Digital Image Processing Module 5: Part 2

Image Description and Representation

Image Representation and Description?

Objective:

To represent and describe information embedded in an image in other forms that are more suitable than the image itself.

Benefits:

- Easier to understand
- Require fewer memory, faster to be processed
- More "ready to be used"

What kind of information we can use?

- Boundary, shape
- Region
- Texture
- Relation between regions

Shape Representation by Using Chain Codes

Why we focus on a boundary?

The boundary is a good representation of an object shape and also requires a few memory.

Chain codes: represent an object boundary by a connected sequence of straight line segments of specified length and direction.



chain code

The First Difference of a Chain Codes

a chain code sequence depends on a starting point. **Solution:** treat a chain code as a circular sequence and redefine the starting point so that the resulting sequence of numbers forms an

Problem of a chain code:

integer of minimum magnitude.

8-directional (Images chain code Wood, I

Examples of Chain Codes



Polygon Approximation



polygon

Minimum perimeter polygon consists of line segments that minimize distances between boundary pixels.

(Images from Rafael C. Gonzalez and Richard E Wood, Digital Image Processing, 2nd Edition.



The first difference of a chain code: counting the number of direction change (in counterclockwise) between 2 adjacent elements of the code.



Distance-Versus-Angle Signatures

Represent an 2-D object boundary in term of a 1-D function of radial distance with respect to θ .



Boundary Segments

Concept: Partitioning an object boundary by using vertices of a convex hull.



(Images from Rafael C. Gonzalez and Richard E Wood, Digital Image Processing, 2nd Edition.

Convex Hull Algorithm

Input : A set of points on a cornea boundary

Output: A set of points on a boundary of a convex hull of a cornea 1. Sort the points by x-coordinate to get a sequence p₁, p₂, ..., p_n

For the upper side of a convex hull

2. Put the points \boldsymbol{p}_1 and \boldsymbol{p}_2 in a list \boldsymbol{L}_{upper} with \boldsymbol{p}_1 as the first point

3. For i = 3 to n

6.

- 4.
- Do append p_i to $L_{_{upper}}$ While $L_{_{upper}}$ contains more than 2 points and the last 3 5. points in L_{upper} do not make a right turn

Do delete the middle point of the last 3 points from Lupper



Convex Hull Algorithm (cont.)

- For the lower side of a convex hull
- 7. Put the points p_n and p_{n-1} in a list L_{lower} with p_n as the first point
- 8. For i = n-2 down to 1
- **Do** append \mathbf{p}_{i} to \mathbf{L}_{lower} 9.
- While L_{lower} contains more than 2 points and the last 3 points 10. in L_{lower} do not make a right turn
- Do delete the middle point of the last 3 points from L 11.
- 12. Remove the first and the last points from L
- 13. Append L_{lower} to L_{upper} resulting in the list L
- 14. Return L



Skeletons

Obtained from thinning or skeletonizing processes



(Images from Rafael C. Gonzalez and Richard E Wood, Digital Image Processing, 2st Edition.

Thinning Algorithm

Con	Concept: 1. Do not remove end points									
	2. Do not break connectivity									
	3. Do not cause excessive erosion									
App neig	Apply only to contour pixels: pixels "1" having at least one of its 8 neighbor pixels valued "0"						8 8			
Not	ation:									
		p_9	p_2	p_3		Neighb	orhood			
	Let	p _o	p_1	p_A	=	arrange	ment			
		10	11	1 4		for the	thinning			
		p_7	p_6	<i>p</i> ₅		algorith	im	Ex	amp	ole
Let	$N(p_1) = p$	2+	p_{3} -	+	+ p	$p_{8} + p_{9}$		0	0	1
	$T(p_1) = $ the n	umbe	er of	trans	sitior	0-1 in		1	p_1	0
	the o	rdere	ed se	quen	$\operatorname{ce} p_2$, <i>p</i> ₃ ,		1	0	1
	, p ₈ , j	p_{9}, p_{2}	•					N()	() =	= 4
								T(j	$(p_1) =$	3

Thinning Algorithm (cont.)

Step 1. Mark pixels for deletion if the following conditions are true.

a) $2 \le N(p_1) \le 6$ b) $T(p_1) = 1$	(Apply to all border pixels
c) $p_2 \cdot p_4 \cdot p_6 = 0$	
d) $p_4 \cdot p_6 \cdot p_8 = 0$	

5)	p_9	p_2	<i>p</i> ₃
	p_8	p_1	p_4
	p_7	p_6	p_5

Step 2. Delete marked pixels and go to Step 3.

Step 3. Mark pixels for deletion if the following conditions are true.

a) $2 \le N(p_1) \le 6$ b) $T(p_1) = 1$ c) $p_2 \cdot p_4 \cdot p_8 = 0$ (Apply to all border pixels)

 $d) p_2 \cdot p_6 \cdot p_8 = 0$

Step 4. Delete marked pixels and repeat Step 1 until no change occurs.

Example: Skeletons Obtained from the Thinning Alg.



Boundary Descriptors

1. Simple boundary descriptors:

- we can use
 - Length of the boundary
 - The size of smallest circle or box that can totally enclosing the object
- 2. Shape number
- 3. Fourier descriptor
- 4. Statistical moments

(Images from Rafael C. Gonzalez and Richard E Wood, Digital Image Processing, 2nd Edition.



Shape Number (cont.)



(Images from Rafael C. Gonzalez and Richard E. Wood, Digital Image Processing, 2nd Edition.

Example: Shape Number **Fourier Descriptor** Fourier descriptor: view a coordinate (x,y) as a complex number (x = real part and y = imaginary part) then apply the Fourier transform to a sequence of boundary points. Let s(k) be a coordinate s(k) = x(k) + jy(k)2. Find the smallest rectangle of a boundary point k : that fits the shape 1. Original boundary $a(u) = \frac{1}{K} \sum_{k=0}^{K-1} s(k) e^{-2\pi u k/K}$ Fourier descriptor : Chain code: $0 \ 0 \ 0 \ 0 \ 3 \ 0 \ 0 \ 3 \ 2 \ 2 \ 3 \ 2 \ 2 \ 1 \ 2 \ 1 \ 1$ Reconstruction formula First difference: Imaginary axis 300031033013003130 $s(k) = \frac{1}{\kappa} \sum_{i=1}^{\kappa-1} a(u) e^{2\pi u k/\kappa}$ Shape No. 000310330130031303 4. Find the nearest Boundary 3. Create grid Grid. points (Images from Rafael C. Gonzalez and Richard E Wood, Digital Image Processing, 2nd Edition.

Example: Fourier Descriptor

Examples of reconstruction from Fourier descriptors



Fourier Descriptor Properties

Real axis

Some properties of Fourier descriptors

Transformation	Boundary	Fourier Descriptor
Identity	s(k)	a(u)
Rotation	$s_r(k) = s(k)e^{j\theta}$	$a_r(u) = a(u)e^{j\theta}$
Translation	$s_t(k) = s(k) + \Delta_{xy}$	$a_t(u) = a(u) + \Delta_{xy}\delta(u)$
Scaling	$s_s(k) = \alpha s(k)$	$a_s(u) = \alpha a(u)$
Starting point	$s_p(k) = s(k - k_0)$	$a_p(u) = a(u)e^{-j2\pi k_0 u/K}$

(Images from Rafael C. Gonzalez and Richard E Wood, Digital Image Processing, 2^{sd} Edition.

(Images from Rafael C. Wood Digital Image P

Statistical Moments



- Convert a boundary segment into 1D graph 1.
- 2. View a 1D graph as a PDF function
- 3. Compute the nth order moment of the graph