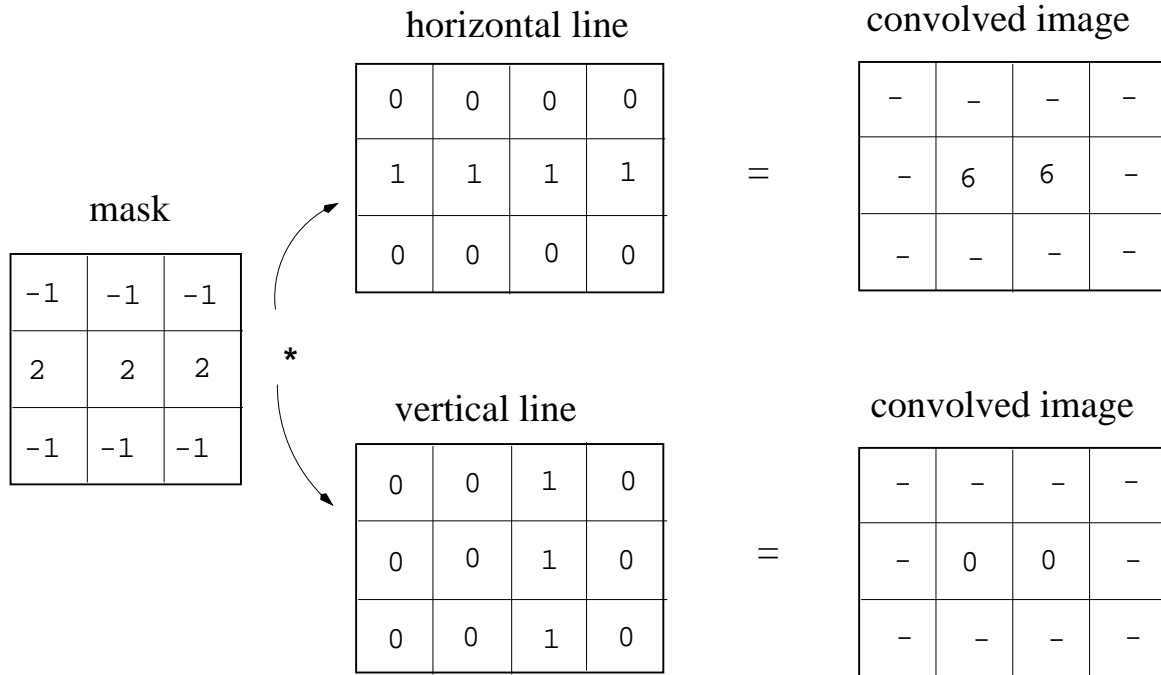
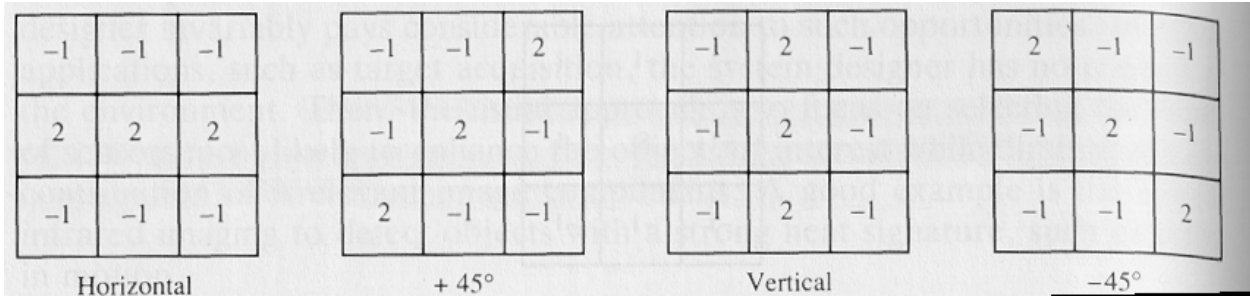


Line detection

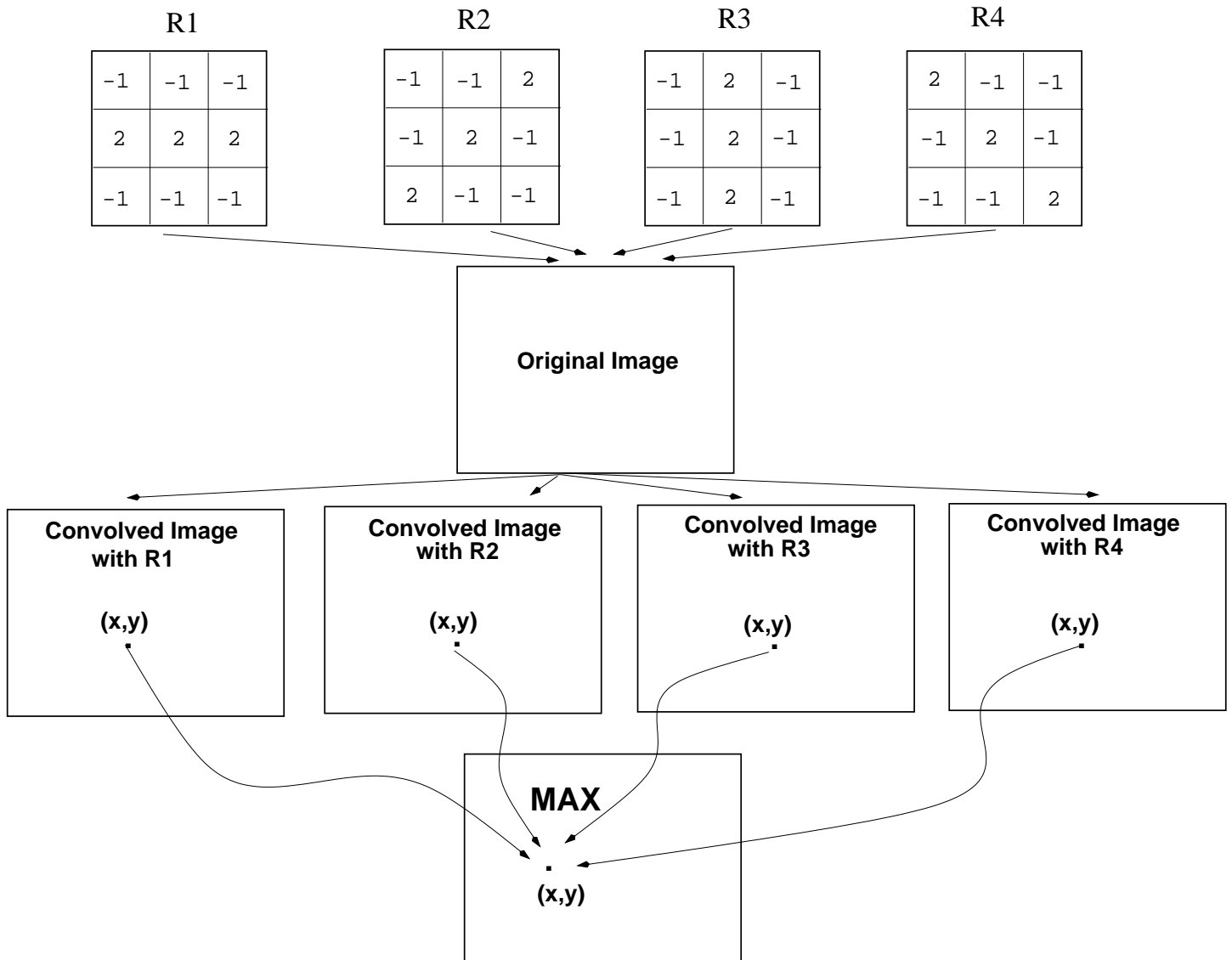
- The masks shown below can be used to detect lines at various orientations



- In practice, we run every mask over the image and we combine the responses:

$$R(x, y) = \max(|R_1(x, y)|, |R_2(x, y)|, |R_3(x, y)|, |R_4(x, y)|)$$

If $R(x, y) > T$, then discontinuity



Using Hough Transform to detect lines

(Trucco, Chapt. 5)

- Consider the slope-intercept equation of line

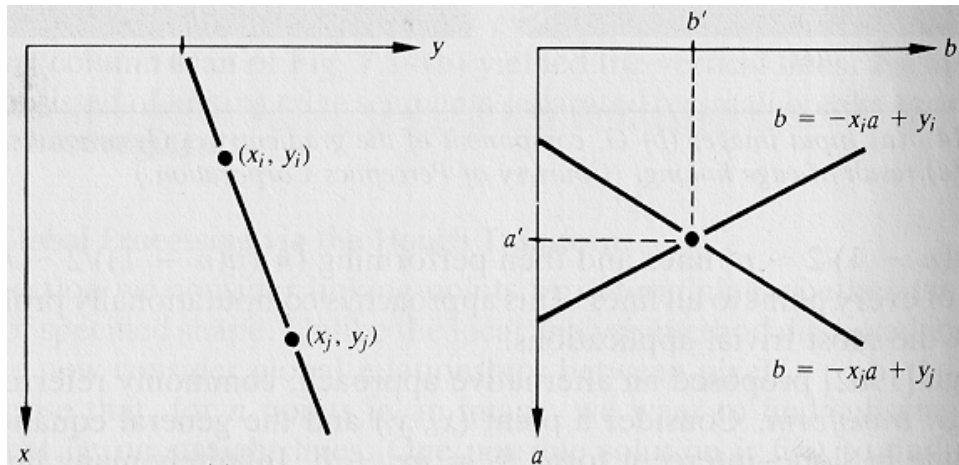
$$y = ax + b,$$

(a, b are constants, x is a variable, y is a function of x)

- Rewrite the equation as follows:

$$b = -xa + y$$

(now, x, y are constants, a is a variable, b is a function of a)



- The following properties are true:

Each point (x_i, y_i) defines a line in the $a - b$ space (parameter space)

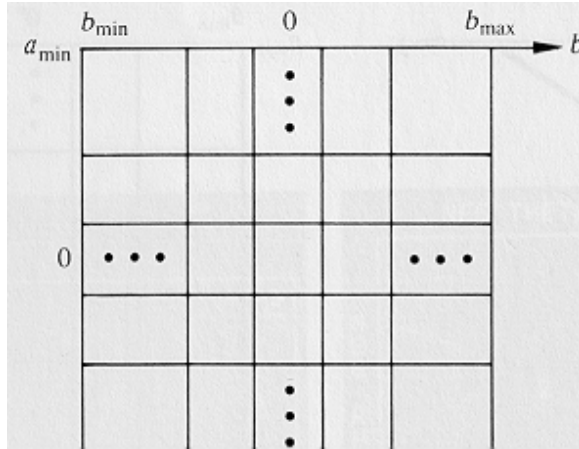
Points lying on the same line in the $x - y$ space, define lines in the parameter space which all intersect at the same point

The coordinates of the point of intersection define the parameters of the line in the $x - y$ space

Algorithm

1. Quantize the parameter space

$P[a_{\min}, \dots, a_{\max}][b_{\min}, \dots, b_{\max}]$ (accumulator array)



2. For each edge point (x, y)

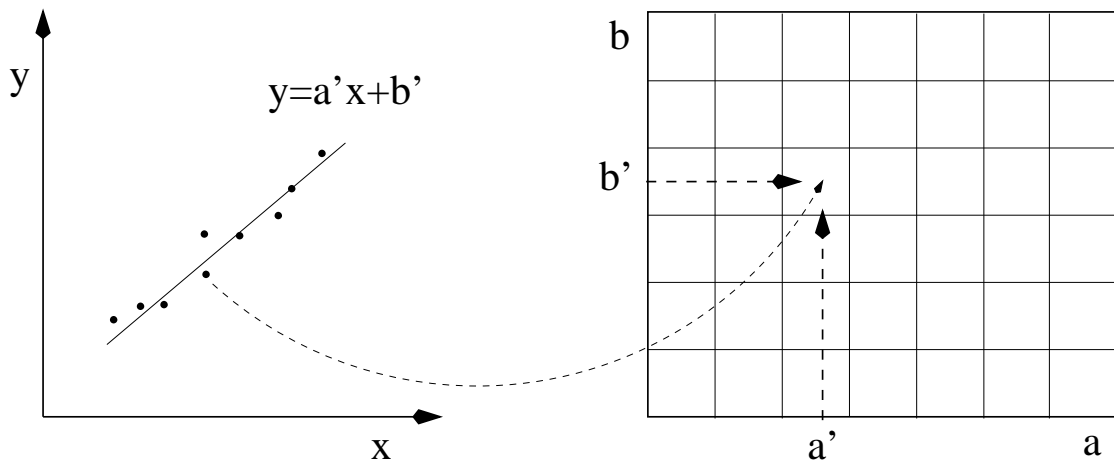
```
For( $a = a_{\min}; a \leq a_{\max}; a++$ ) {  
     $b = -xa + y$ ; /* round off if needed */  
     $(P[a][b])++$ ; /* voting */  
}
```

3. Find local maxima in $P[a][b]$

(If $P[a_j][b_k]=M$, then M points lie on the line $y = a_jx + b_k$)

- **Effects of quantization**

- The parameters of a line can be estimated more accurately using a finer quantization of the parameter space
- Finer quantization increases space and time requirements
- For noise tolerance, however, a coarser quantization is better



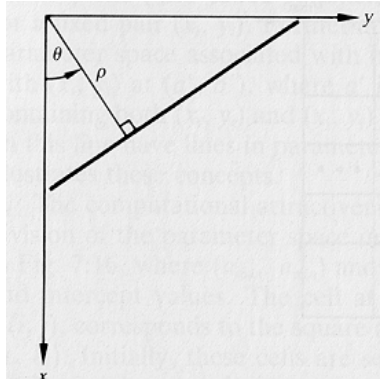
(it is very likely that every point will cast a vote in the (a', b') cell)

- **Problem with slope-intercept equation**

- The slope can become very large or even infinity !!
- It will be impossible to quantize such a large space

• **Polar representation of lines**

$$x \cos \theta + y \sin \theta = \rho \text{ (if the line is vertical, } \theta=0, x = \rho \text{)}$$

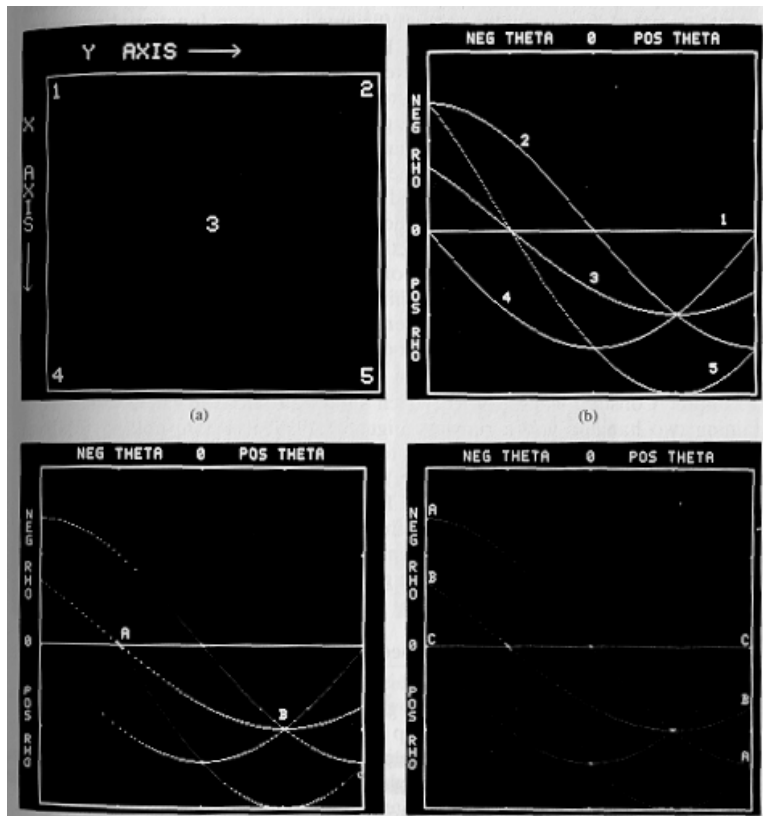


- The following properties are true:

Each point (x_i, y_i) defines a sinusoidal curve in the $\rho - \theta$ space (parameter space)

Points lying on the same line in the $x - y$ space, define curves in the parameter space which all intersect at the same point

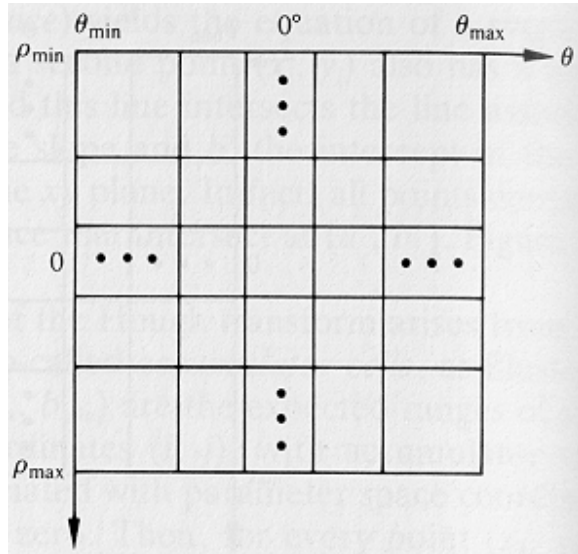
The coordinates of the point of intersection define the parameters of the line in the $x - y$ space



Algorithm

1. Quantize the parameter space

$P[\rho_{\min}, \dots, \rho_{\max}][\theta_{\min}, \dots, \theta_{\max}]$ (accumulator array)



2. For each edge point (x, y)

For($\theta = \theta_{\min}; \theta \leq \theta_{\max}; \theta++$) {

$\rho = x \cos \theta + y \sin \theta;$ /* round off if needed *

$(P[\rho][\theta])++;$ /* voting */

}

3. Find local maxima in $P[\rho][\theta]$

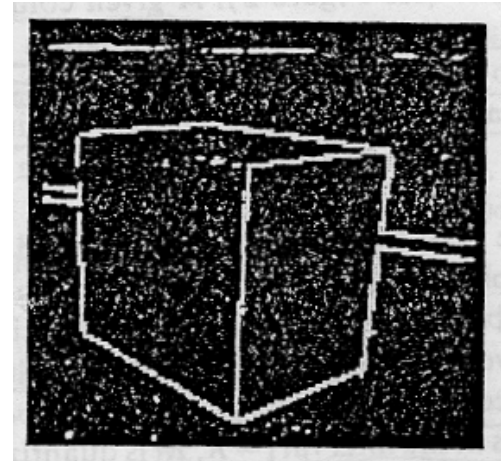
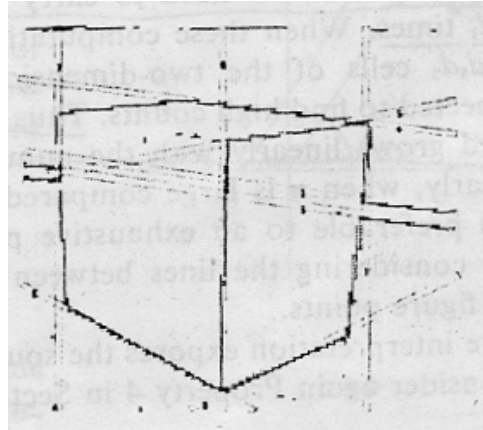
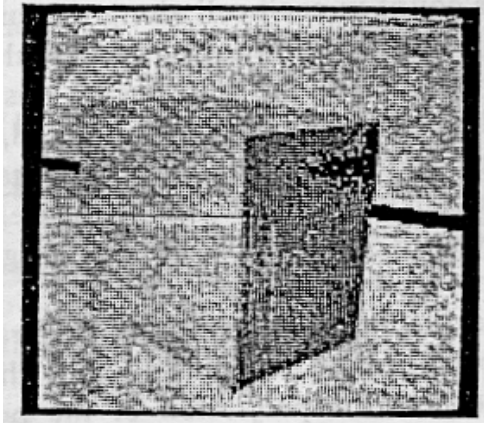


Table I. Accumulator Array for Figure 3(c)

θ ρ	0°	20°	40°	60°	80°	100°	120°	140°	160°
85									2
83							1		2
81							4		4
79			2				6		5
77				2			8	4	2
75							6	6	2
73							4	3	3
71		2				1	4	2	3
69			1	4		12	4	3	5
67			3		2	14	2	3	4
65			1			11	1	2	4
63					5	2		2	4
61						1		3	9
59	4	1			11	9	1	8	12
57	4	3		3	10	12	3	10	15
55	9	5		4	5	4	5	11	12
53	6	6		4	10		11	9	14
51	4	9		4	20	2	11	10	8
49	5	6		2	10	3	11	13	8
47	8	4	4	4		2	13	10	10
45	4	7	14	3			11	6	8
43	4	18	21	5		1	12	10	8
41	9	17	21	15		25	18	7	8
39	8	20	21	13		22	11	11	7
37	12	17	22	17		9	10	9	10
35	18	14	17	17	38	8	7	9	6
33	37	16	22	21	42	10	5	9	9
31	35	11	21	23	23	8	11	9	10
29	13	18	18	23	20	14	13	9	9
27	7	16	12	30	20	20	7	9	6
25	7	18	12	32	19	27	8	7	8
23	8	12	11	20	17	32	11	6	7
21	7	17	12	23	8	11	15	11	10
19	9	14	12	16	7	7	14	6	7
17	9	12	12	16	6	9	16	12	7
15	8	13	13	11	7	10	16	14	10
13	10	9	15	11	7	10	16	13	6
11	12	11	13	14	8	10	16	13	13
9	10	10	16	14	8	9	14	21	22
7	10	8	22	12	41	6	7	12	21
5	11	12	15	11	23	6	11	14	14
3	13	15	15	8	18	7	11	16	15
1	10	14	17	11	7	8	9	10	12

θ ρ	0°	20°	40°	60°	80°	100°	120°	140°	160°
-85									
-83									
-81									
-79			3		1				
-77			1		3				
-75									
-73									
-71			2						
-69			2						
-67									
-65				1		2			
-63			1				2		
-61									
-59	5								
-57	7		2		1				
-55	6						2		1
-53	0	10	6				1		1
-51	16	13	12			18	4		1
-49	32	18	11			15	16		21
-47	10	16	11		37	14	16		21
-45	7	17	11		11	16	18		41
-43	8	12	14		10	13	17		12
-41	6	7	14		11	14	14		7
-39	7	10	9		8	12	8		11
-37	7	7	14		8	17	9		12
-35	8	9	17		8	10	7		10
-33	6	12	15		8	12	9		11
-31	5	9	19		9	8	11		16
-29	9	10	12		9	8	9		18
-27	7	12	10		8	6	9		18
-25	5	10	8		8	7	7		22
-23	6	11	9		9	6	11		19
-21	7	15	9		7	10	10		16
-19	6	13	8		16	9	11		17
-17	7	17	9		15	7	11		16
-15	6	15	10		17	8	13		10
-13	10	15	9		15	9	17		11
-11	10	13	10		7	8	17		9
-9	7	14	8		7	8	23		8
-7	9	15	12		7	8	21		7
-5	13	15	9		7	7	14		10
-3	26	14	14		6	8	12		9
-1	10	13	18		9	8	8		11

Extending Hough Transform

- Hough transform can also be used for detecting circles, ellipses, etc.

- For example, the equation of circle is:

$$(x - x_0)^2 + (y - y_0)^2 = r^2$$

- In this case, there are three parameters: $(x_0, y_0), r$

- In general, we can use hough transform to detect any curve which can be described analytically by an equation of the form:

$$g(v, C) \quad (v: \text{vector of coordinates}, C: \text{parameters})$$

- Detecting arbitrary shapes, with no analytical description, is also possible
(Generalized Hough Transform)