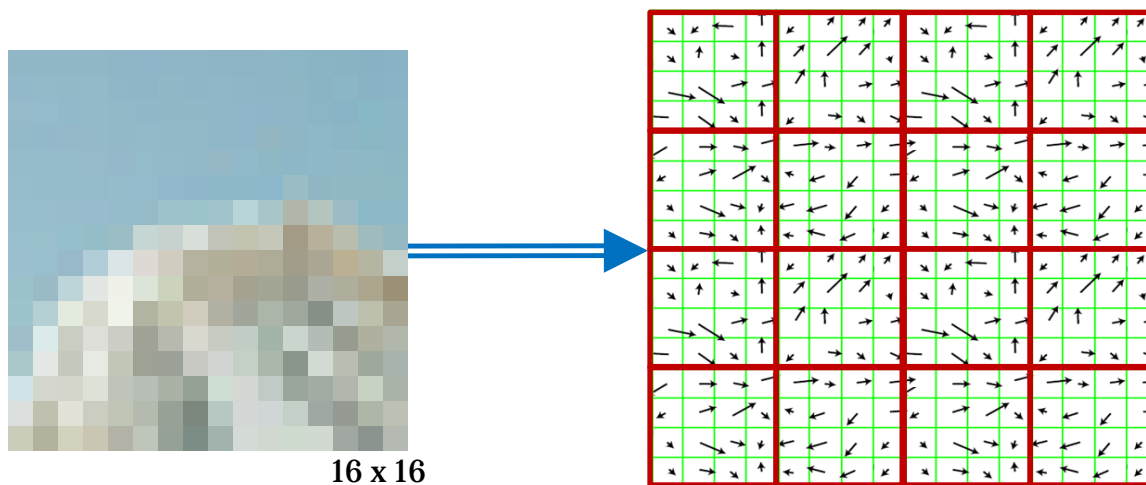




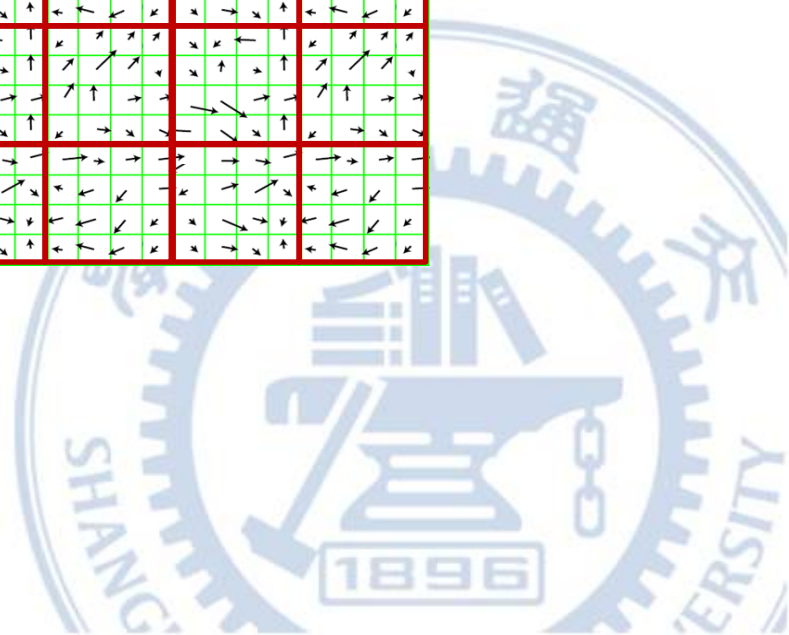


# Descriptors - SIFT (A brief introduction)

**Step 2: Compute the gradient for each pixel (direction and magnitude)**



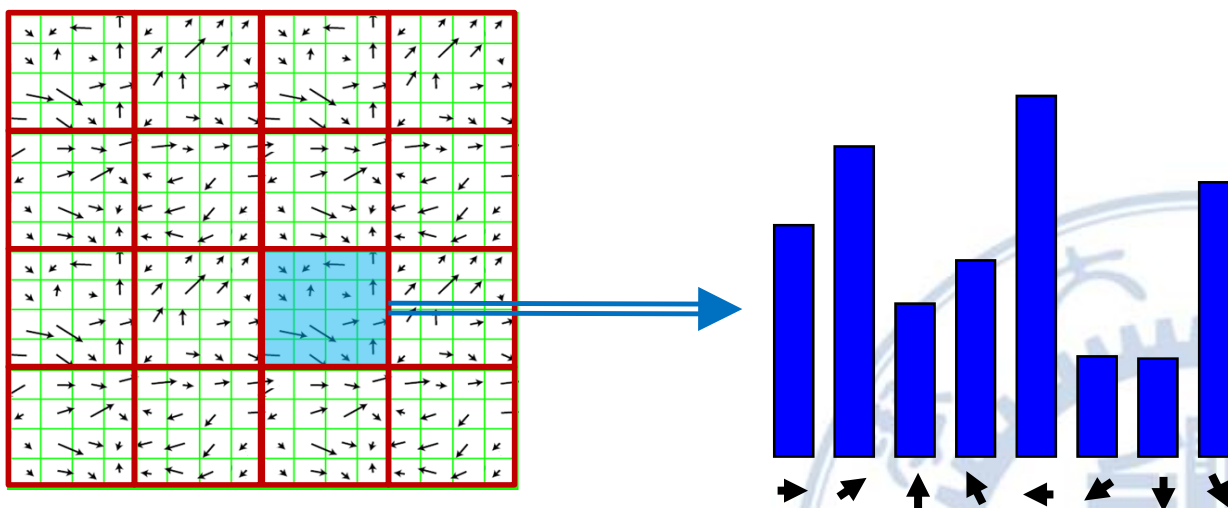
**Step 3: Divide the pixels into 4×4 squares**



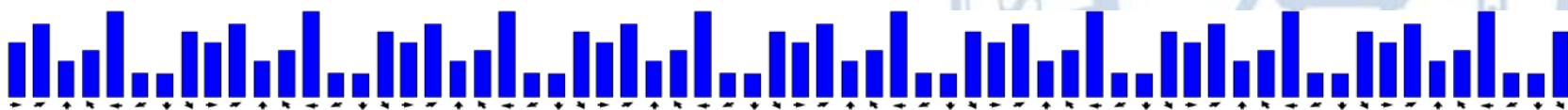


# Descriptors - SIFT (A brief introduction)

**Step 4: For each square, compute gradient direction histogram over 8 directions. Choose the peak of that histogram as Keypoint orientation and rotate all of the directions so that Keypoint orientation=0°.**



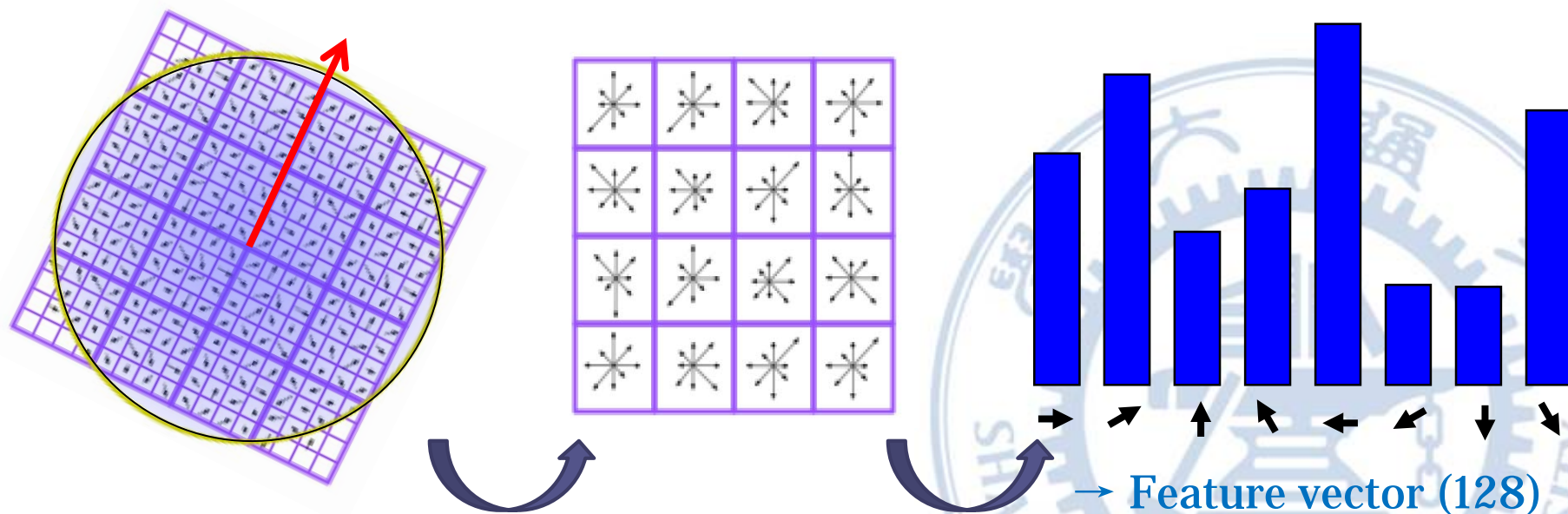
**The result: 128 dimensions feature vector.**





# Conclusion

- Warp the feature into  $16 \times 16$  square.
- Divide into  $4 \times 4$  squares.
- For each square, compute an histogram of the gradient directions.





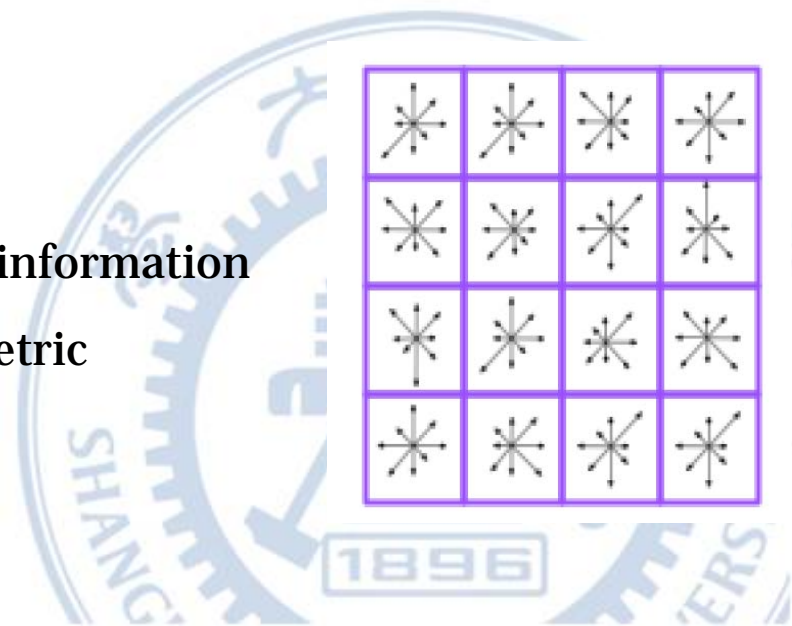
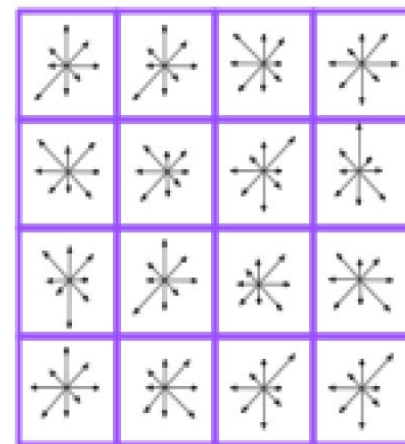
# Conclusion

## Invariance to illumination

- Gradient are invariant to Light intensity shift (i.e. add a scalar to all the pixels)
- Normalization to unit length add invariance to light intensity change (i.e. multiply all the pixels by a scalar)

## Invariance to shift and rotation

- Histograms does not contains any geometric information
- Using 16 histograms allows to preserve geometric information





# Image Descriptors

- Chose a representation scheme
  - Chain codes (4- and 8-directional chain codes)
  - Polygonal approximation
  - Signature (1-D functional representation)
- Describe the region based on the scheme
  - Boundary descriptors
    - Length, diameter, shape numbers, Fourier descriptors
  - Region descriptors
    - Area, compactness, principal axes
    - Moment , Texture
- It is important that the features selected as descriptors be as insensitive as possible to such variations as changes in scale, translation, and rotation



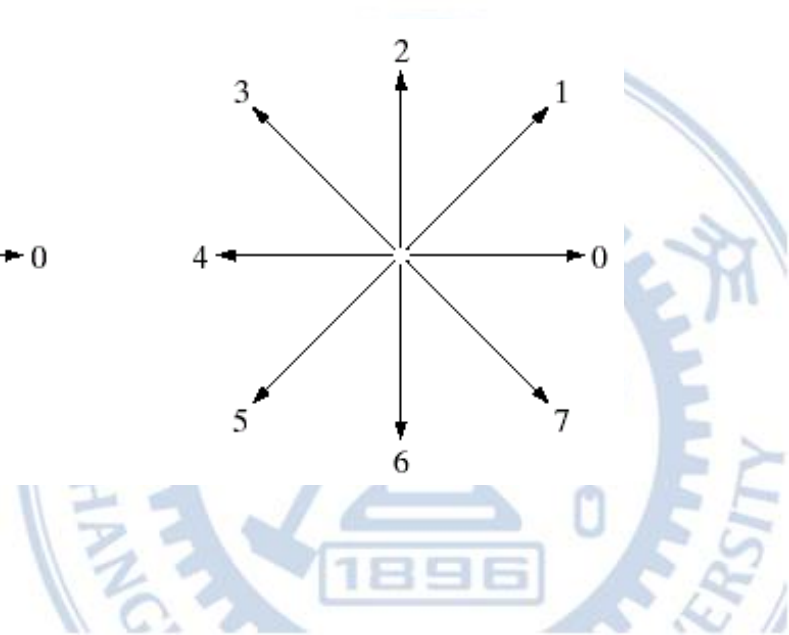
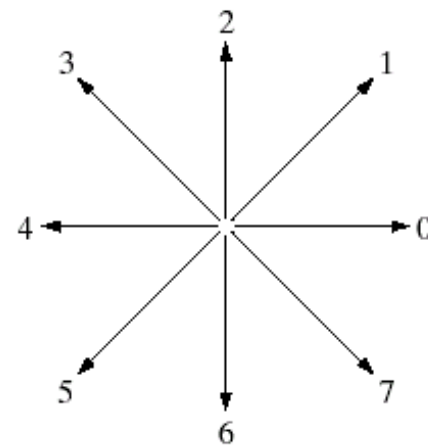
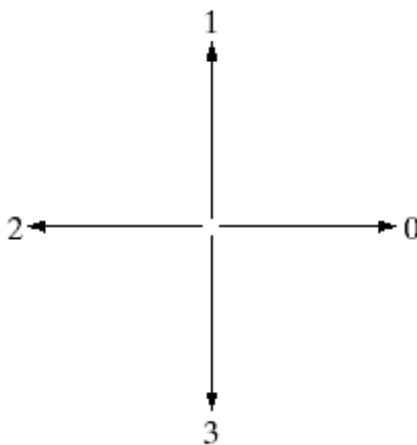


# Chain codes

- Represent a boundary by a connected sequence of straight-line segments of specified length and direction
- 4-directional chain codes
- 8-directional chain codes

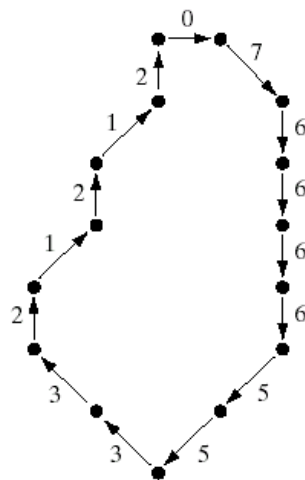
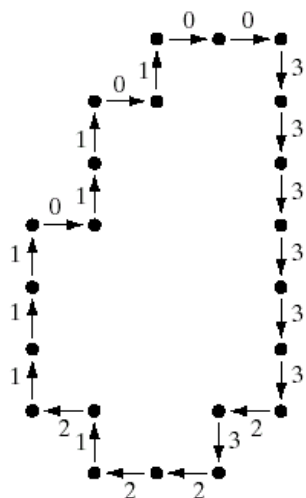
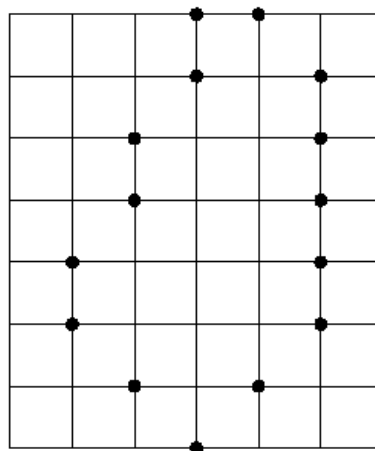
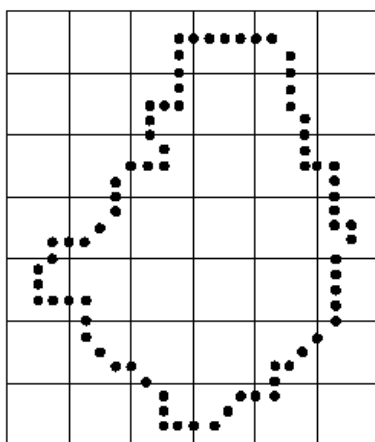
a b

**FIGURE 11.1**  
 Direction numbers for  
 (a) 4-directional chain code, and  
 (b) 8-directional chain code.





# Chain codes - example



a	b
c	d

**FIGURE 11.2**

(a) Digital boundary with resampling grid superimposed.  
 (b) Result of resampling.  
 (c) 4-directional chain code.  
 (d) 8-directional chain code.





# Chain codes normalization

- **Problem 1: different starting points result in different chain codes**
  - **normalization** → redefine the starting point such that the chain code forms a smallest number

code = {3,4,3,4,4,5,4,6,7,7,7,0,0,1,1,2} (a) Initial chain code	code = {4,3,4,4,4,5,4,6,7,7,7,0,0,1,1,2,3} (b) Result of one shift
code = {3,4,4,5,4,6,7,7,7,0,0,1,1,2,3,4} (c) Result of two shifts	code = {0,0,1,1,2,3,4,3,4,4,5,4,6,7,7,7} (d) Minimum integer chain code

- **Problem 2: object rotation results in different chain codes**
  - **difference code** → coding with the difference of directions (counter-clockwise)
  - E.g.: 0000655332 → 0006706076 → 0006706076







# Polygonal approximations

- Approximate a boundary using a polygon
  - Minimum perimeter polygons
  - Merging techniques
  - Splitting techniques





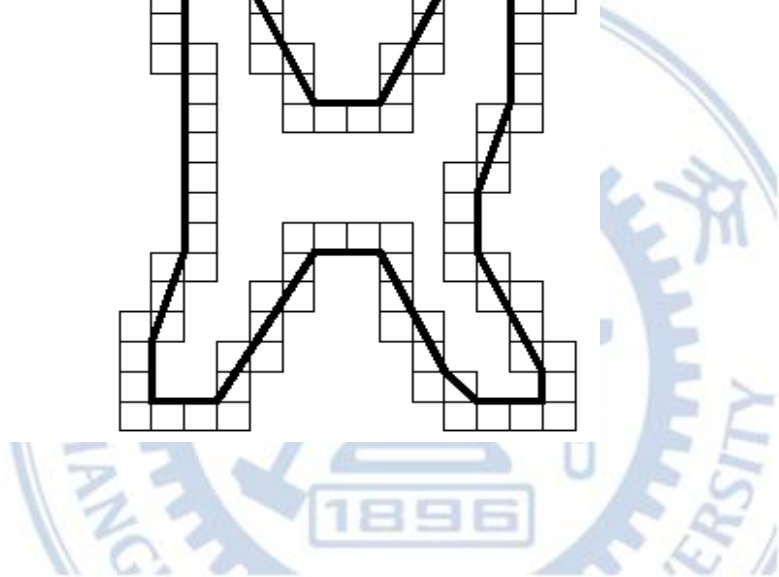
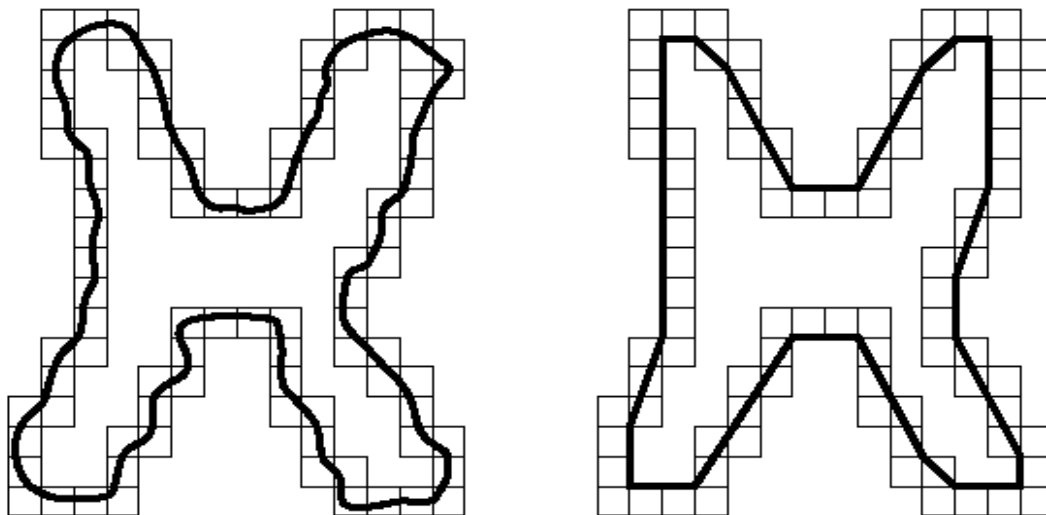
# Minimum perimeter polygons

- Choose an appropriate grid → The boundary is enclosed by a set of concatenated cells
- Allow the boundary to shrink as a rubber band

a b

**FIGURE 11.3**

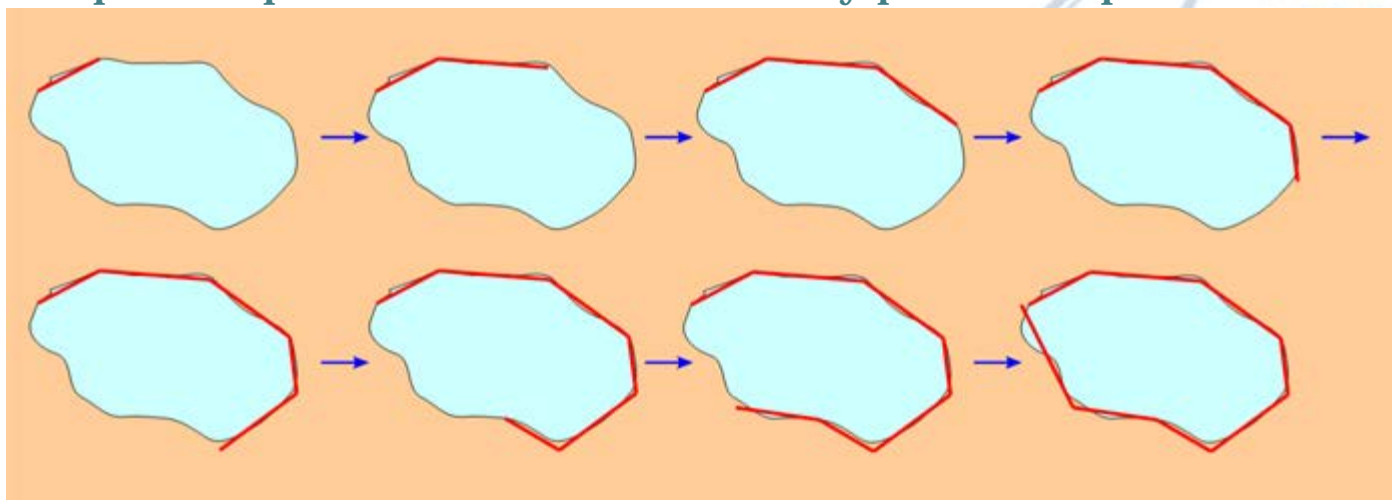
(a) Object boundary enclosed by cells.  
 (b) Minimum perimeter polygon.





# Merging techniques

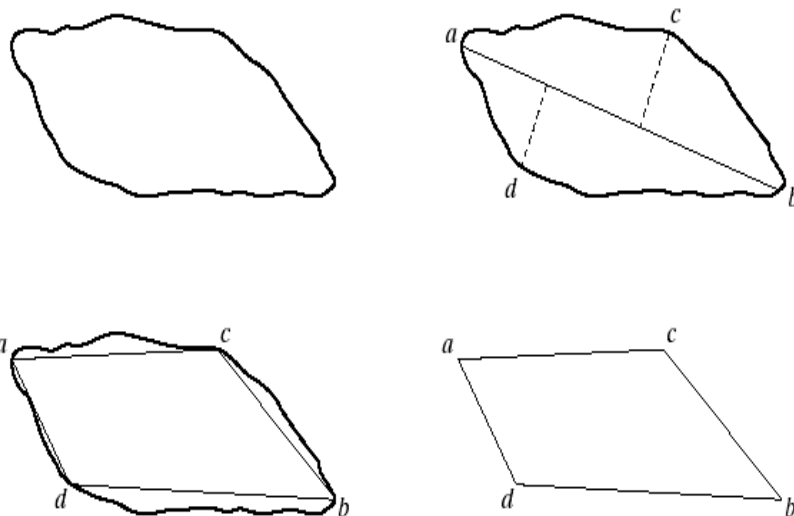
- Step 1:
  - Merge points along a boundary until the least square error line fit of the points merged so far exceeds a threshold
- Step 2:
  - Record the two end point of the line
- Step 3:
  - Repeat Steps 1 and 2 until all boundary points are processed





# Splitting technique

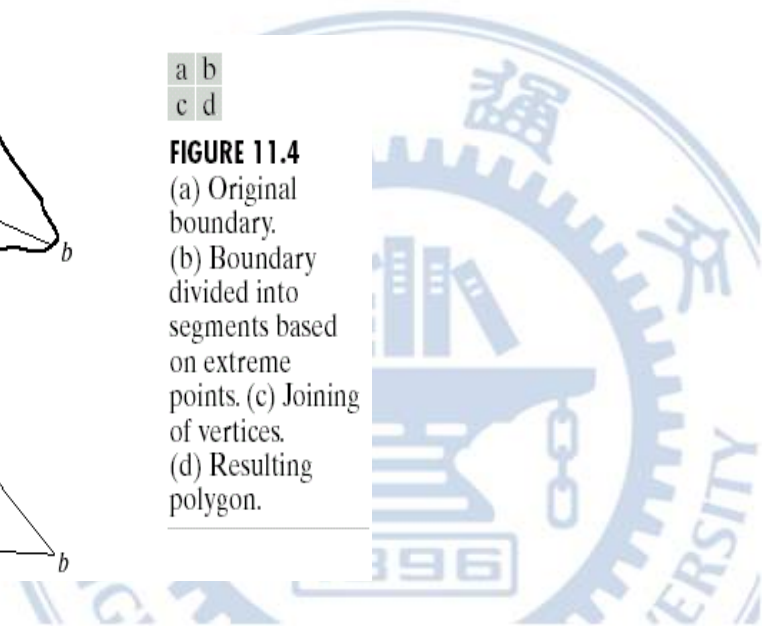
- Successively subdivide a segment into two parts until a given criterion is satisfied.
- One can use the major axis as the first subdivided line.
- The stopping criterion can be taken as 0.25 times the length of the major axis



a	b
c	d

**FIGURE 11.4**

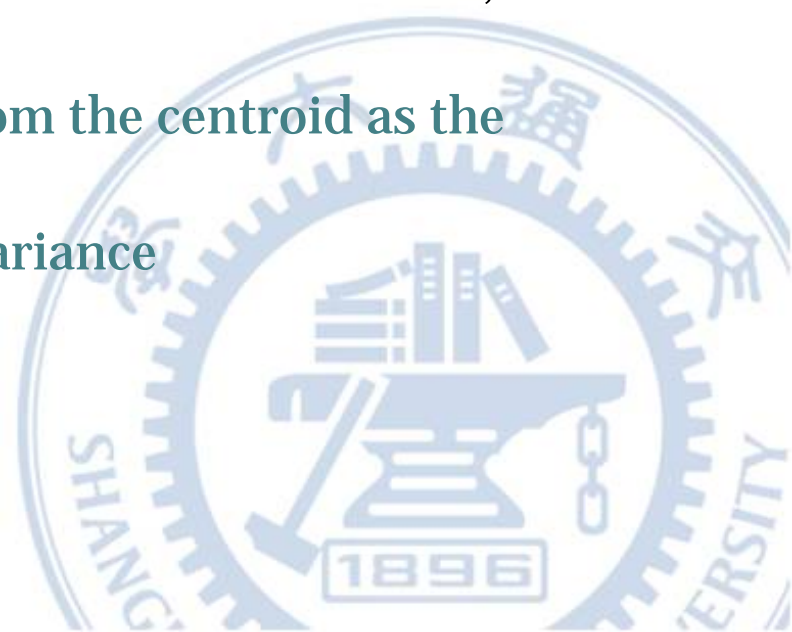
(a) Original boundary.  
 (b) Boundary divided into segments based on extreme points. (c) Joining of vertices.  
 (d) Resulting polygon.





# Signature

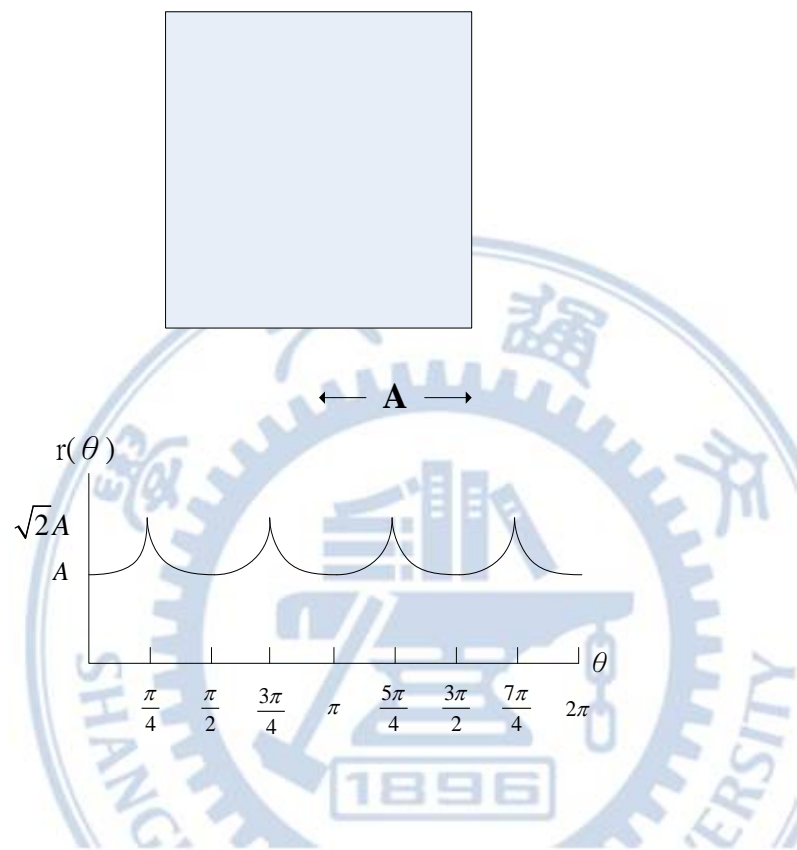
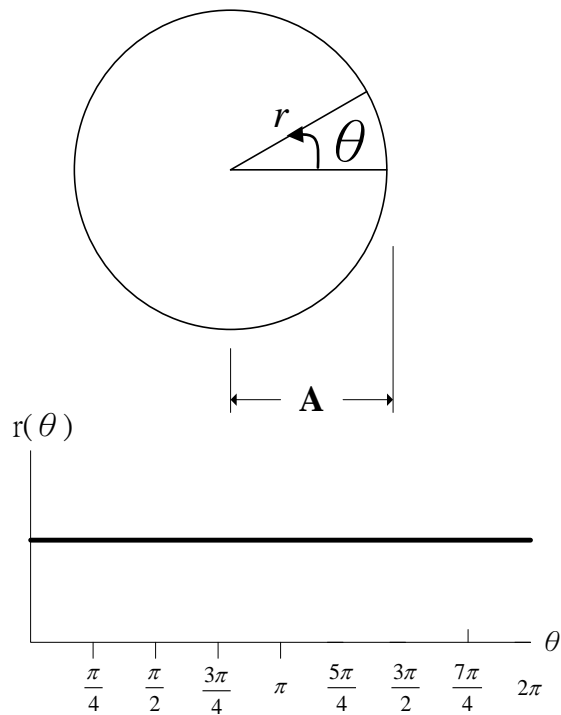
- A 1-D functional representation of a boundary
- Basic idea : reduce the boundary representation to a 1-D function, which might be easier to describe than a 2-D boundary
- One simple approach: use the distance from the centroid to the boundary as a function of angle. It is invariant to translation, but not to rotation and scaling.
  - **Rotation** : select the farthest point from the centroid as the starting point
  - **Scaling** : normalize the function by variance





# Signature

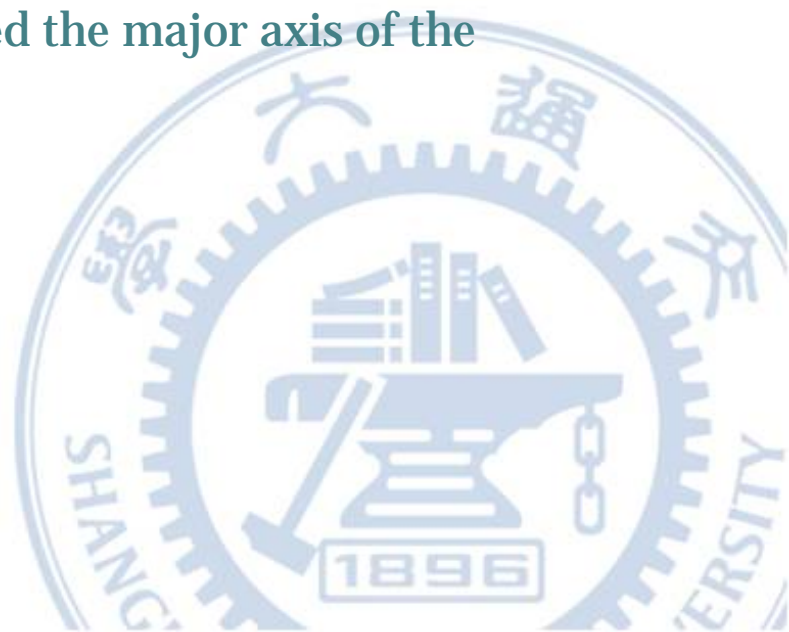
- Distance signature of circle shapes
- Distance signature of rectangular





# Boundary descriptors

- **Length**
  - For a chain-coded curve with unit spacing
  - Length = the number of vertical and horizontal components + the number of diagonal components  $\times 2^{1/2}$
- **Diameter**
  - Maximum distance between any two points on the boundary
  - The line formed by this two points is called the major axis of the boundary
- **Fourier descriptors**





# Fourier descriptors

- Points on the boundary can be treated as a complex number

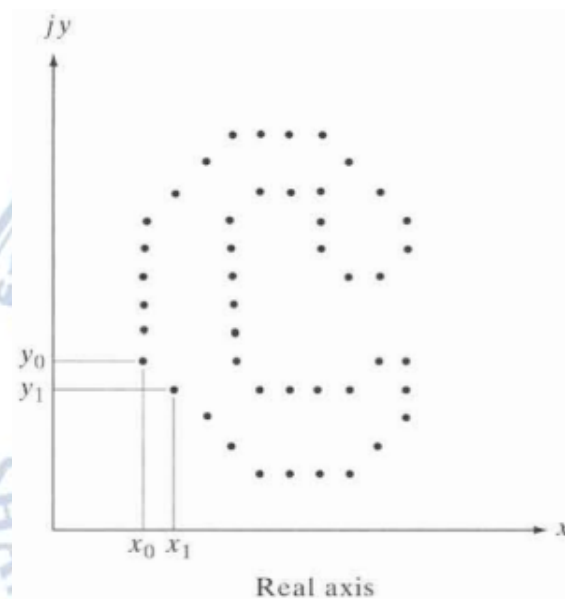
$$s(k) = x(k) + jy(k)$$

- Fourier descriptor : the discrete Fourier transform (DFT) of the  $s(k)$

$$a(u) = \frac{1}{N} \sum_{k=0}^{N-1} s(k) \exp[-j2\pi \cdot uk/N]$$

- **Reconstruction:** if  $a(u) = 0$ , for  $u > N - 1$

$$\hat{s}(k) = \sum_{u=0}^{N-1} a(u) \exp[j2\pi \cdot uk/N]$$





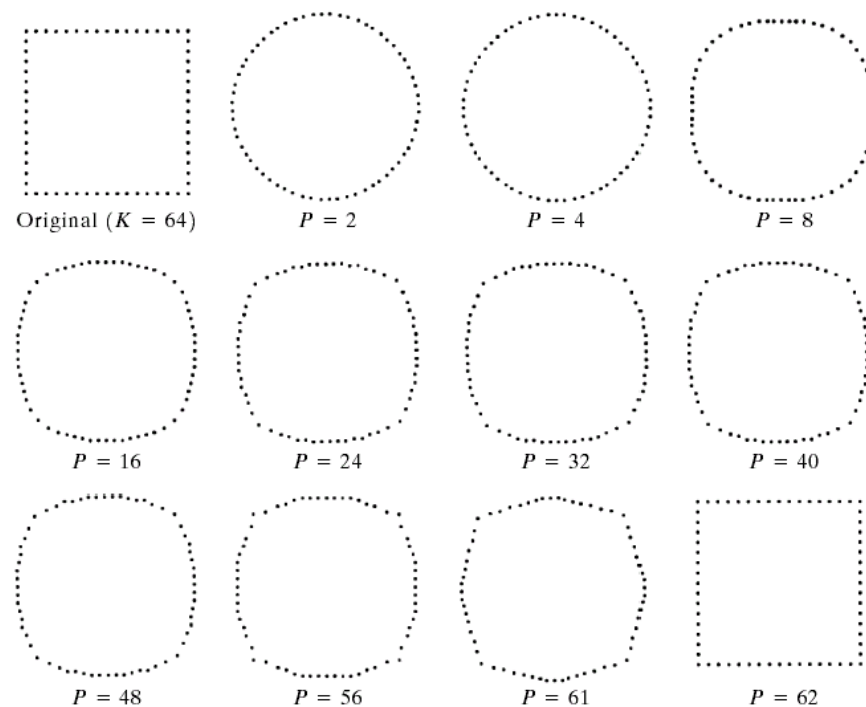


# Fourier descriptors

- Usually, only the first few coefficients are used to represent the shape

- Disadvantage:
  - Just for closed boundaries

**FIGURE 11.14**  
Examples of reconstruction from Fourier descriptors.  $P$  is the number of Fourier coefficients used in the reconstruction of the boundary.





# Simple region descriptor

- **Area**

- The number of pixels contained within its boundary

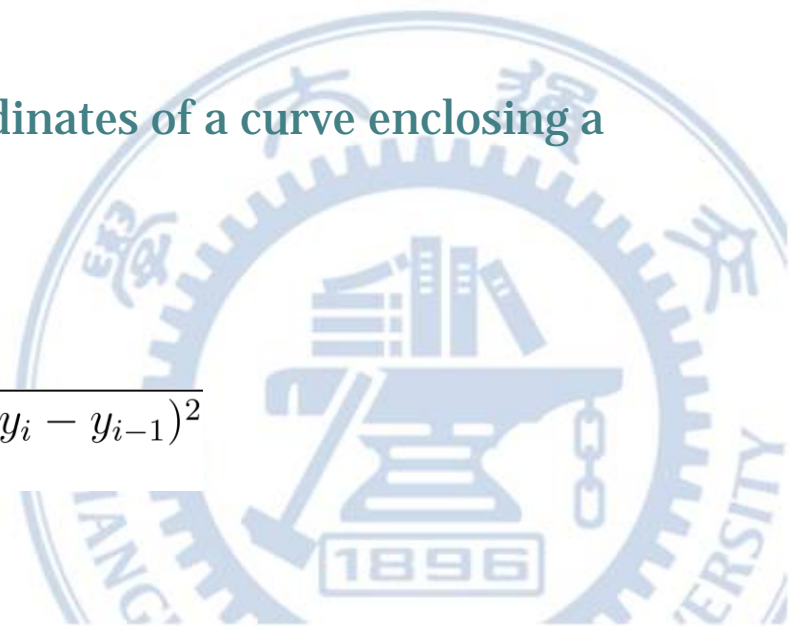
$$A(S) = \sum_{x,y} 1 \quad (x, y) \in S$$

- **Perimeter**

- The length of its boundary
- $x(t)$  and  $y(t)$  denote the parametric co-ordinates of a curve enclosing a region  $S$

$$P(S) = \int_t \sqrt{x^2(t) + y^2(t)} dt$$

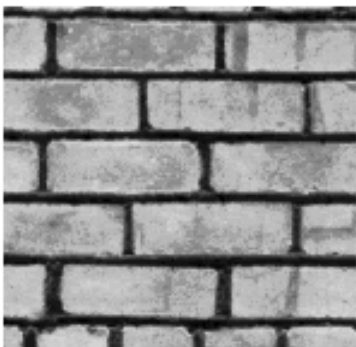
$$P(S) = \sum_i \sqrt{(x_i - x_{i-1})^2 + (y_i - y_{i-1})^2}$$



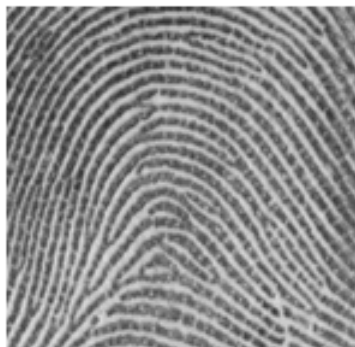


# Texture

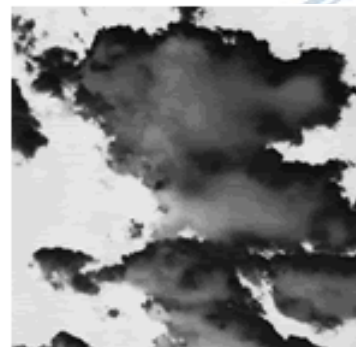
- No formal definition of texture exists
  - repeating patterns of local variations in image intensity, which is too fine to be distinguished
  - relates mostly to a specific, spatially repetitive (micro) structure of surfaces formed by repeating a particular element or several elements in different relative spatial positions.



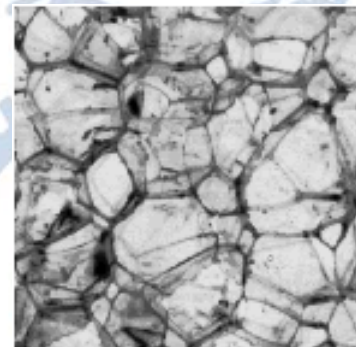
Brick Texture



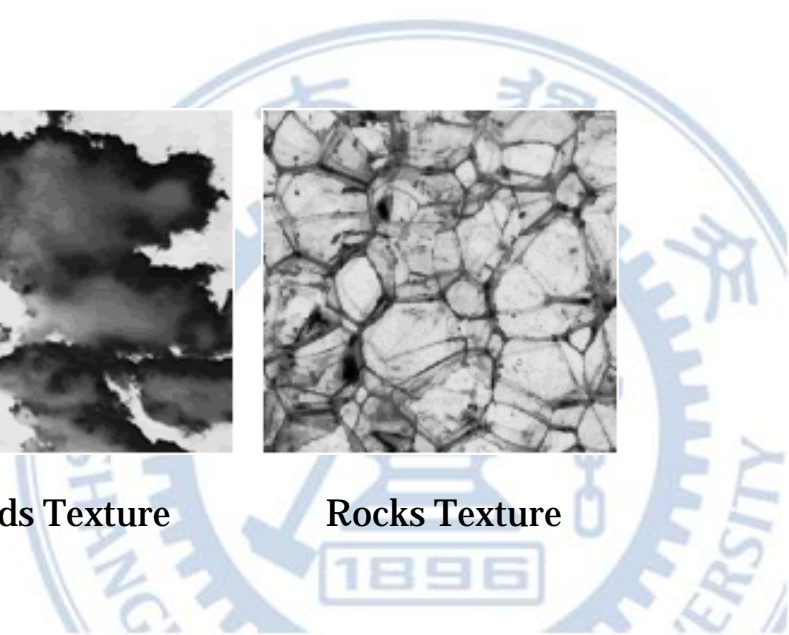
Finger print Texture



Clouds Texture



Rocks Texture





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# Thank You!

