



# Generalized Hough Transform

16-385 Computer Vision

# Hough Circles

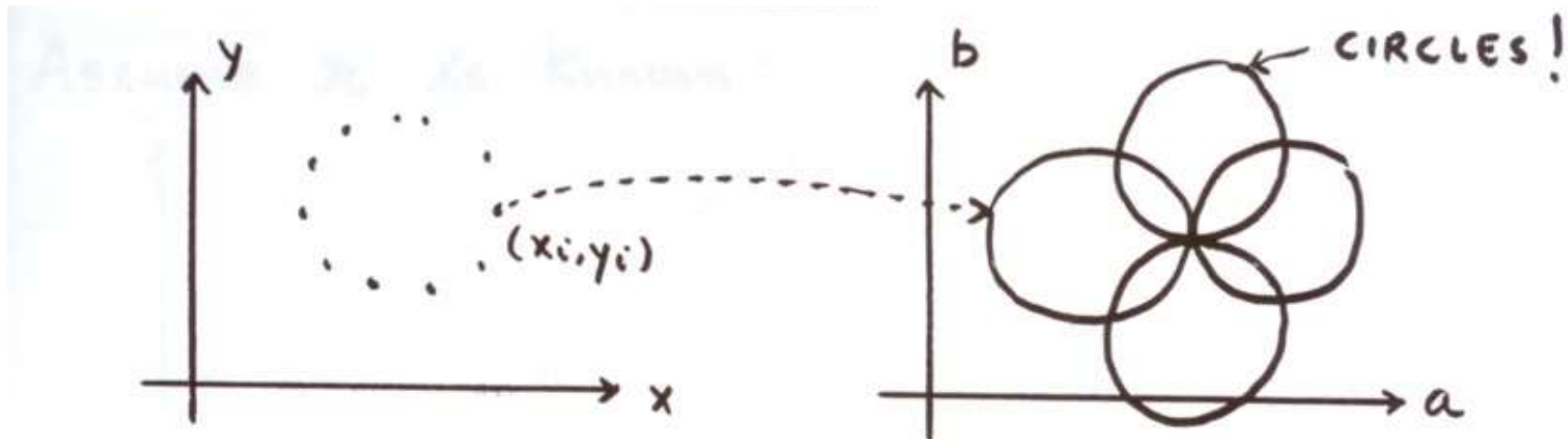
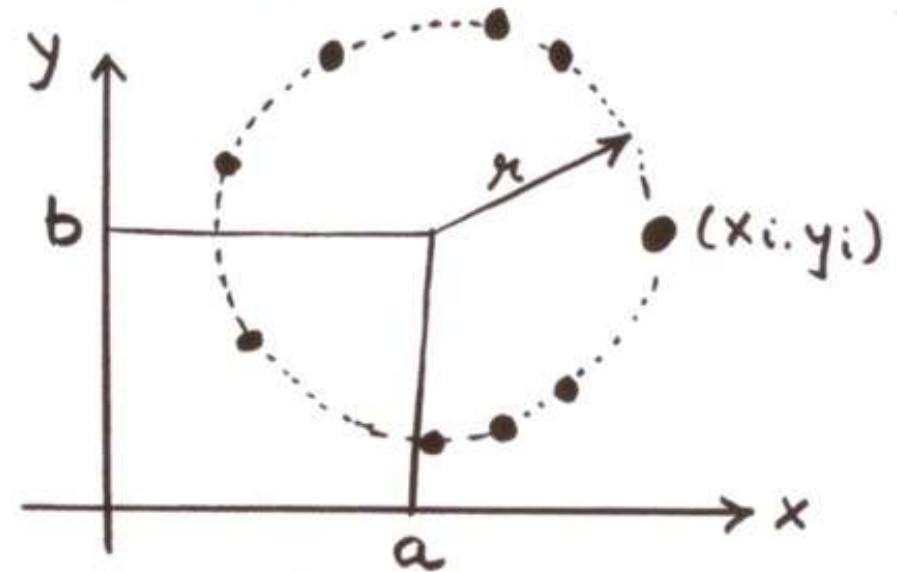
# Finding Circles by Hough Transform

Equation of Circle:

$$(x_i - a)^2 + (y_i - b)^2 = r^2$$

If radius is known: (2D Hough Space)

Accumulator Array  $A(a, b)$



$$(x - a)^2 + (y - b)^2 = r^2$$

parameters

variables

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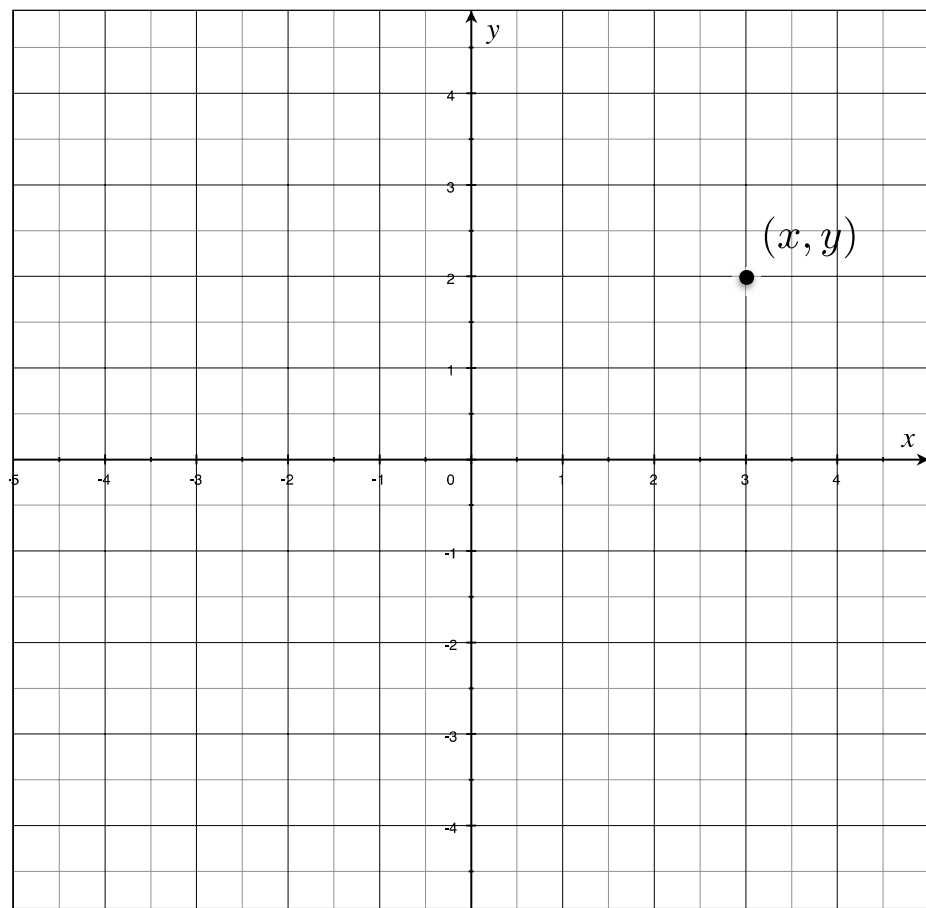
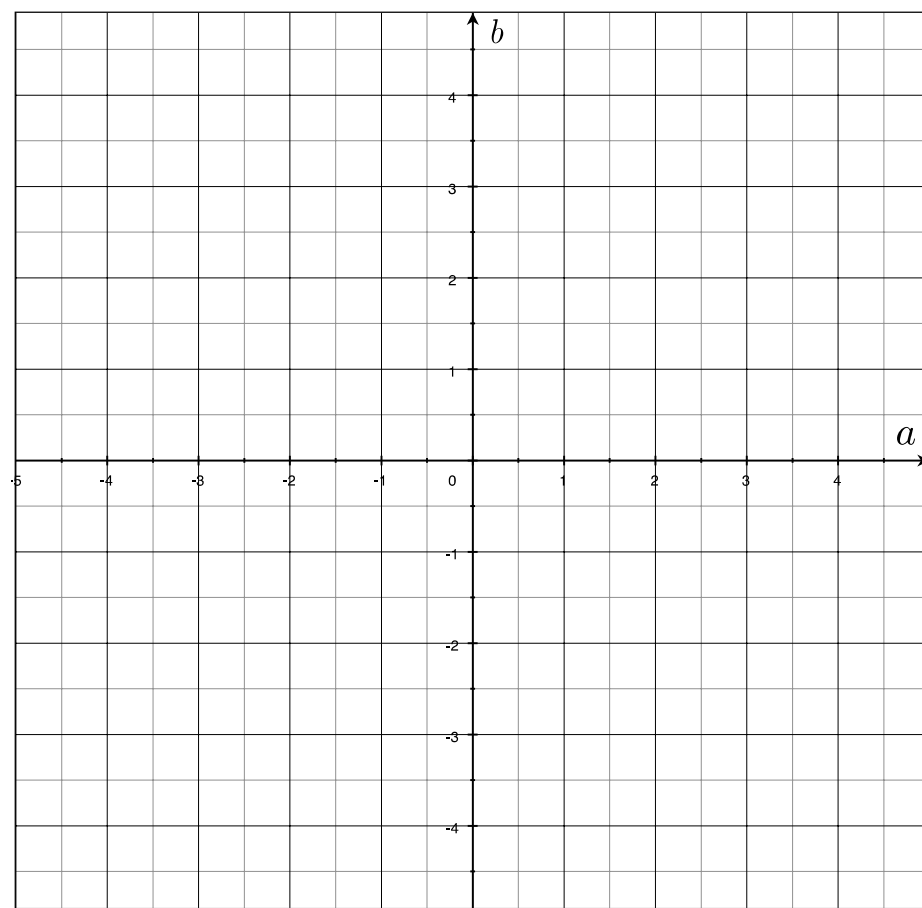


Image space



Parameter space

$$(x - a)^2 + (y - b)^2 = r^2$$

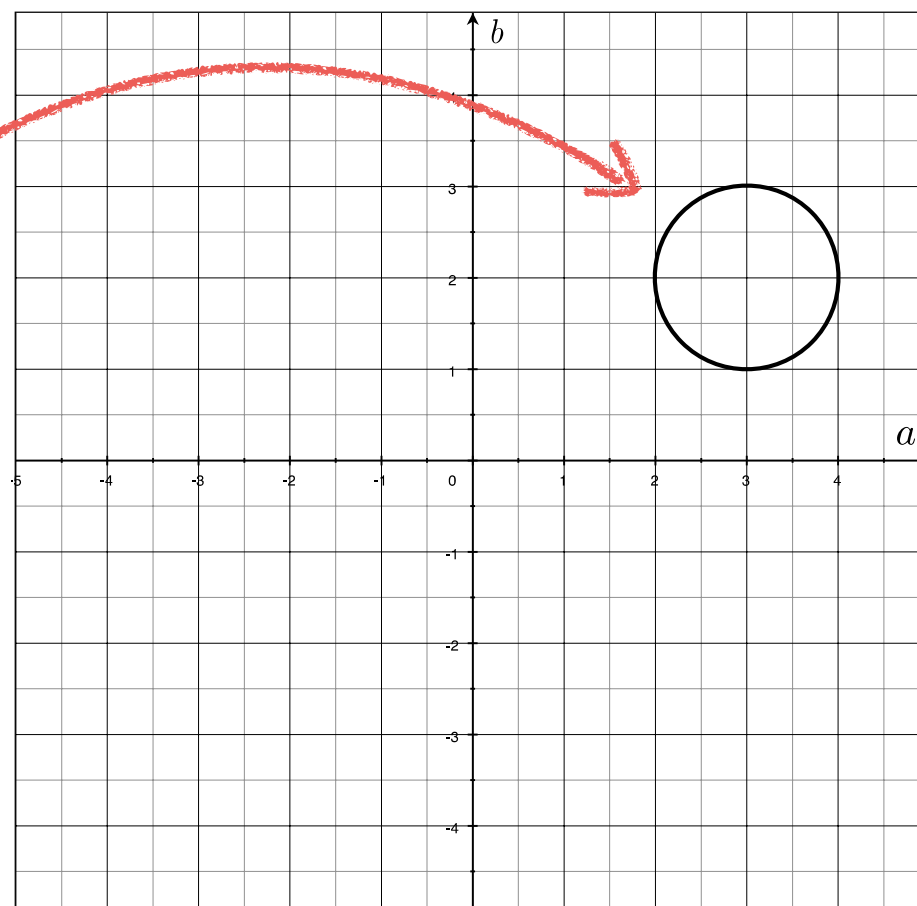
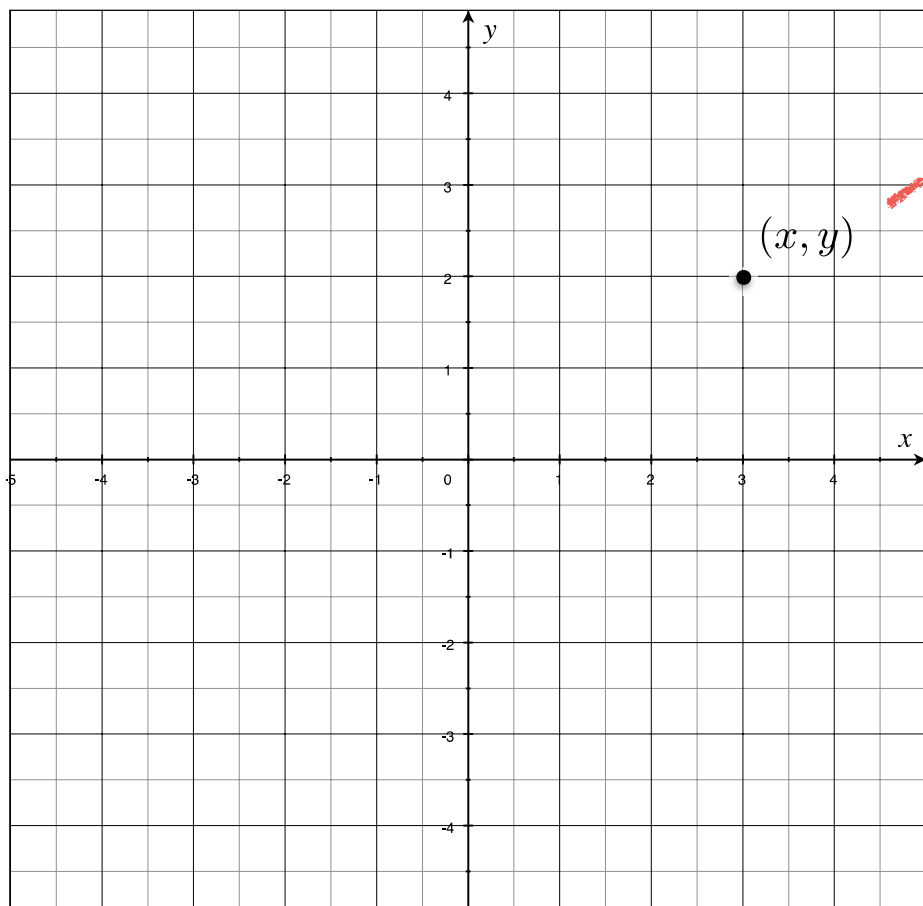
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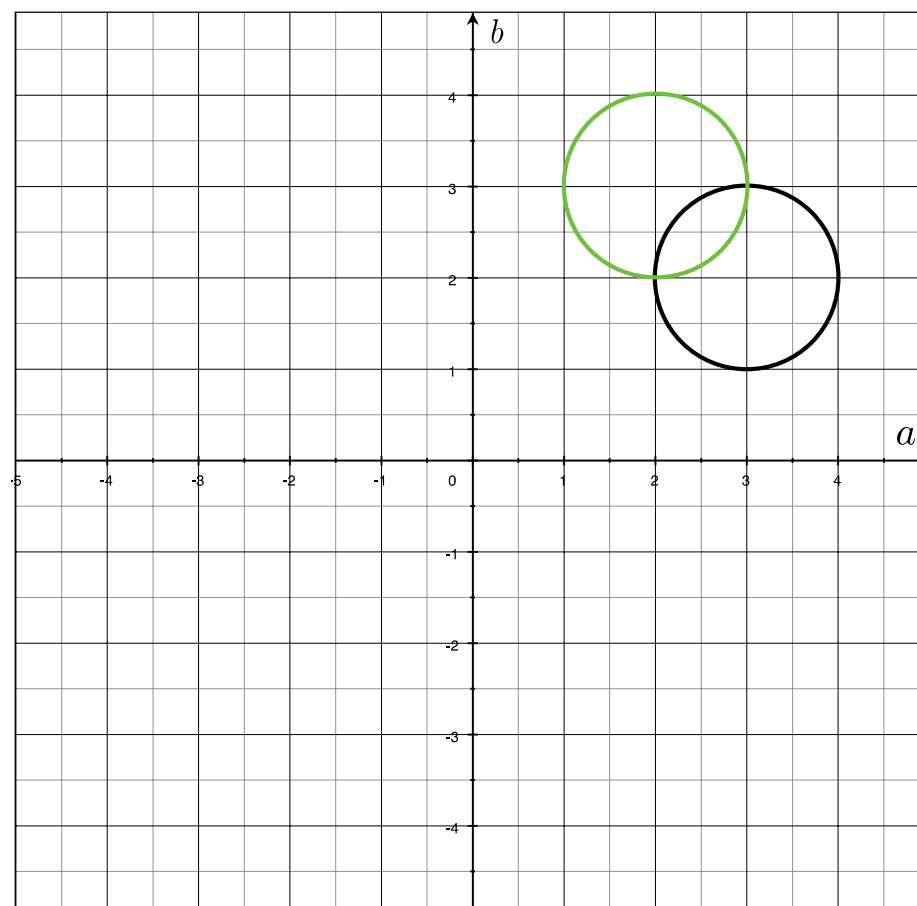
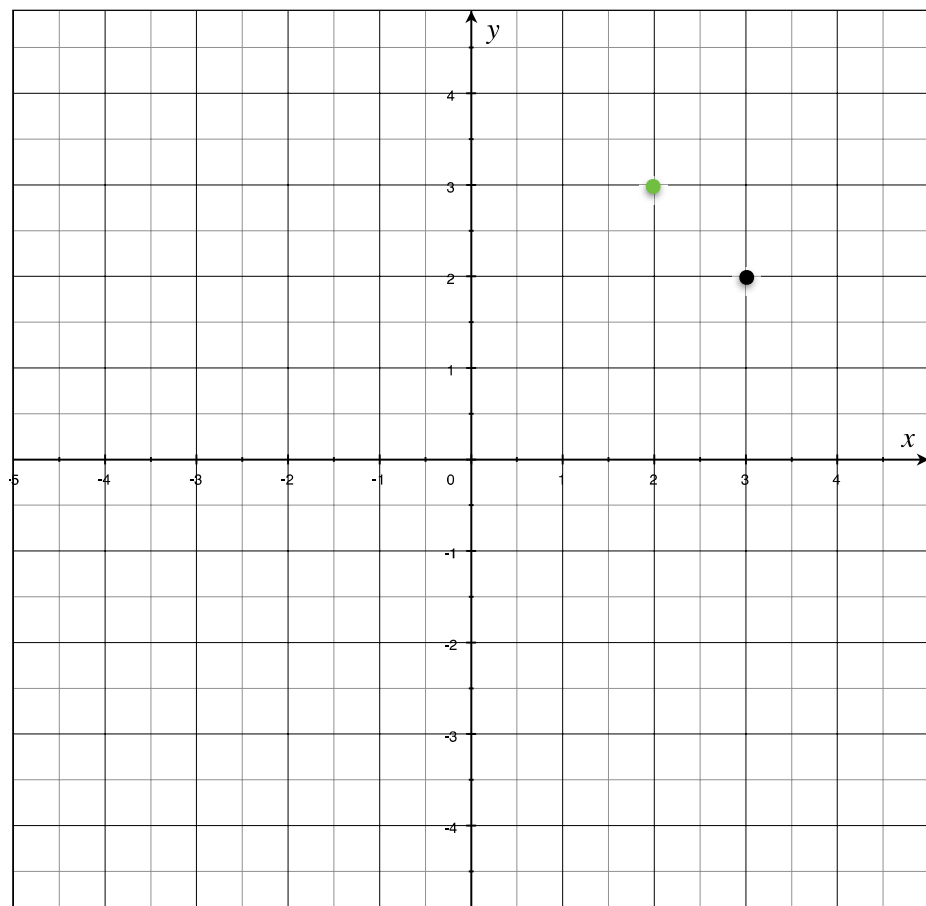
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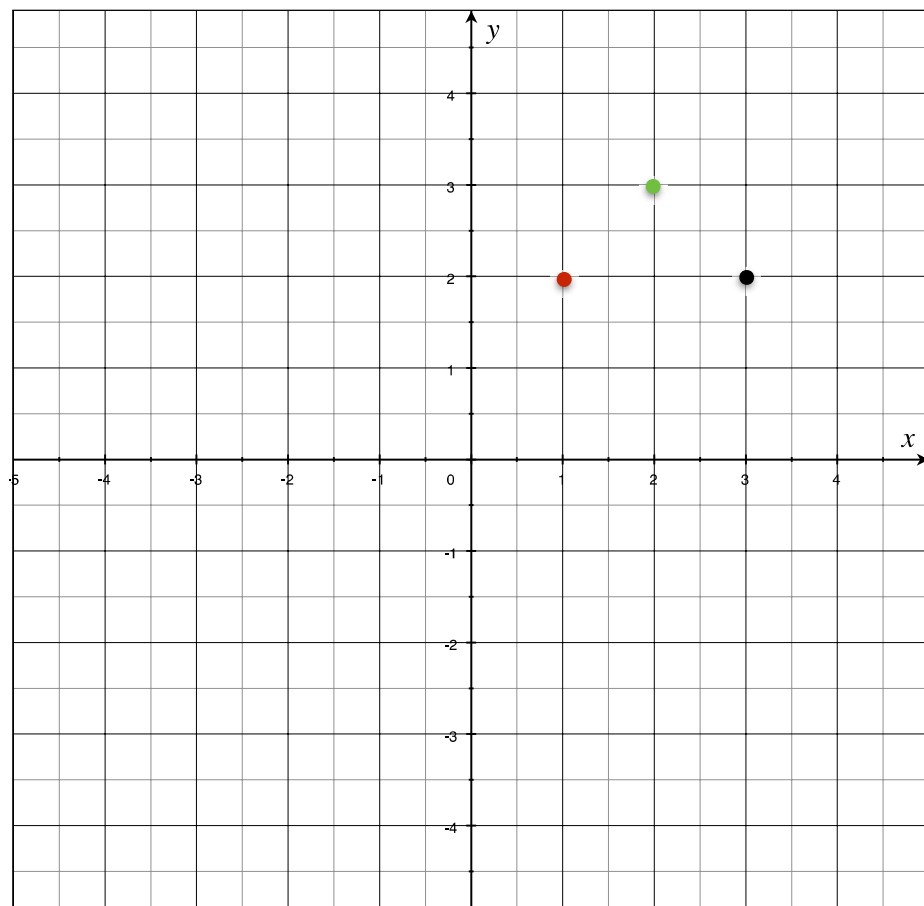
variables



parameters

$$(x - a)^2 + (y - b)^2 = r^2$$

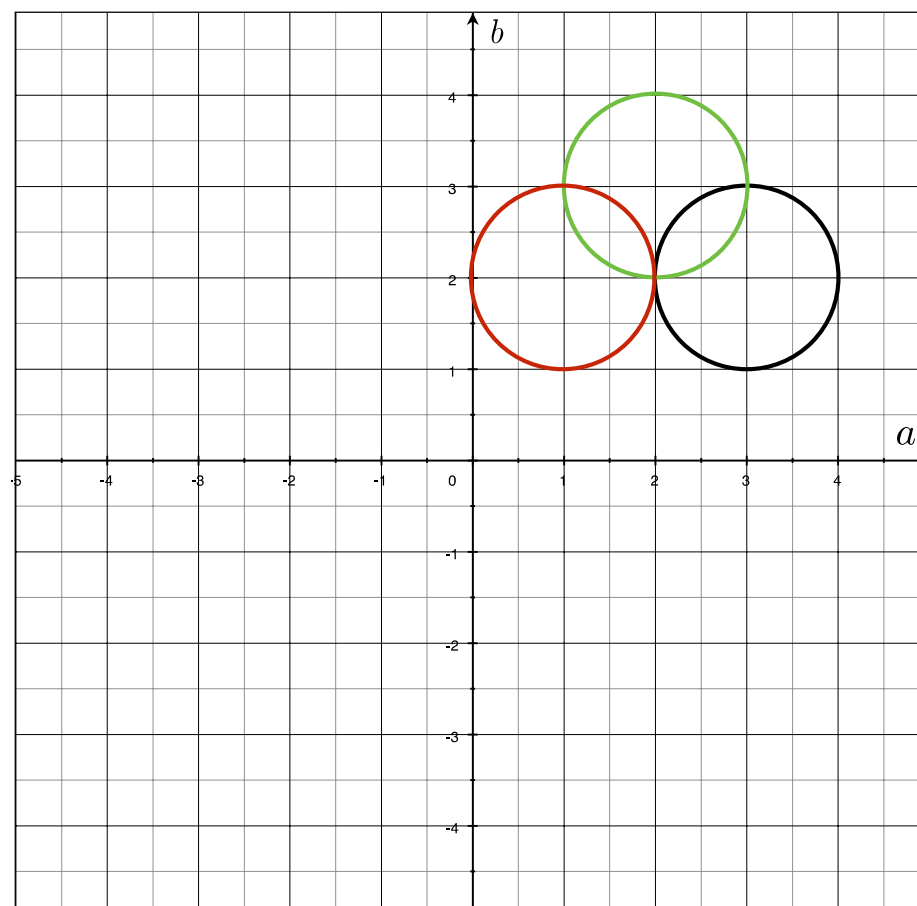
variables



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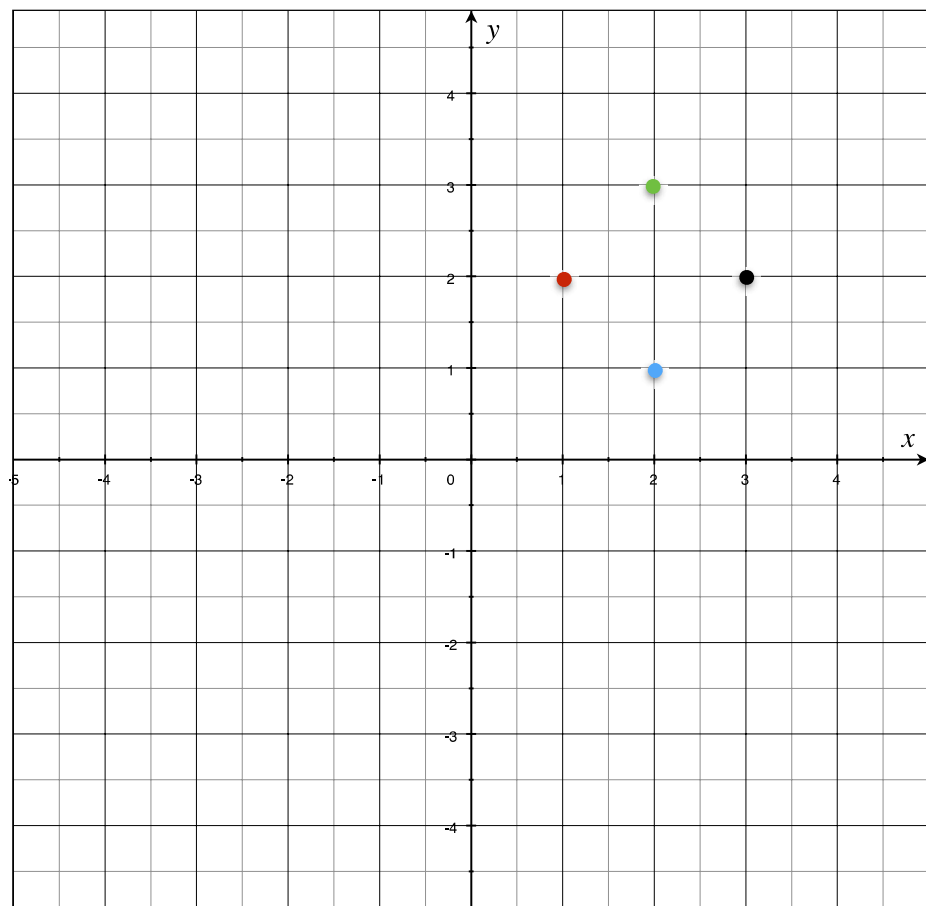
variables



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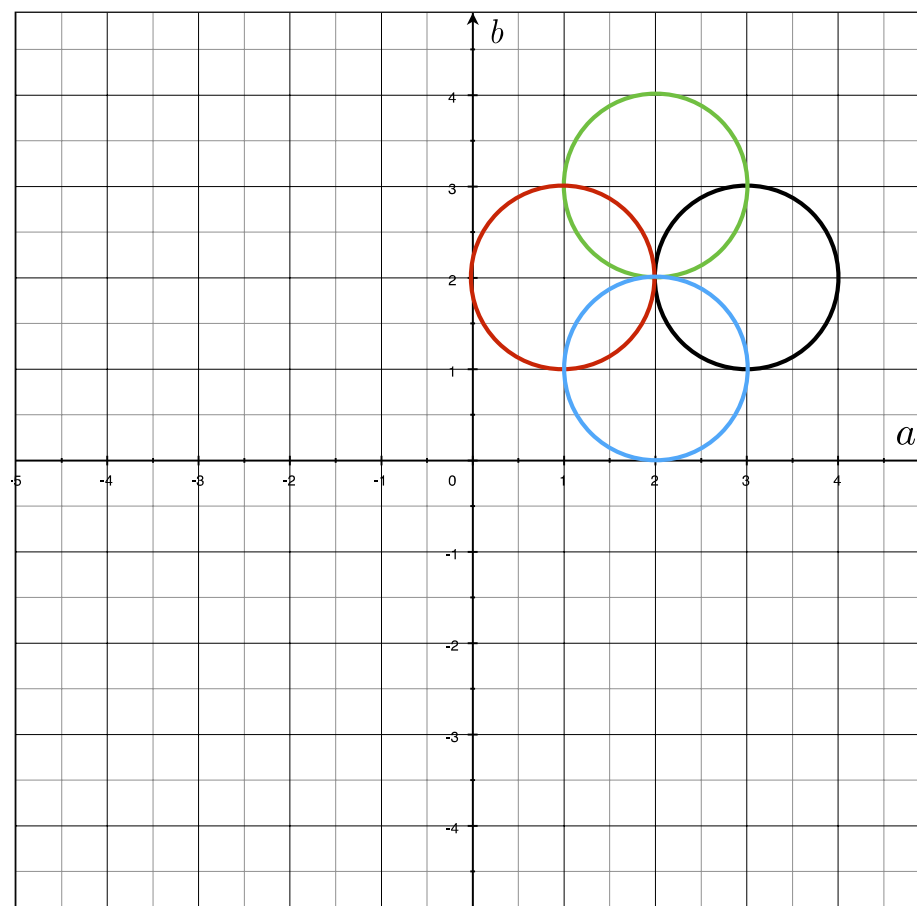
variables



parameters

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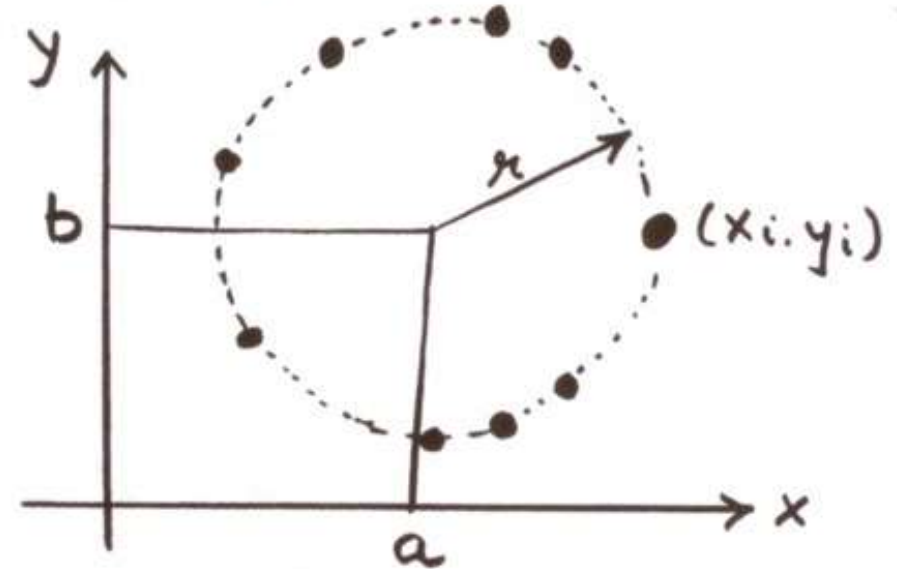




# Finding Circles by Hough Transform

Equation of Circle:

$$(x_i - a)^2 + (y_i - b)^2 = r^2$$



If radius is not known: 3D Hough Space!

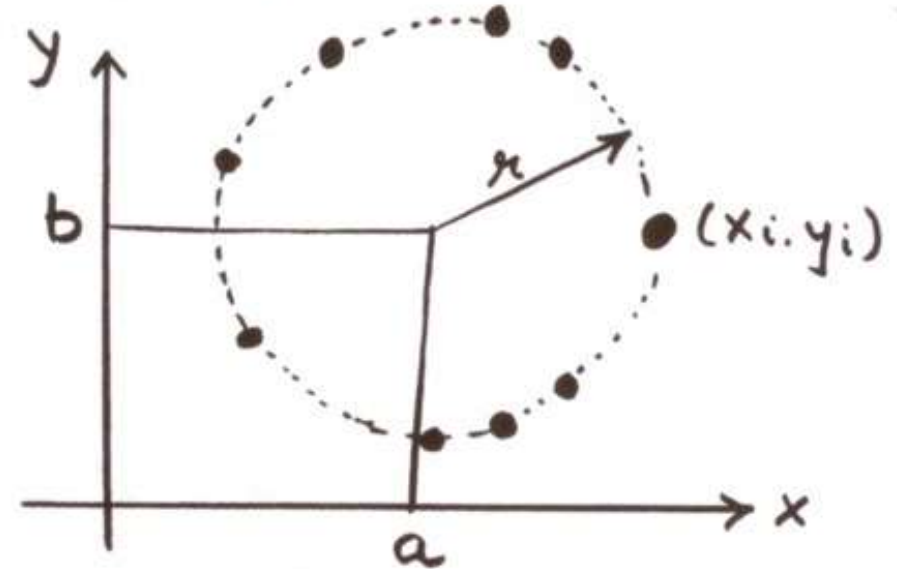
Use Accumulator array  $A(a, b, r)$

*What is the surface in the hough space?*

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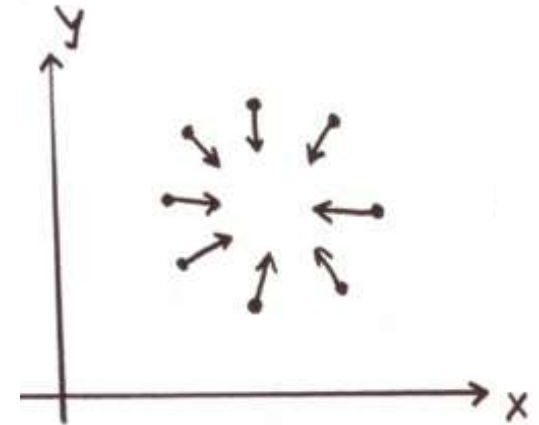


# Using Gradient Information

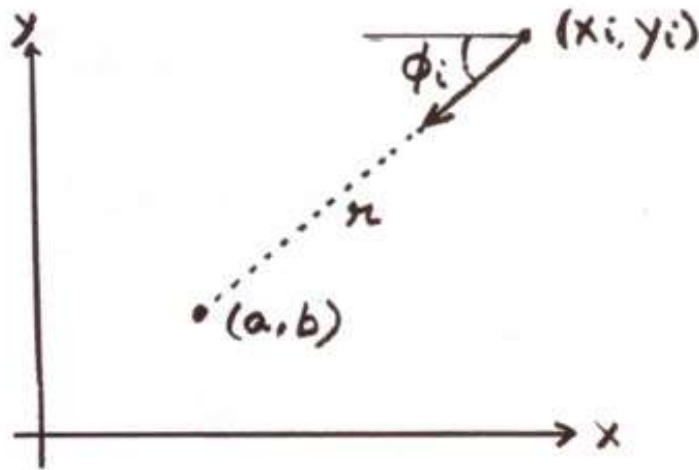
Gradient information can save lot of computation:

Edge Location  $(x_i, y_i)$

Edge Direction  $\phi_i$



Assume radius is known:



$$a = x - r \cos \phi$$

$$b = y - r \sin \phi$$

*Need to increment only one point in accumulator!!*

parameters

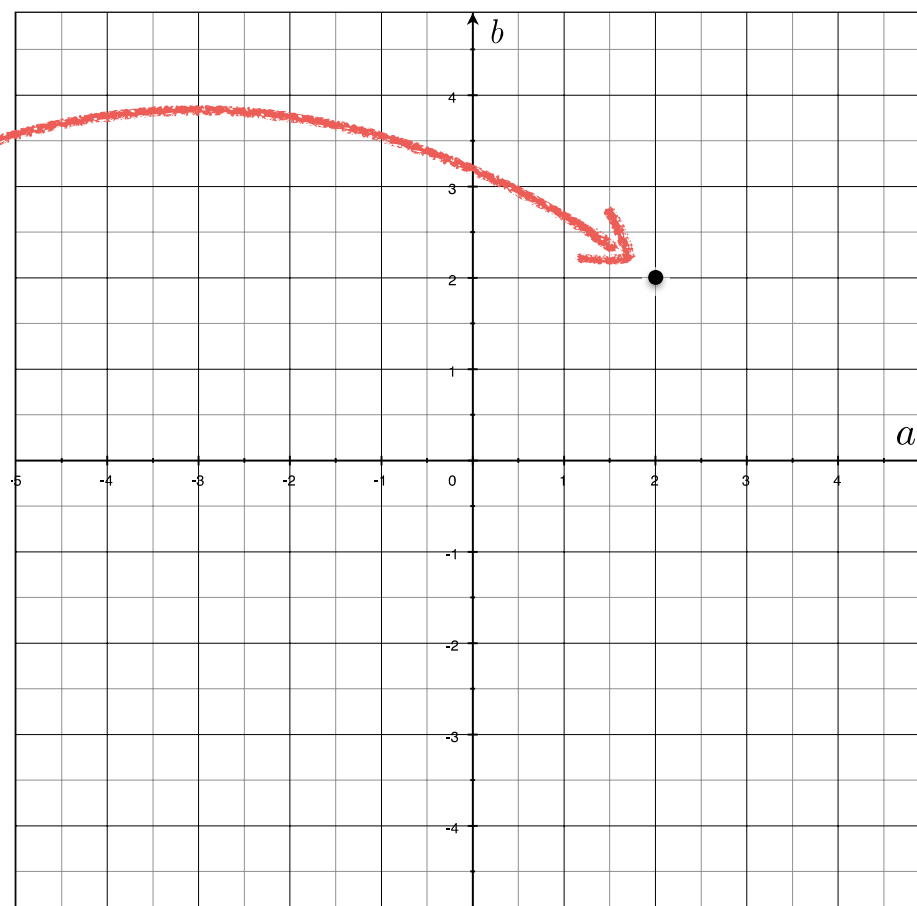
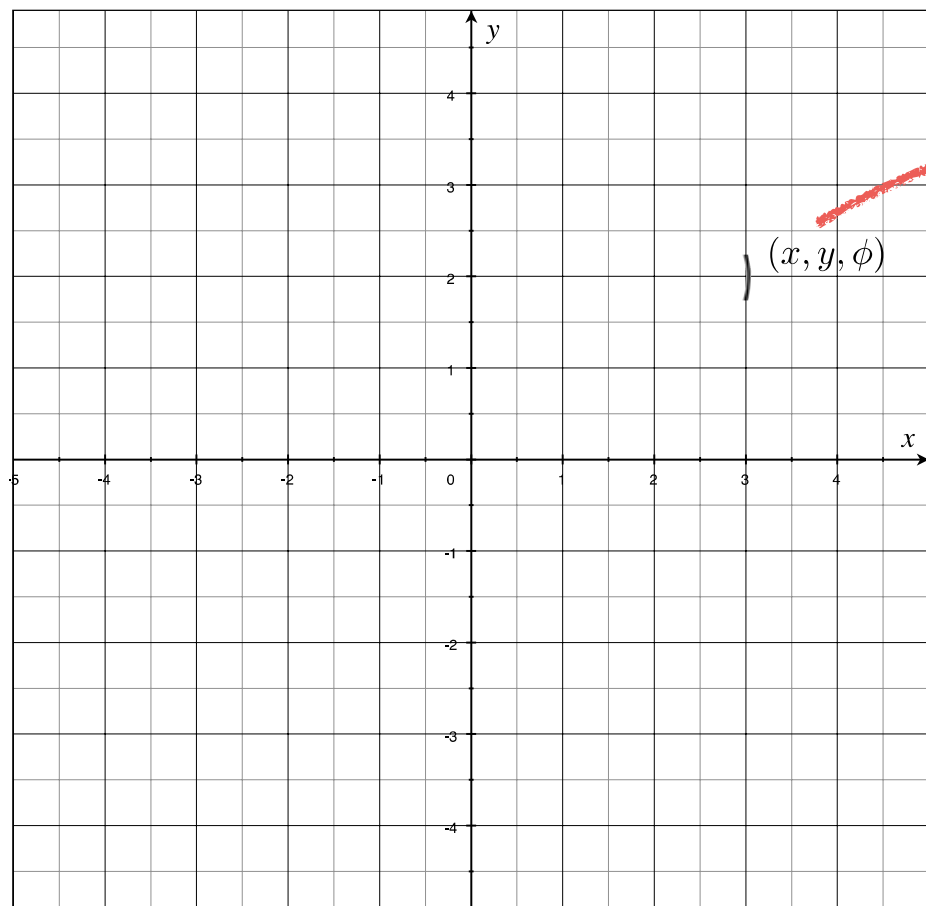
$$(x - a)^2 + (y - b)^2 = r^2$$

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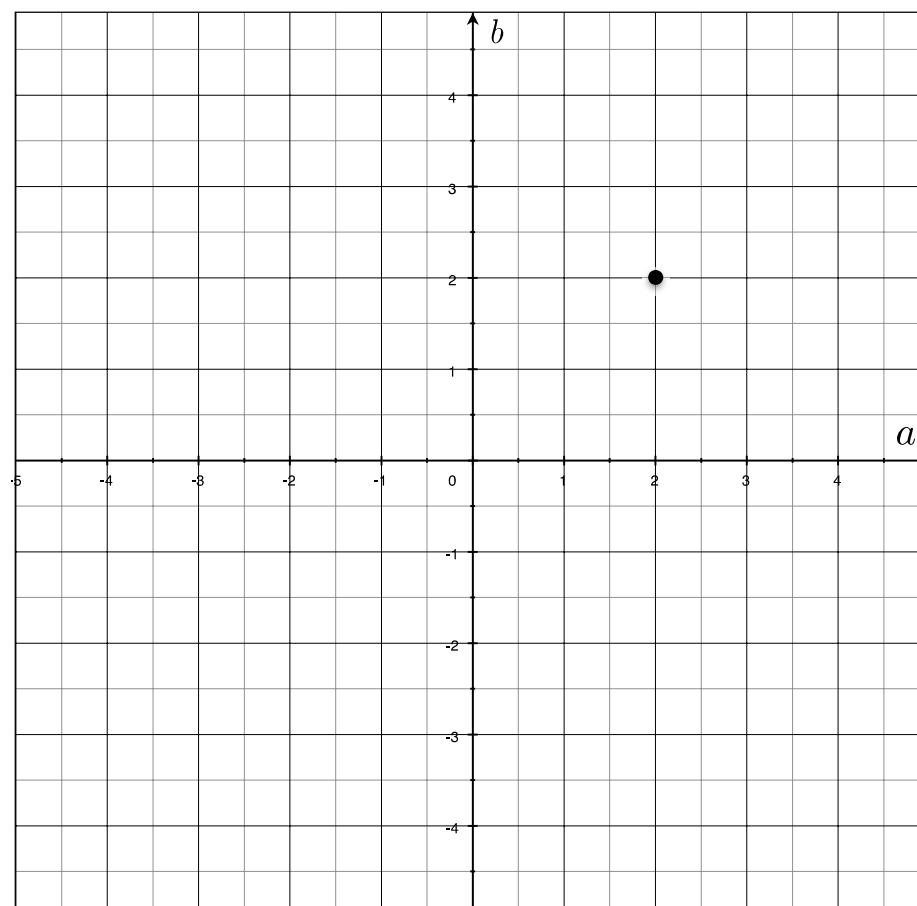
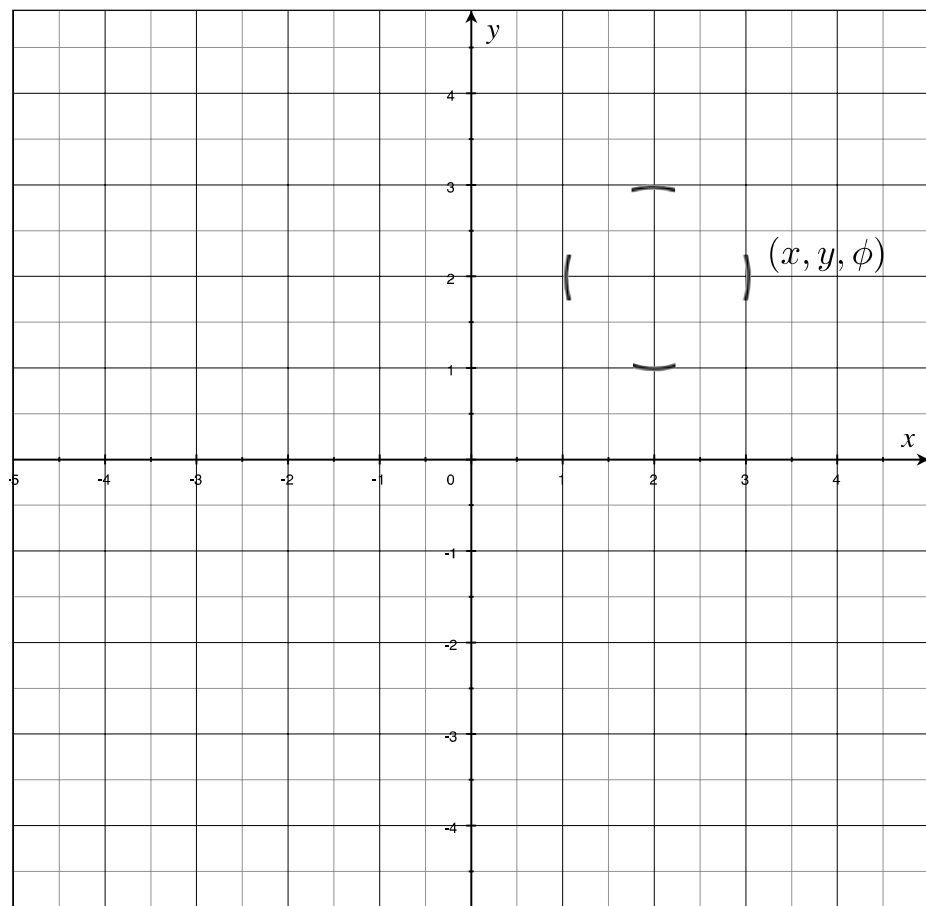
$$(x - a)^2 + (y - b)^2 = r^2$$

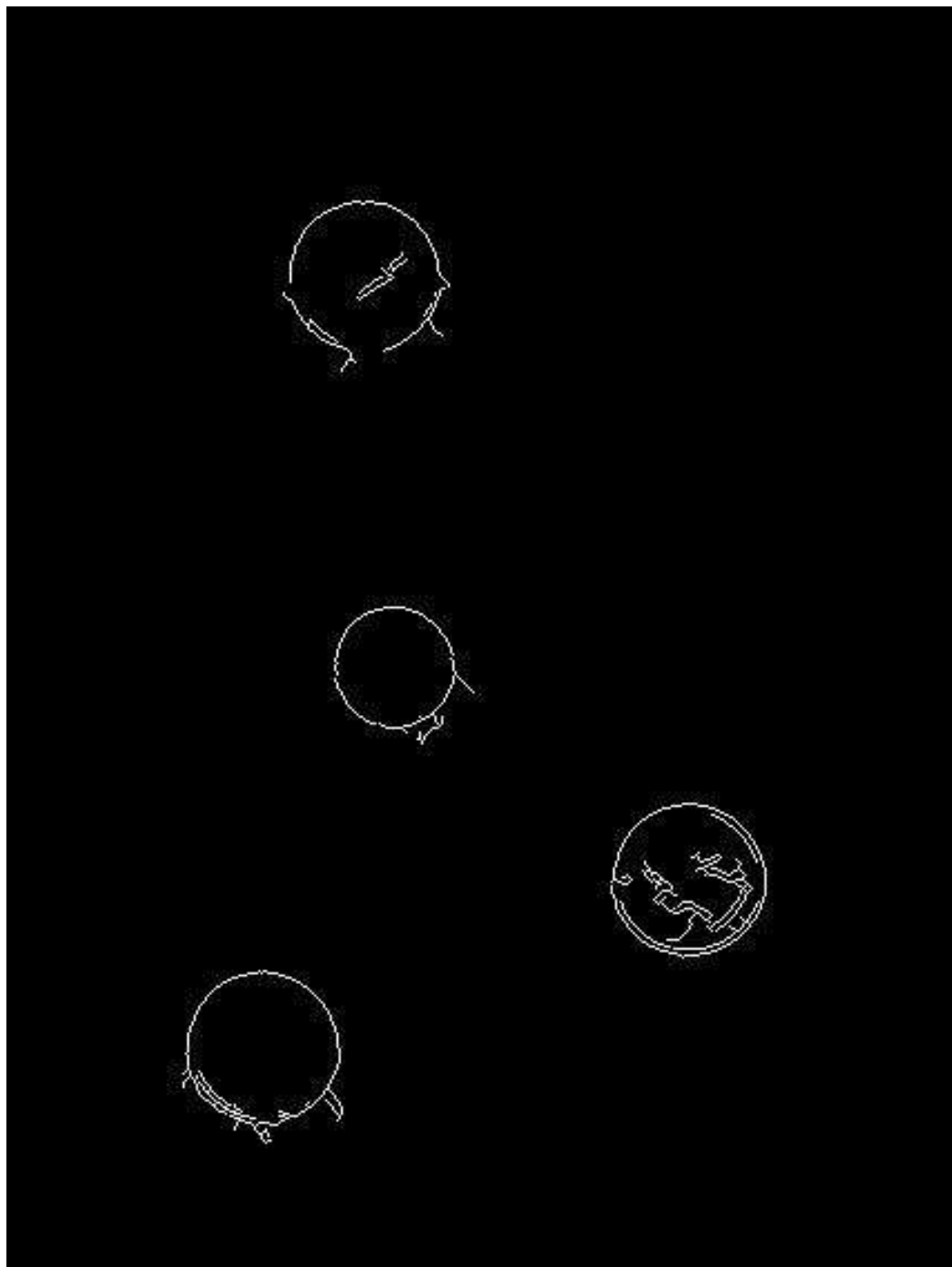
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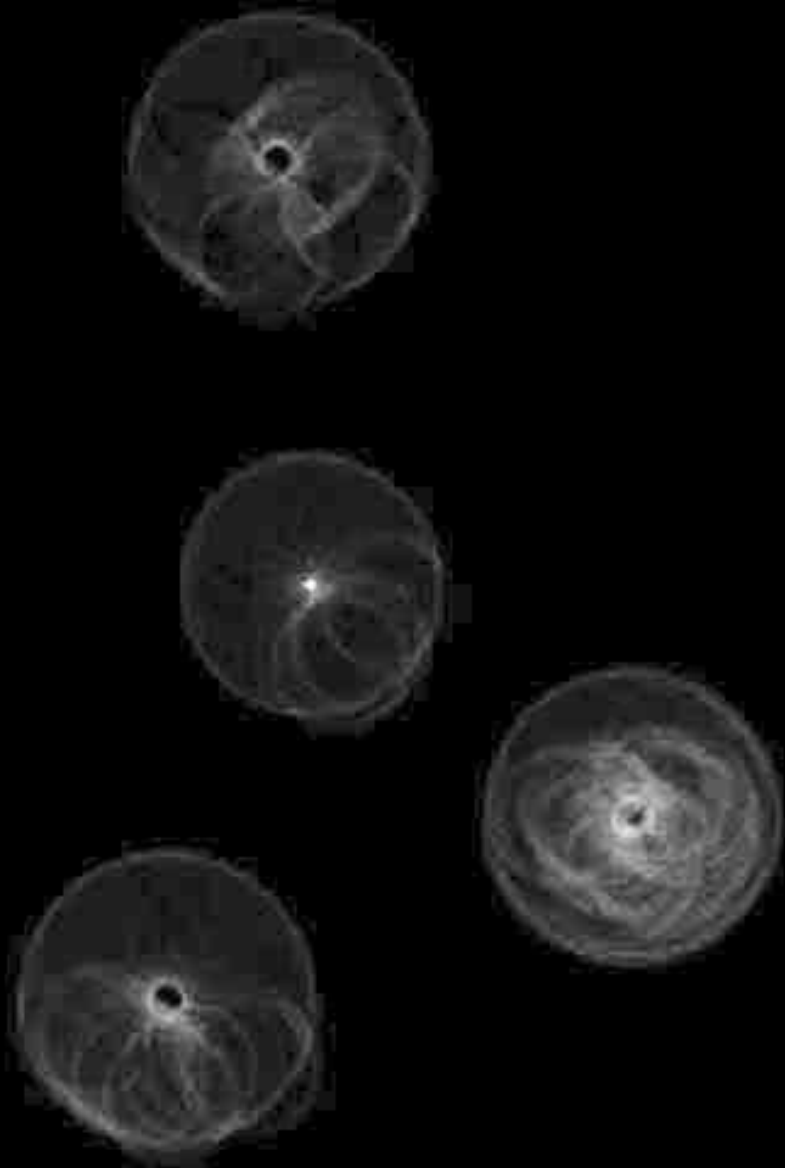
Pennie Hough detector



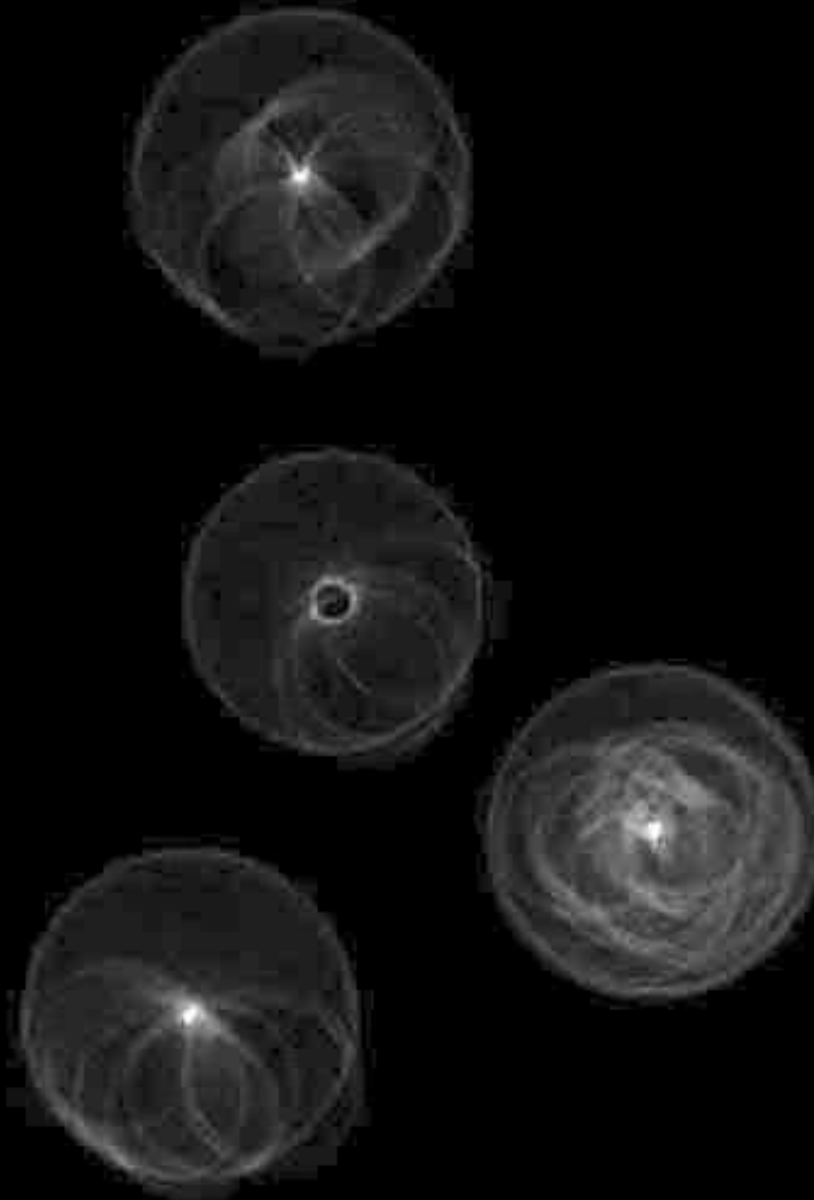
Quarter Hough detector



Pennie Hough detector



Quarter Hough detector

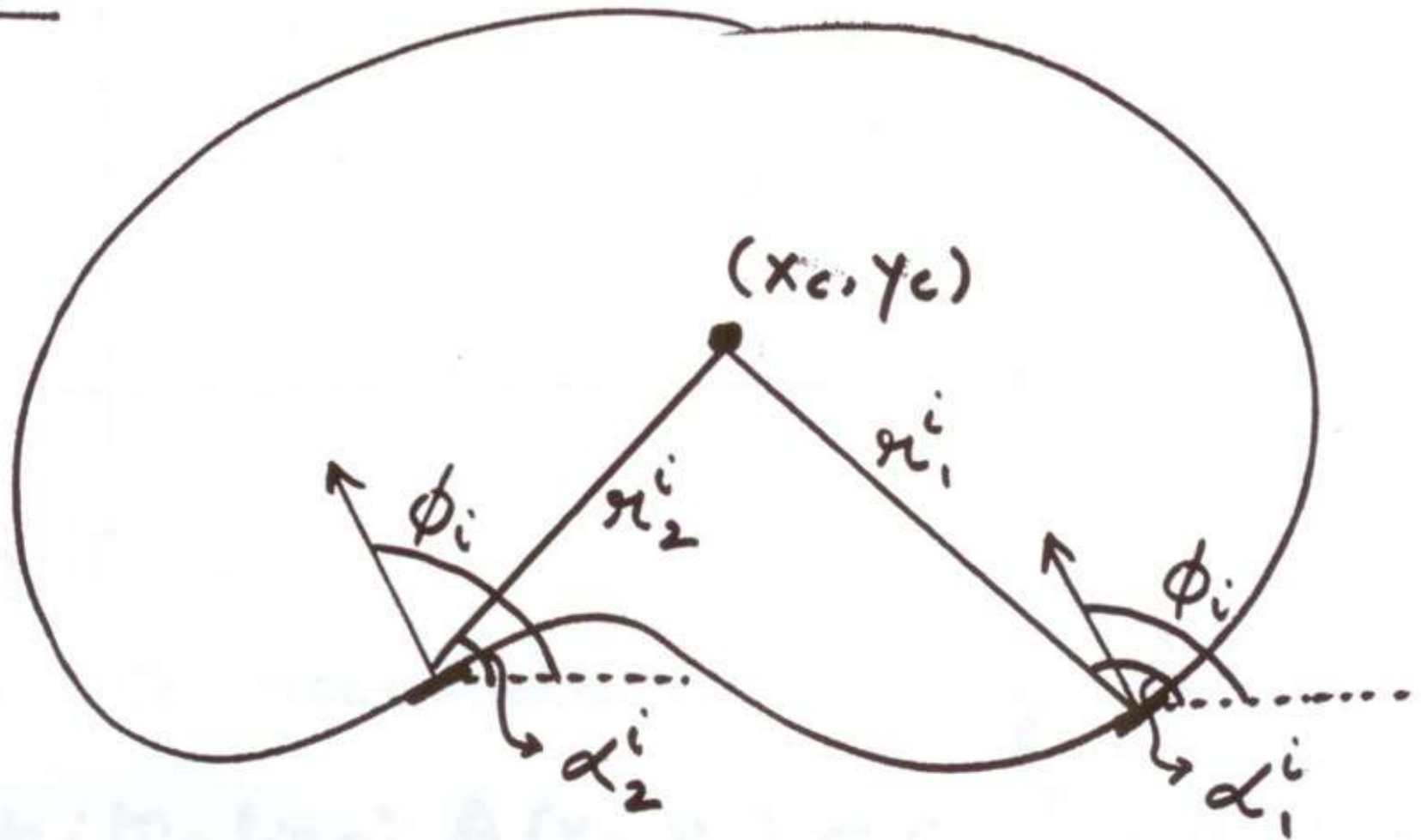




Can you use Hough Transforms for other objects,  
beyond lines and circles?

# Generalized Hough Transform

Model:



## $\phi$ -Table

Edge Direction	$\bar{\pi} = (\pi, \alpha)$
$\phi_1$	$\bar{\pi}'_1, \bar{\pi}'_2, \bar{\pi}'_3$
$\phi_2$	$\bar{\pi}^2_1, \bar{\pi}^2_2$
$\phi_i$	$\bar{\pi}^i_1, \bar{\pi}^i_2$
$\phi_n$	$\bar{\pi}^n_1, \bar{\pi}^n_2$

# Generalized Hough Transform

Find Object Center  $(x_c, y_c)$  given edges  $(x_i, y_i, \phi_i)$

Create Accumulator Array  $A(x_c, y_c)$

Initialize:  $A(x_c, y_c) = 0 \quad \forall (x_c, y_c)$

For each edge point  $(x_i, y_i, \phi_i)$

For each entry  $\bar{r}_k^i$  in table, compute:

$$x_c = x_i + r_k^i \cos \alpha_k^i$$

$$y_c = y_i + r_k^i \sin \alpha_k^i$$

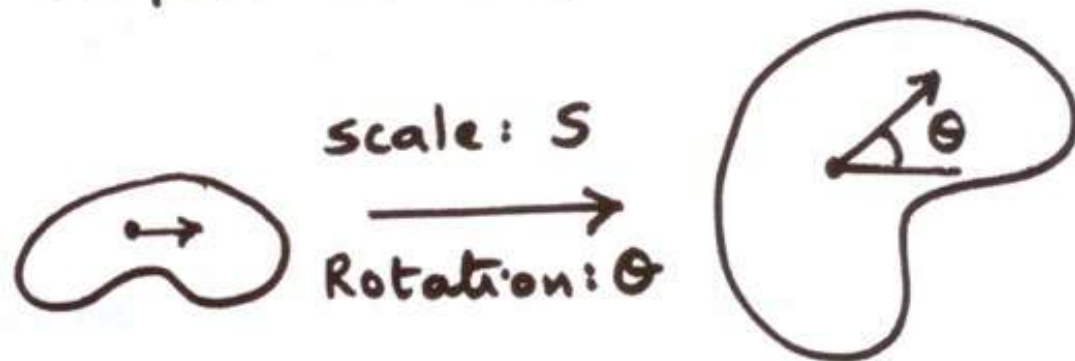
Increment Accumulator:  $A(x_c, y_c) = A(x_c, y_c) + 1$

Find Local Maxima in  $A(x_c, y_c)$

## Scale & Rotation:

Use Accumulator Array:

$$A[x_c, y_c, S, \theta]$$



Use:

$$x_c = x_i + r_k^i S \cos(\alpha_k^i + \theta)$$

$$y_c = y_i + r_k^i S \sin(\alpha_k^i + \theta)$$

$$A(x_c, y_c, S, \theta) = A(x_c, y_c, S, \theta) + 1.$$

Do you have to use edge detectors  
to vote in Hough Space?

## A. Train phase:

1. Get features

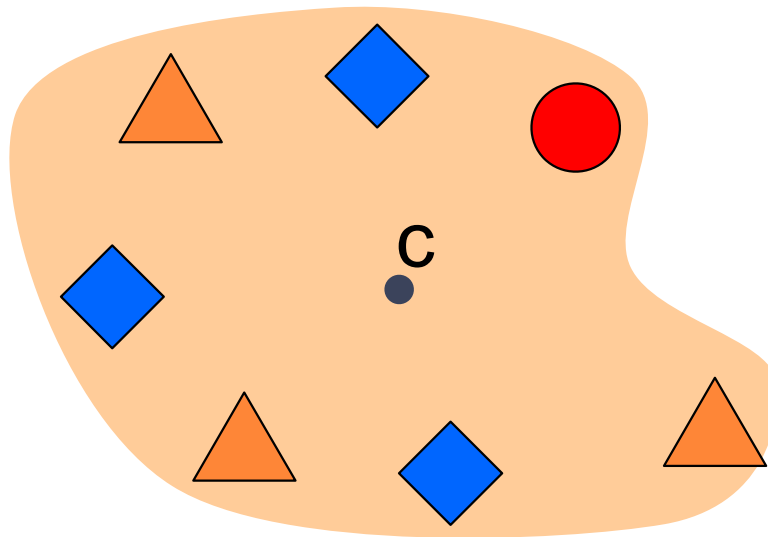
2. Store all displacements of feature from center

## B. Test phase:

1. Get features & lookup displacements

2. Vote for center location

### Template



## A. Train phase:

1. Get features

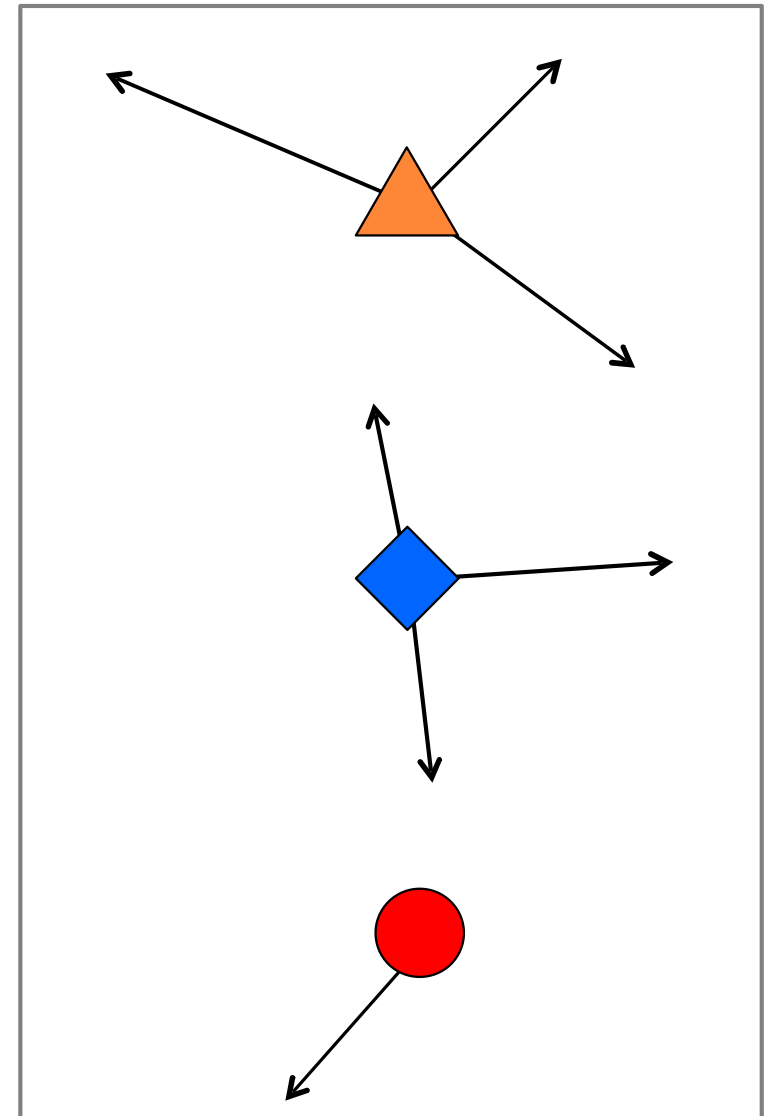
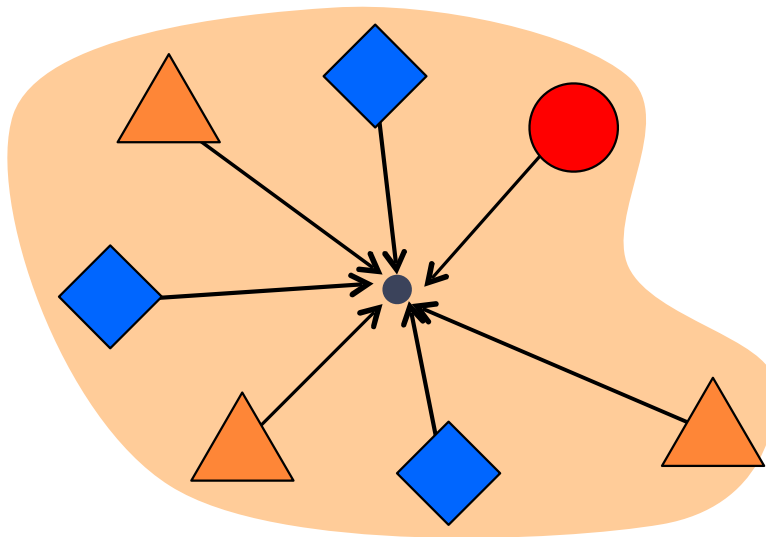
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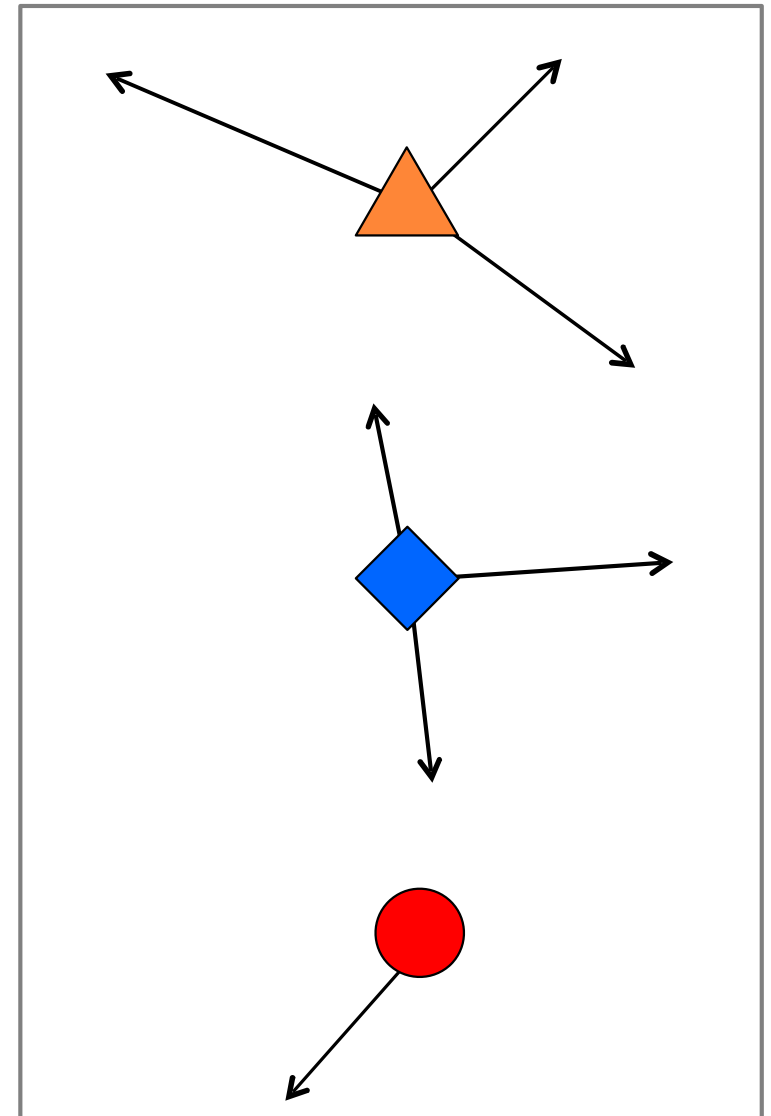
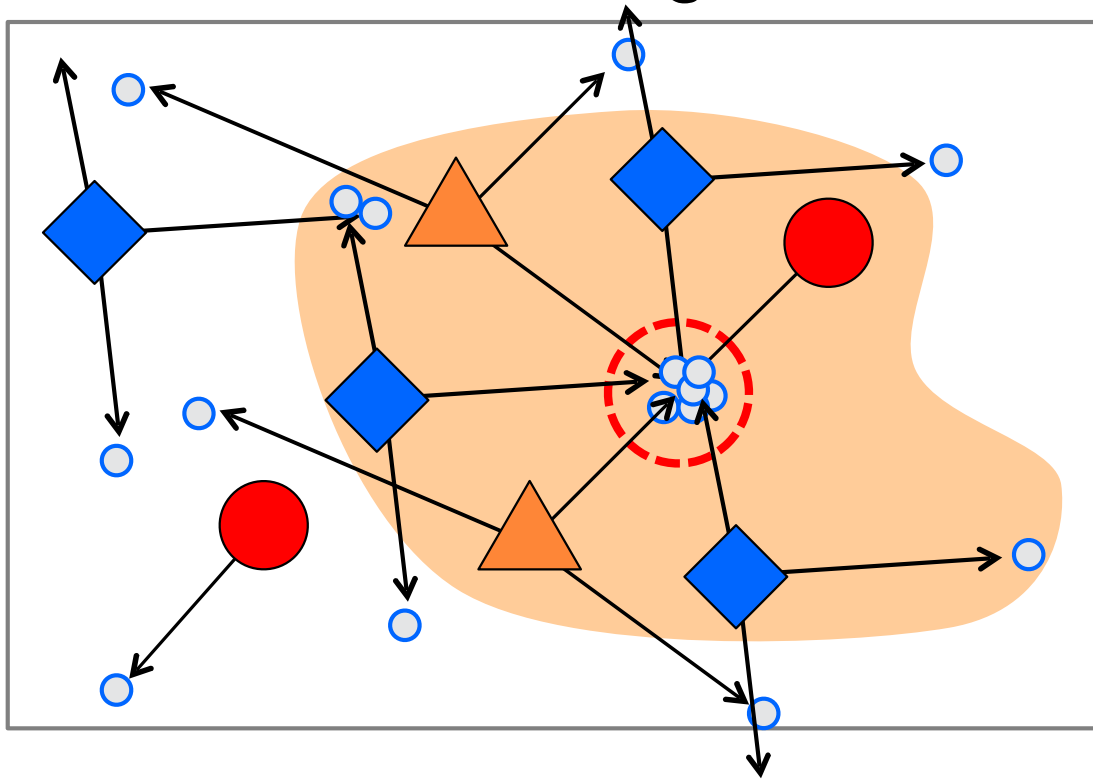
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## B. Test phase:

1. Get features & lookup displacements

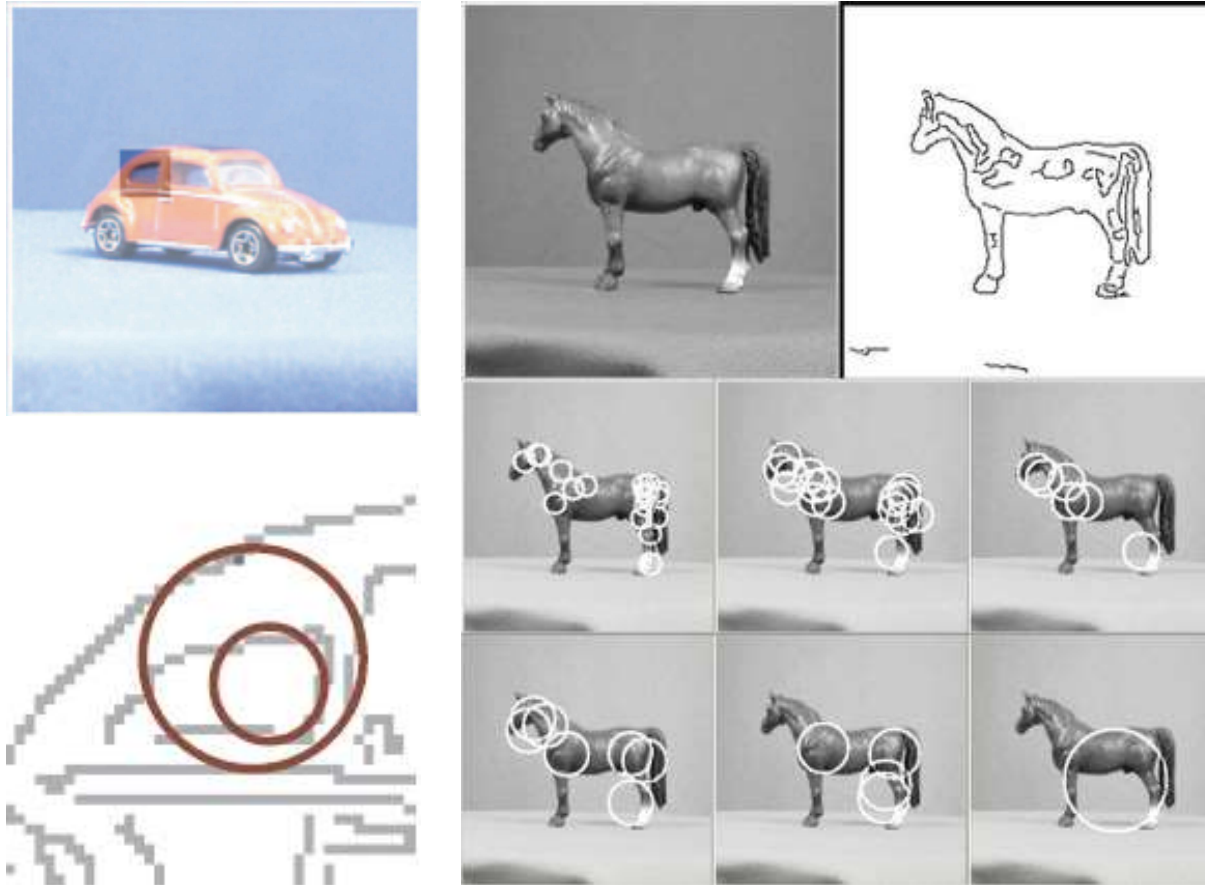
2. Vote for center location

Test image



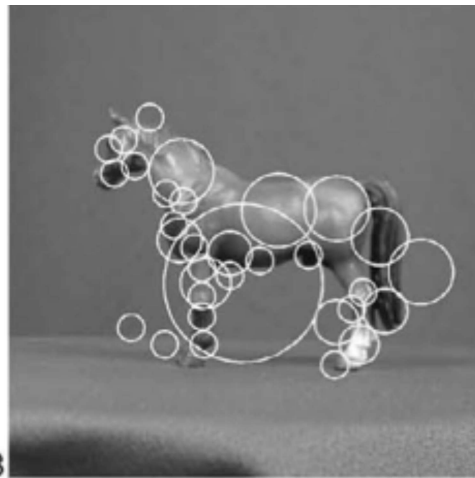
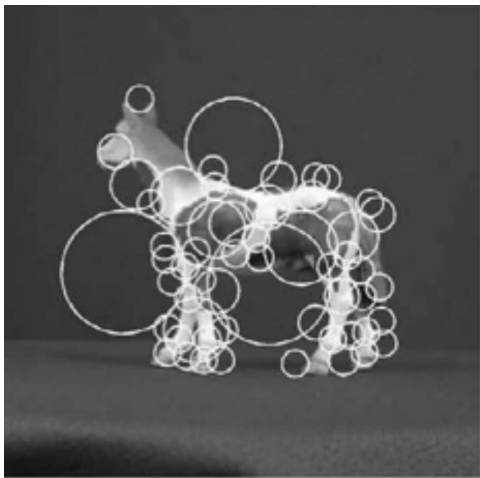
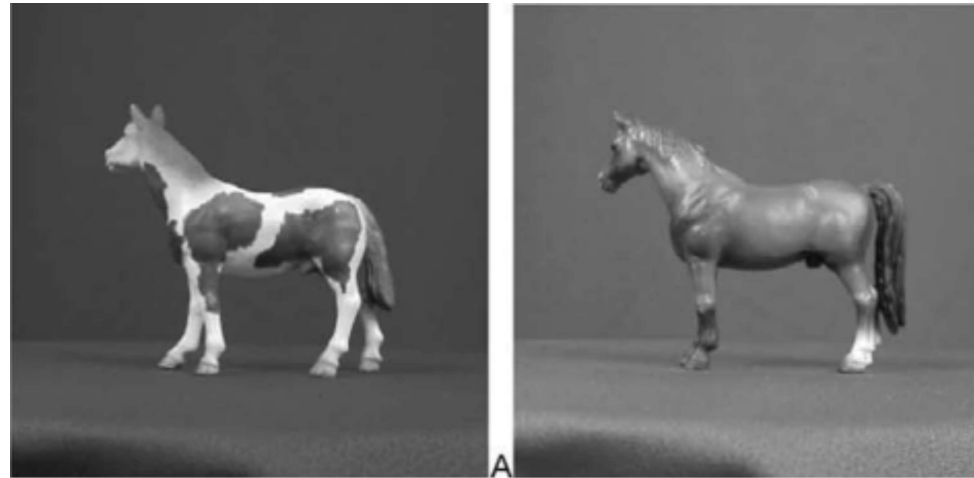
# Application of Hough Transforms

# Detecting shape features

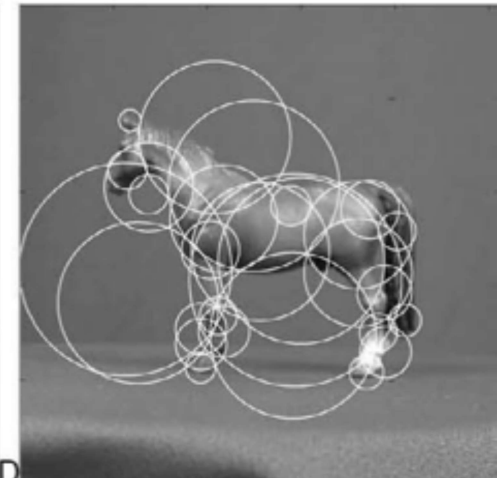
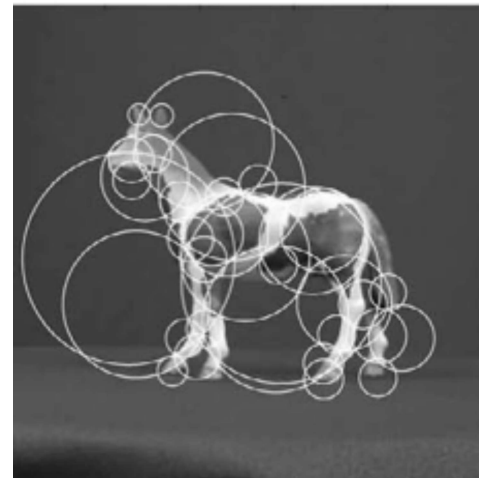


F. Jurie and C. Schmid, Scale-invariant shape features for recognition of object categories, CVPR 2004

Original  
images

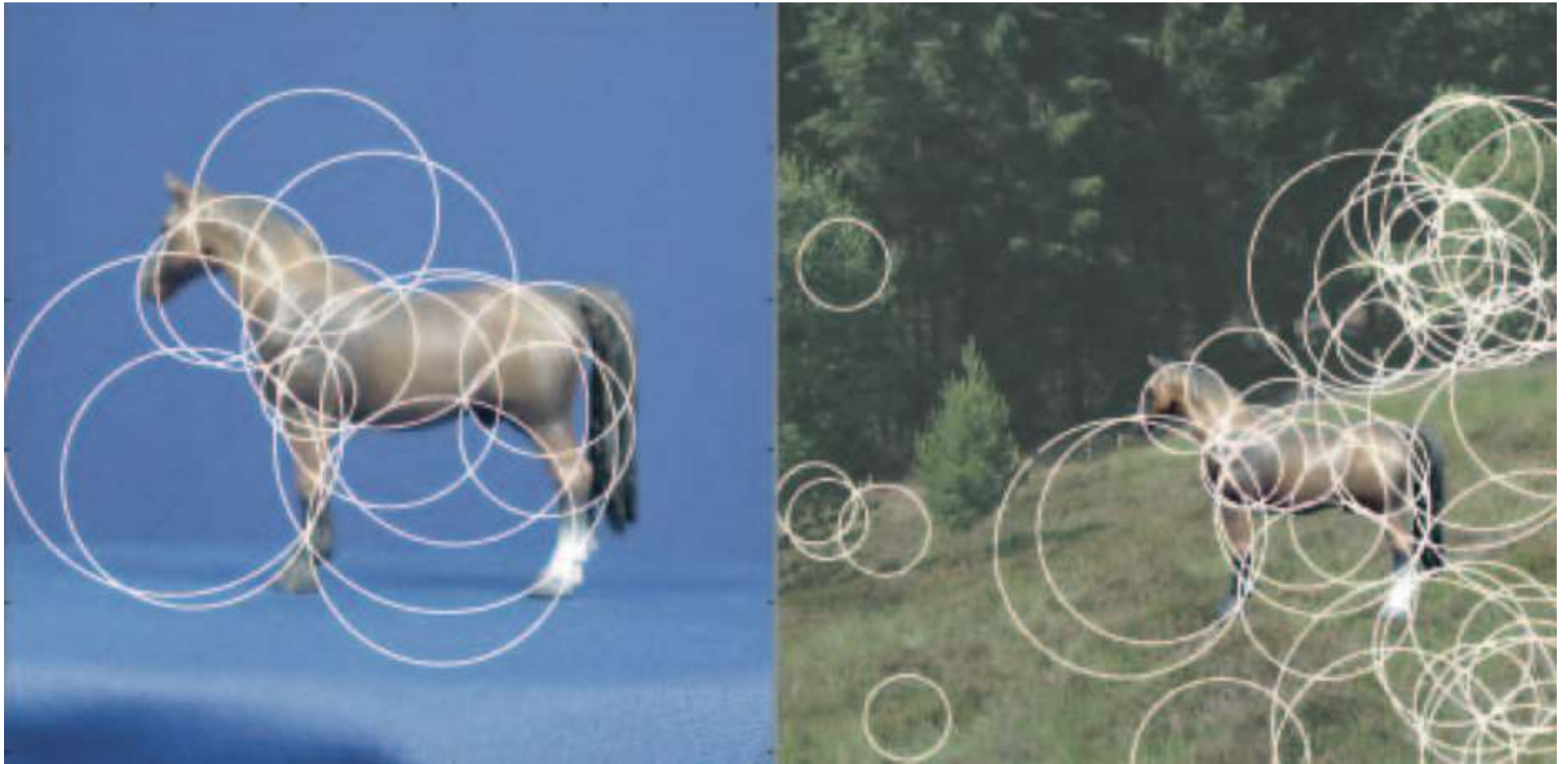


Laplacian circles



Hough-like circles

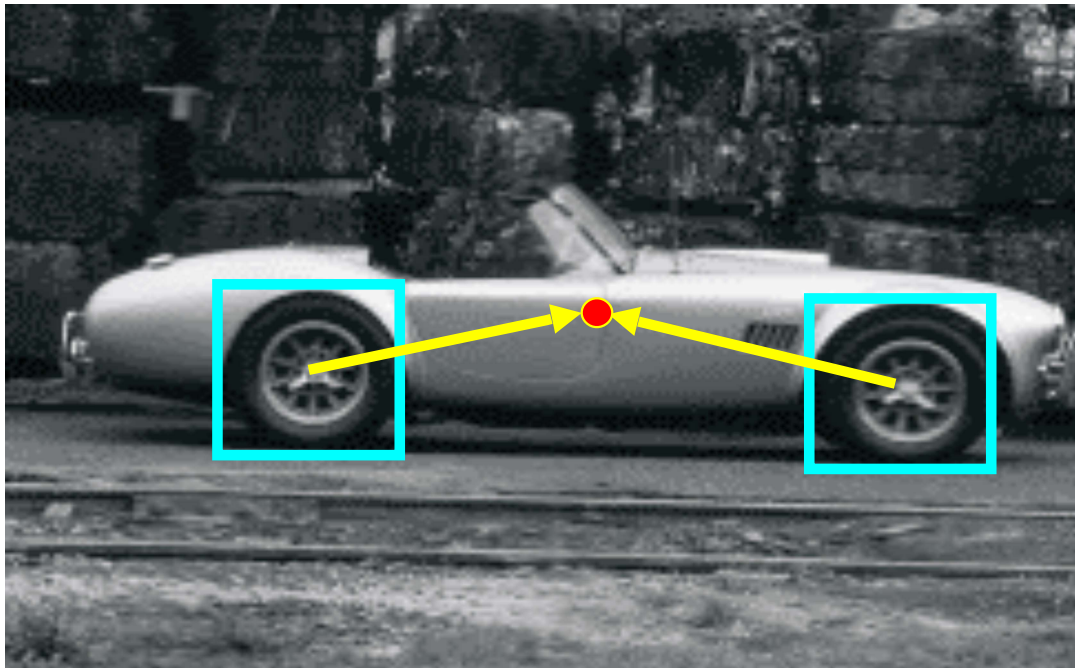
*Which feature detector is more consistent?*



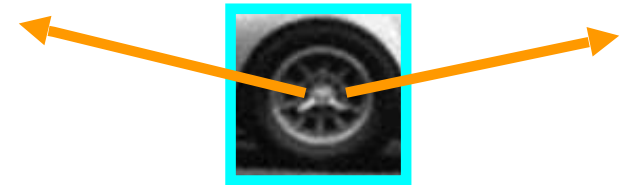
Robustness to scale and clutter

# Object detection

Index displacements by “visual codeword”



training image



visual codeword with  
displacement vectors

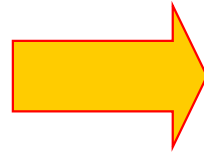
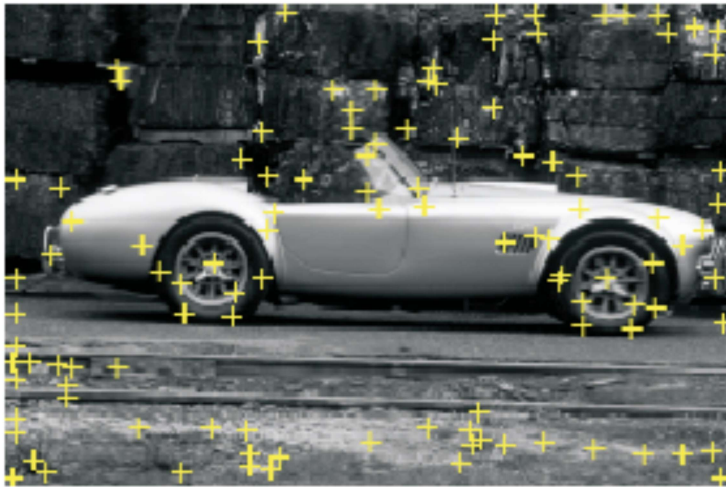
B. Leibe, A. Leonardis, and B. Schiele, Combined Object Categorization and Segmentation with an Implicit Shape Model,  
ECCV Workshop on Statistical Learning in Computer Vision 2004





## Train phase

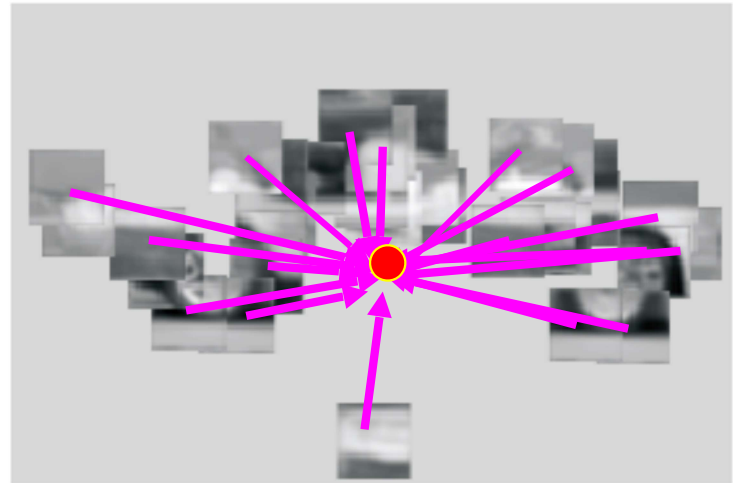
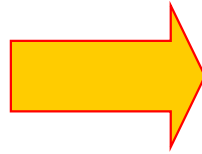
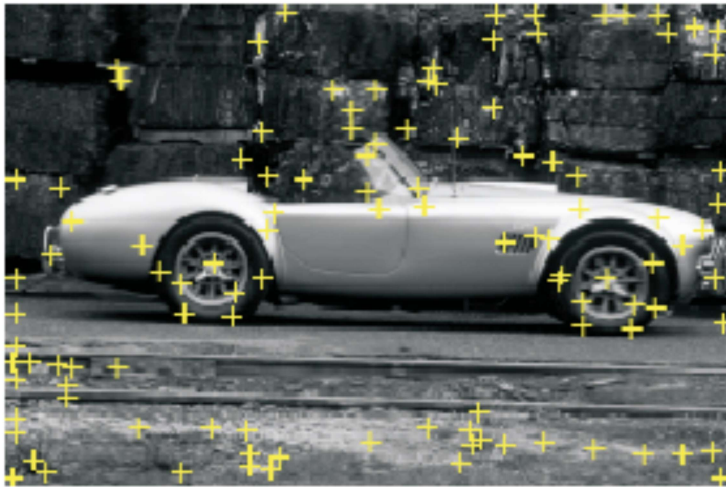
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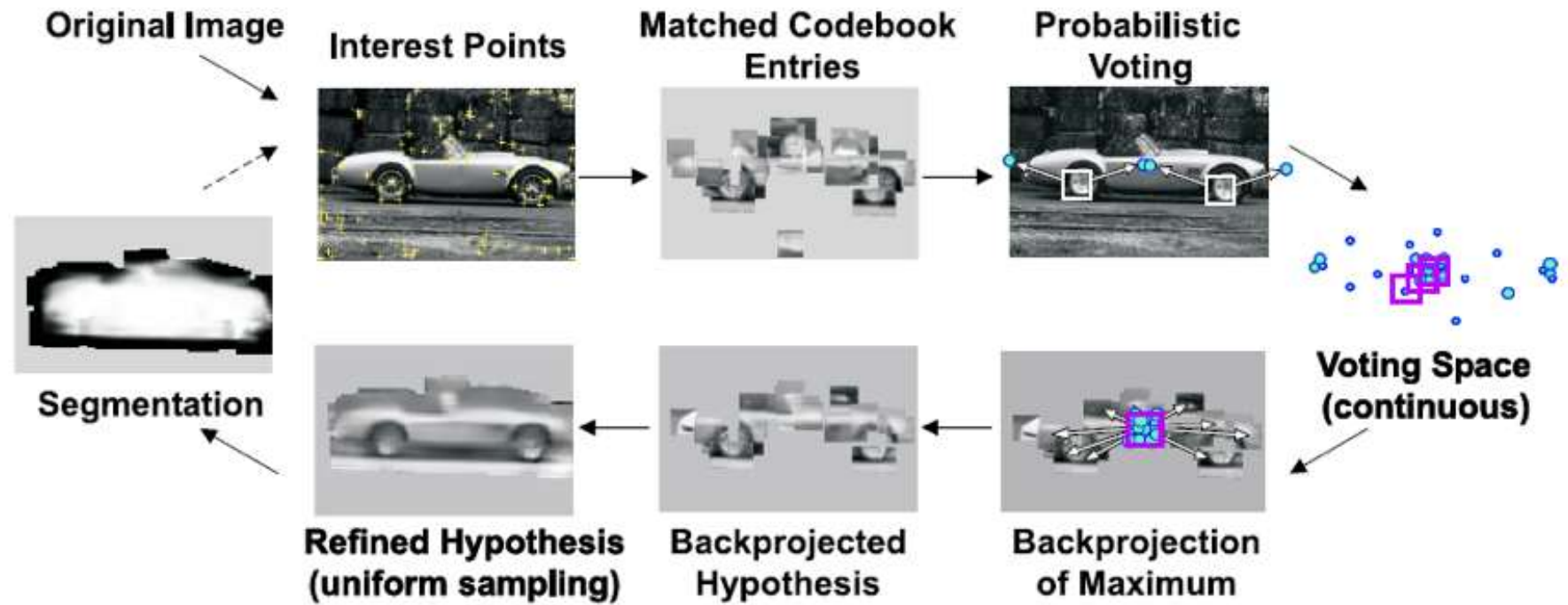


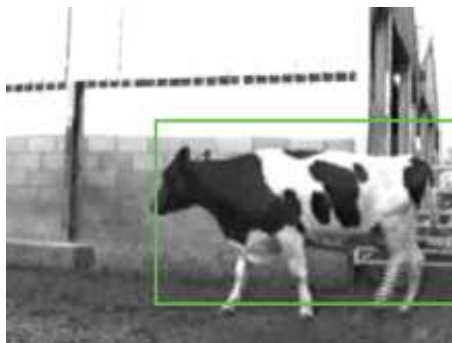
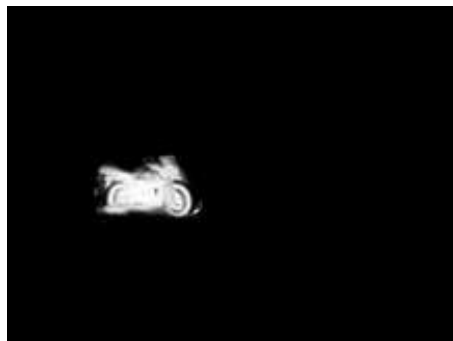
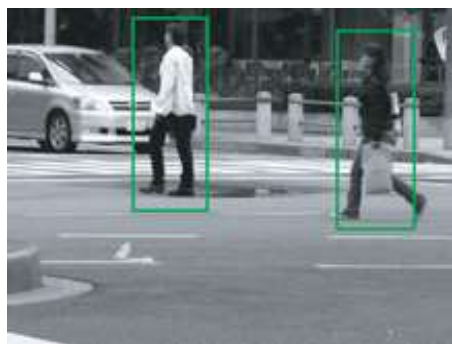
**Train phase**

## 2. store displacements



# Test phase





# The Hough transform ...

Deals with occlusion well?



Detects multiple instances?



Robust to noise?



Good computational complexity?



Easy to set parameters?

