

Hough transform

Hough transform

- The Hough transform (HT) can be used to detect lines.
- It was introduced in 1962 (Hough 1962) and first used to find lines in images a decade later (Duda 1972).
- The goal is to find the location of lines in images.
- **Caveat:** Hough transform can detect lines, circles and other structures **ONLY** if their parametric equation is known.
- It can give robust detection under noise and partial occlusion

Prior to Hough Transform

- Assume that we have performed some edge detection, and a thresholding of the edge magnitude image.
- Thus, we have some pixels that may partially describe the boundary of some objects.



Hough transform

- Generic framework for detecting a parametric model
- Edges don't have to be connected
- Lines can be occluded
- Key idea: edges **vote** for the possible models

Image and parameter space

variables

$$y = mx + b$$

parameters

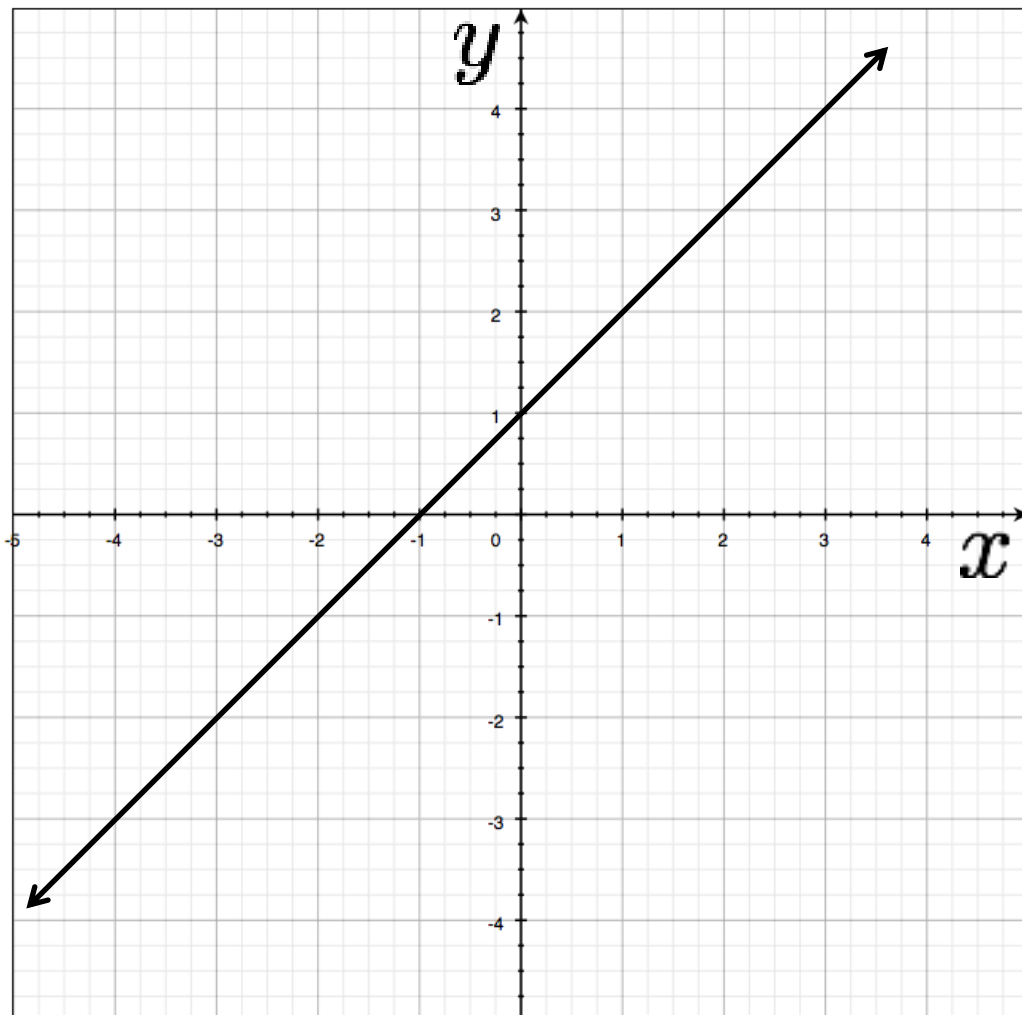


Image space

Image and parameter space

variables

$$y = mx + b$$

parameters

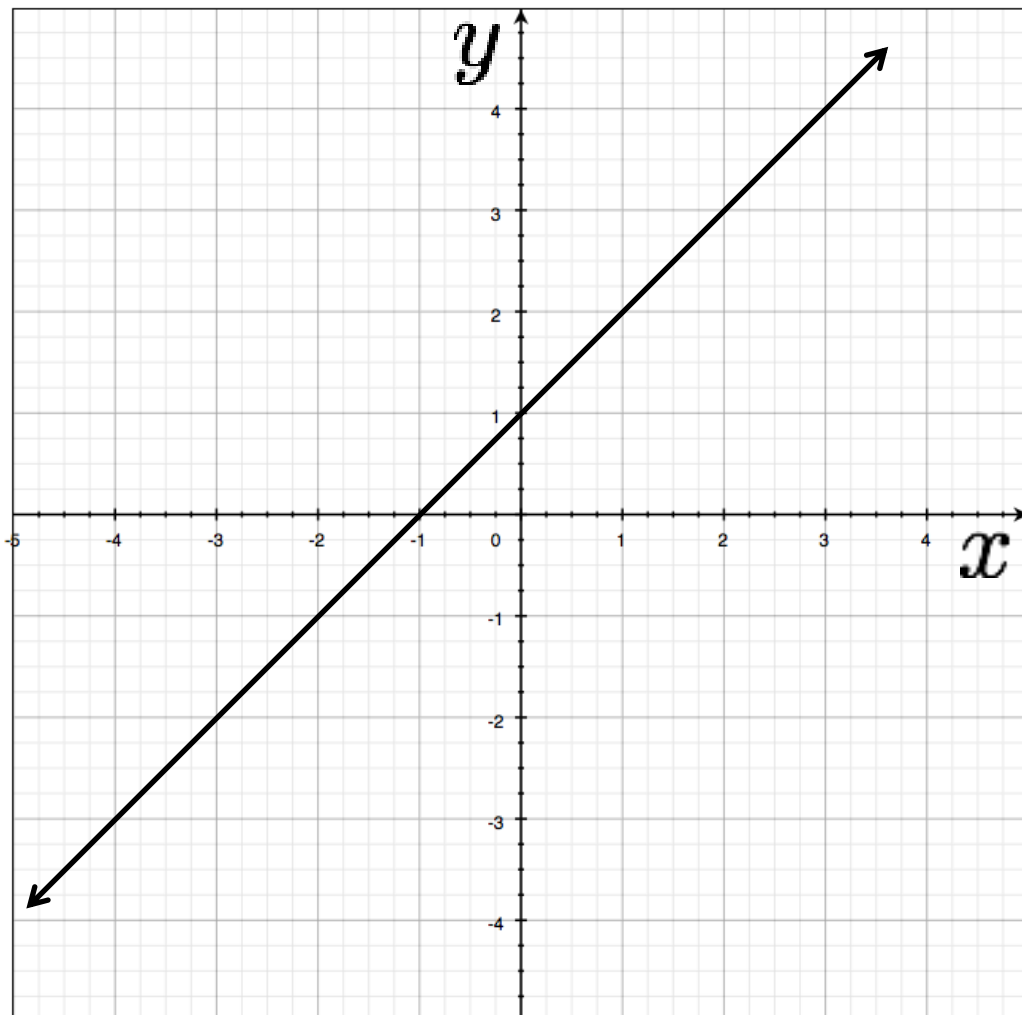
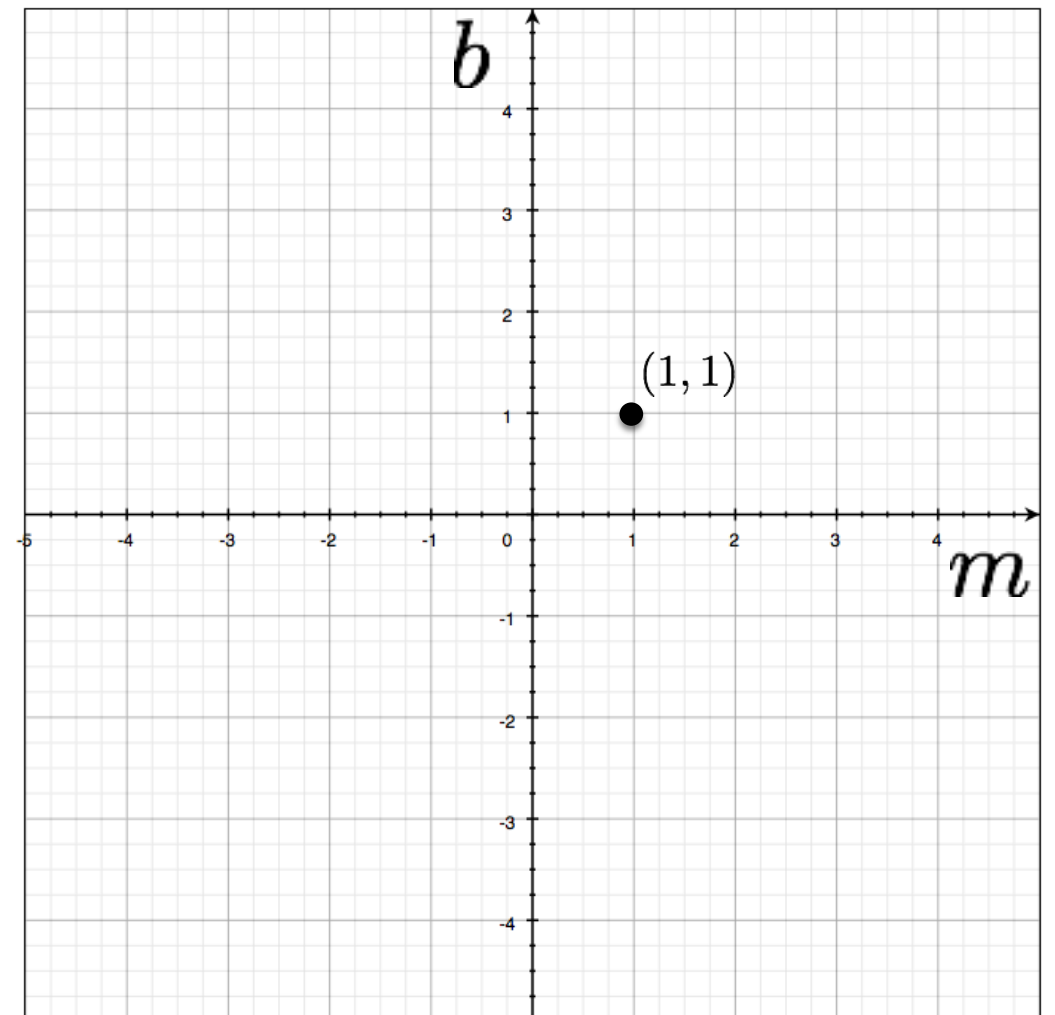


Image space

variables

$$y - mx = b$$

parameters



Parameter space

a line
becomes a
point

Image and parameter space

variables

$$y = mx + b$$

parameters

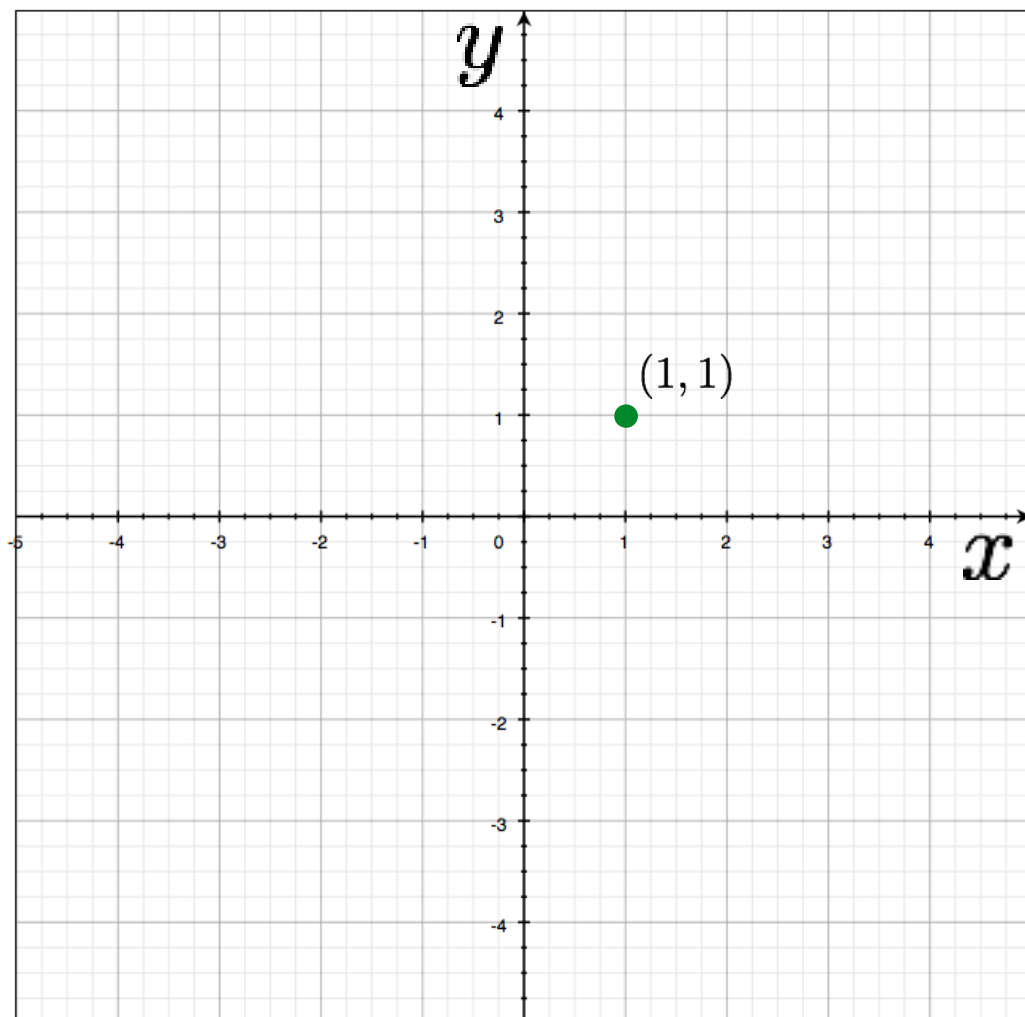


Image space

*What would a point in image space
become in parameter space?*

Image and parameter space

variables

$$y = mx + b$$

parameters

variables

$$y - mx = b$$

parameters

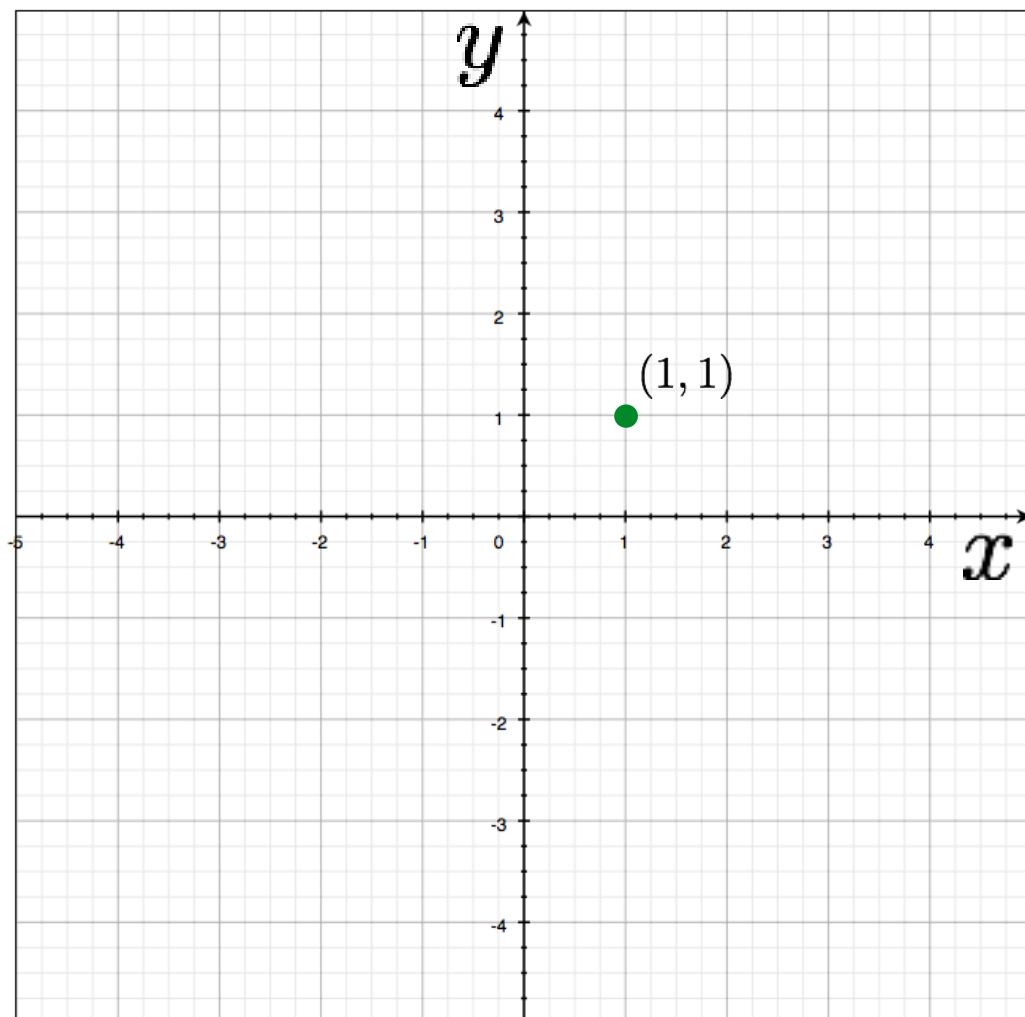
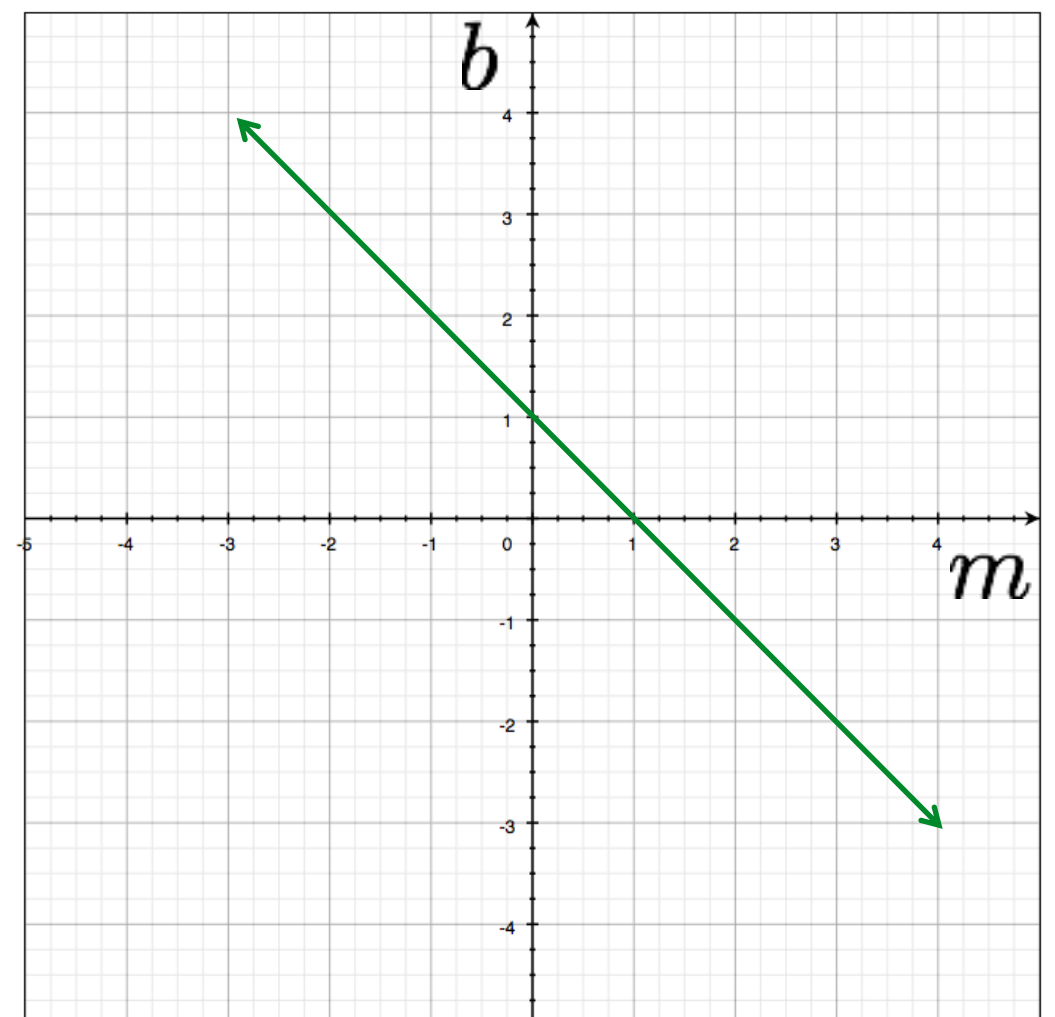


Image space

a point
becomes a
line



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

variables

$$y - mx = b$$

parameters

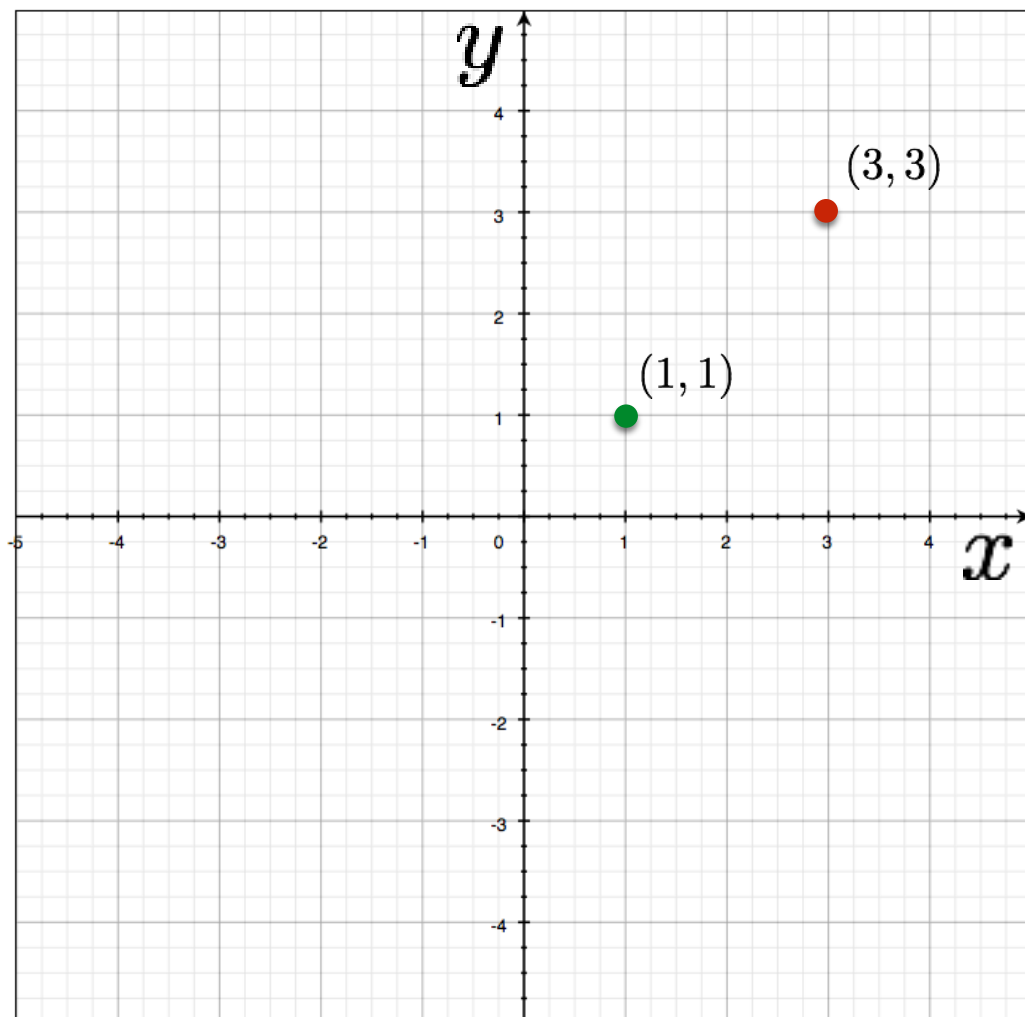
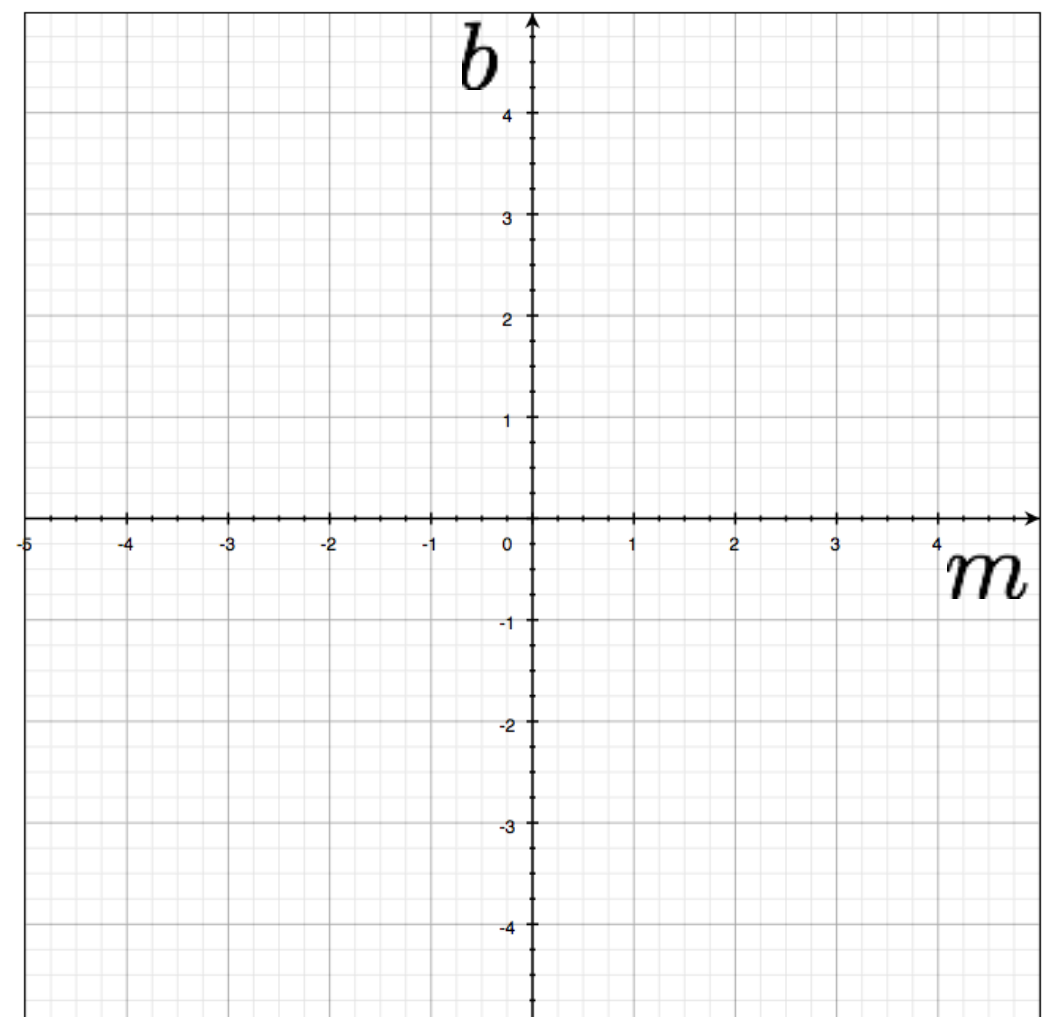


Image space

two points
become
?



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

variables

$$y - mx = b$$

parameters

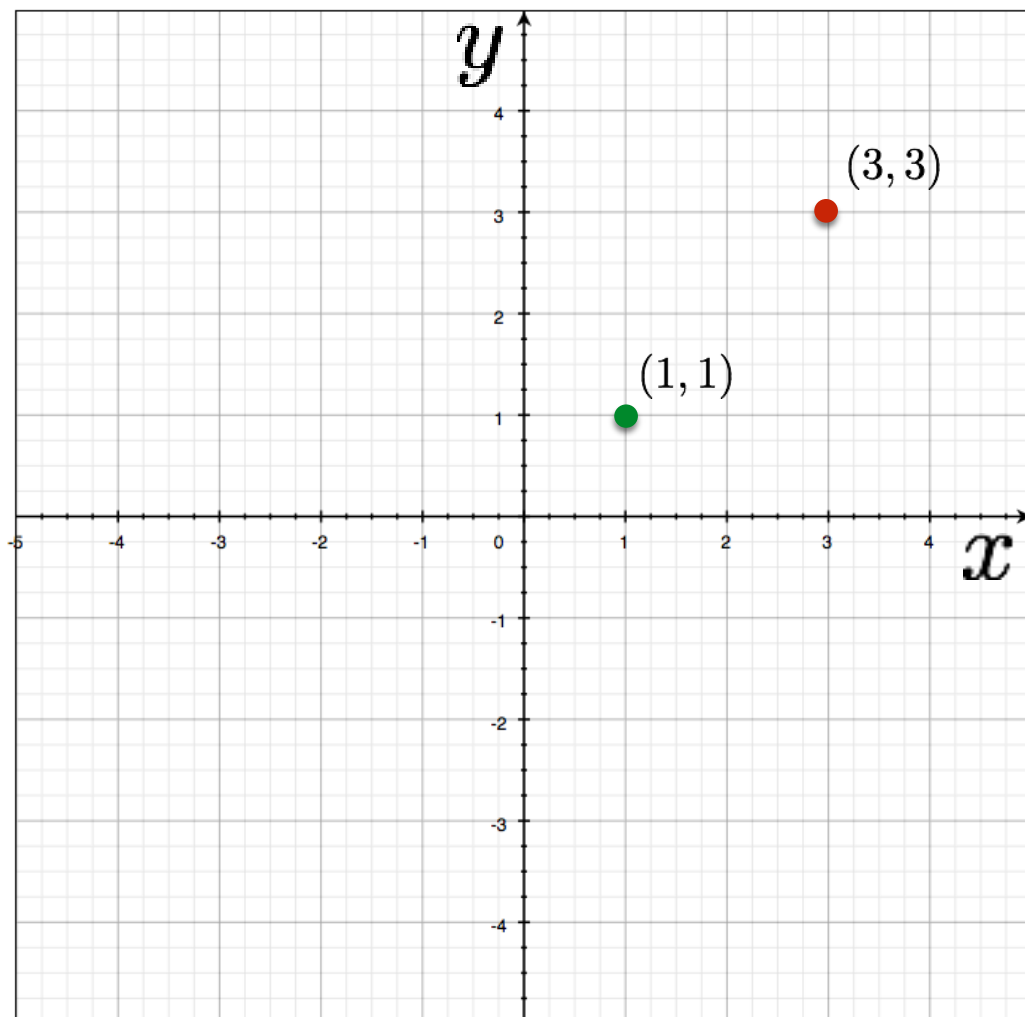
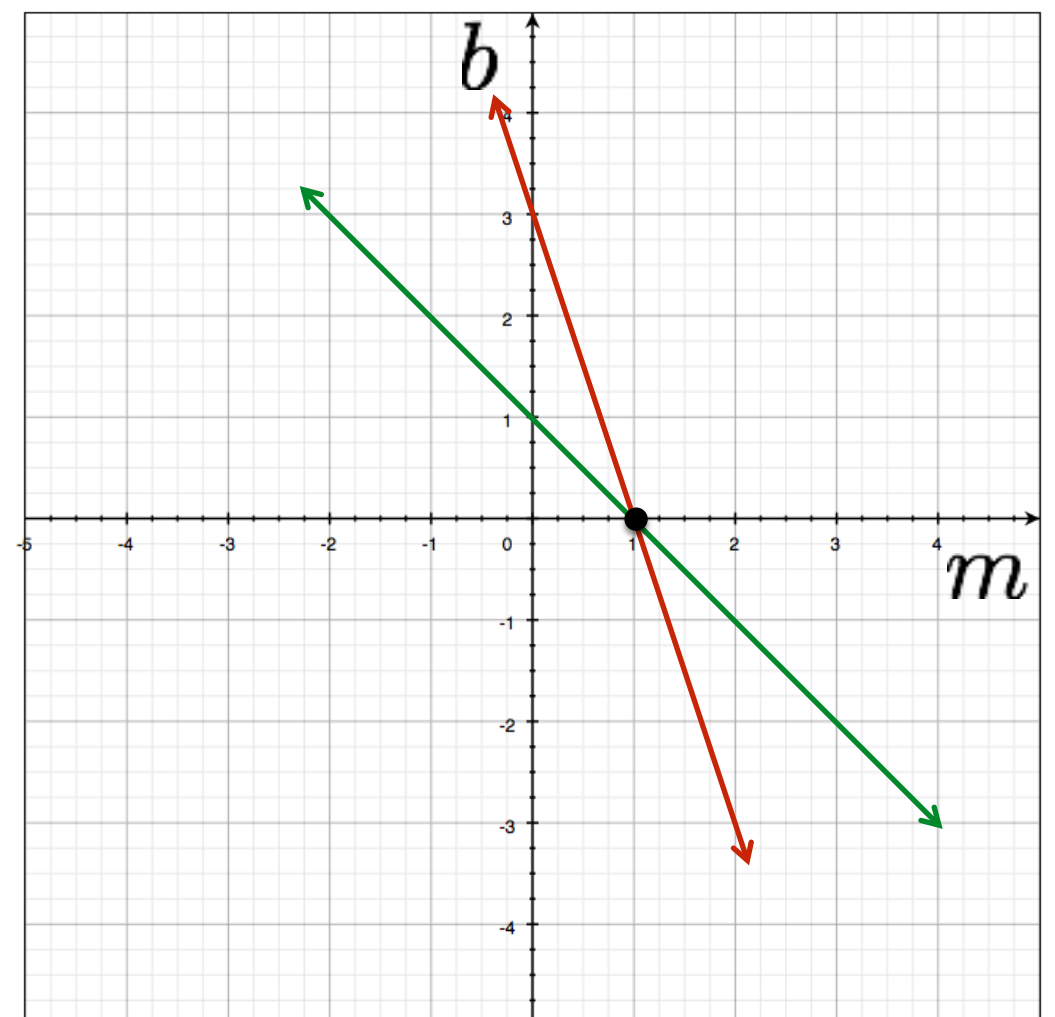


Image space

two points
become
?



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

variables

$$y - mx = b$$

parameters

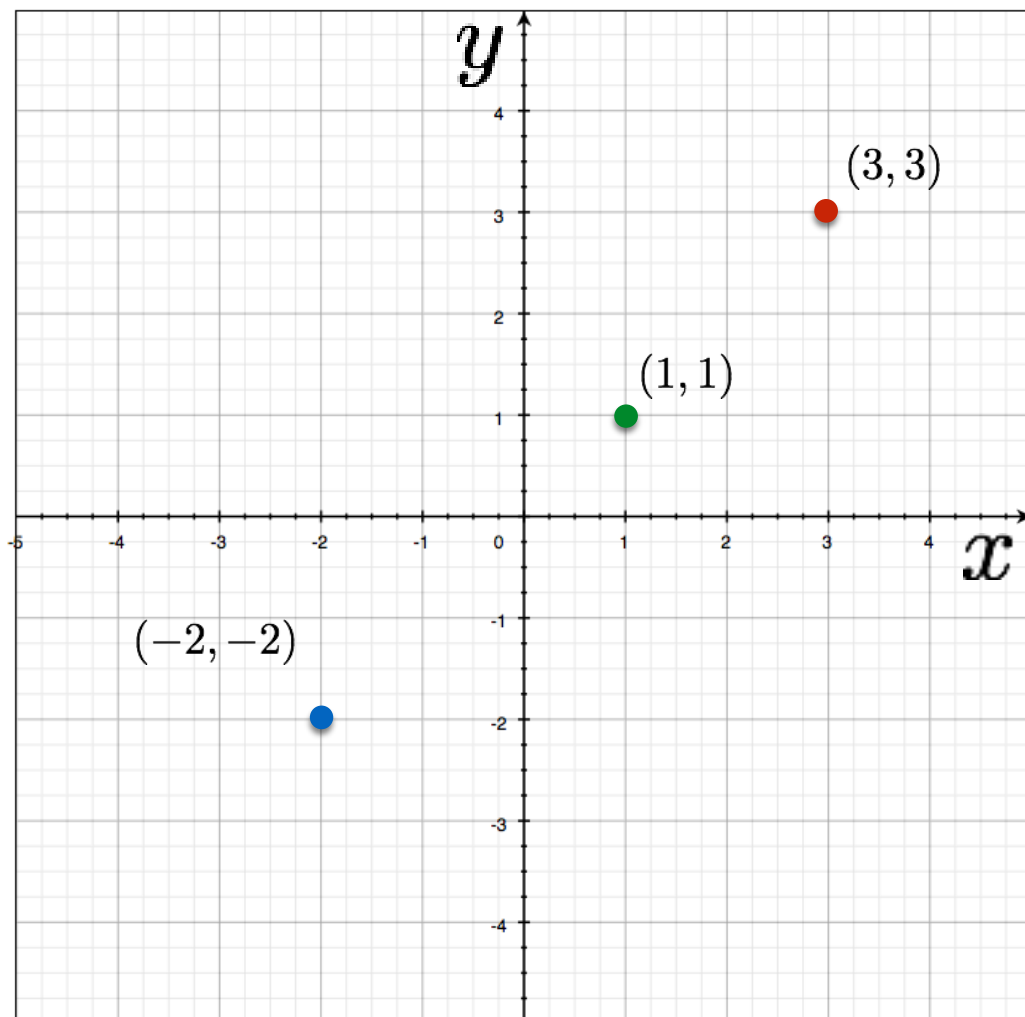
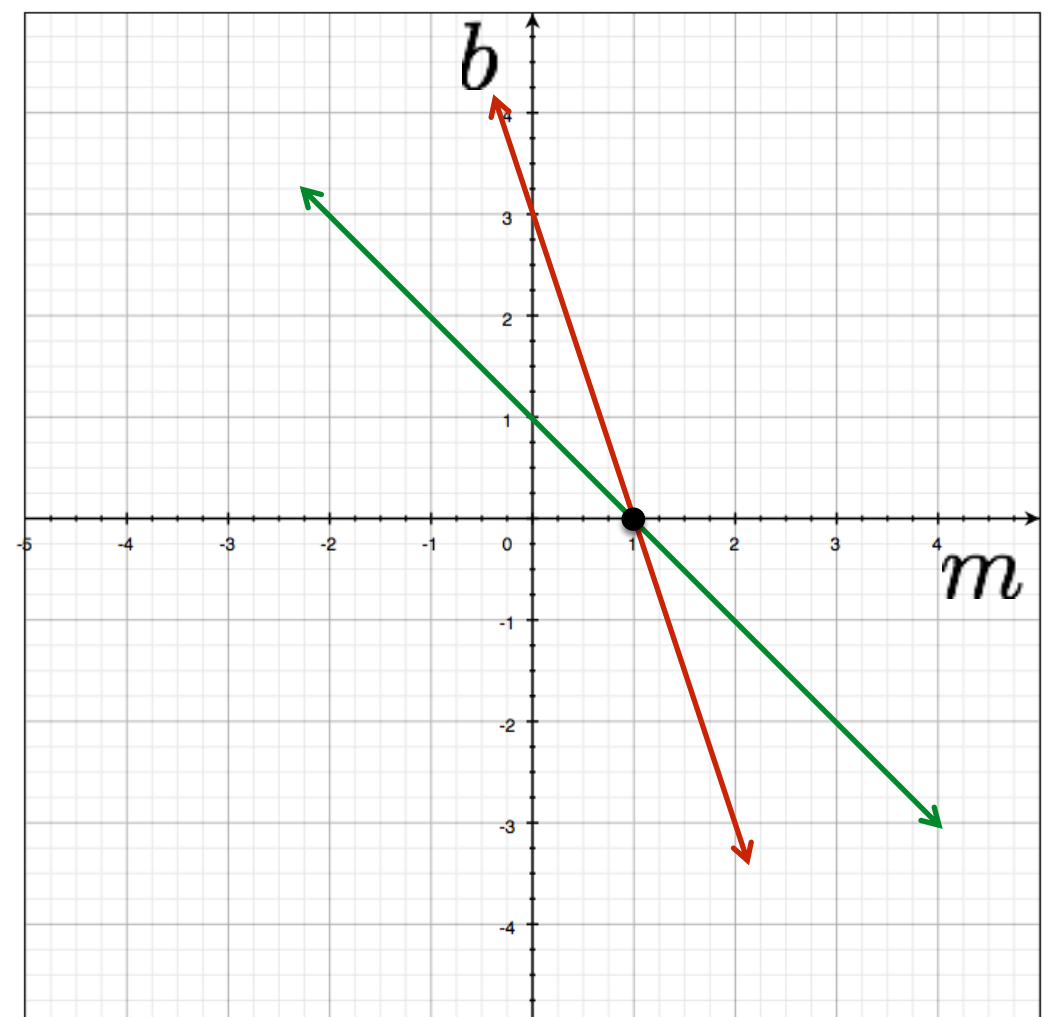


Image space

three points
become
?



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

variables

$$y - mx = b$$

parameters

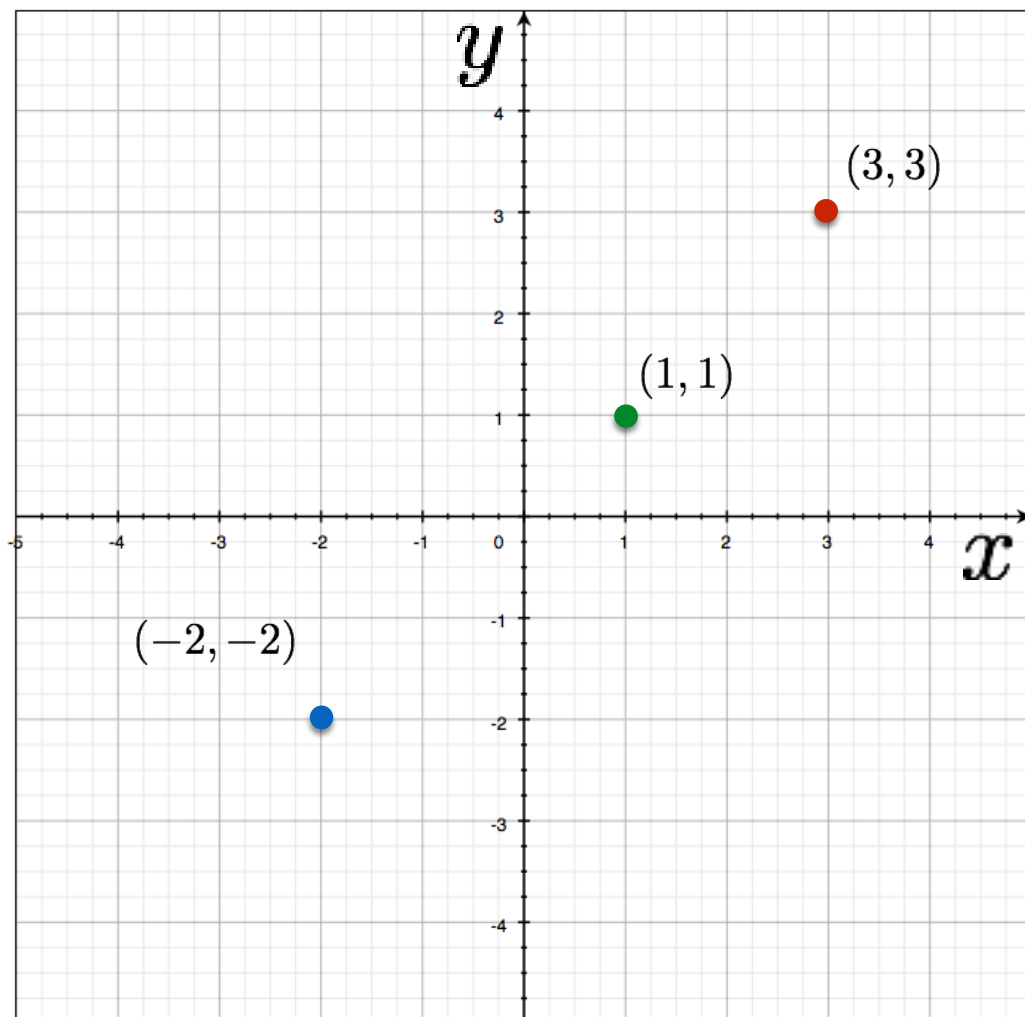
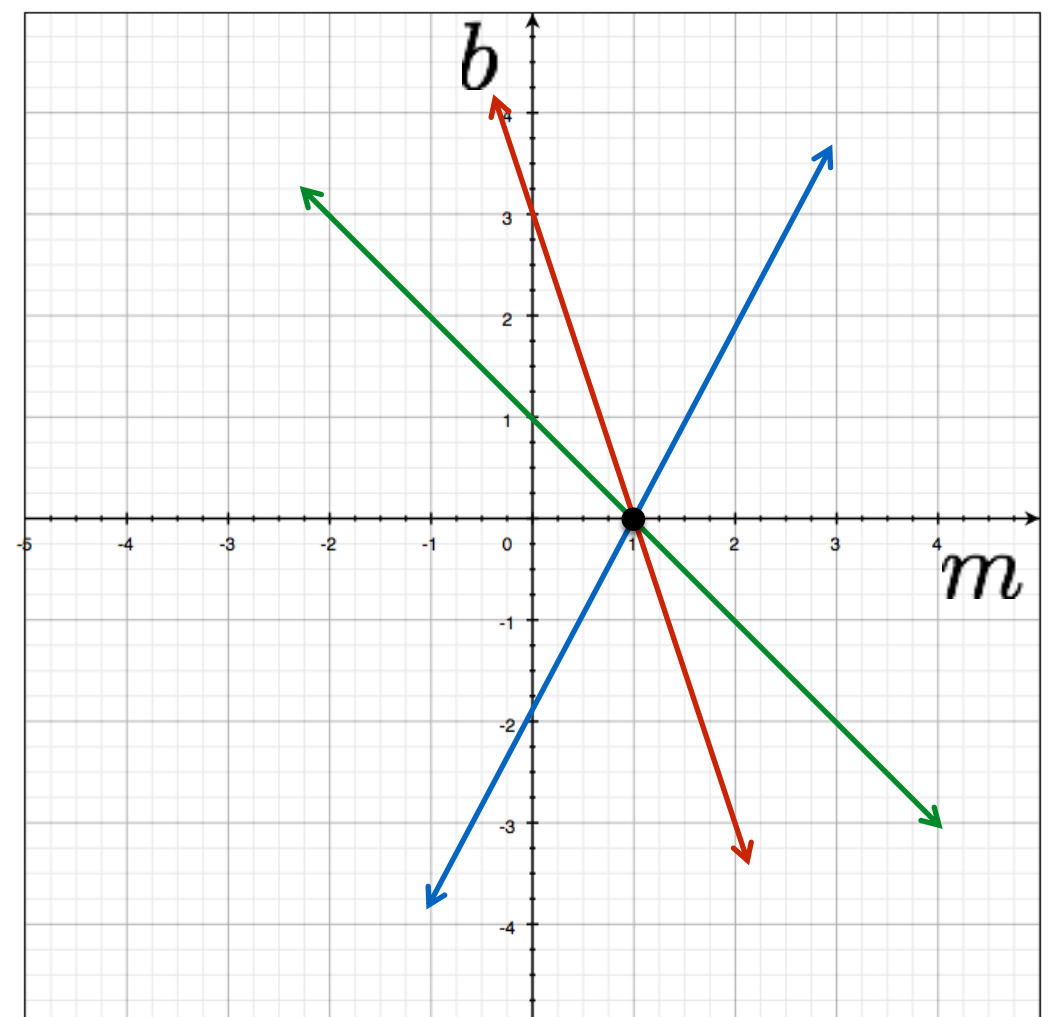


Image space

three points
become
?



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

variables

$$y - mx = b$$

parameters

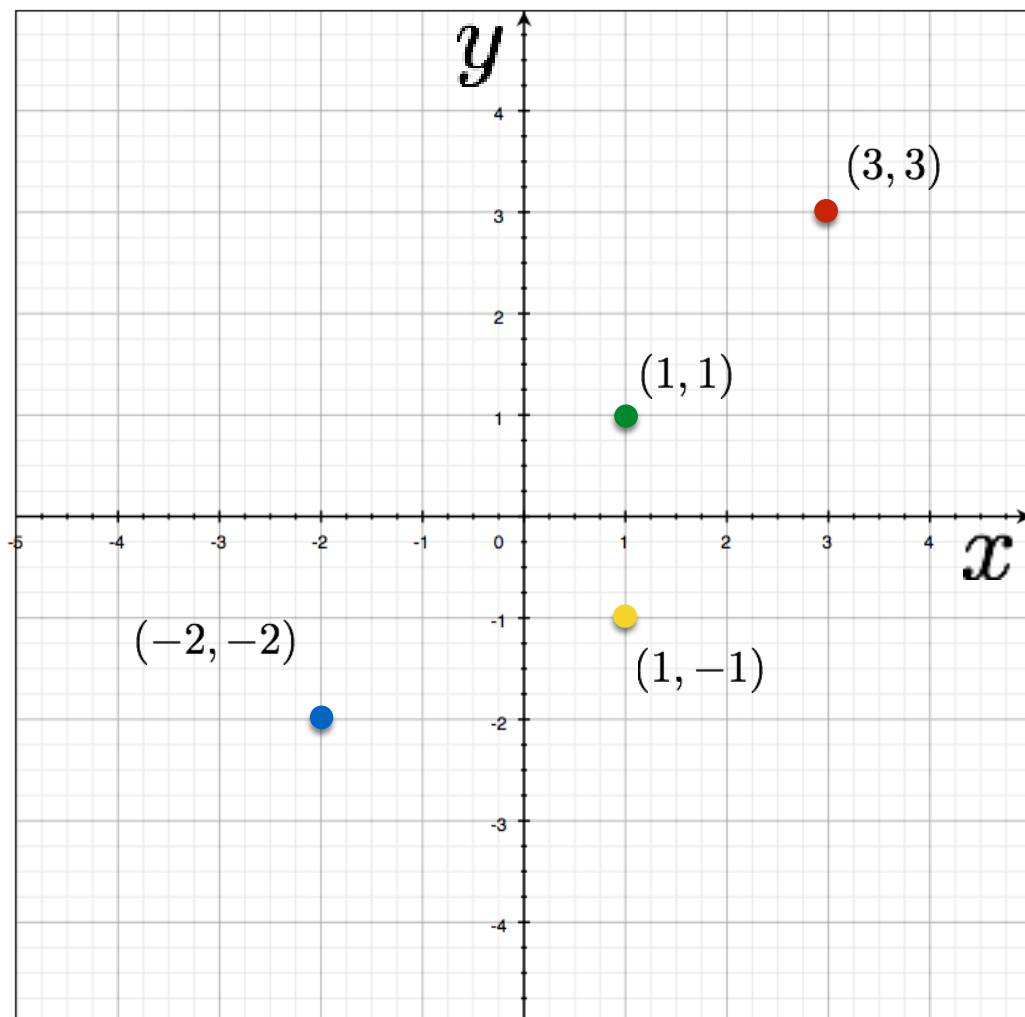
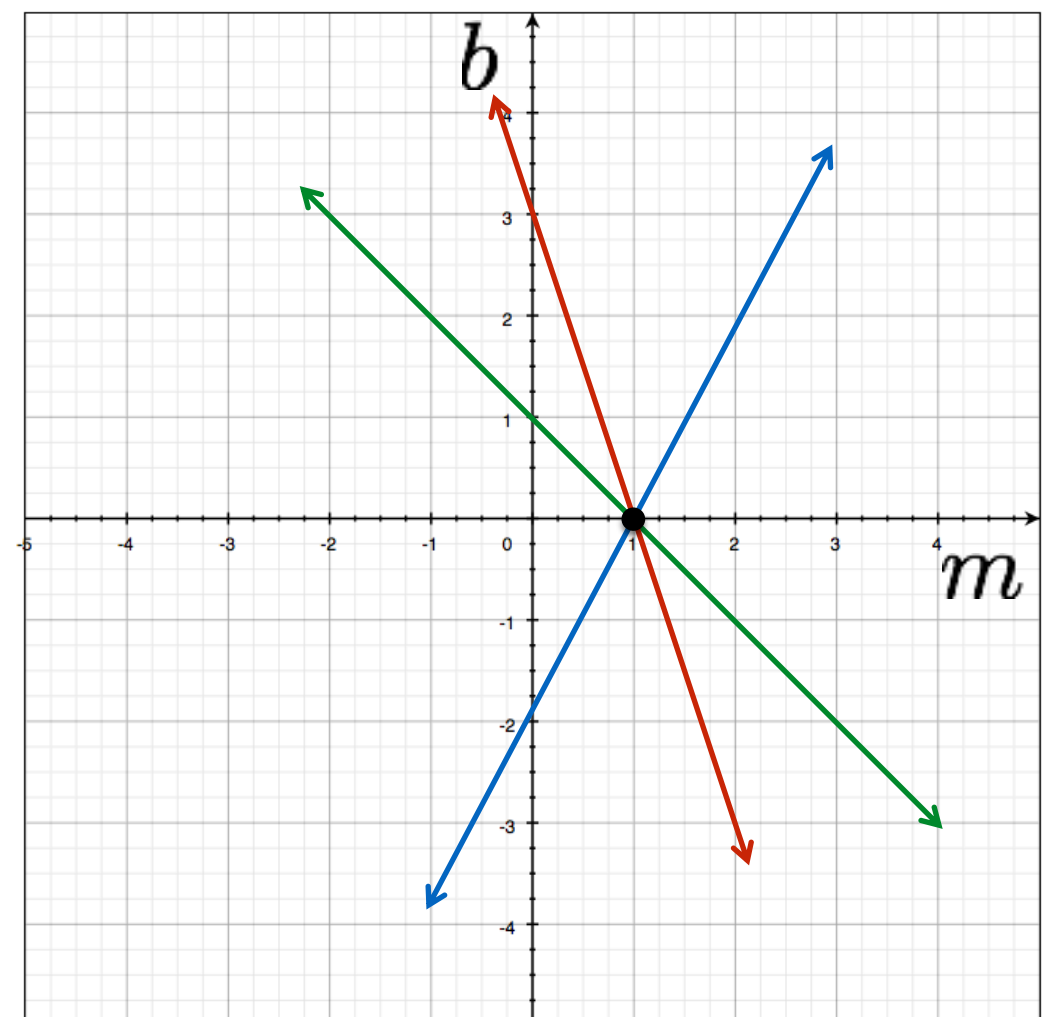


Image space

four points
become
?



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

variables

$$y - mx = b$$

parameters

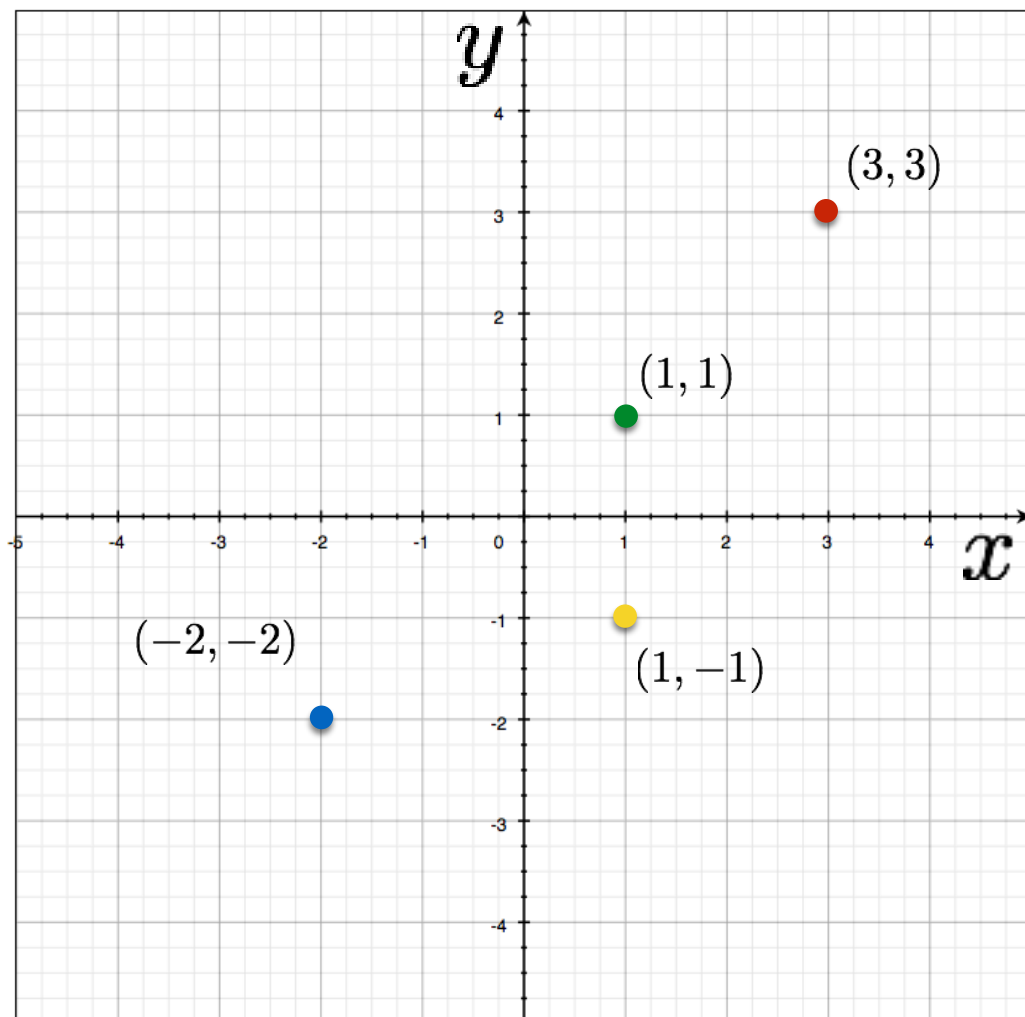
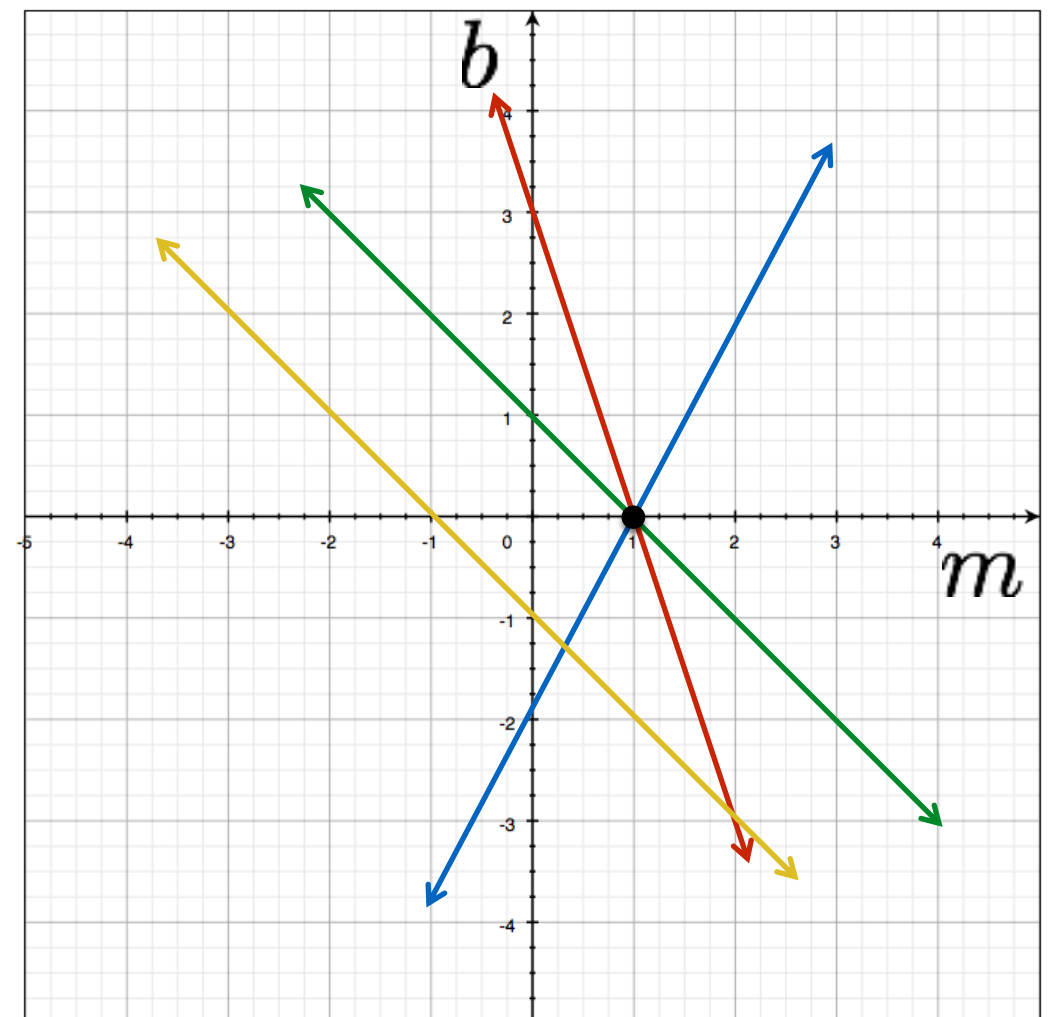


Image space

four points
become
?



Parameter space

How would you find the best fitting line?

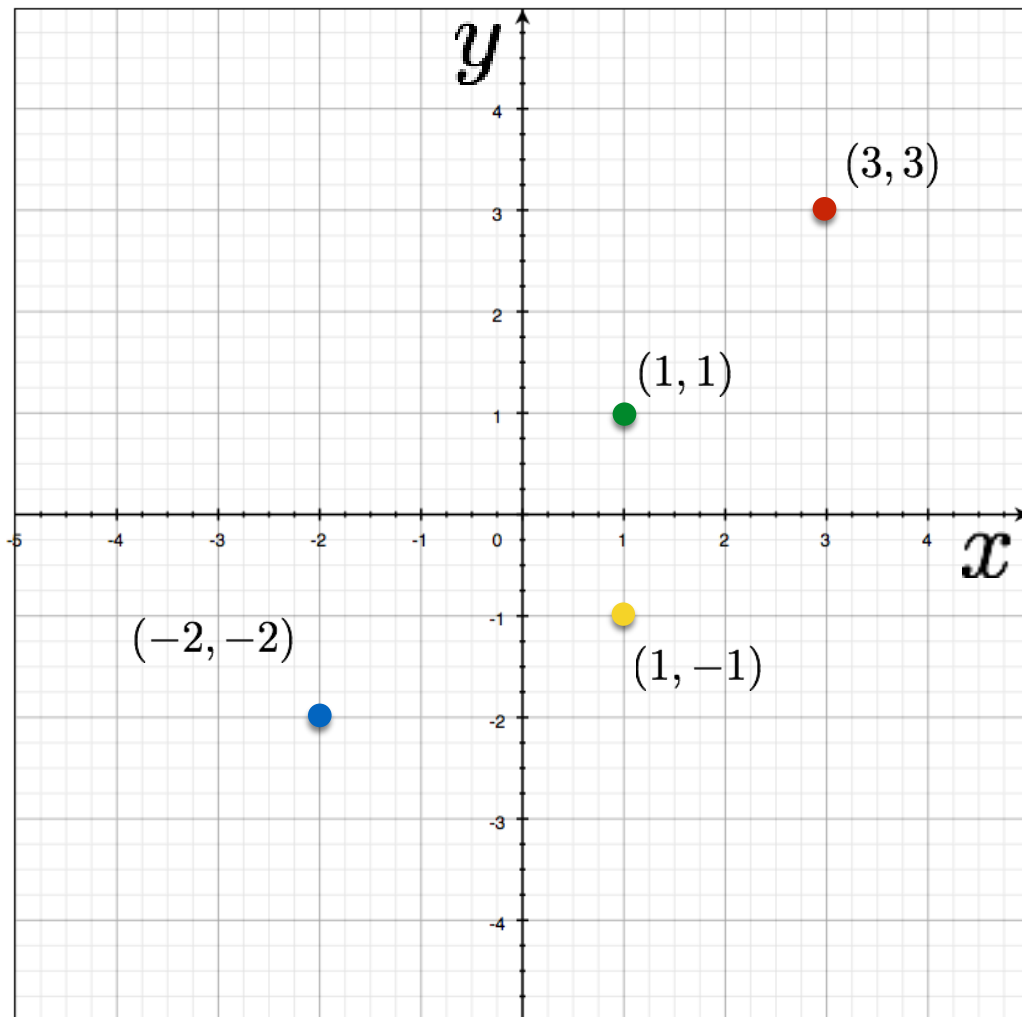
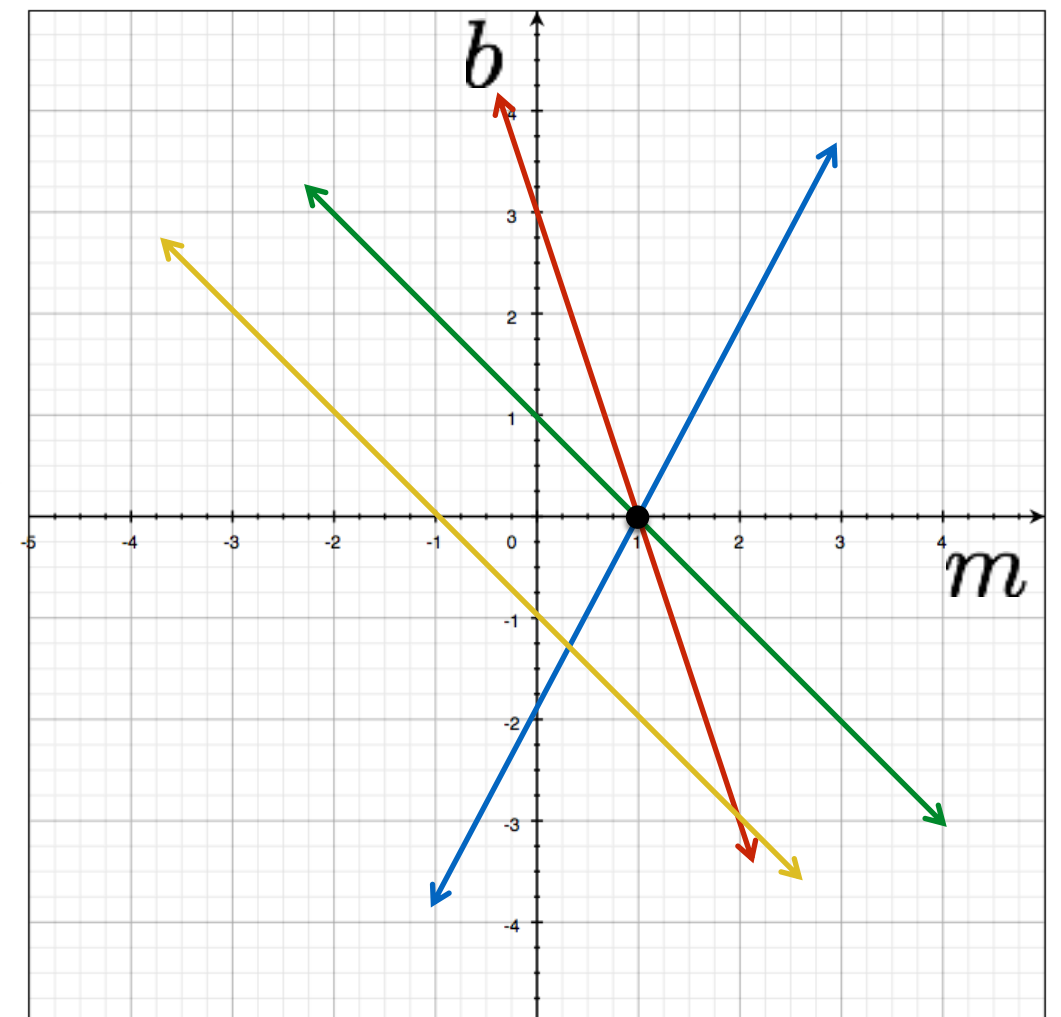


Image space



Parameter space

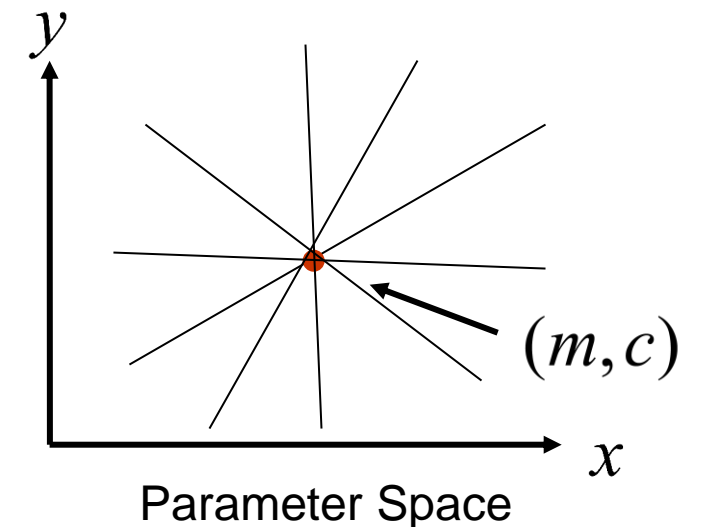
Is this method robust to measurement noise?

Is this method robust to outliers?

Line Detection by Hough Transform

Algorithm:

1. Quantize Parameter Space (m, c)
2. Create Accumulator Array $A(m, c)$
3. Set $A(m, c) = 0 \quad \forall m, c$
4. For each image edge (x_i, y_i)
 For each element in $A(m, c)$
 If (m, c) lies on the line: $c = -x_i m + y_i$
 Increment $A(m, c) = A(m, c) + 1$
5. Find local maxima in $A(m, c)$


$$A(m, c)$$

	1					1		
		1			1			
			1		1			
				2				
			1		1			
		1				1		
	1						1	

Problems with parameterization

How big does the accumulator need to be for the parameterization (m, c) ?

[illegible]

Problems with parameterization

How big does the accumulator need to be for the parameterization (m, c) ?

$A(m, c)$

	1						1		
		1				1			
			1		1				
				2					
			1		1				
		1				1			
	1						1		

The space of m is huge!

The space of c is huge!

$$-\infty \leq m \leq \infty$$

$$-\infty \leq c \leq \infty$$

Better Parameterization

Use normal form:

$$x \cos \theta + y \sin \theta = \rho$$

Given points (x_i, y_i) find (ρ, θ)

Hough Space Sinusoid

$$0 \leq \theta \leq 2\pi$$

$$0 \leq \rho \leq \rho_{\max}$$

(Finite Accumulator Array Size)

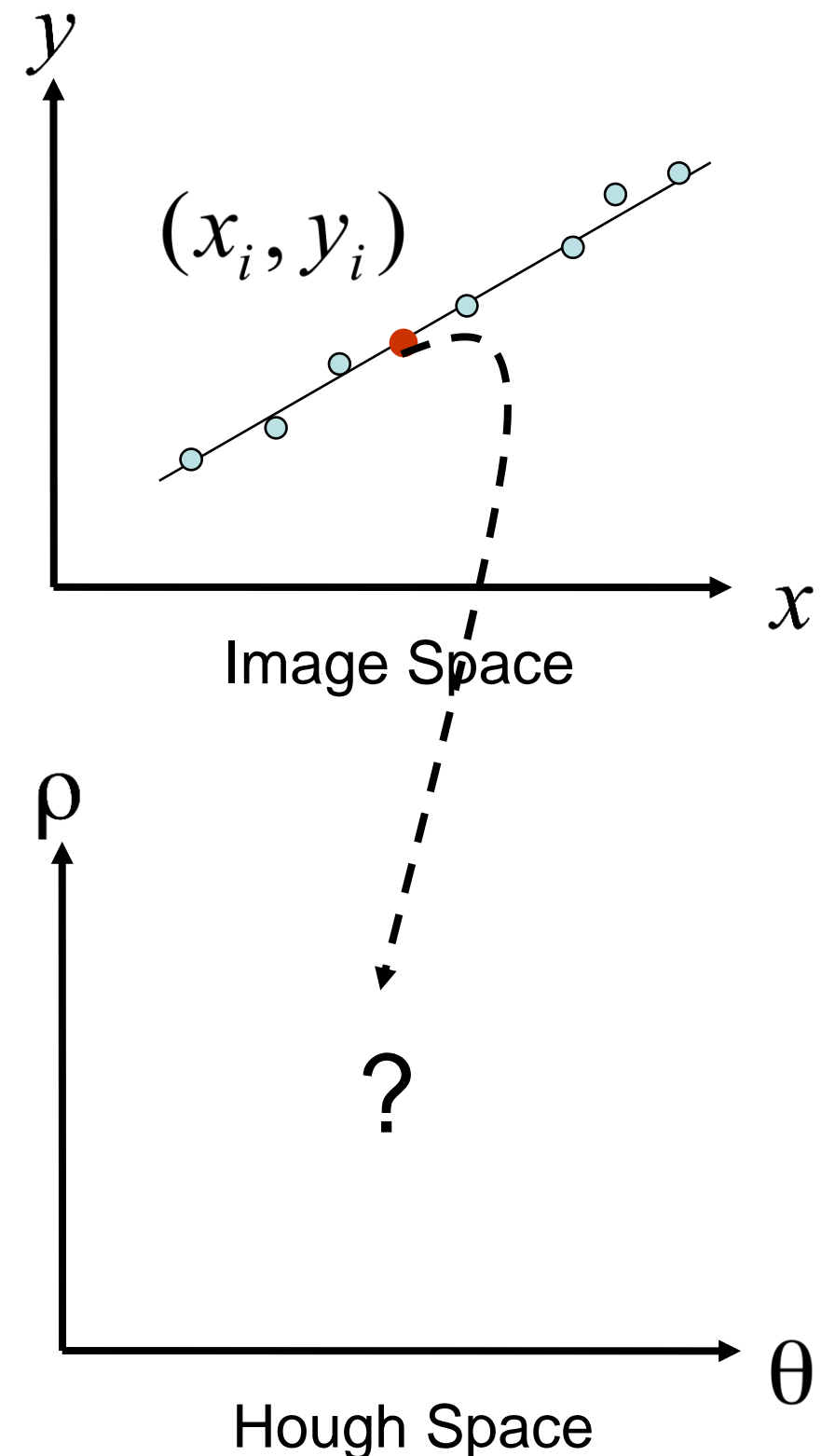


Image and parameter space

variables

$$y = mx + b$$

parameters

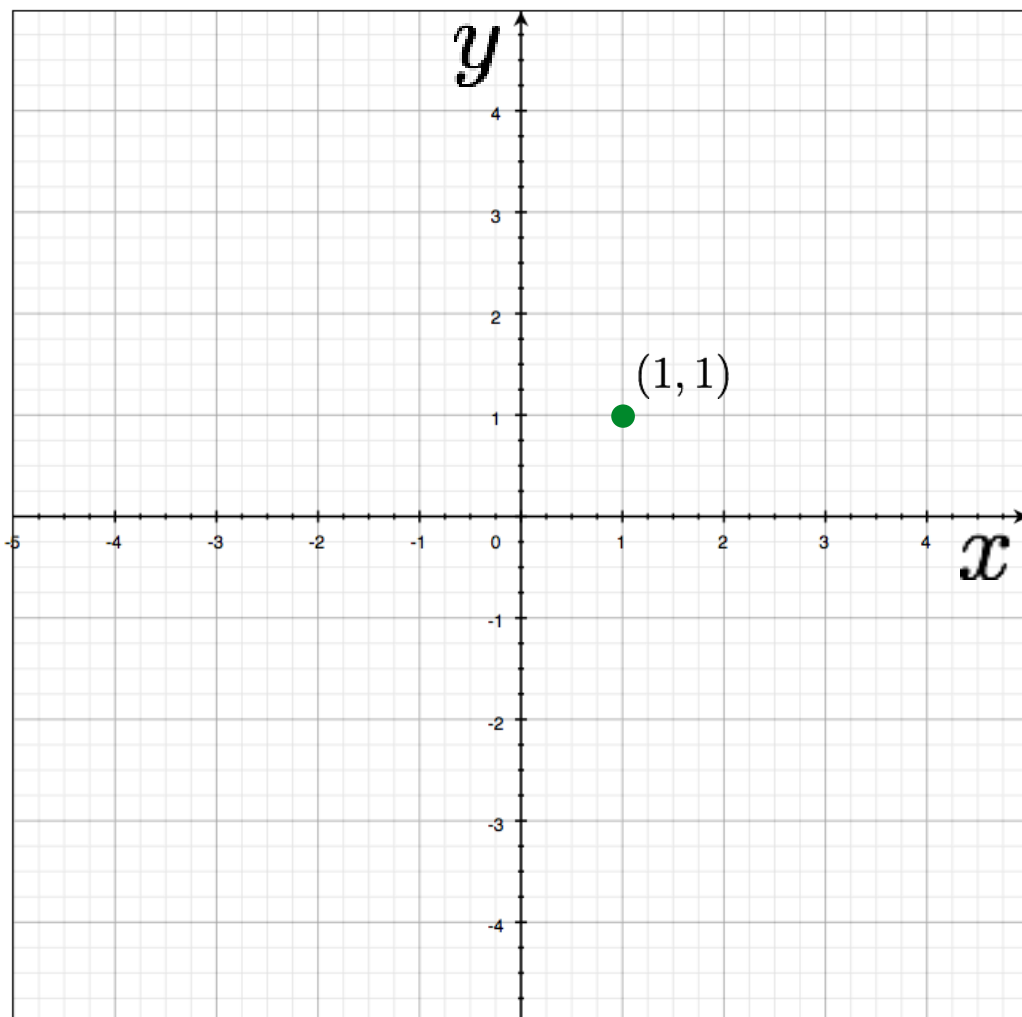


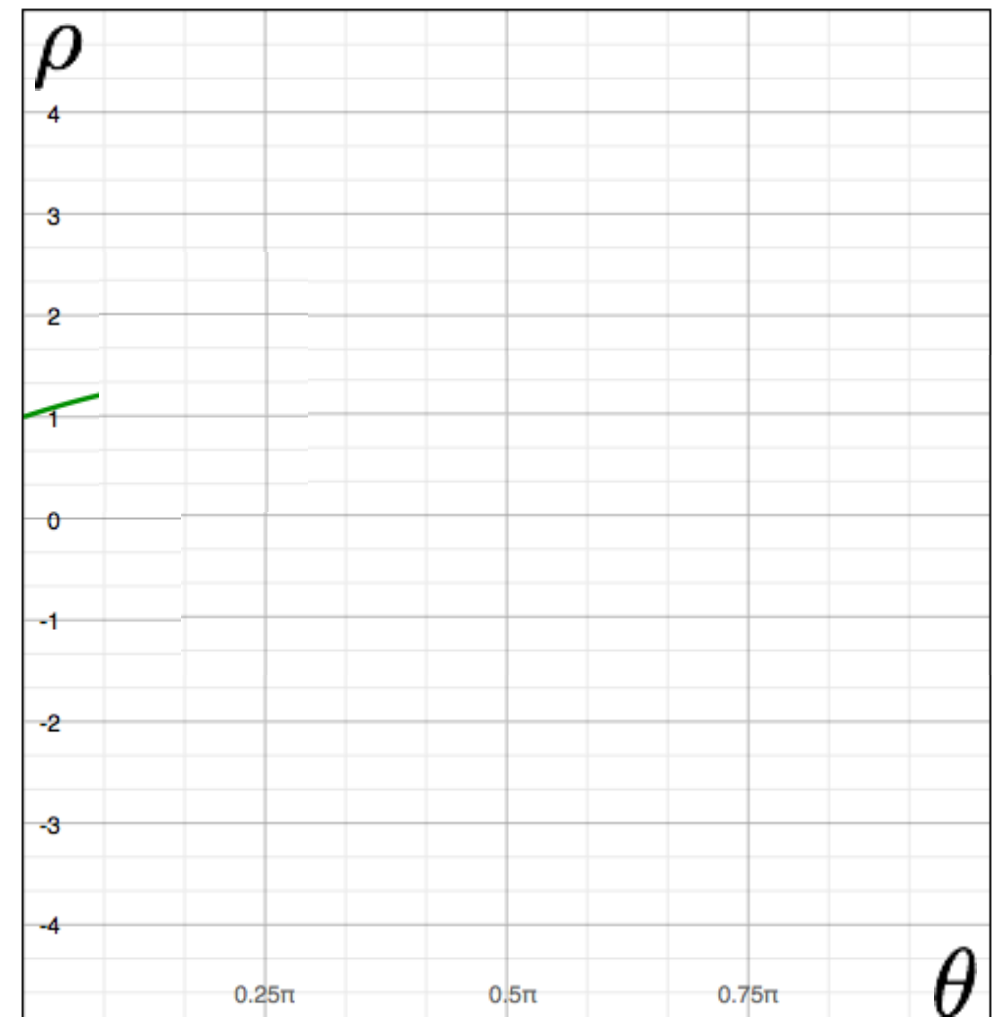
Image space



parameters

$$x \cos \theta + y \sin \theta = \rho$$

variables



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

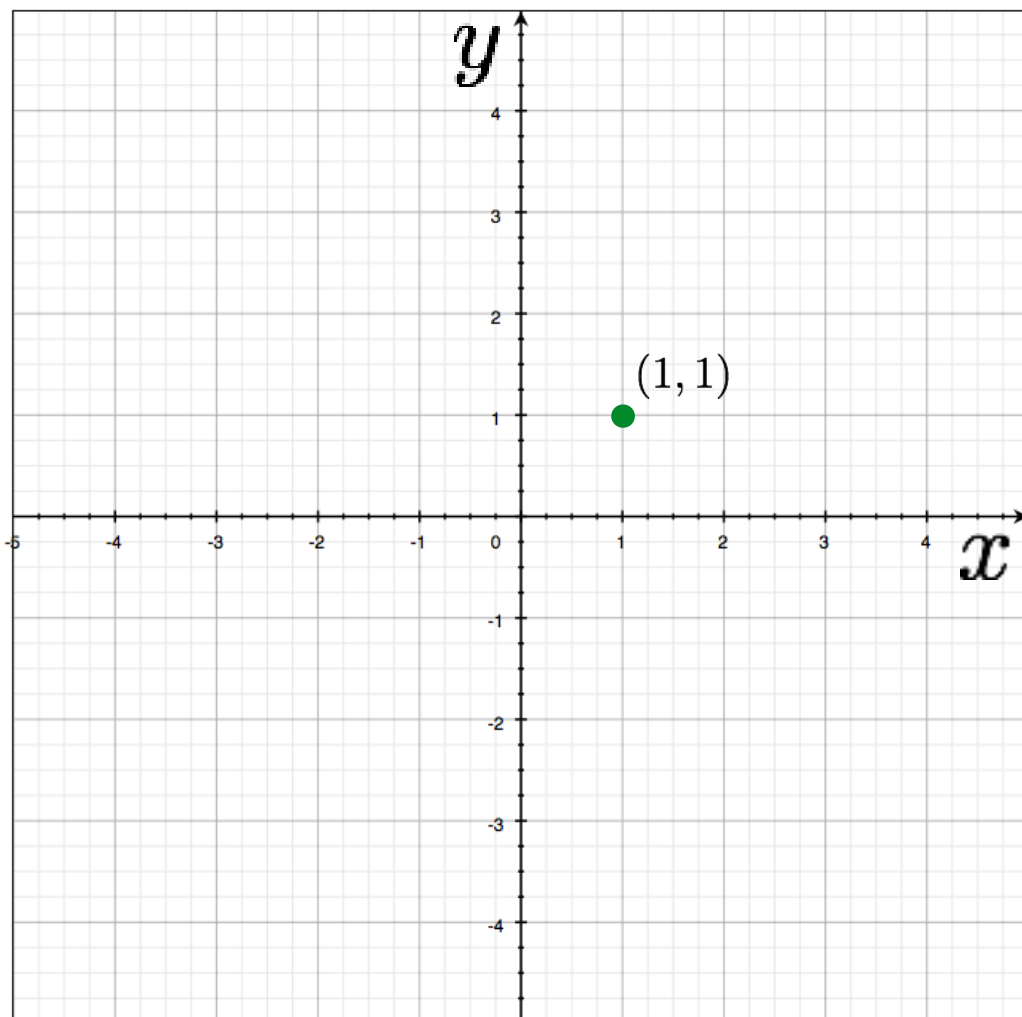


Image space

a point
becomes a
wave

parameters

$$x \cos \theta + y \sin \theta = \rho$$

variables



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

$$x \cos \theta + y \sin \theta = \rho$$

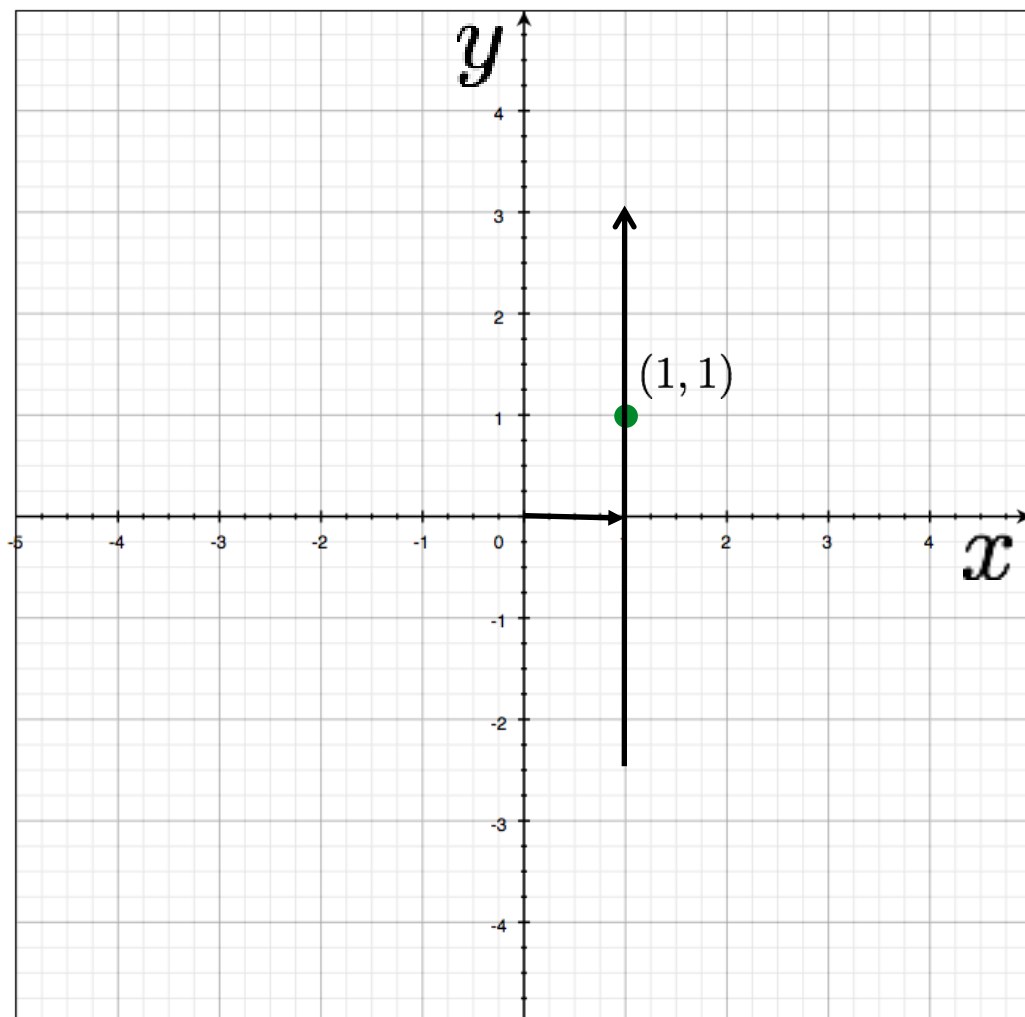
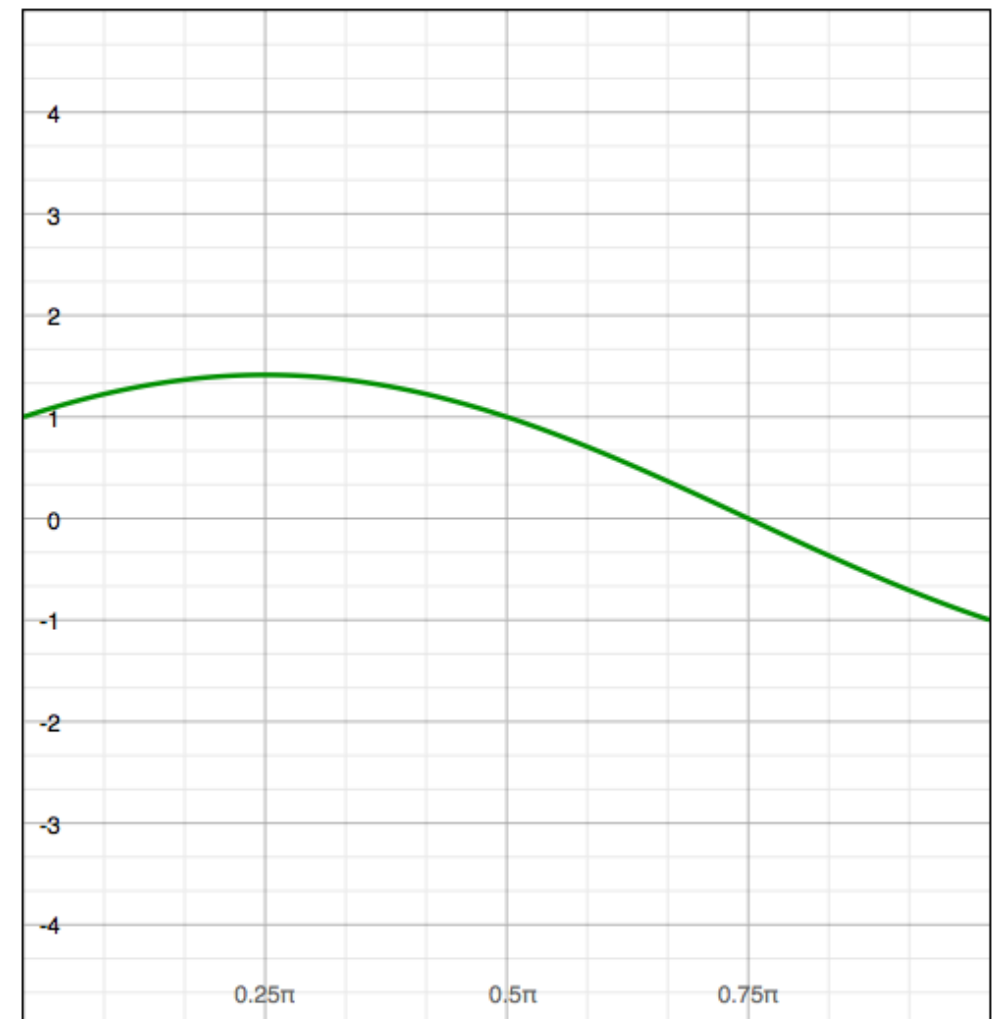


Image space



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

$$x \cos \theta + y \sin \theta = \rho$$

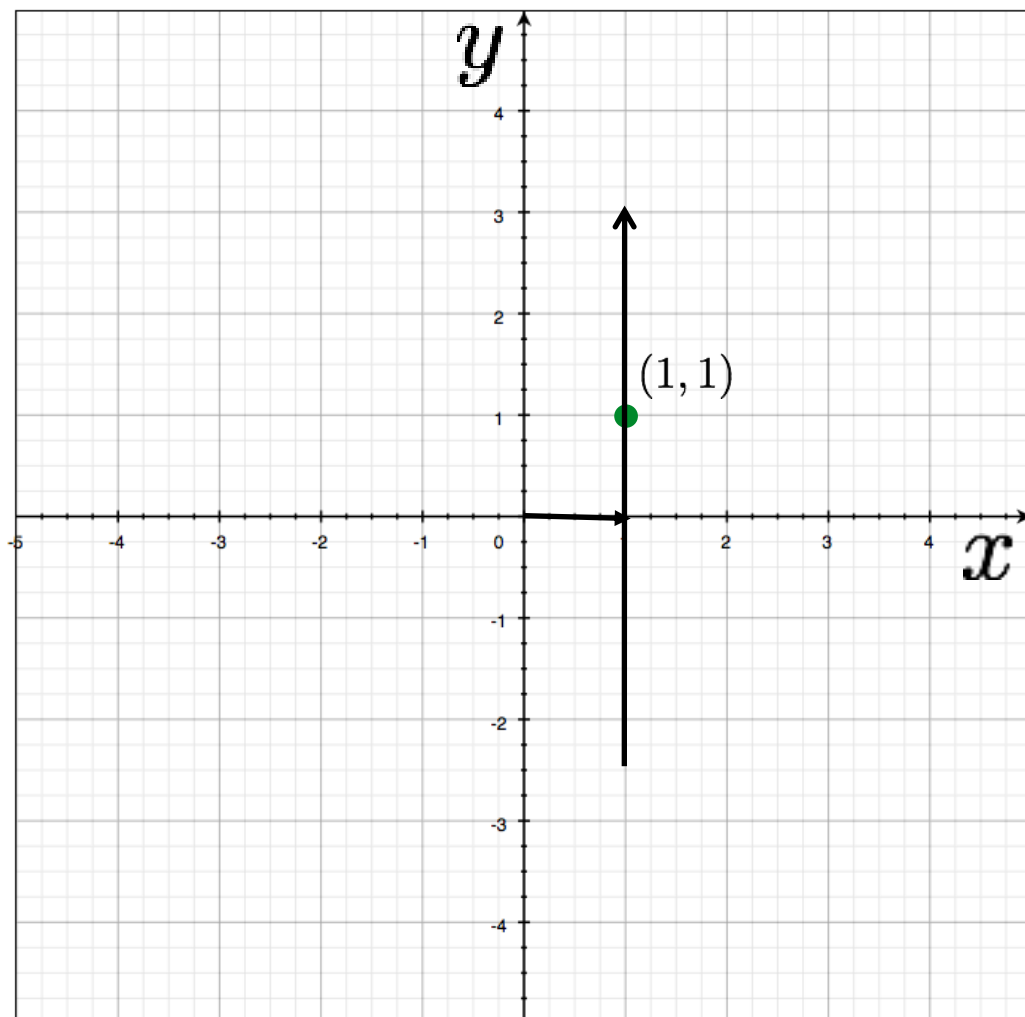
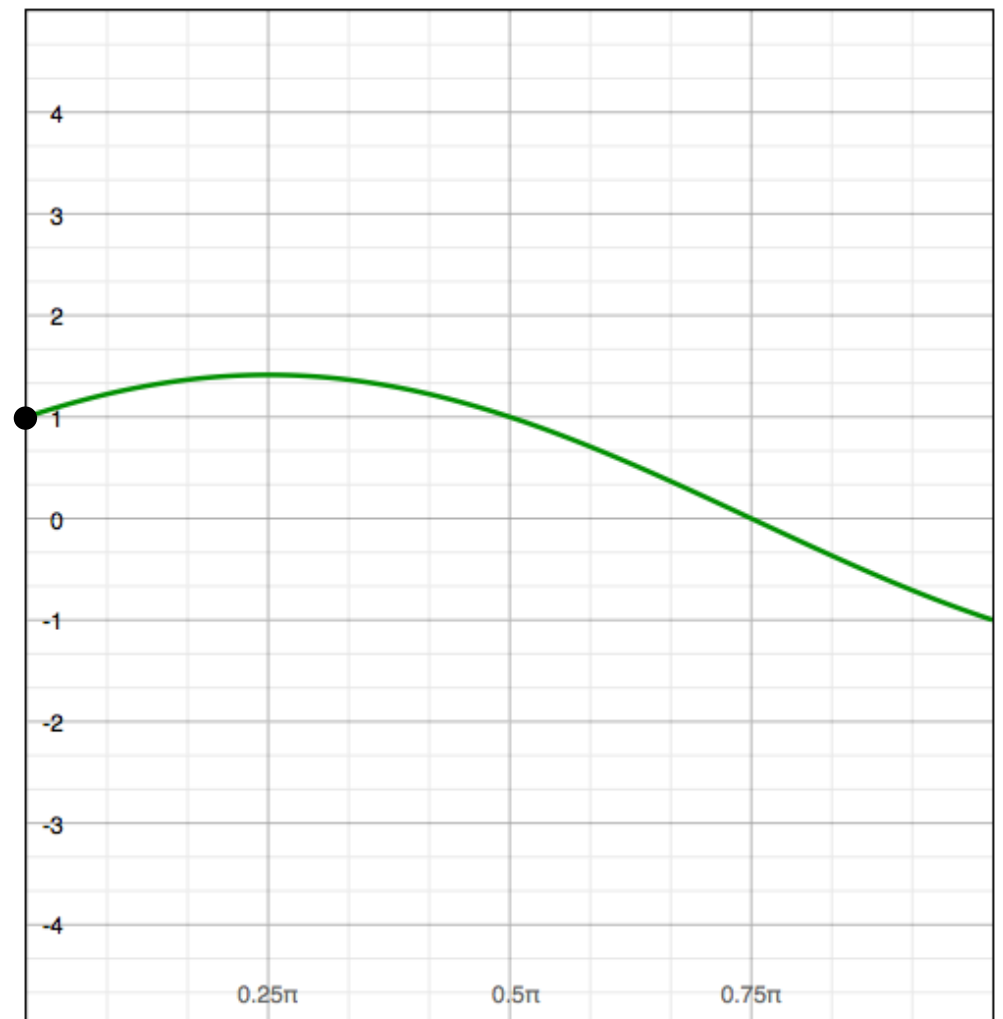


Image space

a line
becomes a
point



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

$$x \cos \theta + y \sin \theta = \rho$$

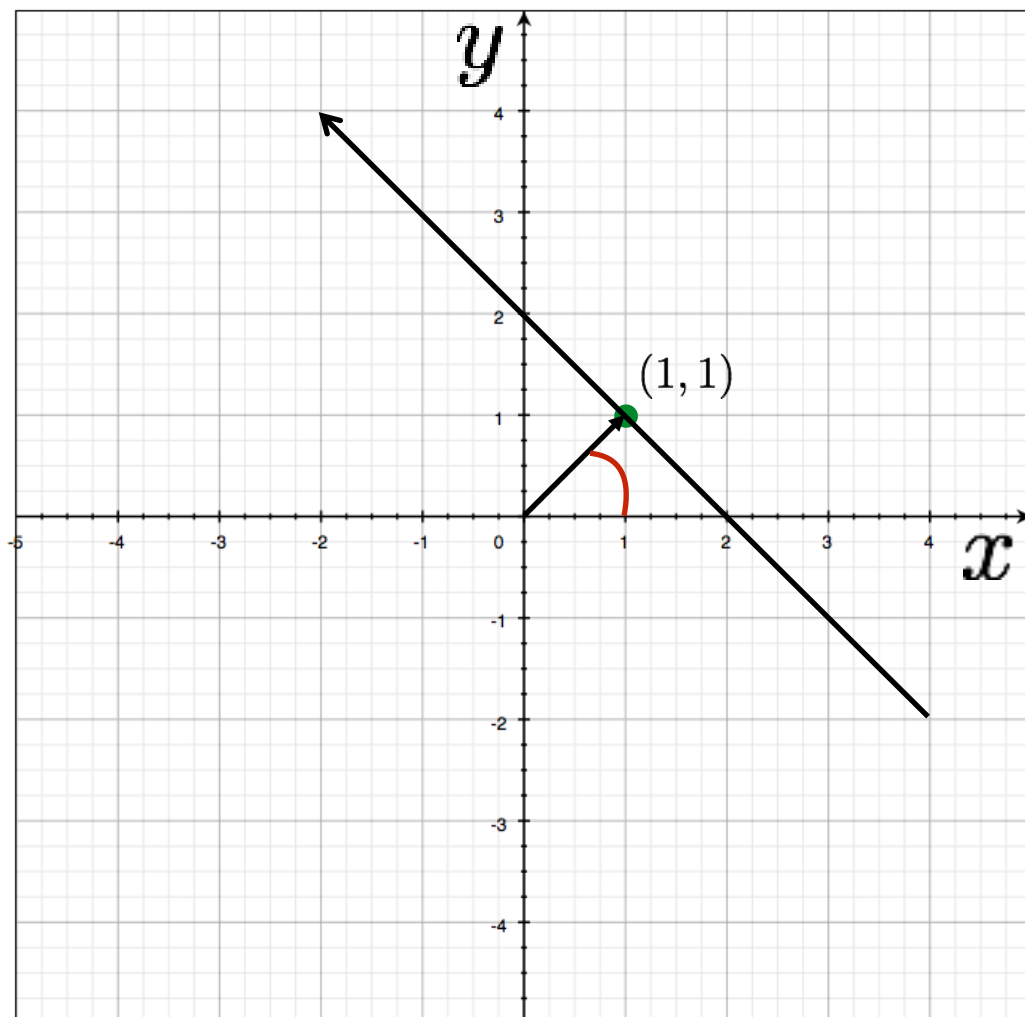
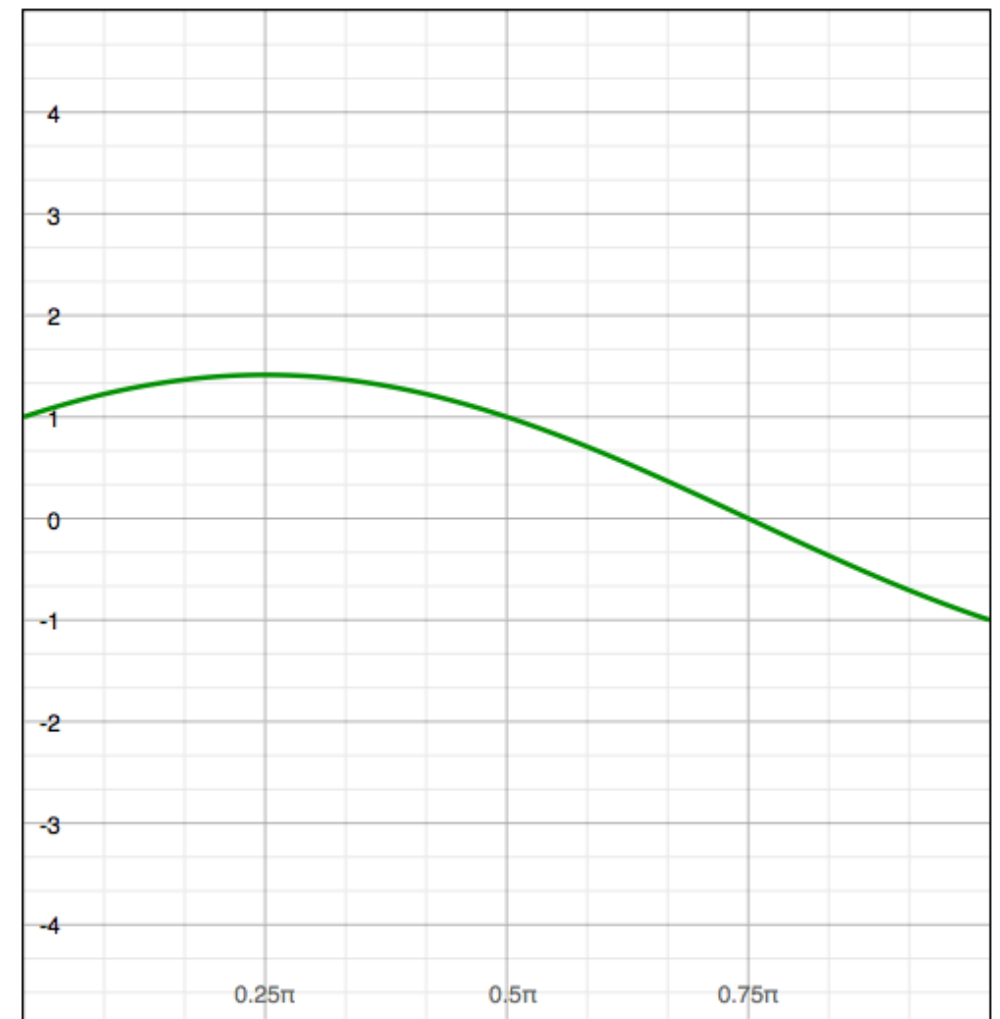


Image space



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

$$x \cos \theta + y \sin \theta = \rho$$

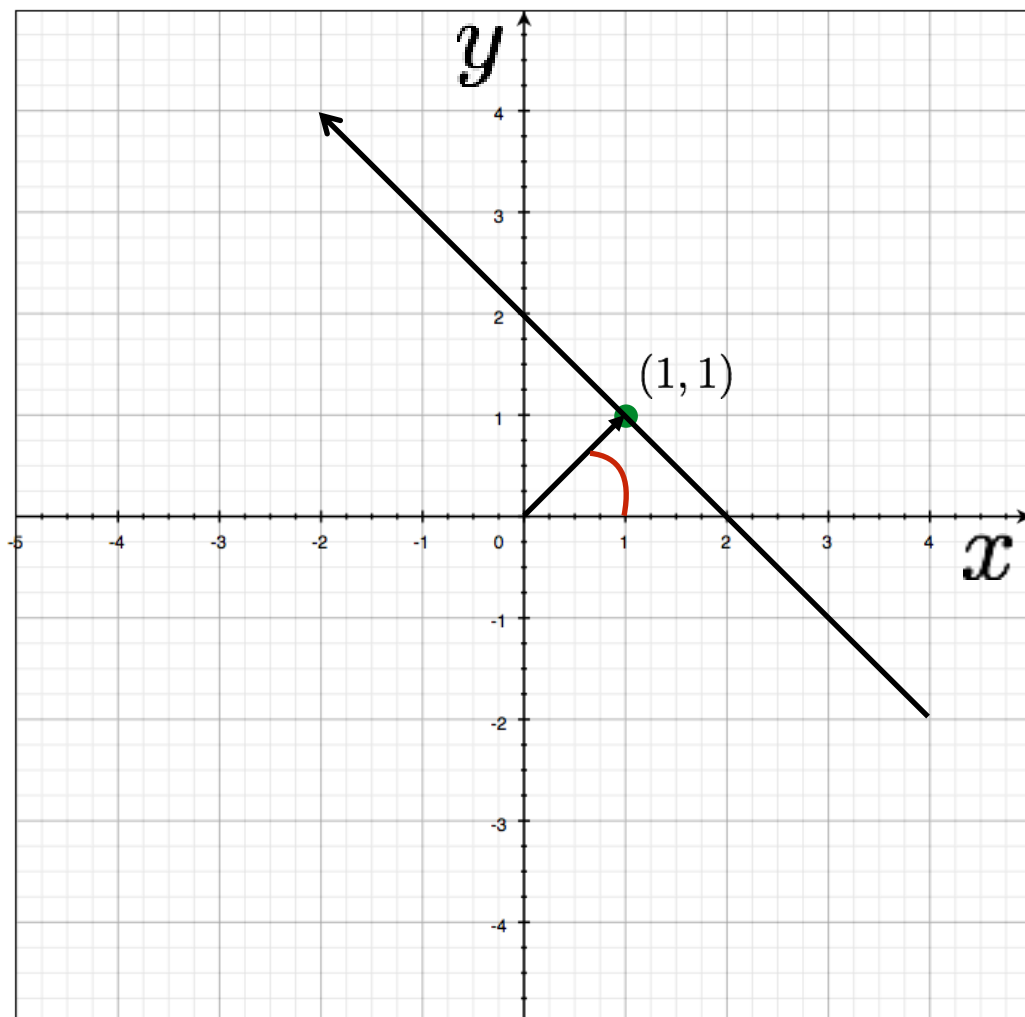
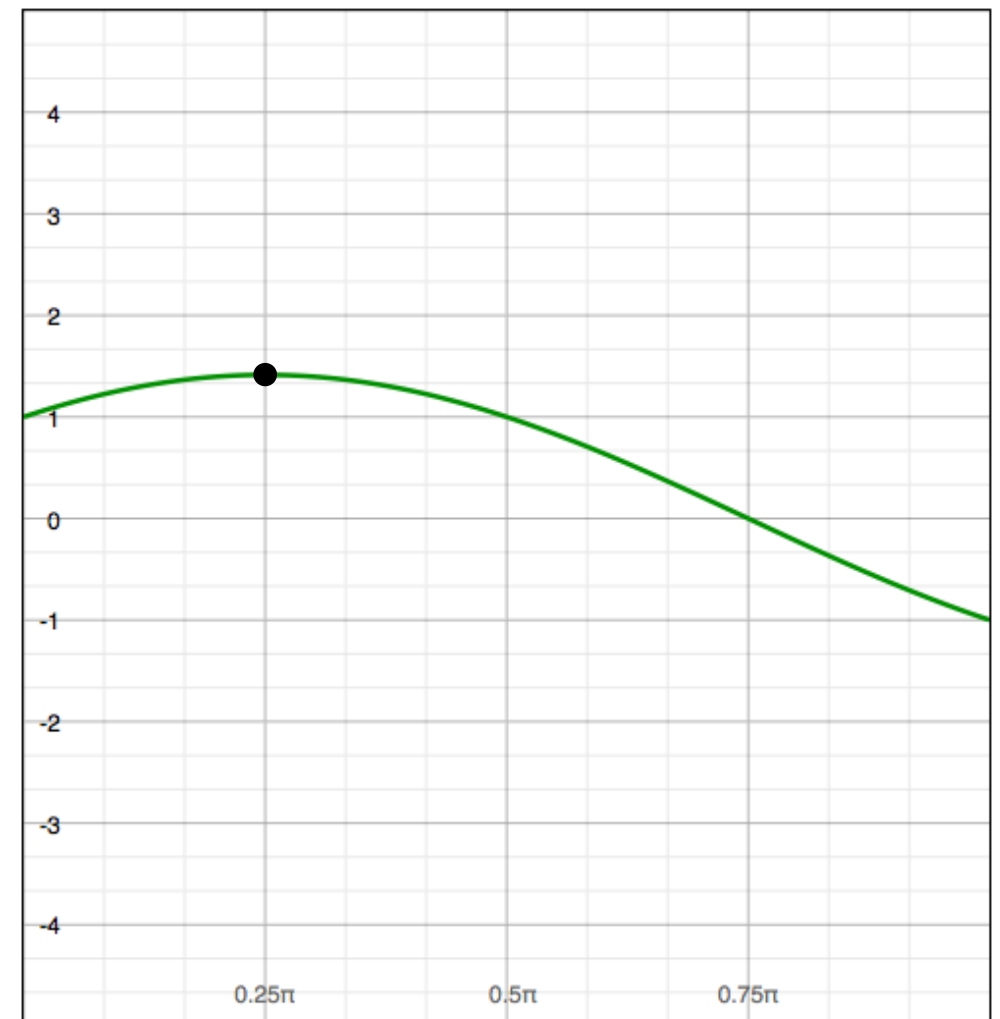


Image space

a line
becomes a
point



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

$$x \cos \theta + y \sin \theta = \rho$$

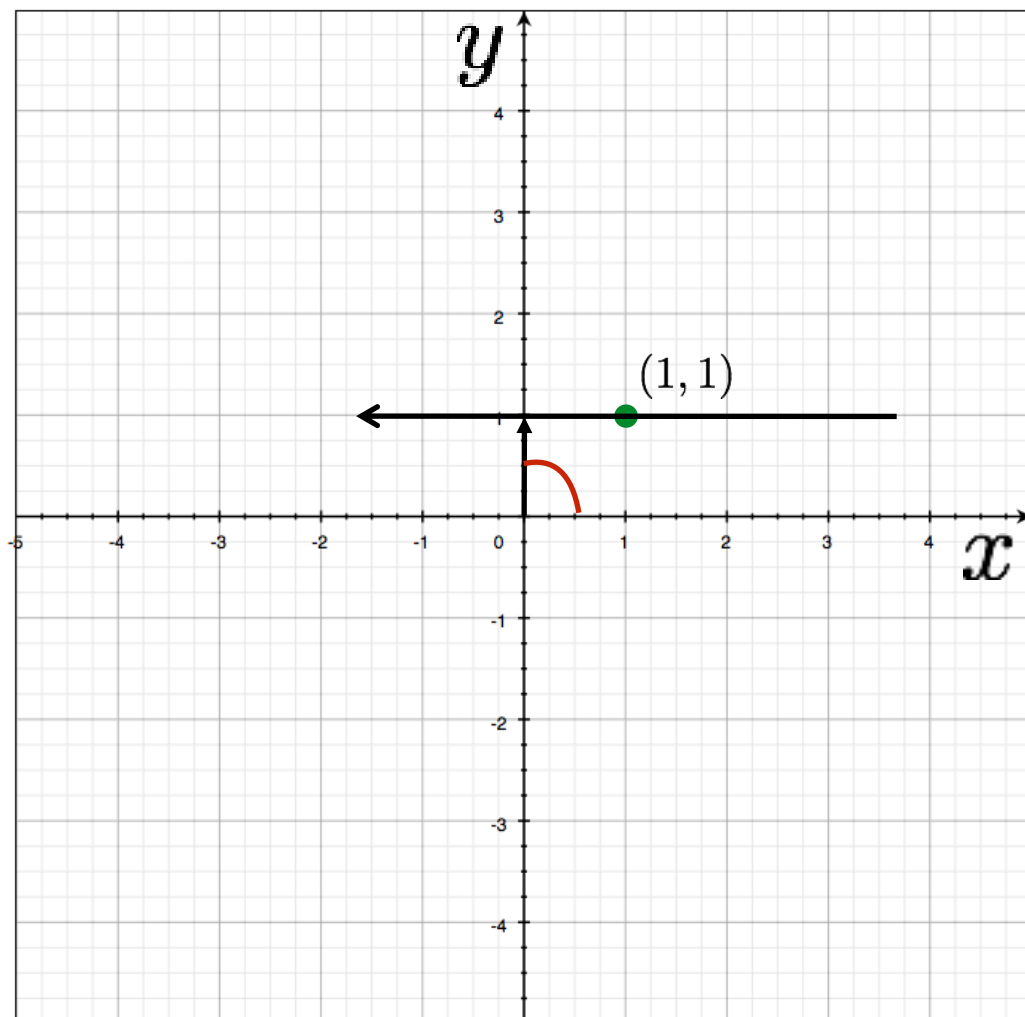
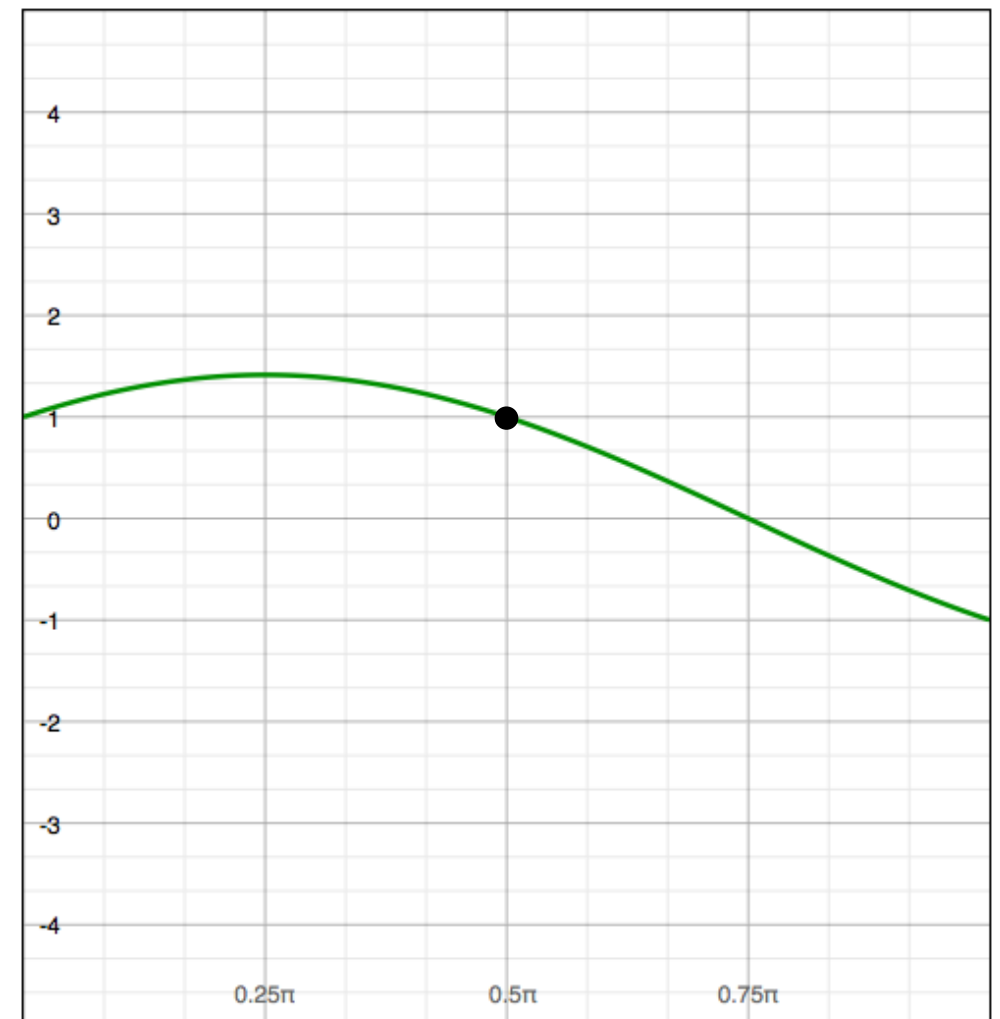


Image space

a line
becomes a
point



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

$$x \cos \theta + y \sin \theta = \rho$$

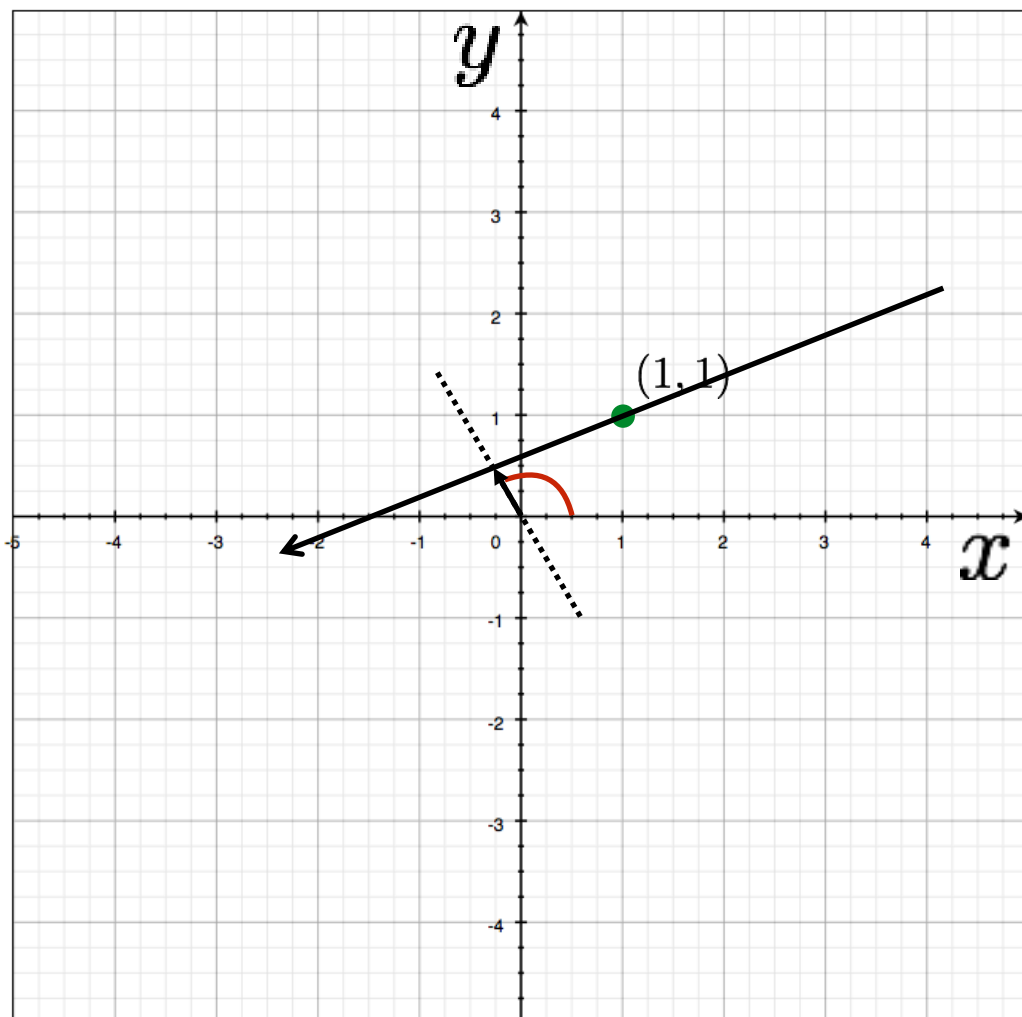
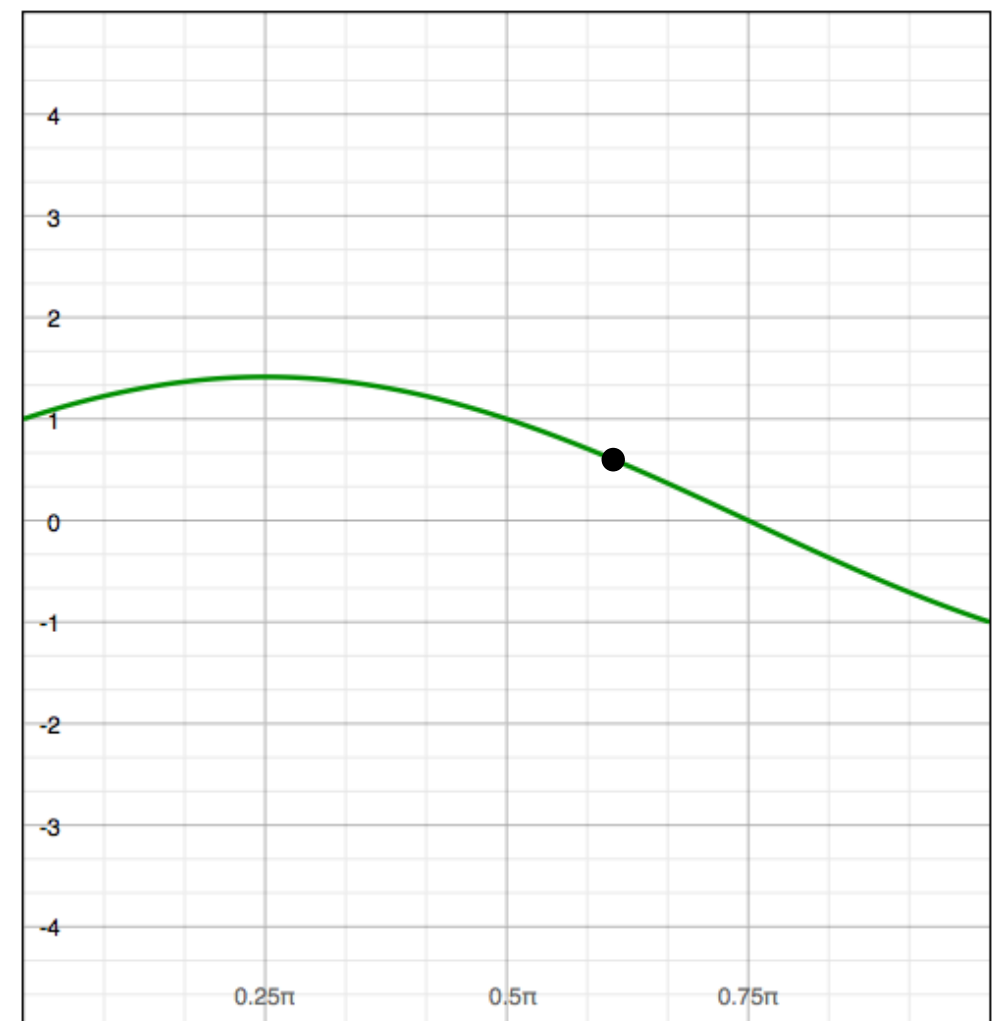


Image space

a line
becomes a
point



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

$$x \cos \theta + y \sin \theta = \rho$$

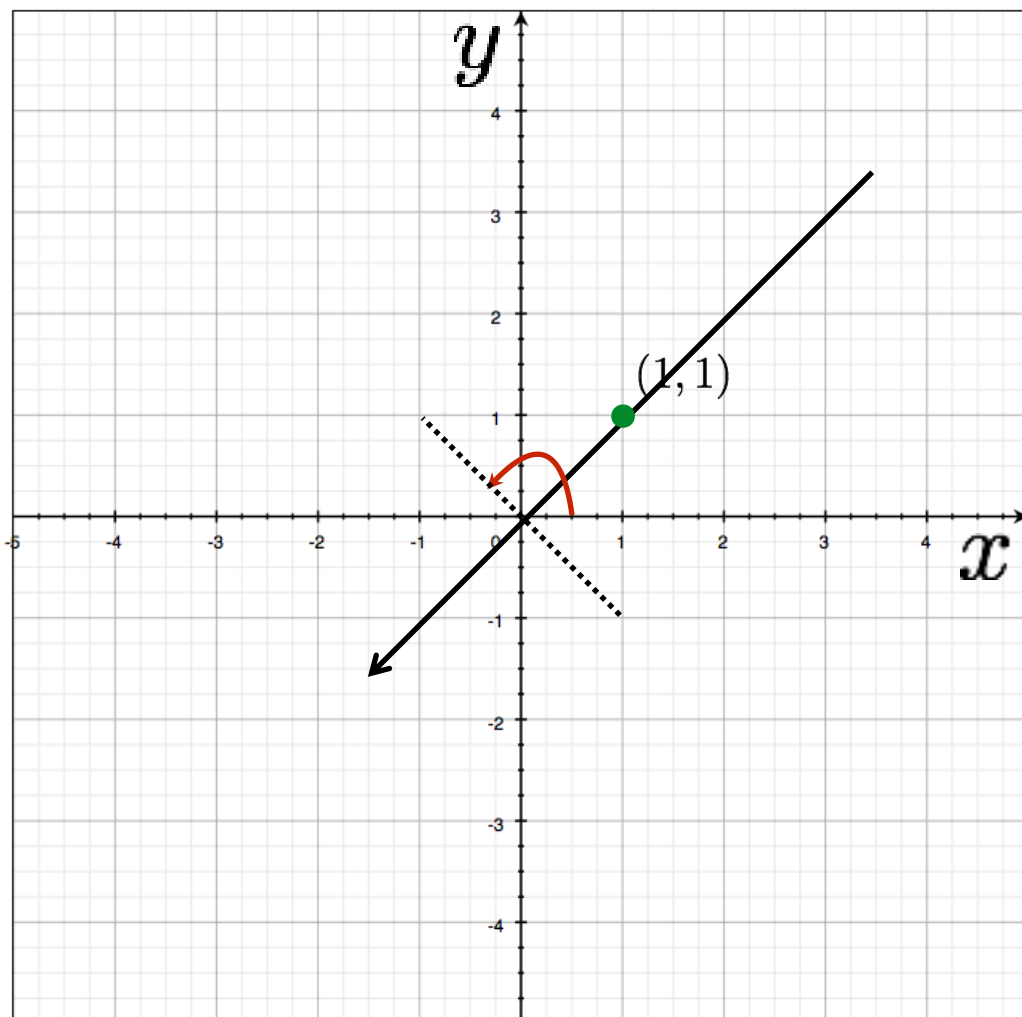
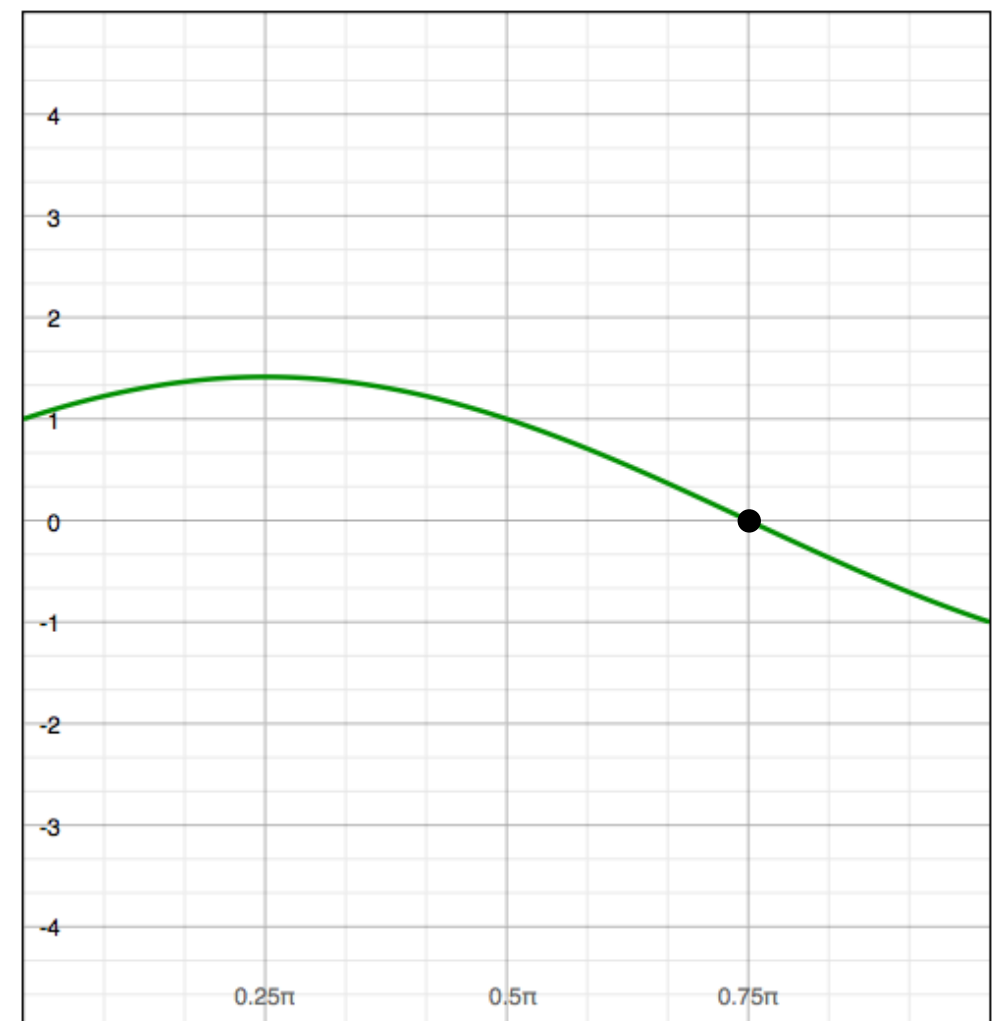


Image space

a line
becomes a
point



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

$$x \cos \theta + y \sin \theta = \rho$$

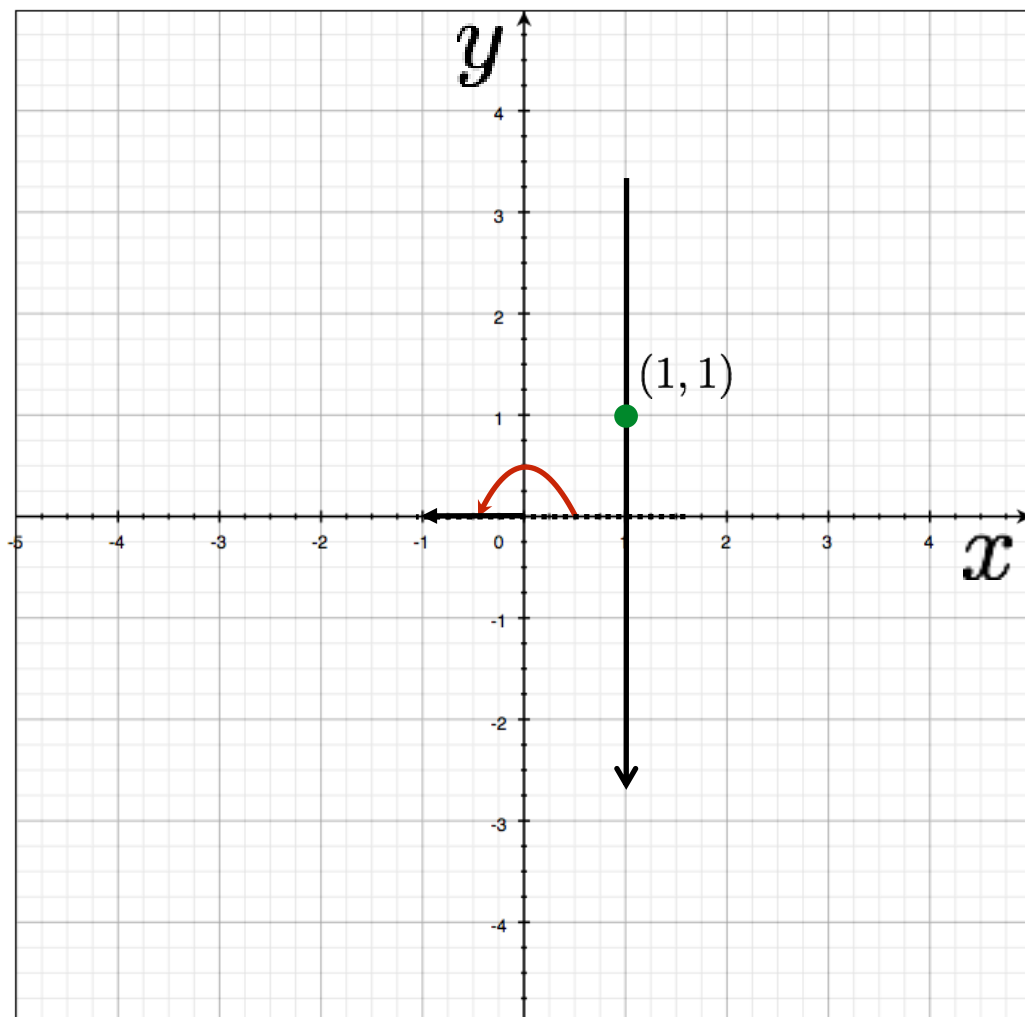
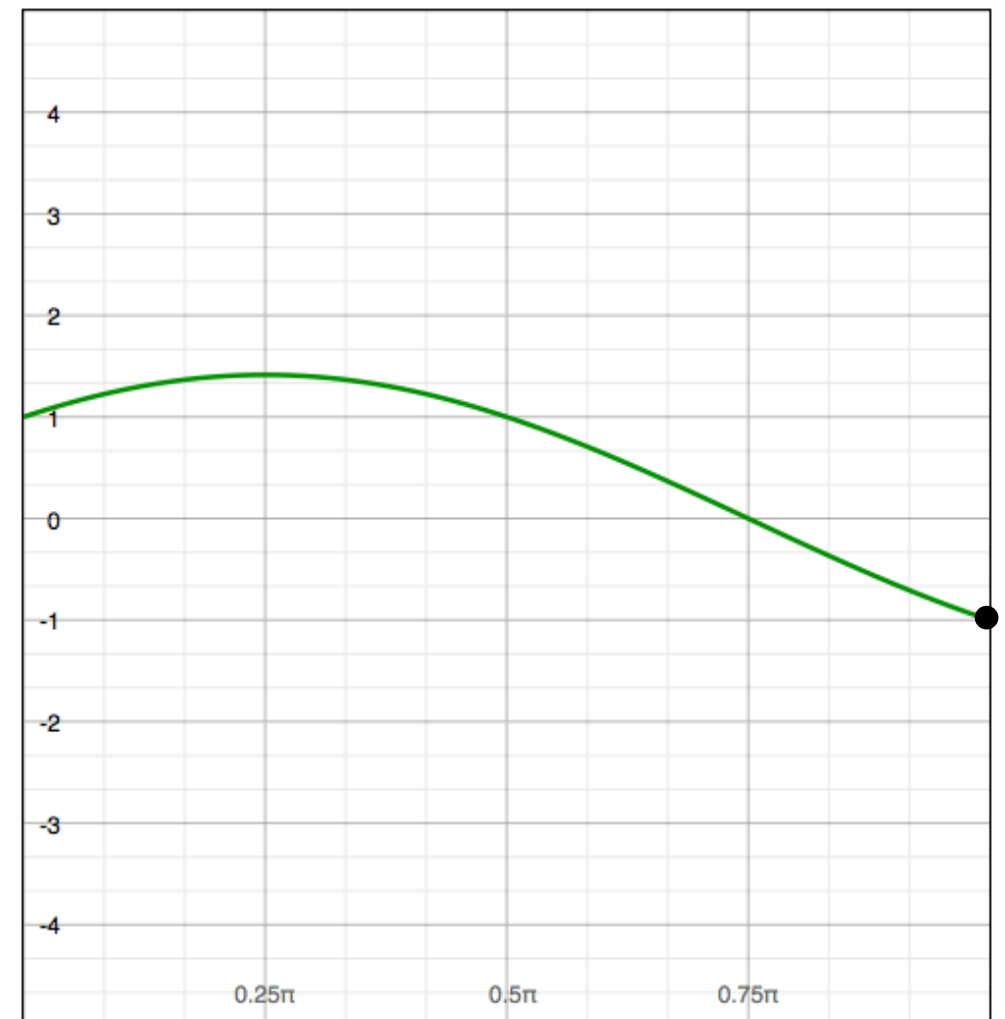


Image space

a line
becomes a
point



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

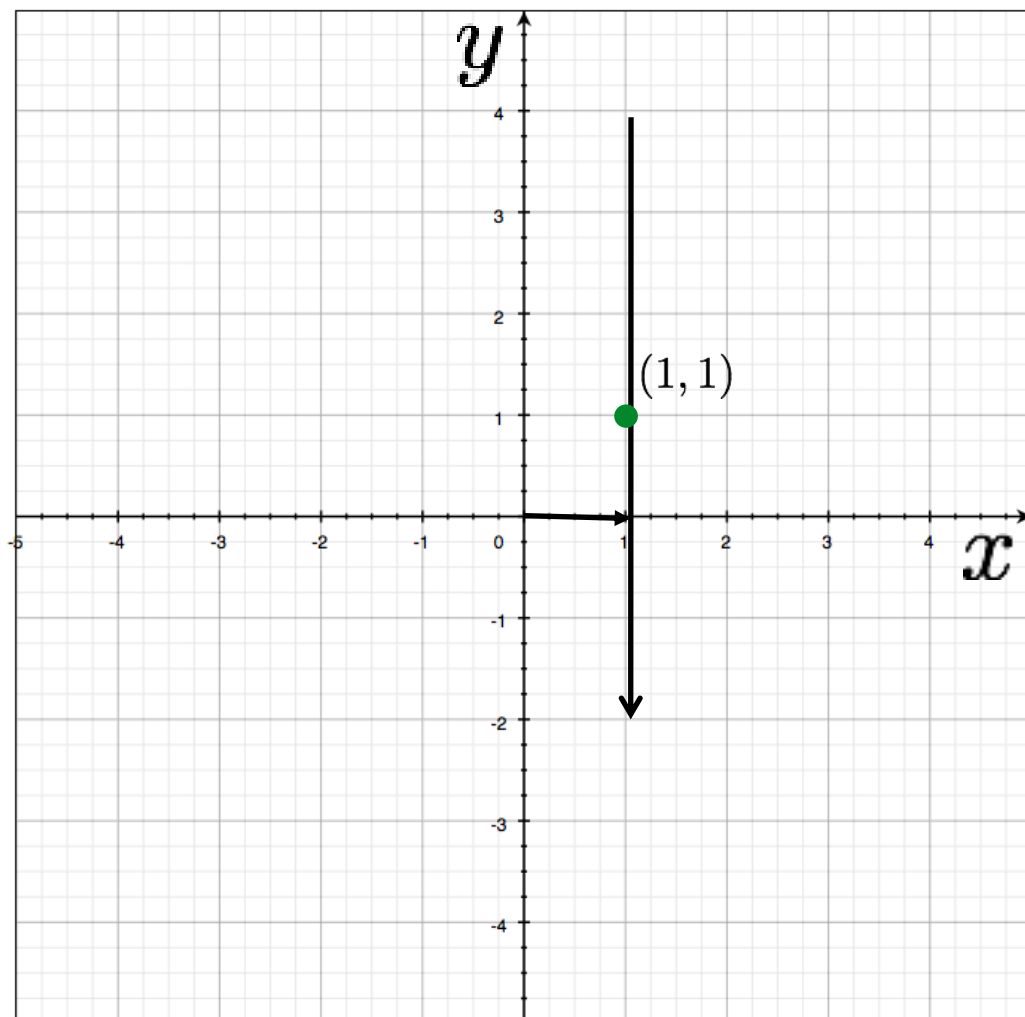
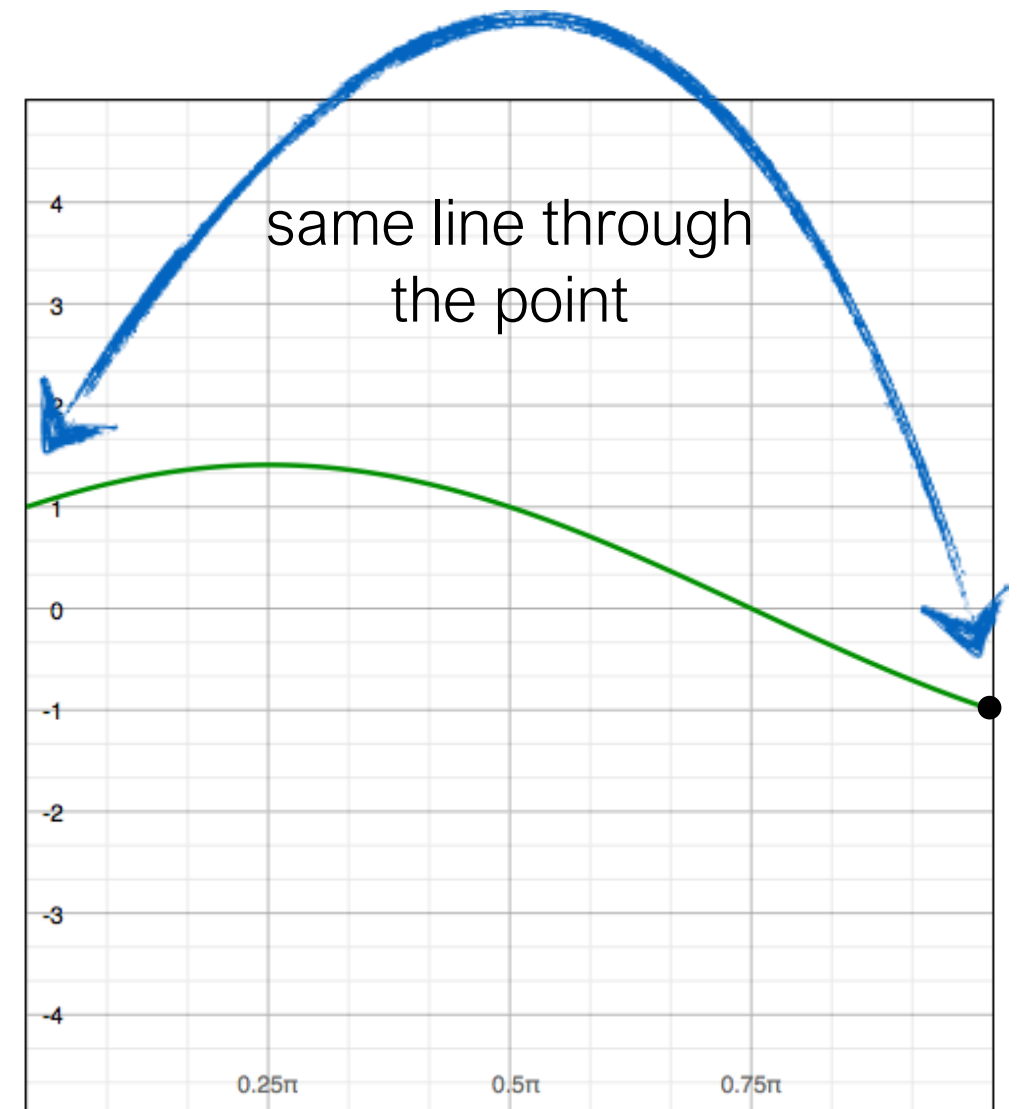


Image space

a line
becomes a
point

$$x \cos \theta + y \sin \theta = \rho$$



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

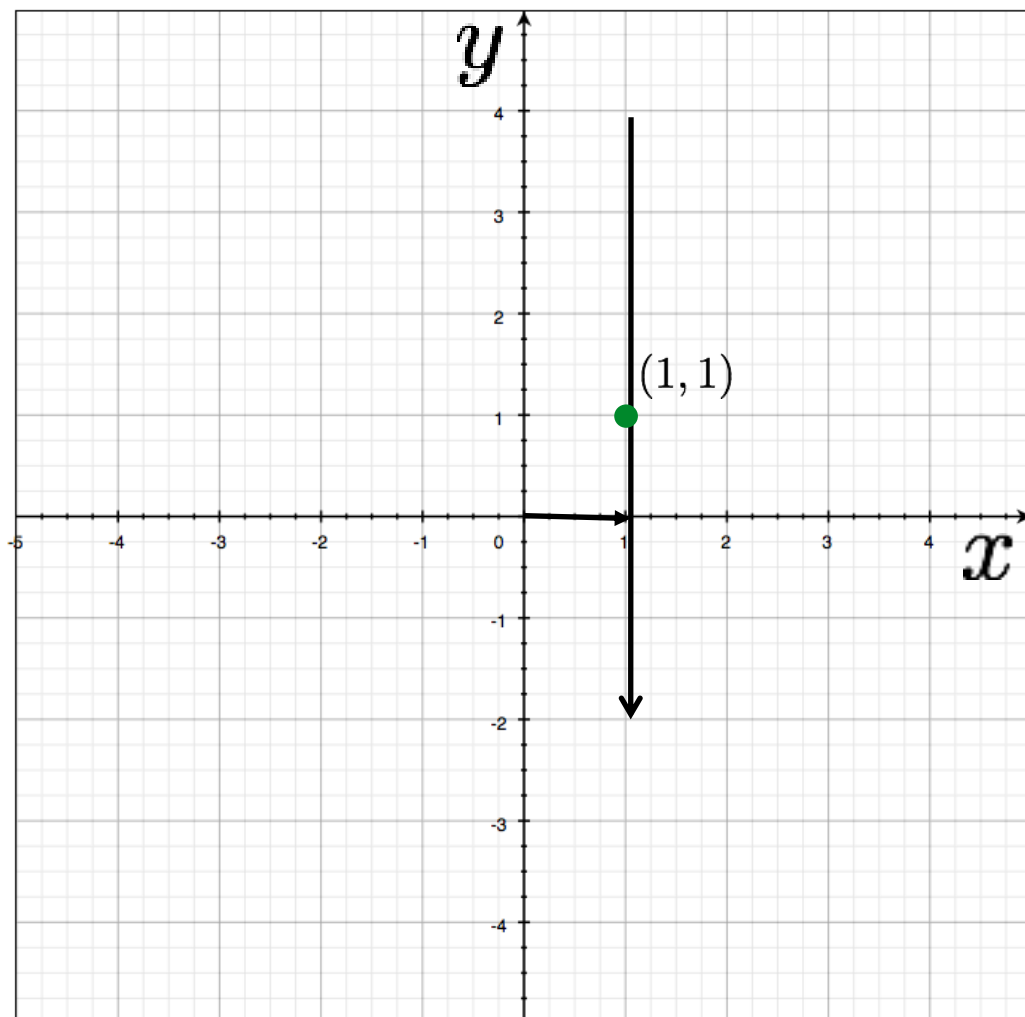
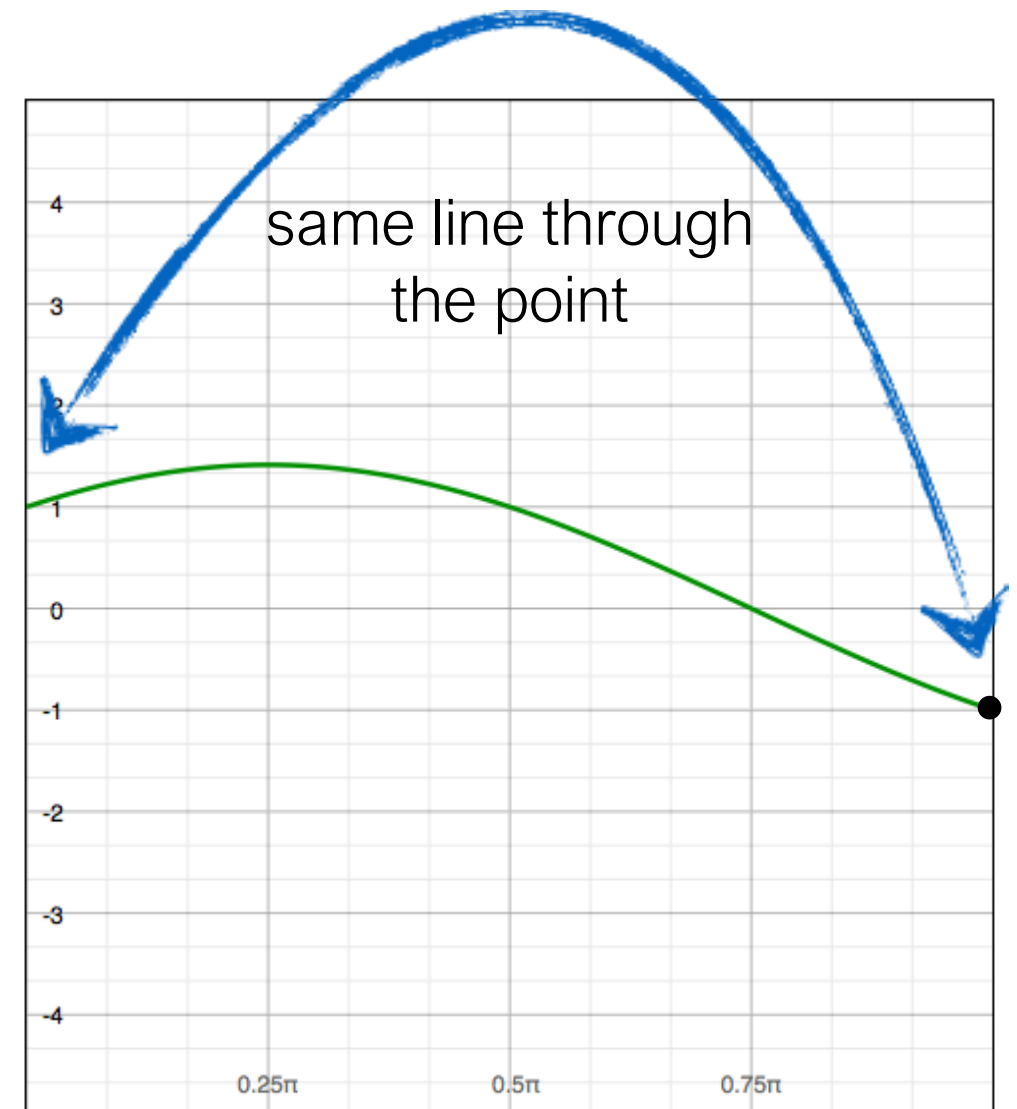


Image space

a line
becomes a
point

$$x \cos \theta + y \sin \theta = \rho$$



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

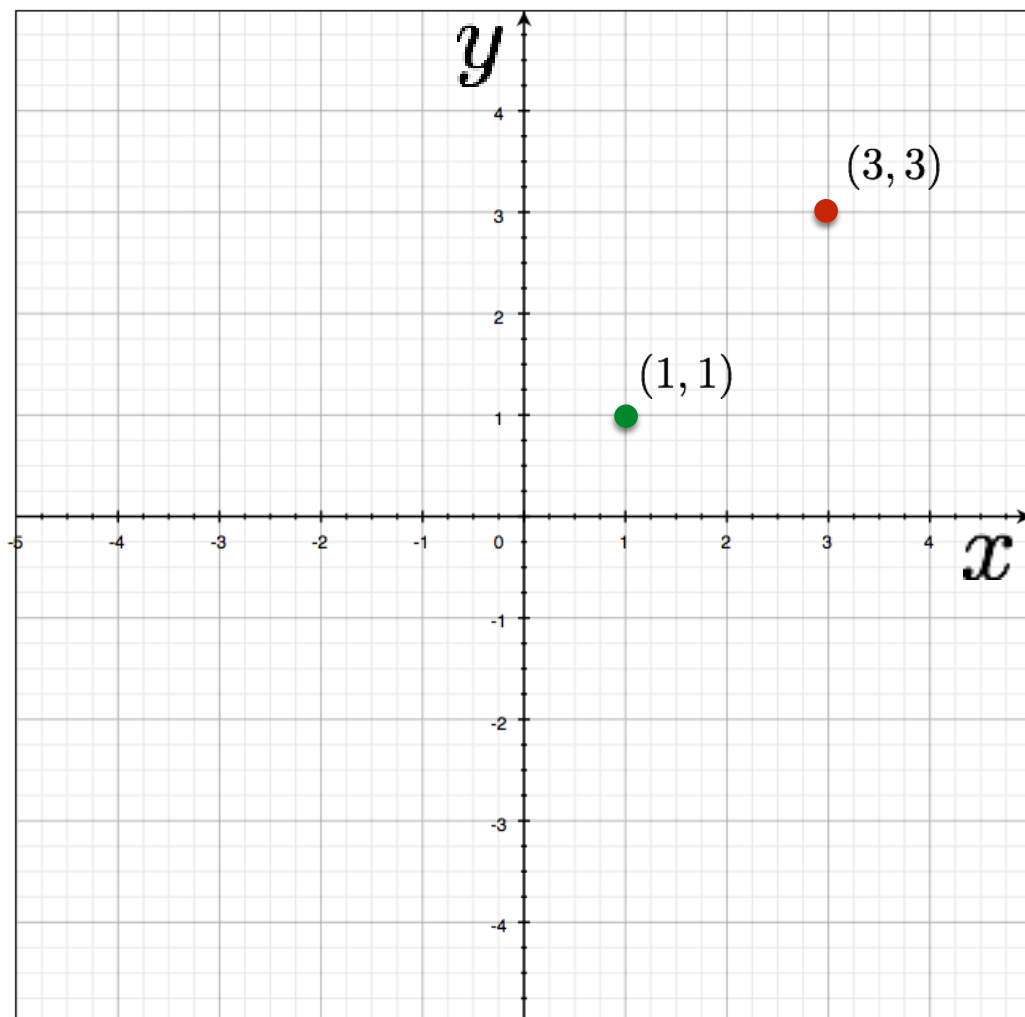
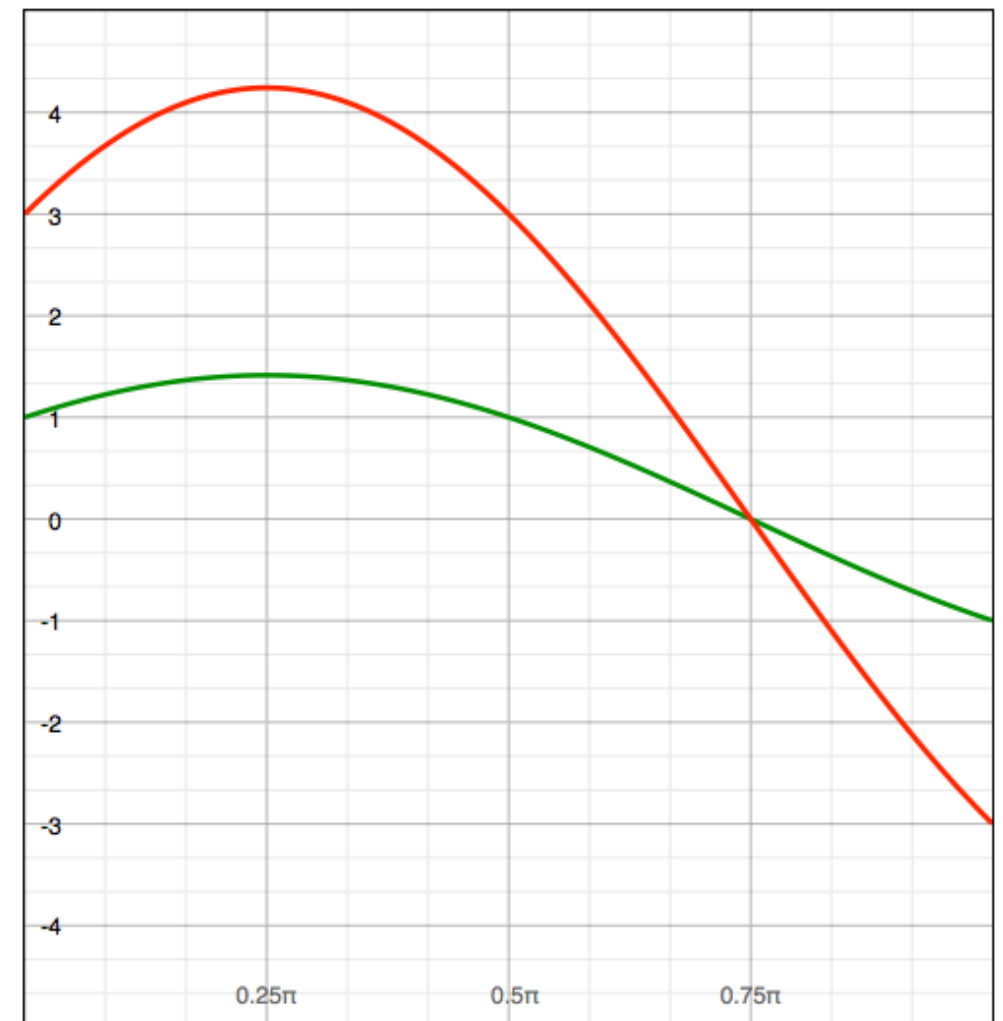


Image space

two points
become
?



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

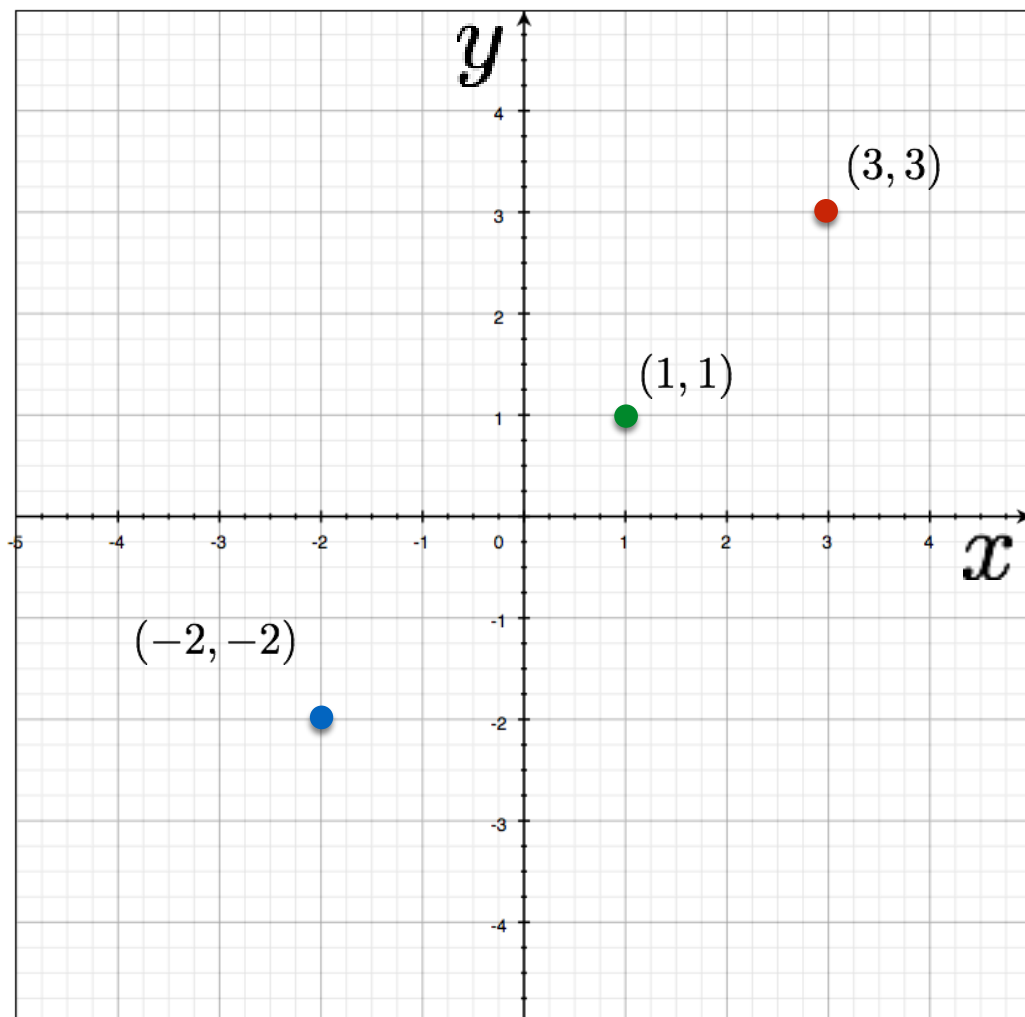
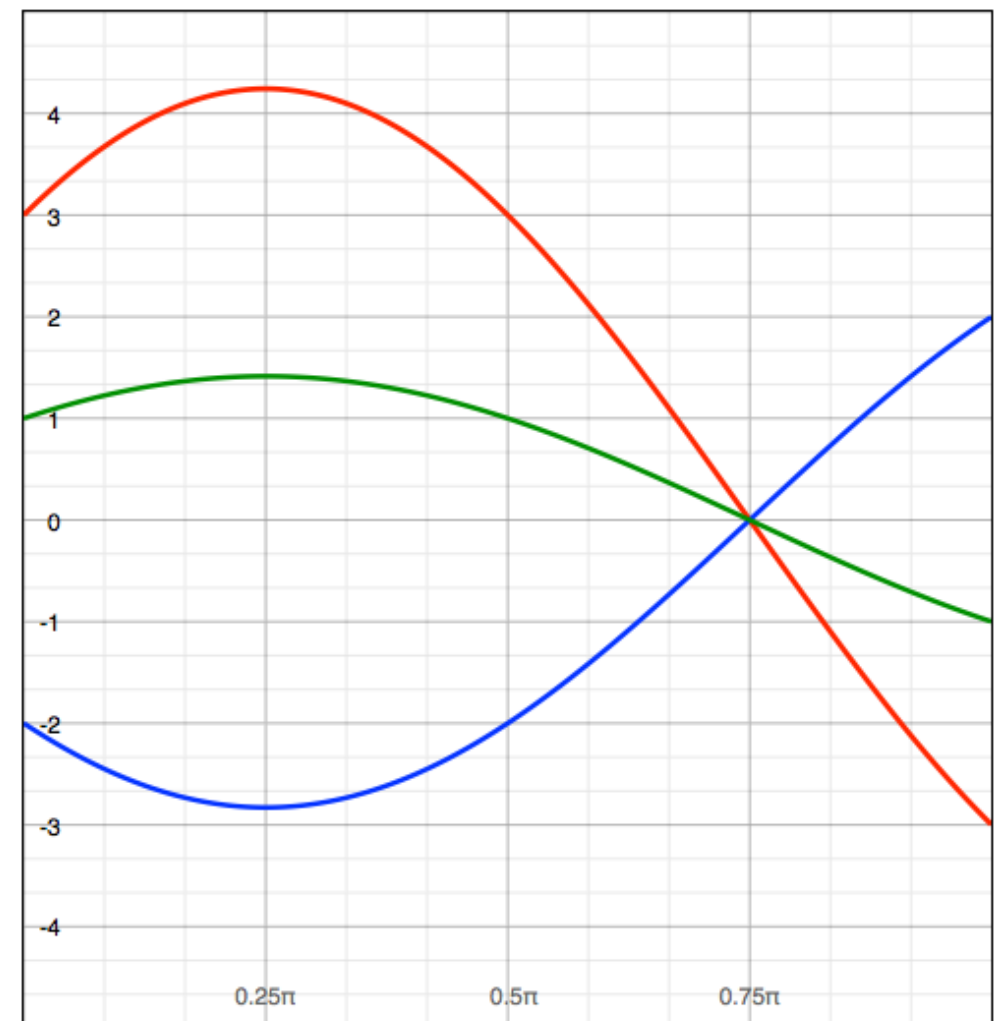


Image space

three points
become
?



Parameter space

Image and parameter space

variables

$$y = mx + b$$

parameters

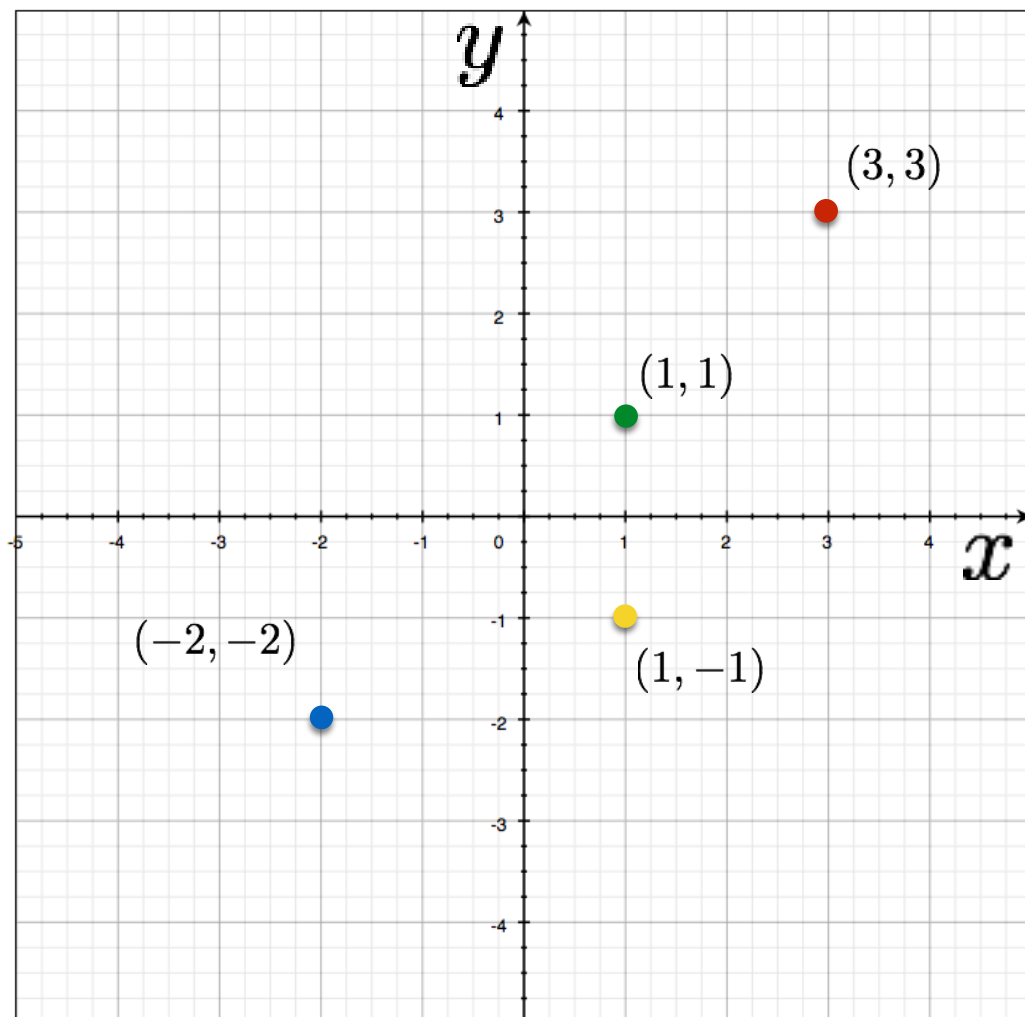
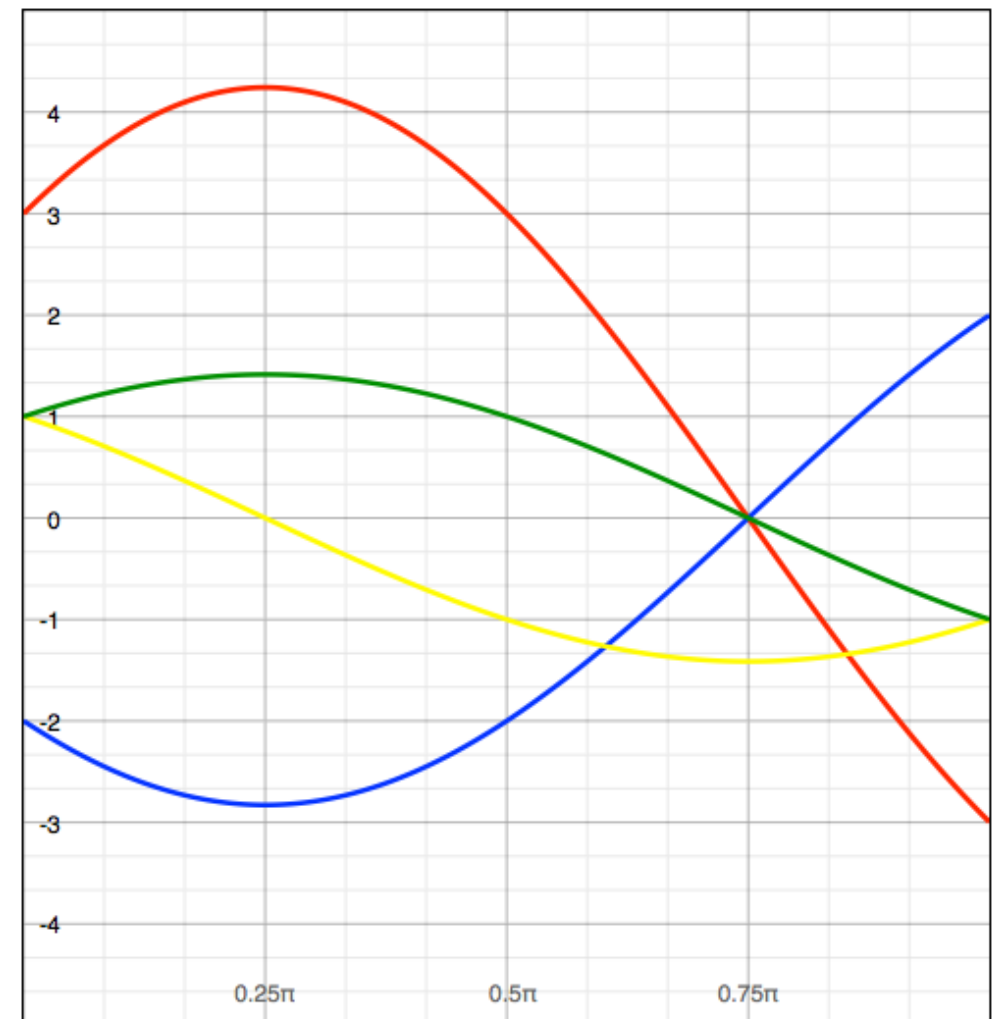


Image space

four points
become
?



Parameter space

Implementation

1. Initialize accumulator H to all zeros

2. For each edge point (x,y) in the image

For $\theta = 0$ to 180

$$\rho = x \cos \theta + y \sin \theta$$

$$H(\theta, \rho) = H(\theta, \rho) + 1$$

end

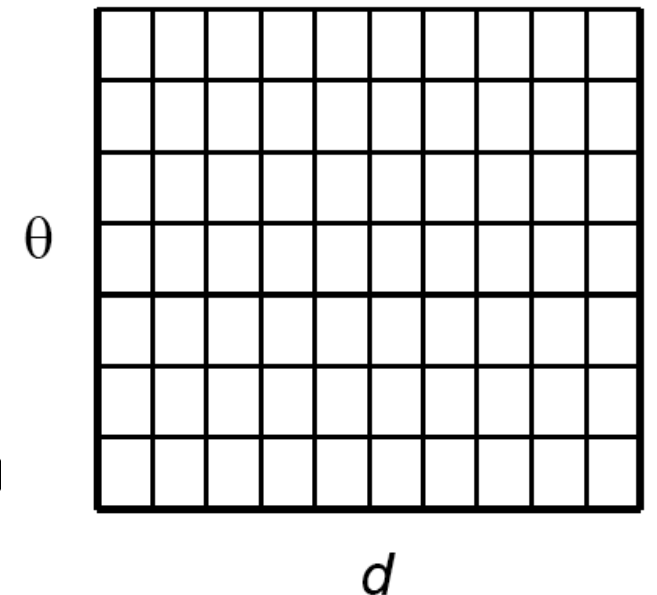
end

3. Find the value(s) of (θ, ρ) where $H(\theta, \rho)$ is a local maximum

4. The detected line in the image is given by

$$\rho = x \cos \theta + y \sin \theta$$

H: accumulator array (votes)



NOTE: Watch your coordinates. Image origin is top left!

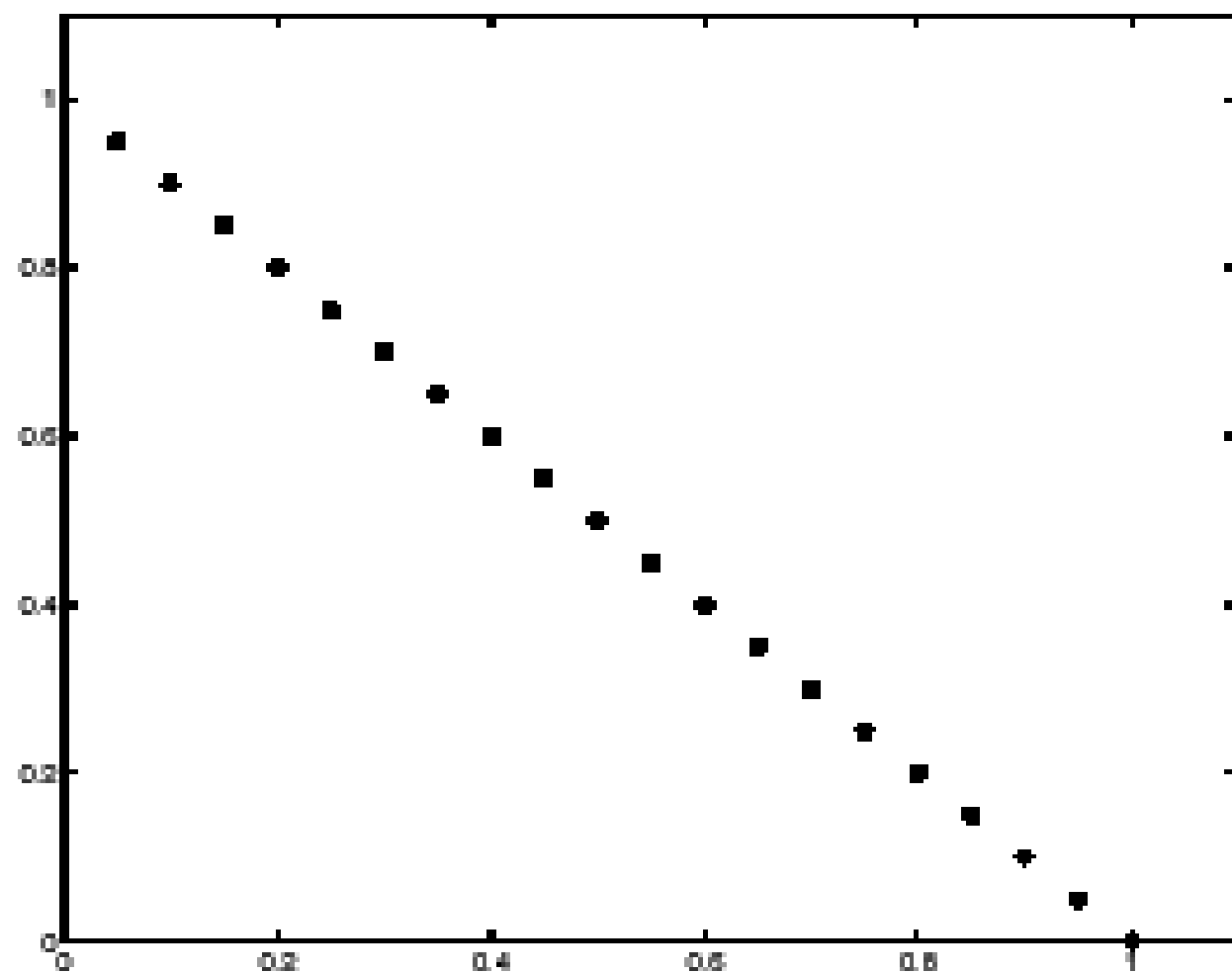
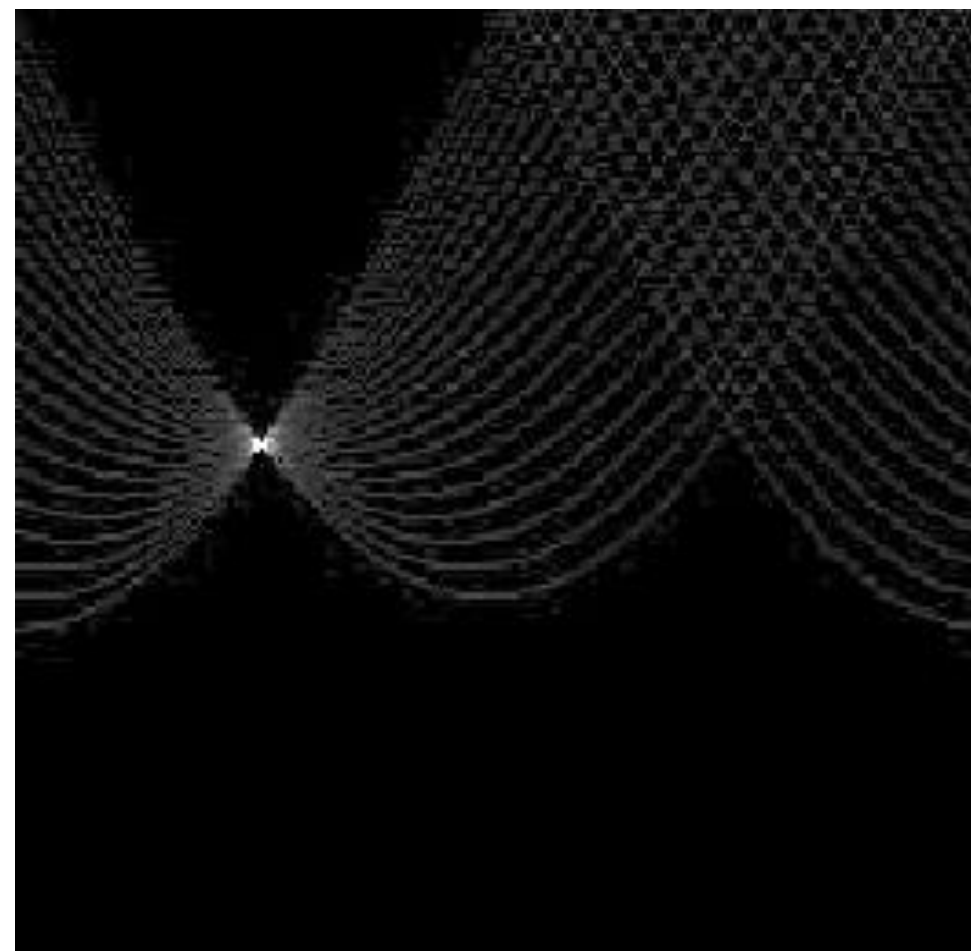


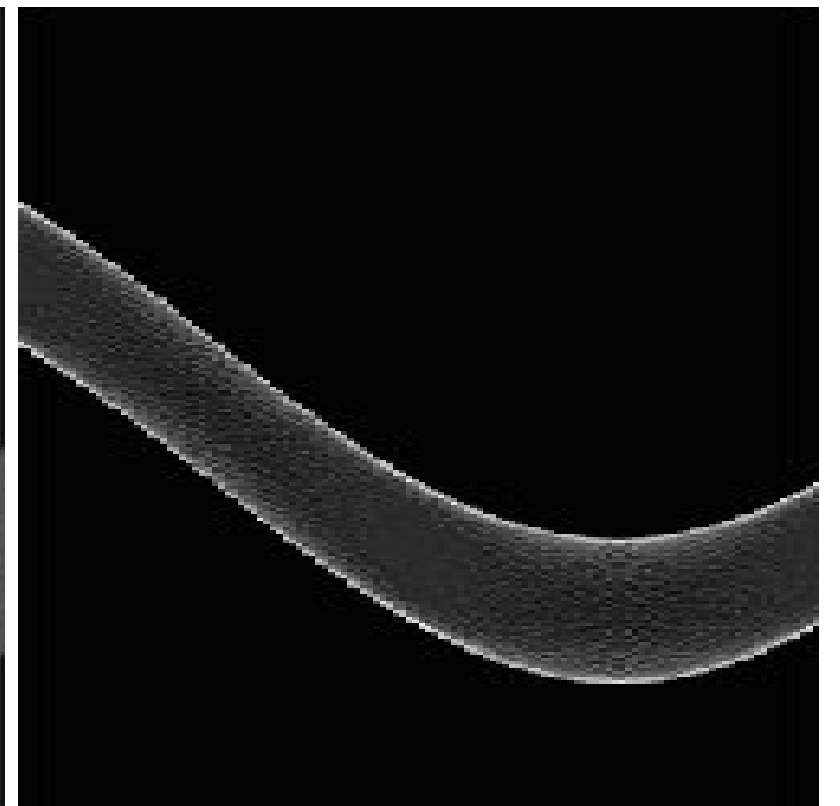
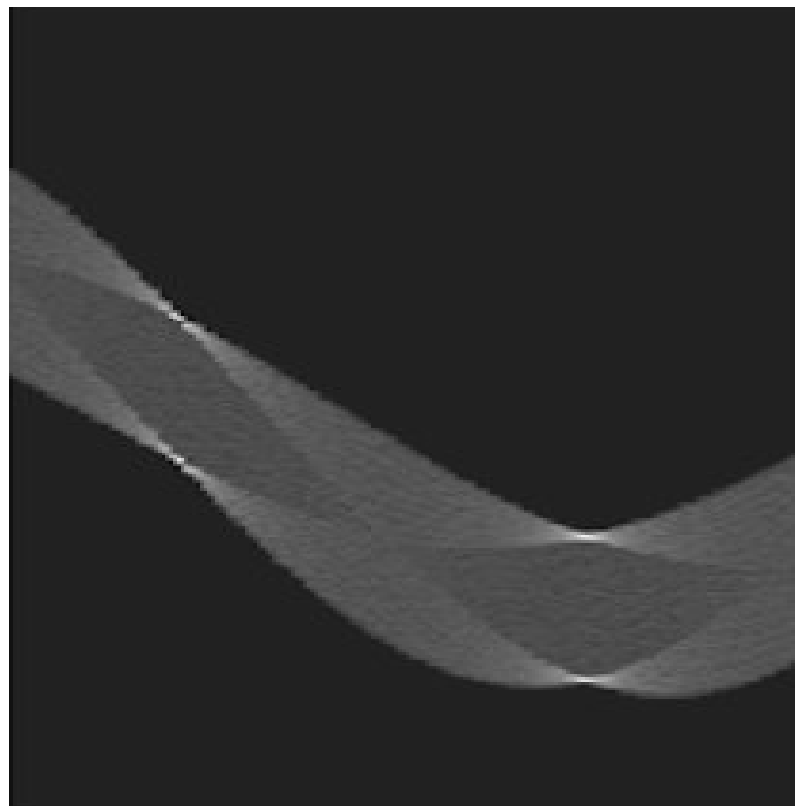
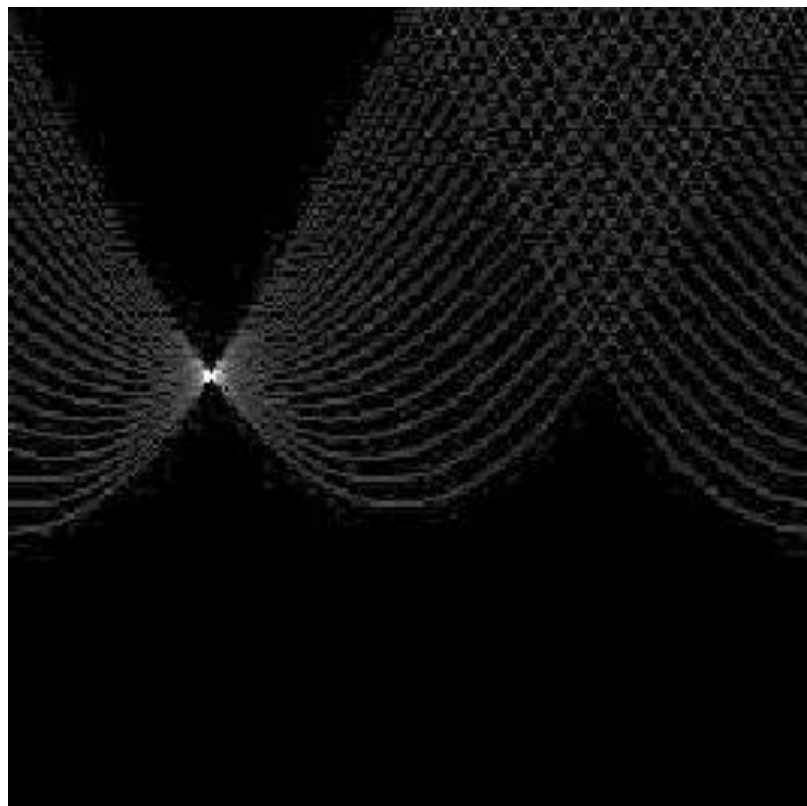
Image space



Votes

Basic shapes

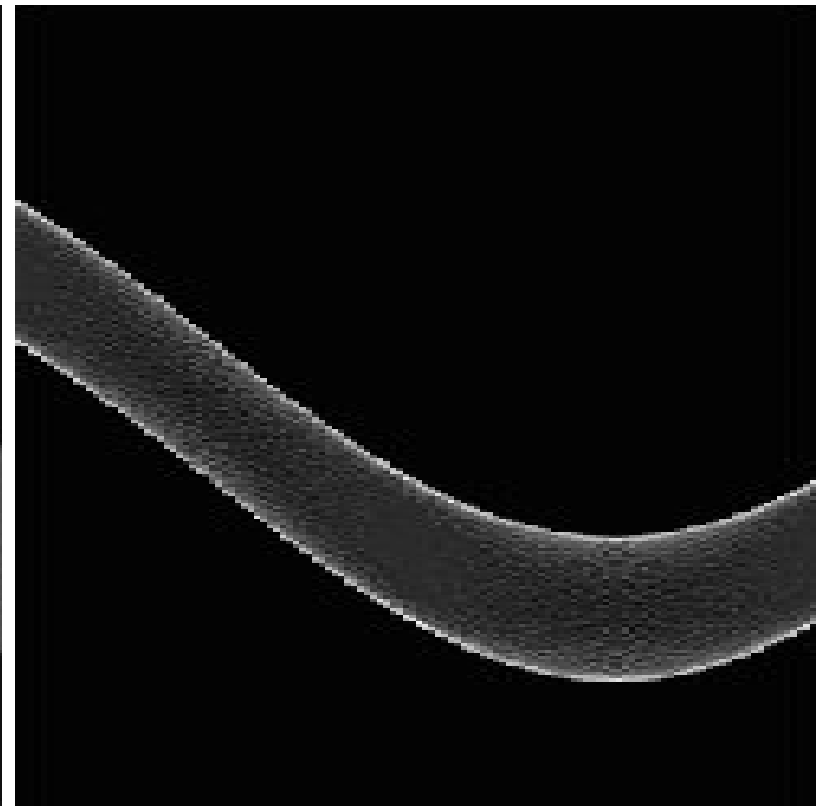
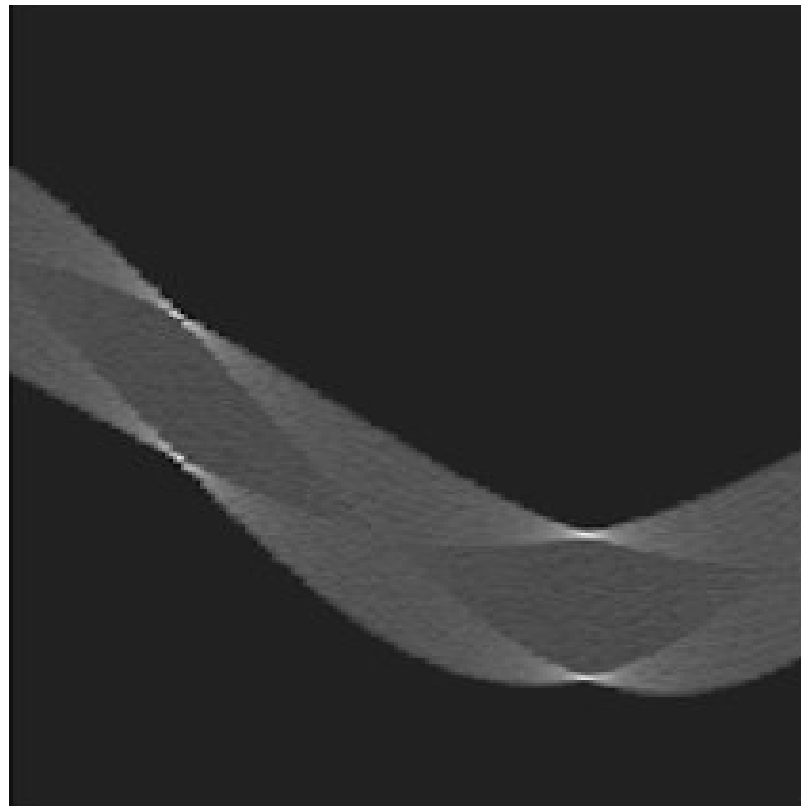
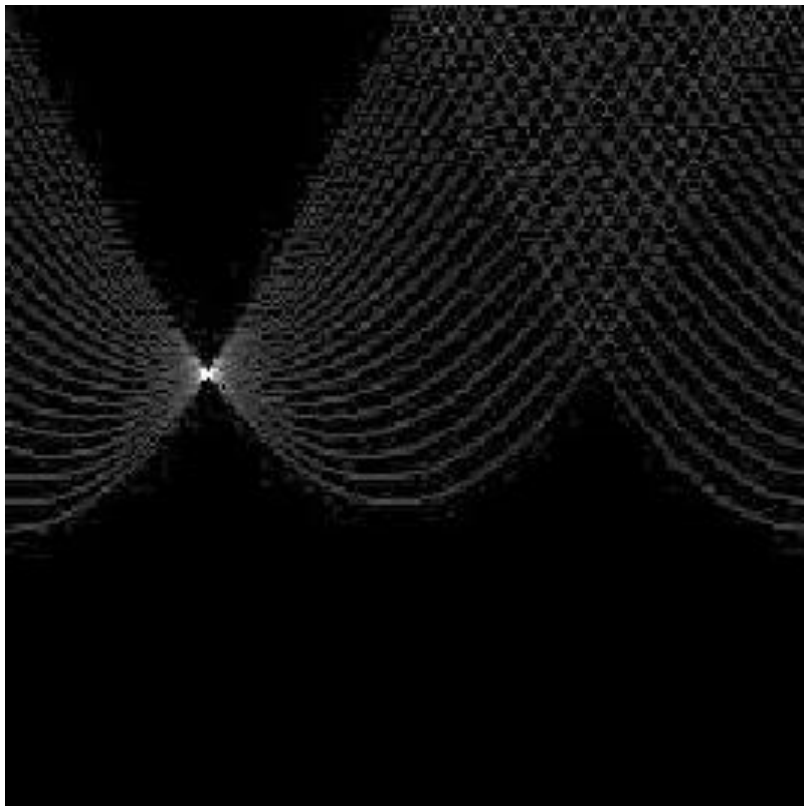
(in parameter space)



can you guess the shape?

Basic shapes

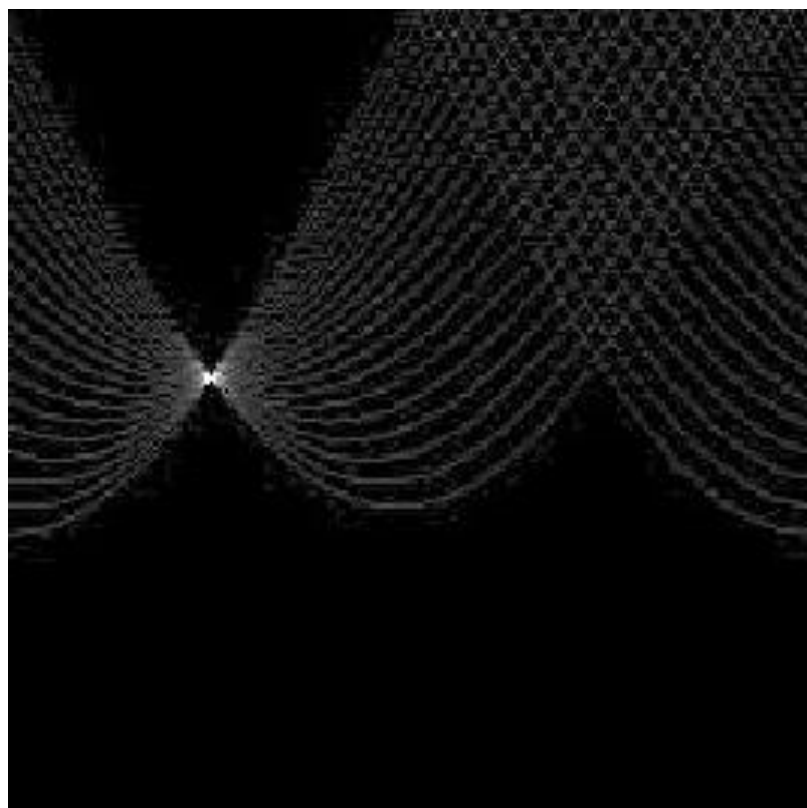
(in parameter space)



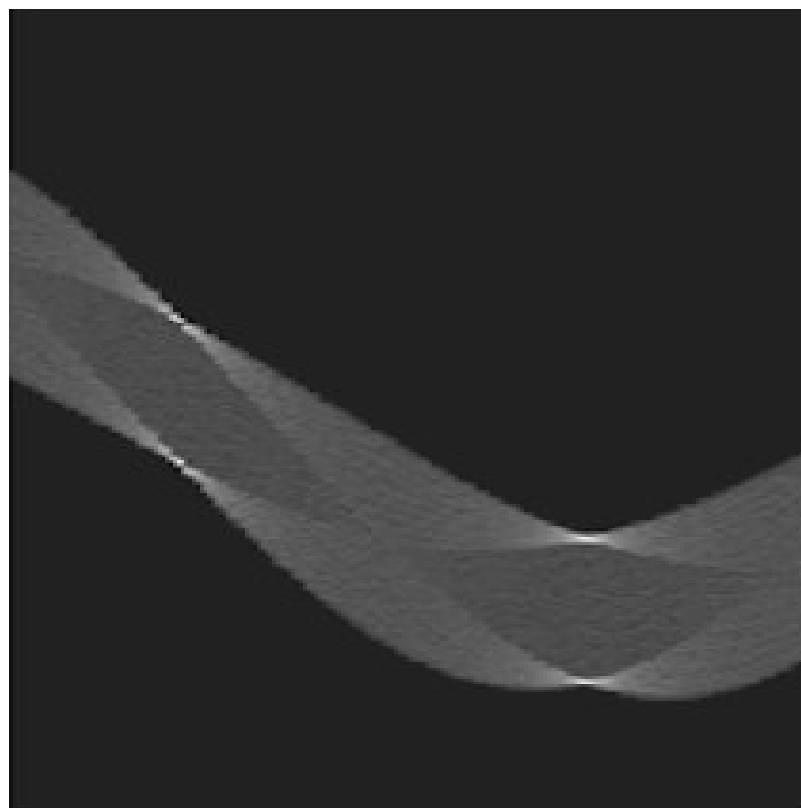
line

Basic shapes

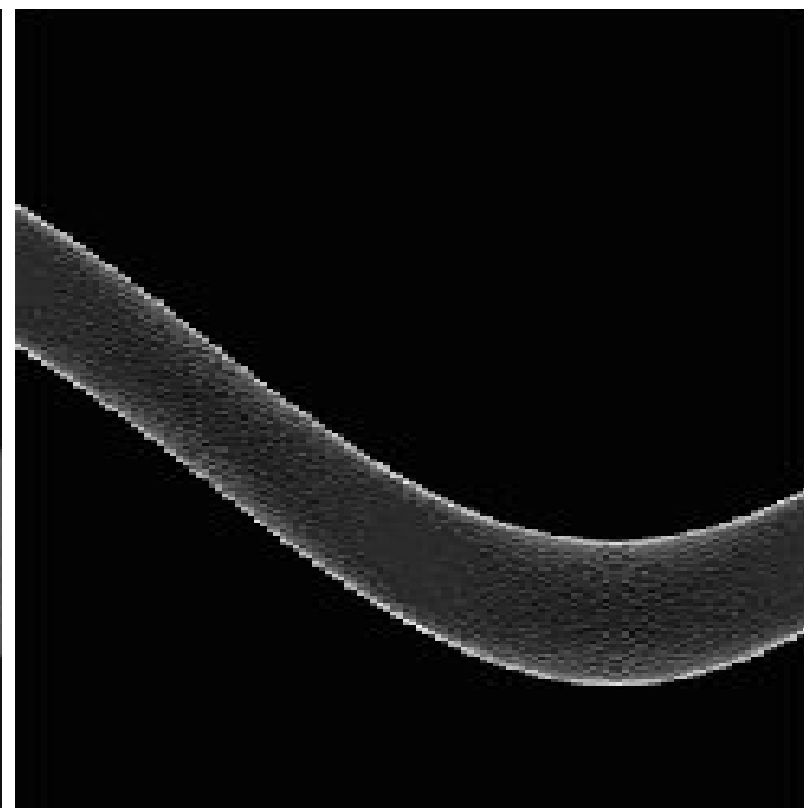
(in parameter space)



line

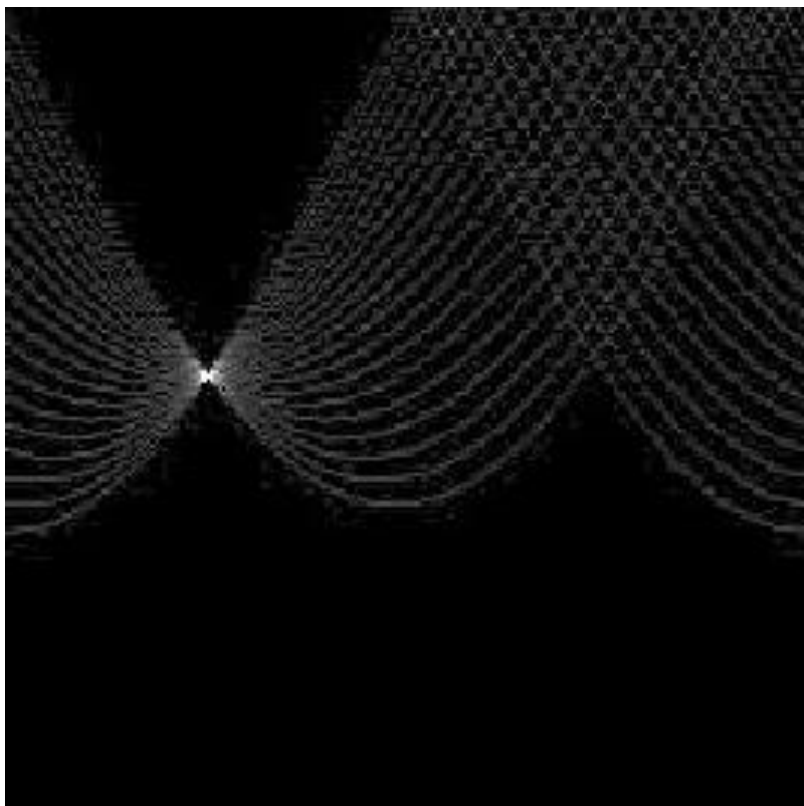


rectangle

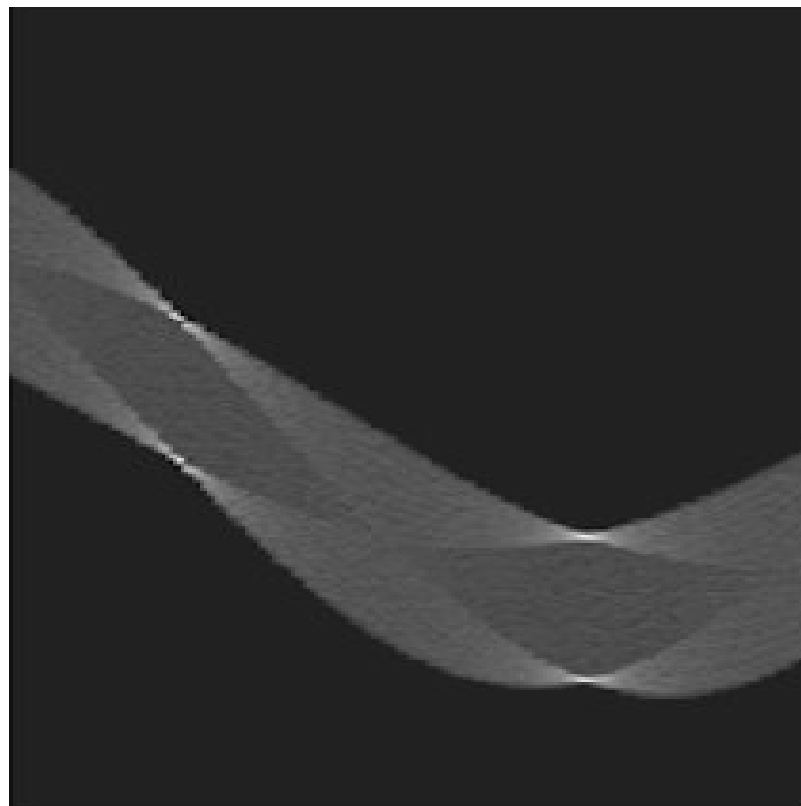


Basic shapes

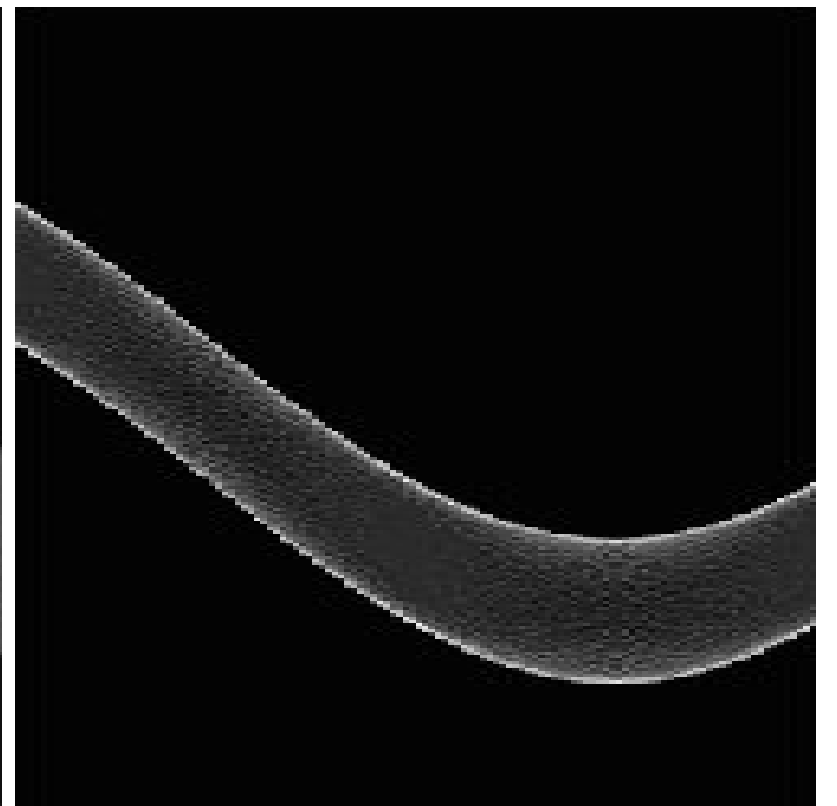
(in parameter space)



line

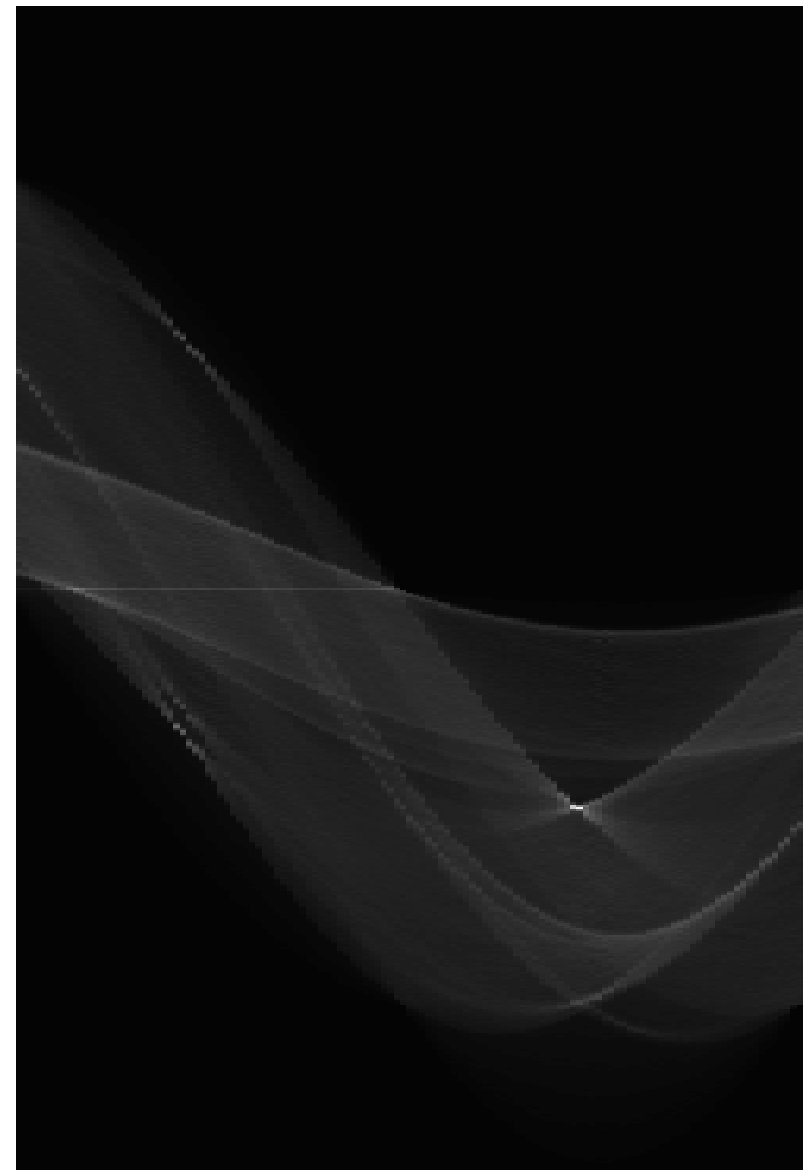
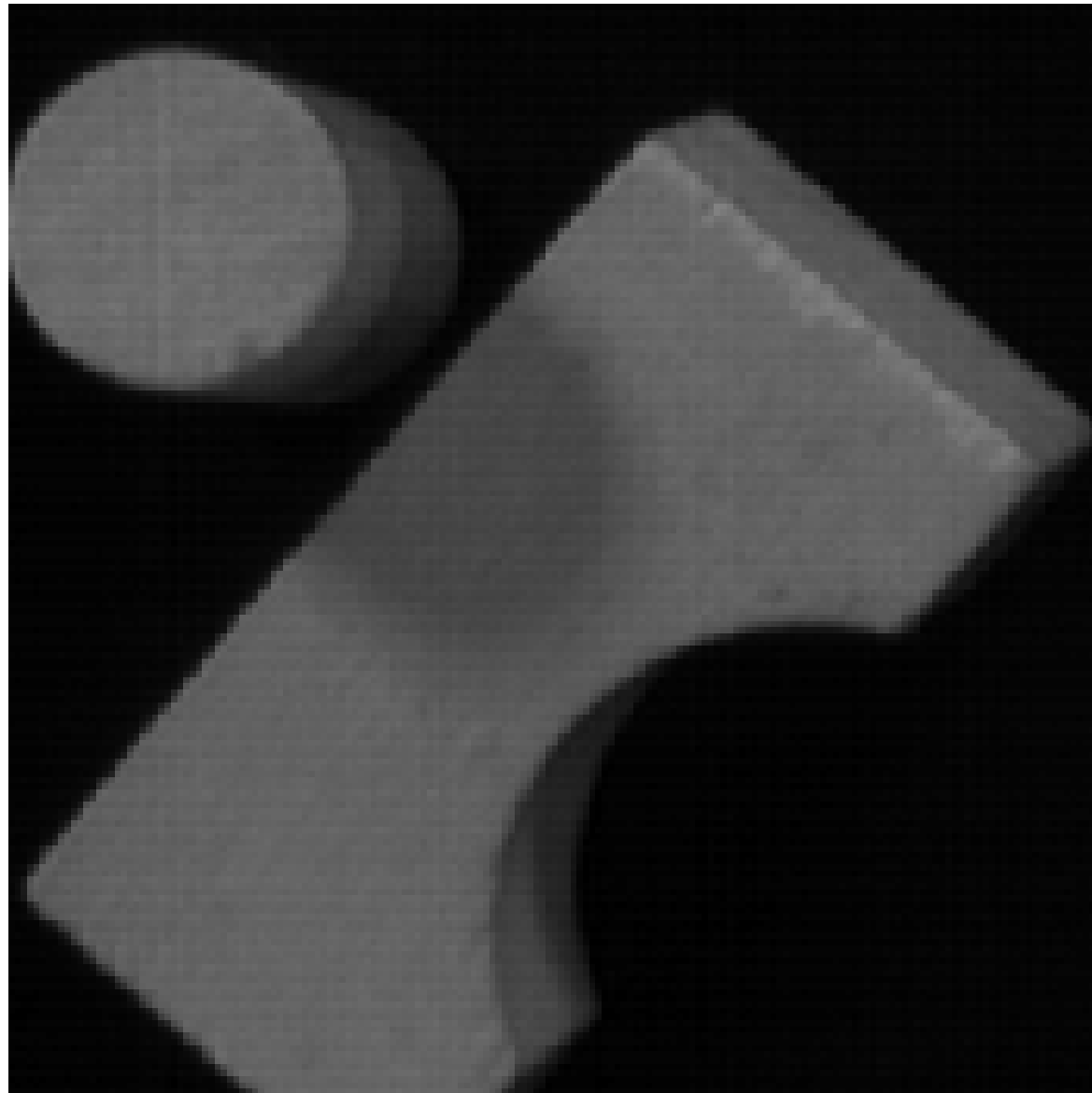


rectangle

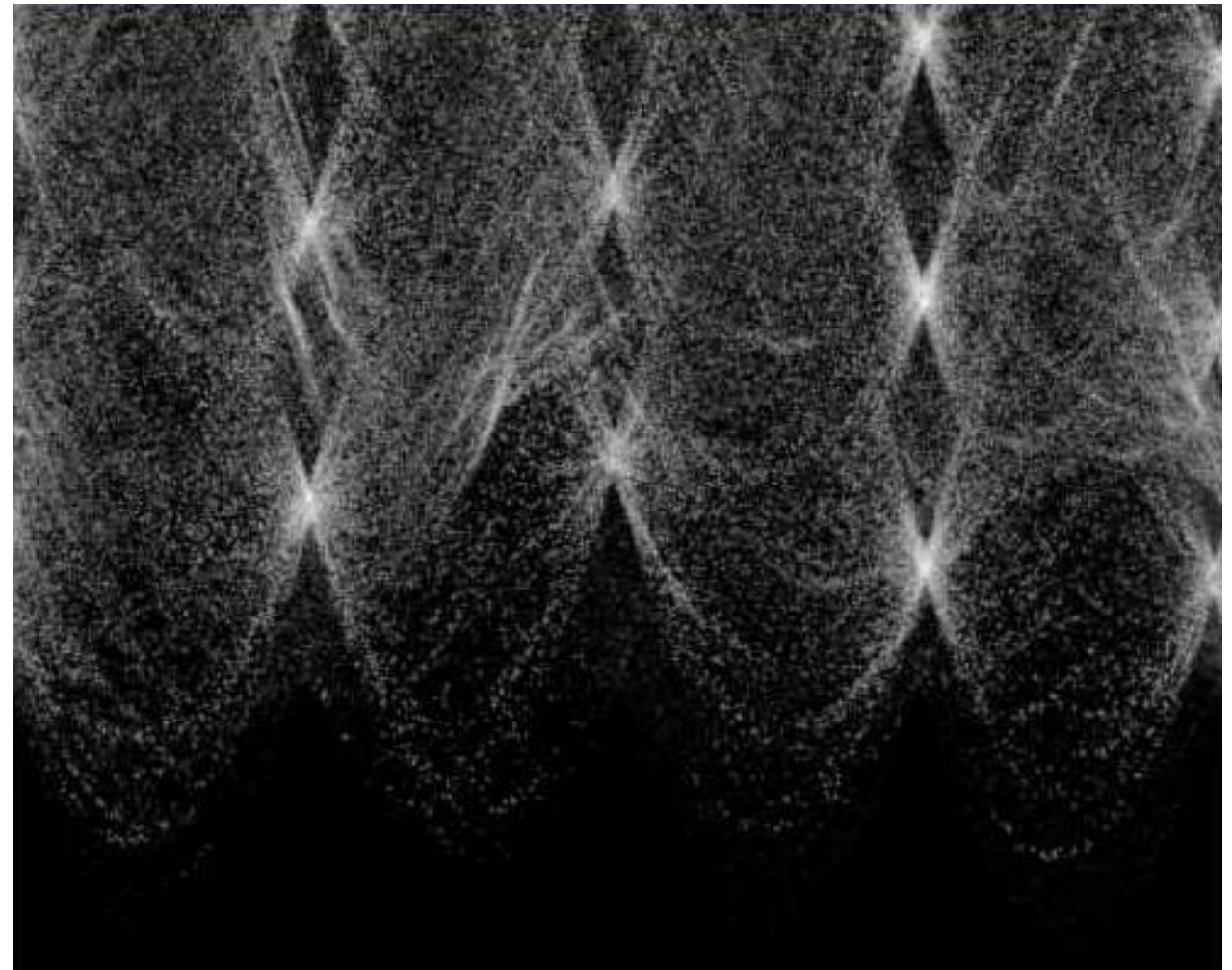


circle

Basic Shapes



More complex image



In practice, measurements are noisy...

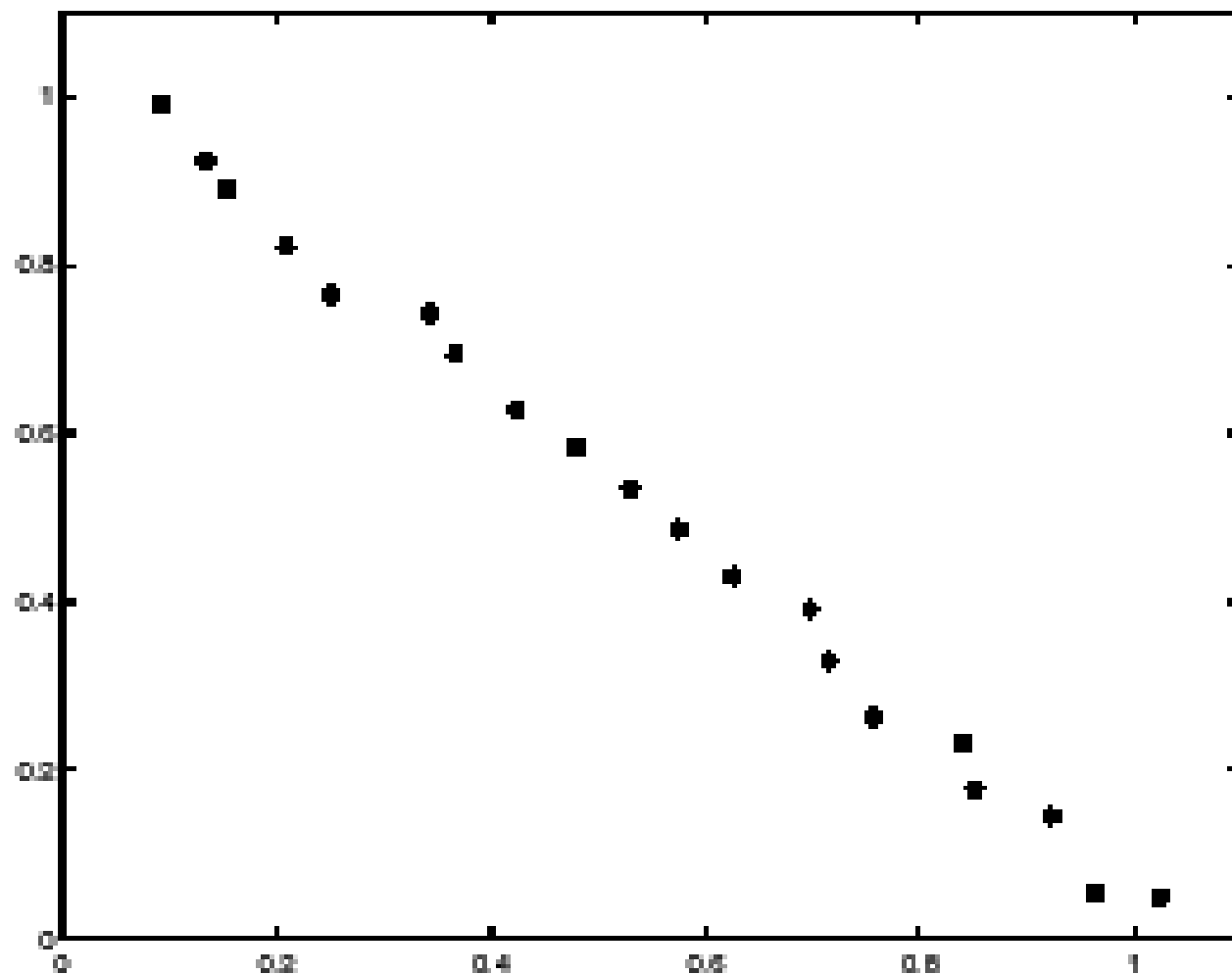
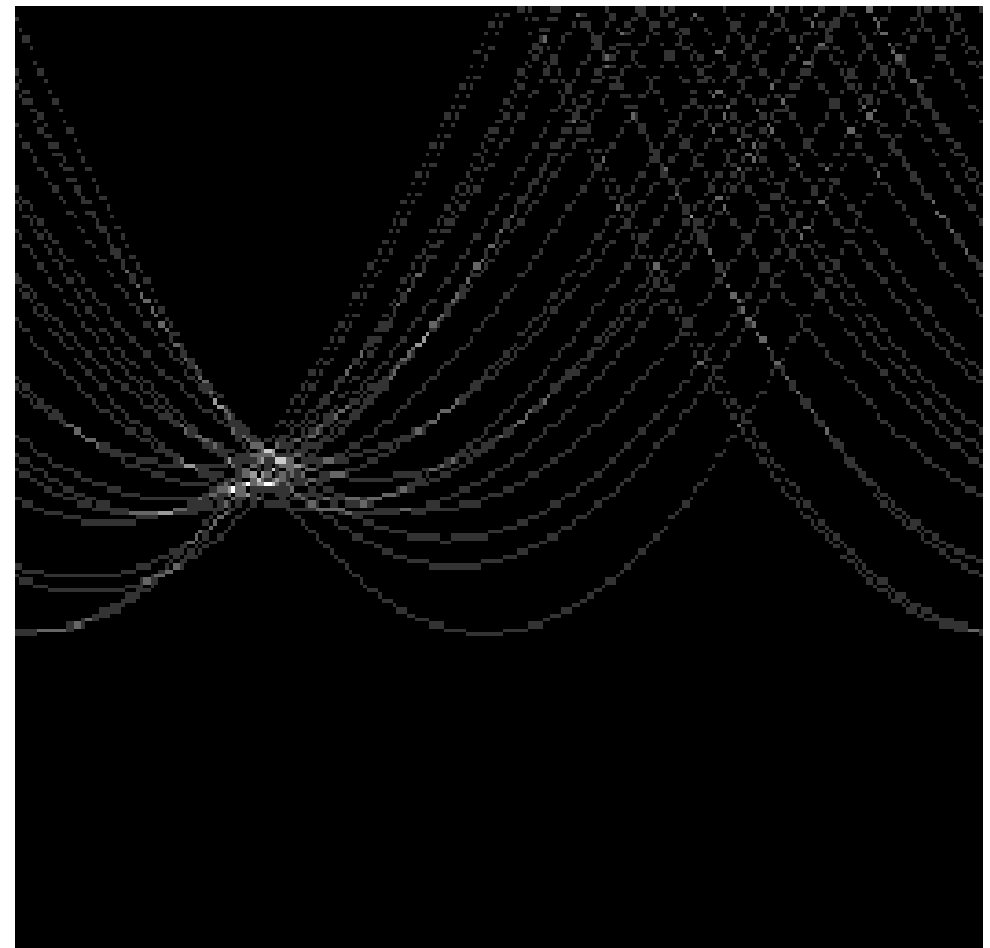


Image space



Votes

Too much noise ...

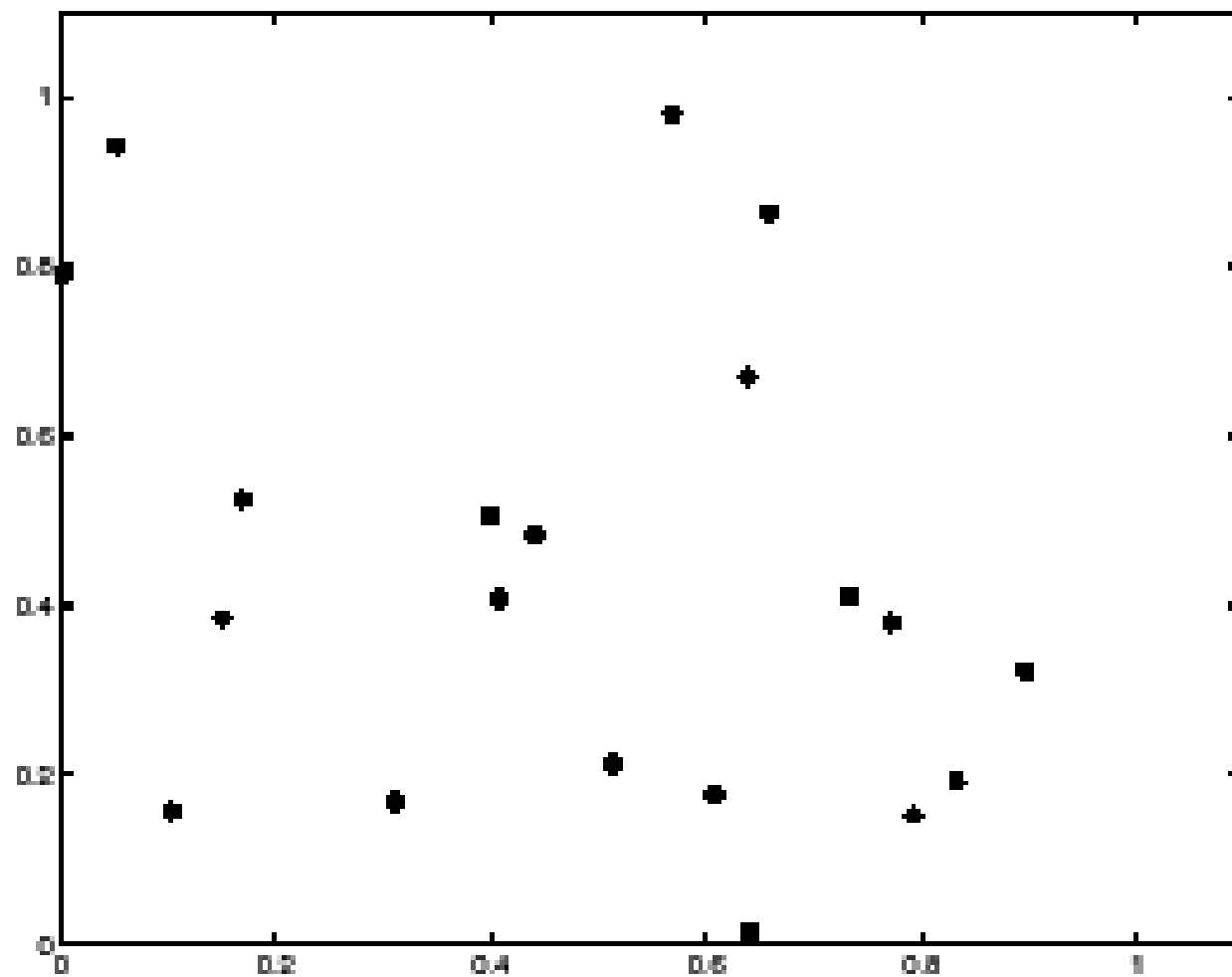
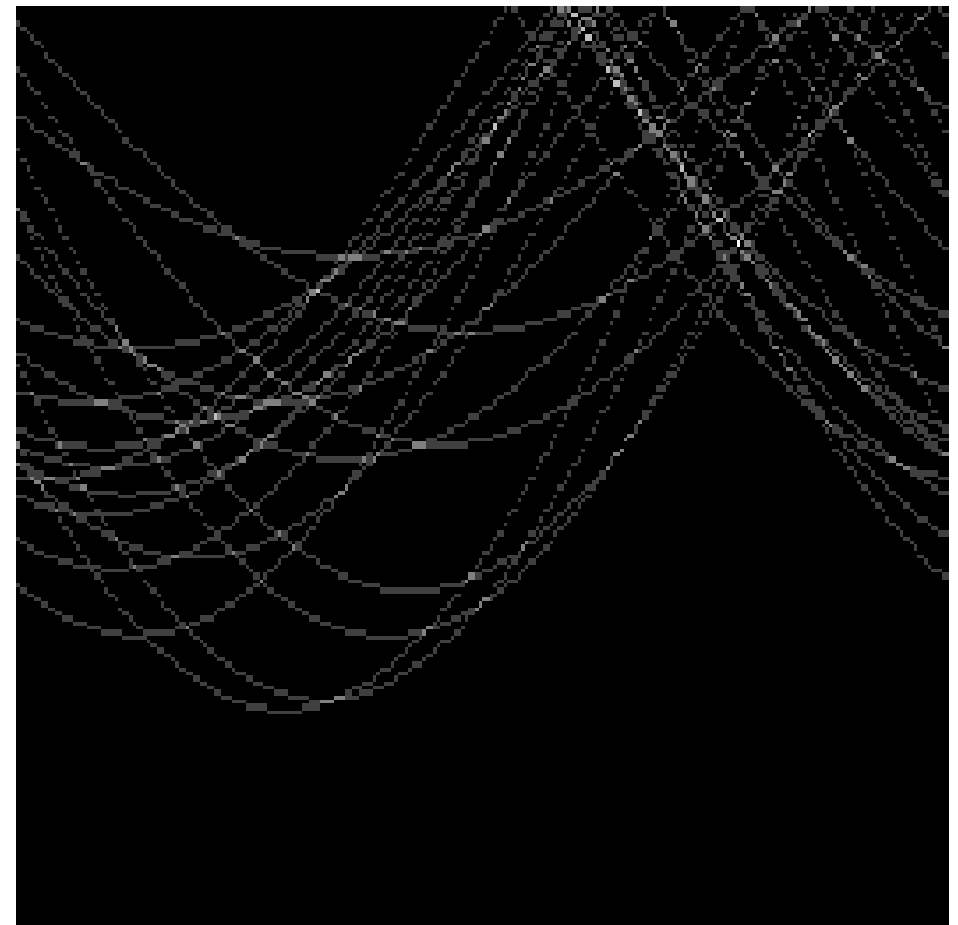


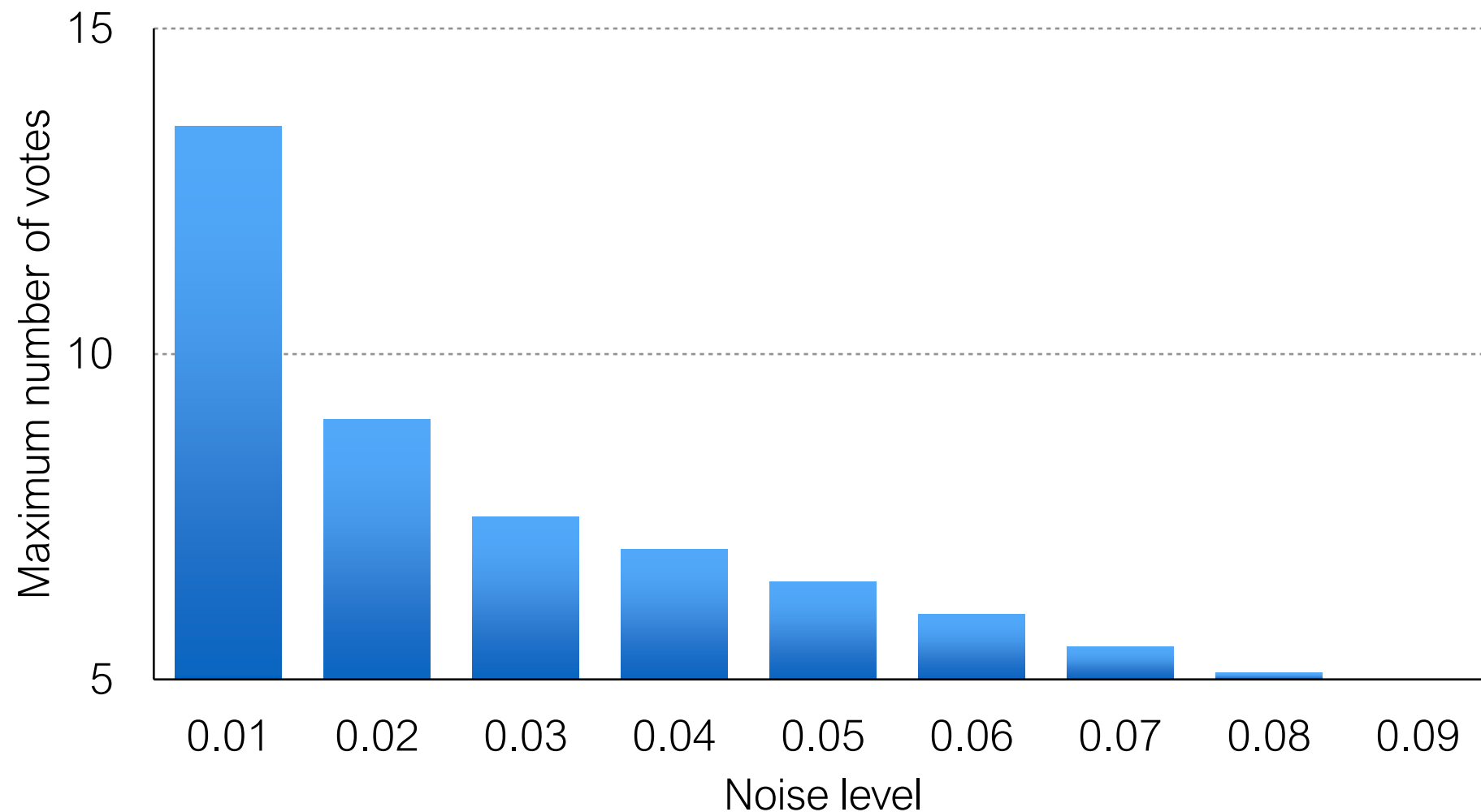
Image space



Votes

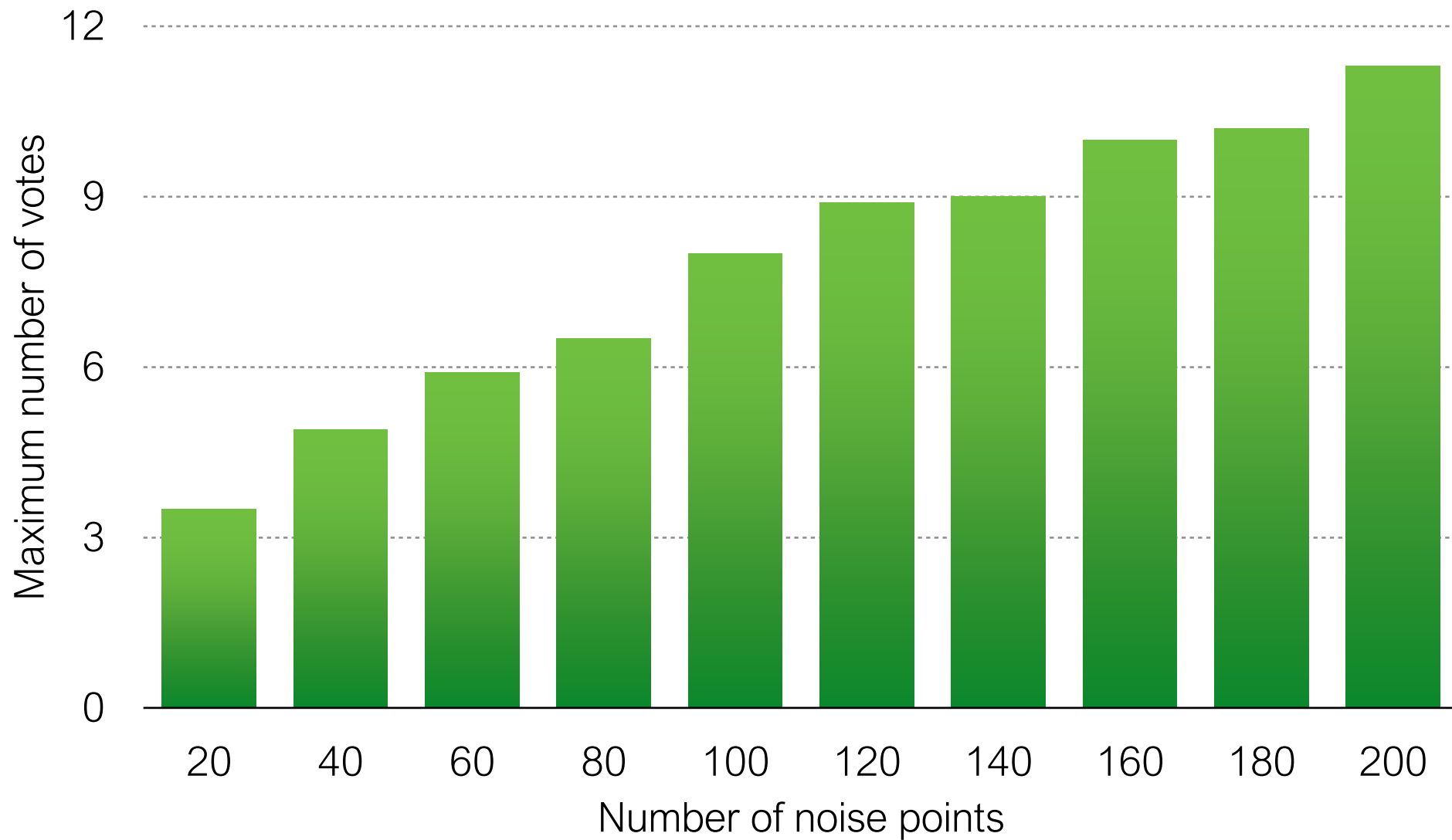
Effects of noise level

Number of votes for a line of 20 points with increasing noise



More noise, fewer votes (in the right bin)

Effect of noise points

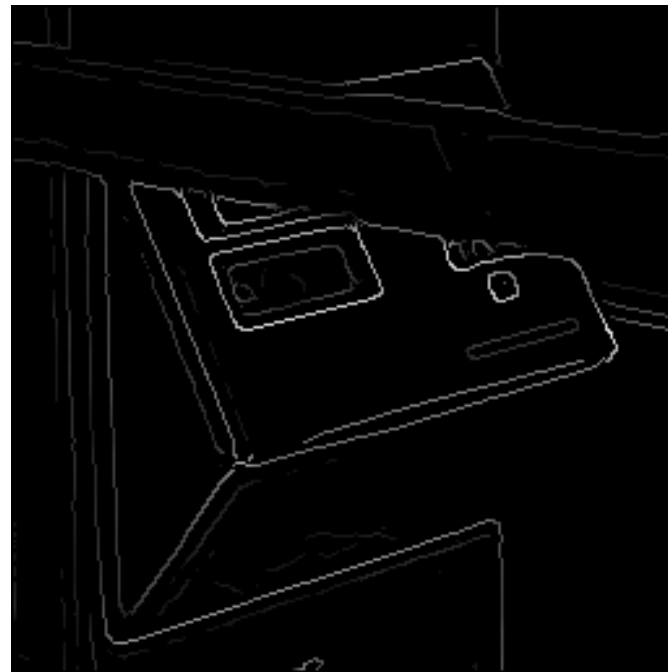


More noise, more votes (in the wrong bin)

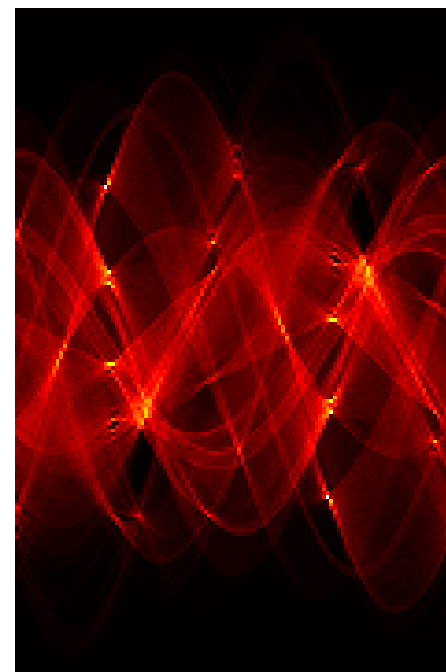
Real-world example



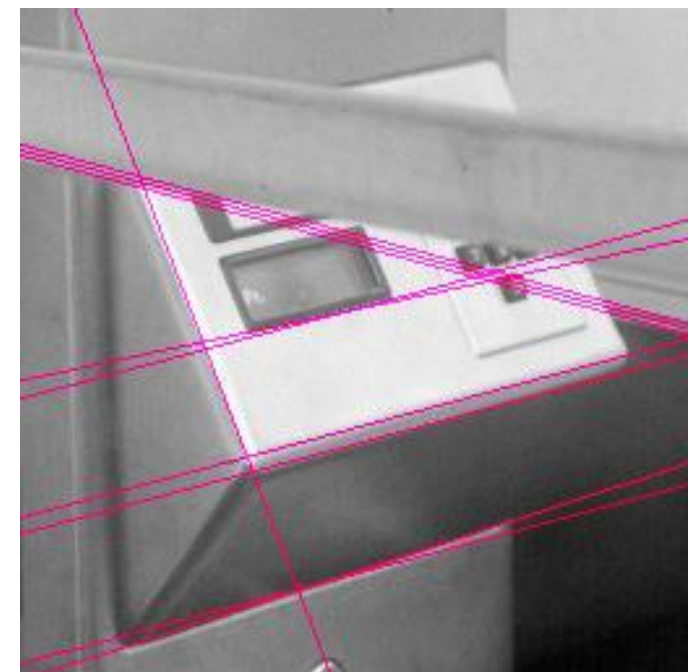
Original



Edges



parameter space



Hough Lines

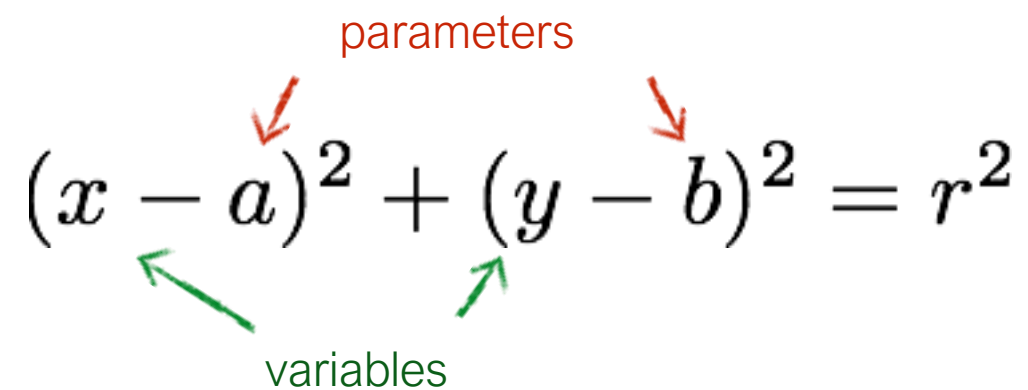
Hough Circles

Let's assume radius known

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

parameters

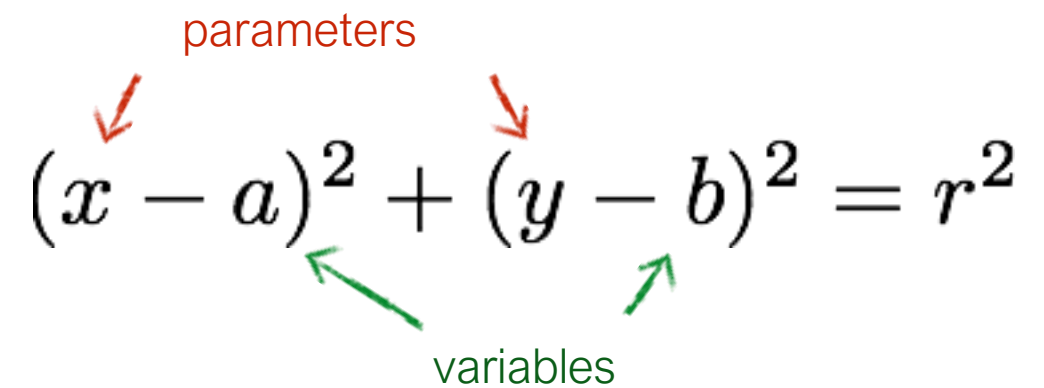
variables

A diagram of the equation (x - a)^2 + (y - b)^2 = r^2. Two red arrows point from the word 'parameters' to the variables 'a' and 'b'. Two green arrows point from the word 'variables' to the variables 'x' and 'y'.

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

parameters

variables

A diagram of the equation (x - a)^2 + (y - b)^2 = r^2. Two red arrows point from the word 'parameters' to the variables 'a' and 'b'. Two green arrows point from the word 'variables' to the variables 'x' and 'y'.

What is the dimension of the parameter space?

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

parameters

variables

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

parameters

variables

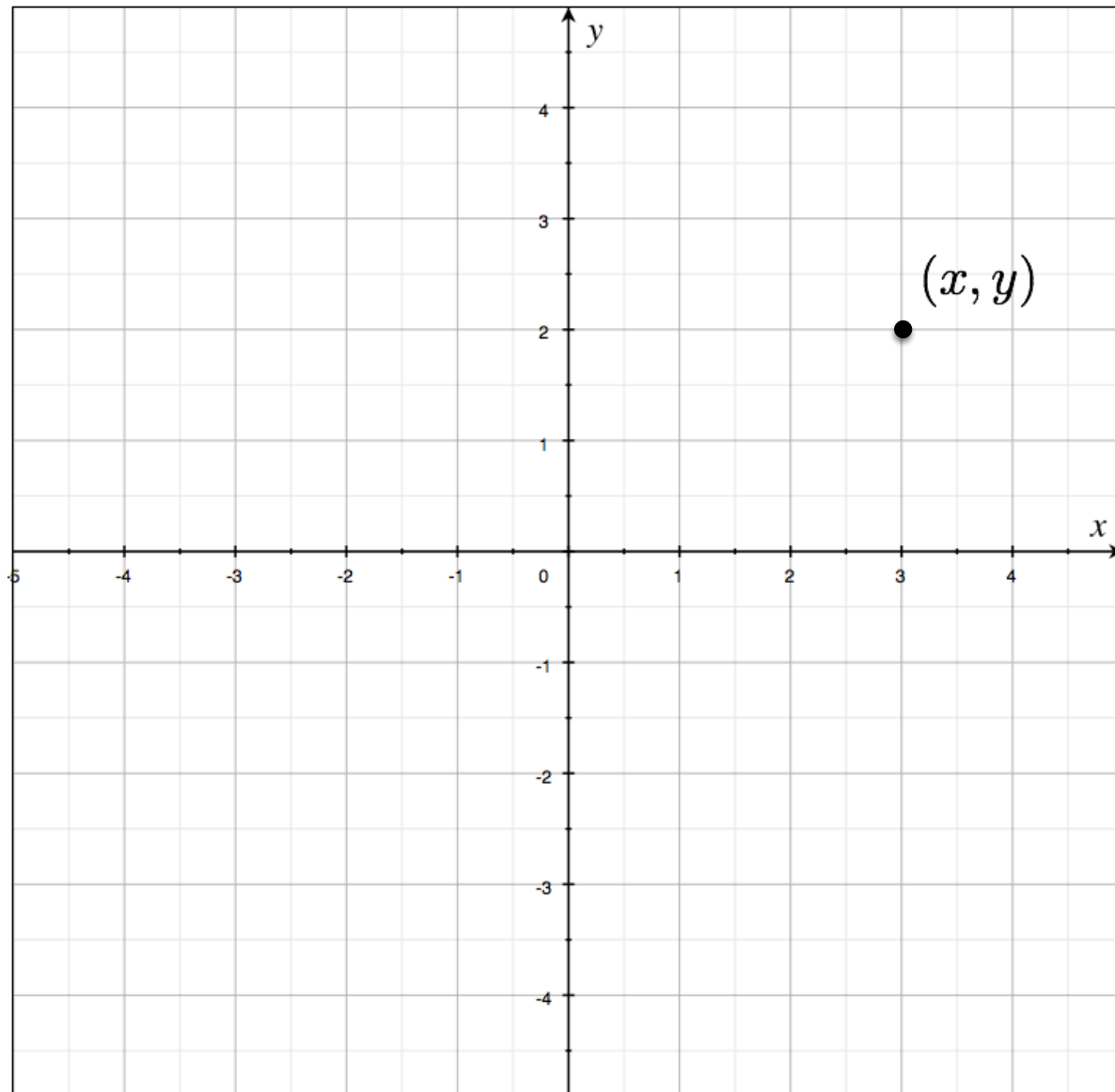
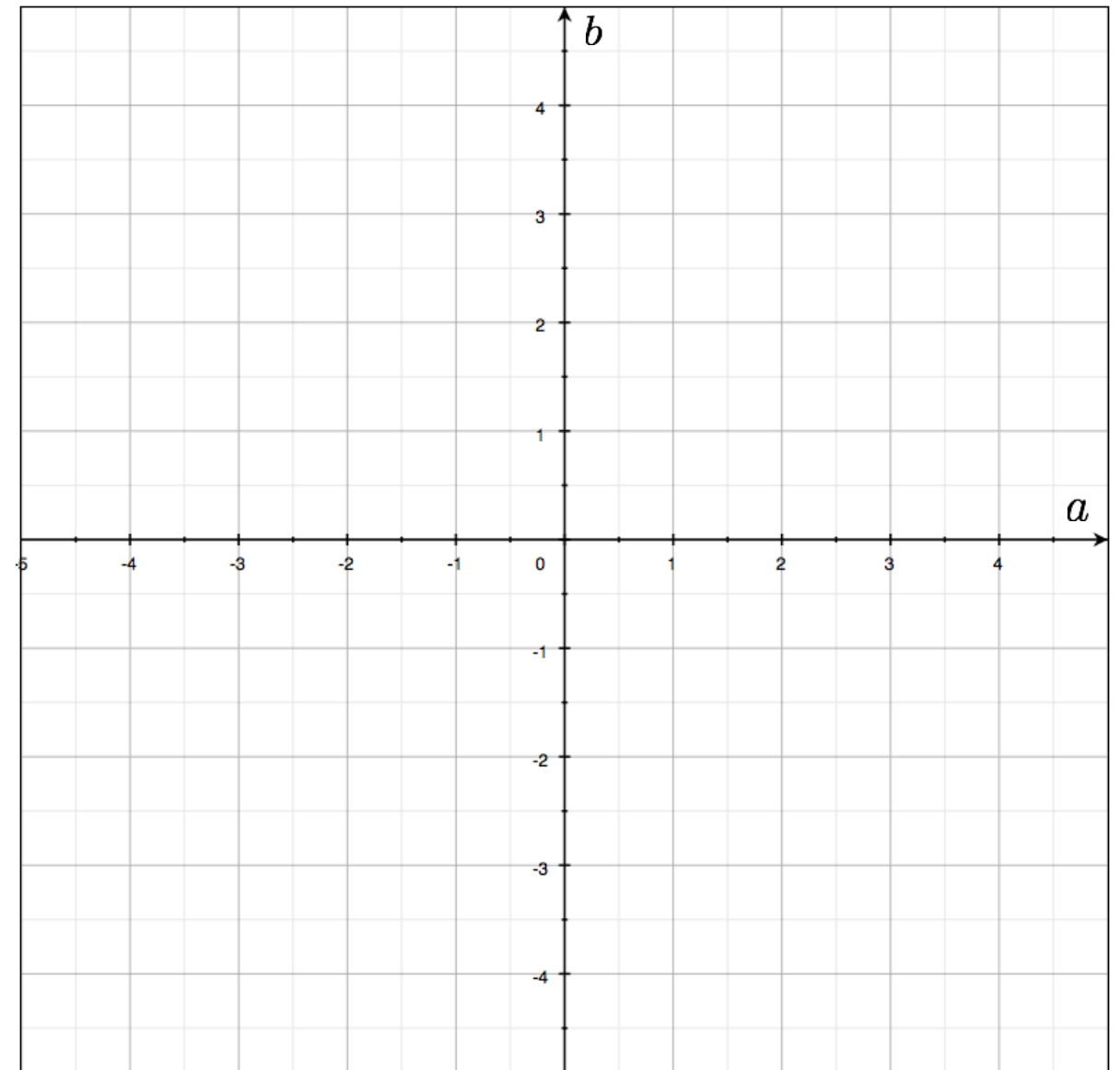


Image space



Parameter space

What does a point in image space correspond to in parameter space?

parameters

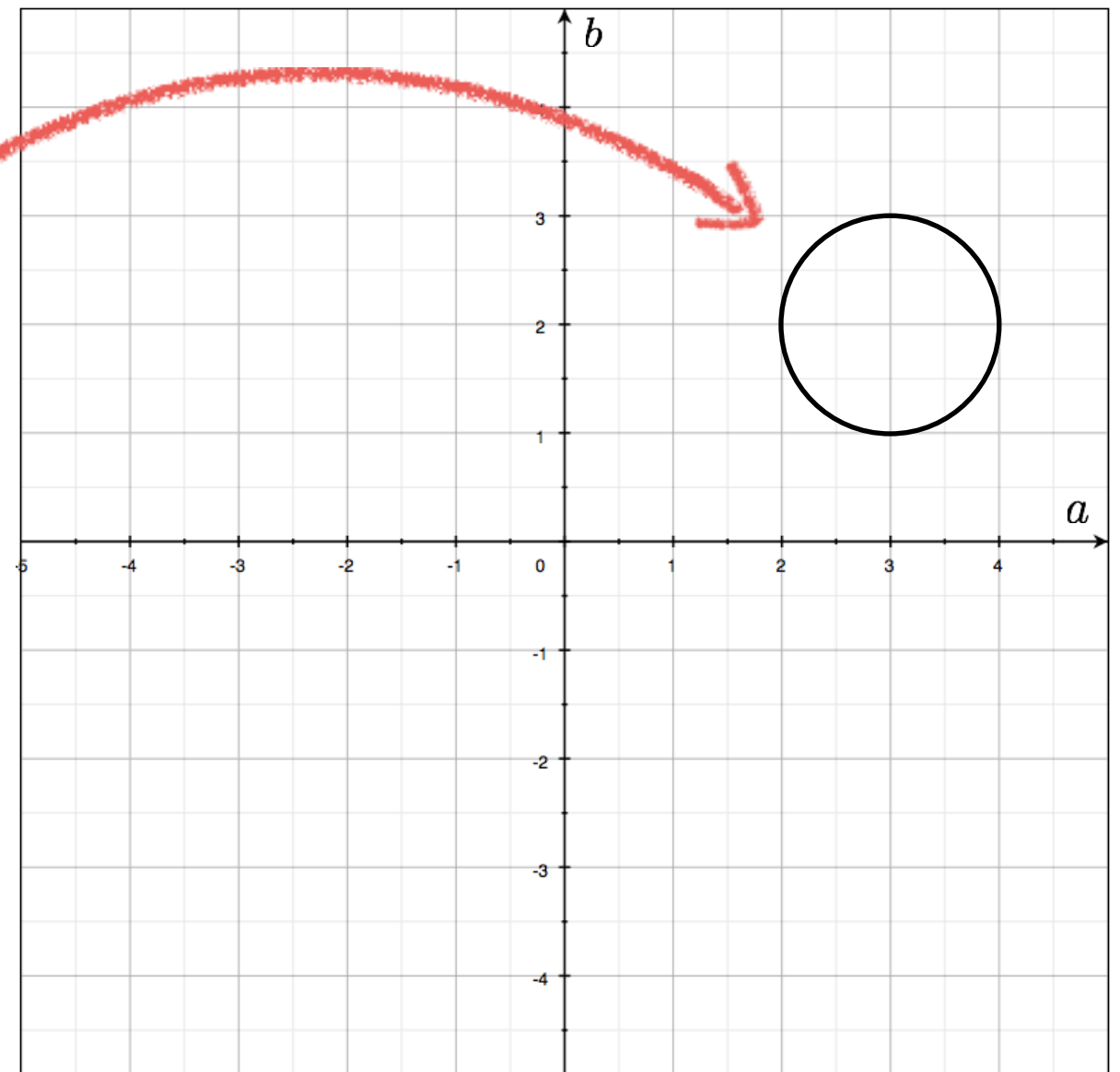
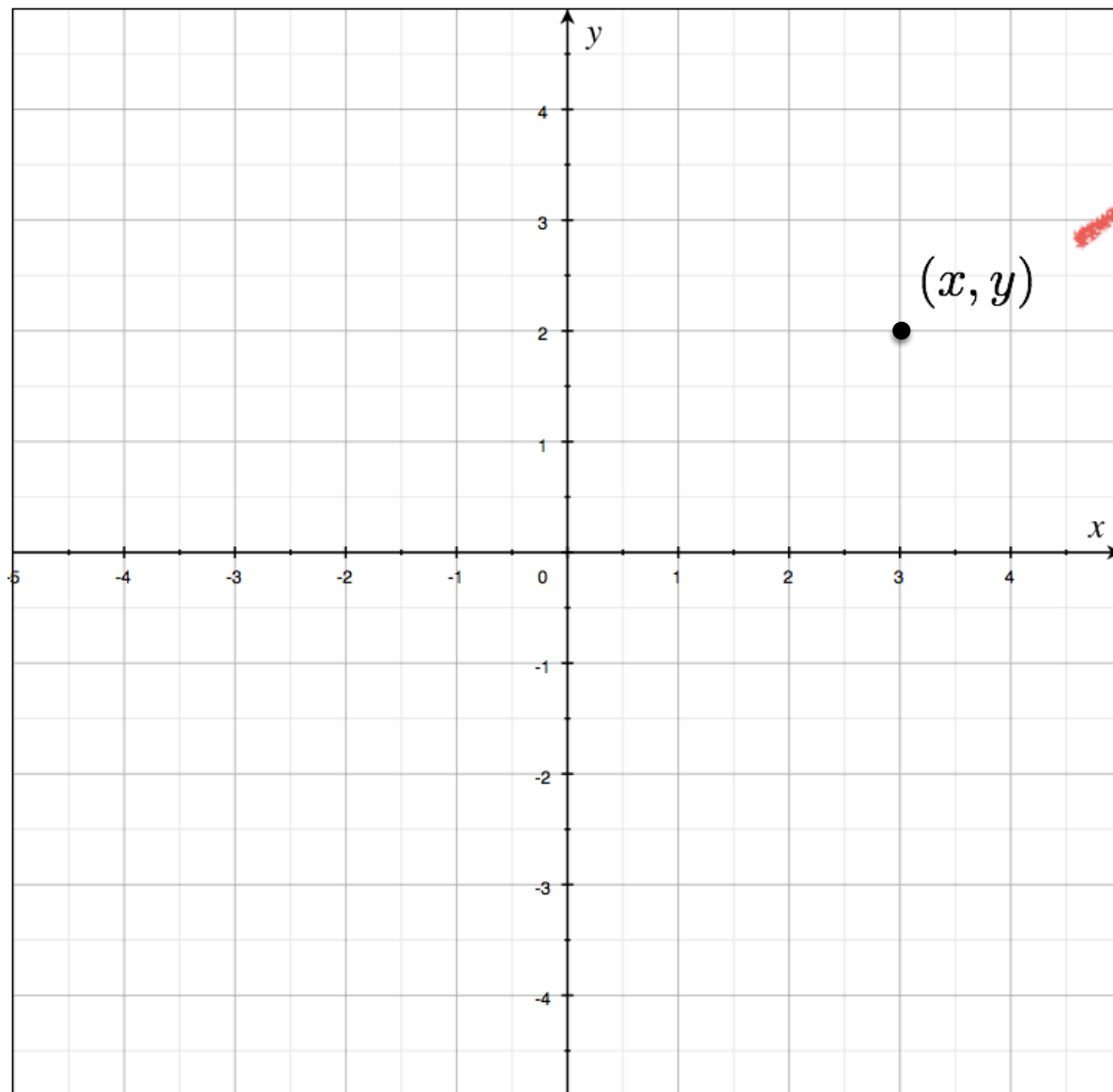
$$(x - a)^2 + (y - b)^2 = r^2$$

variables

parameters

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

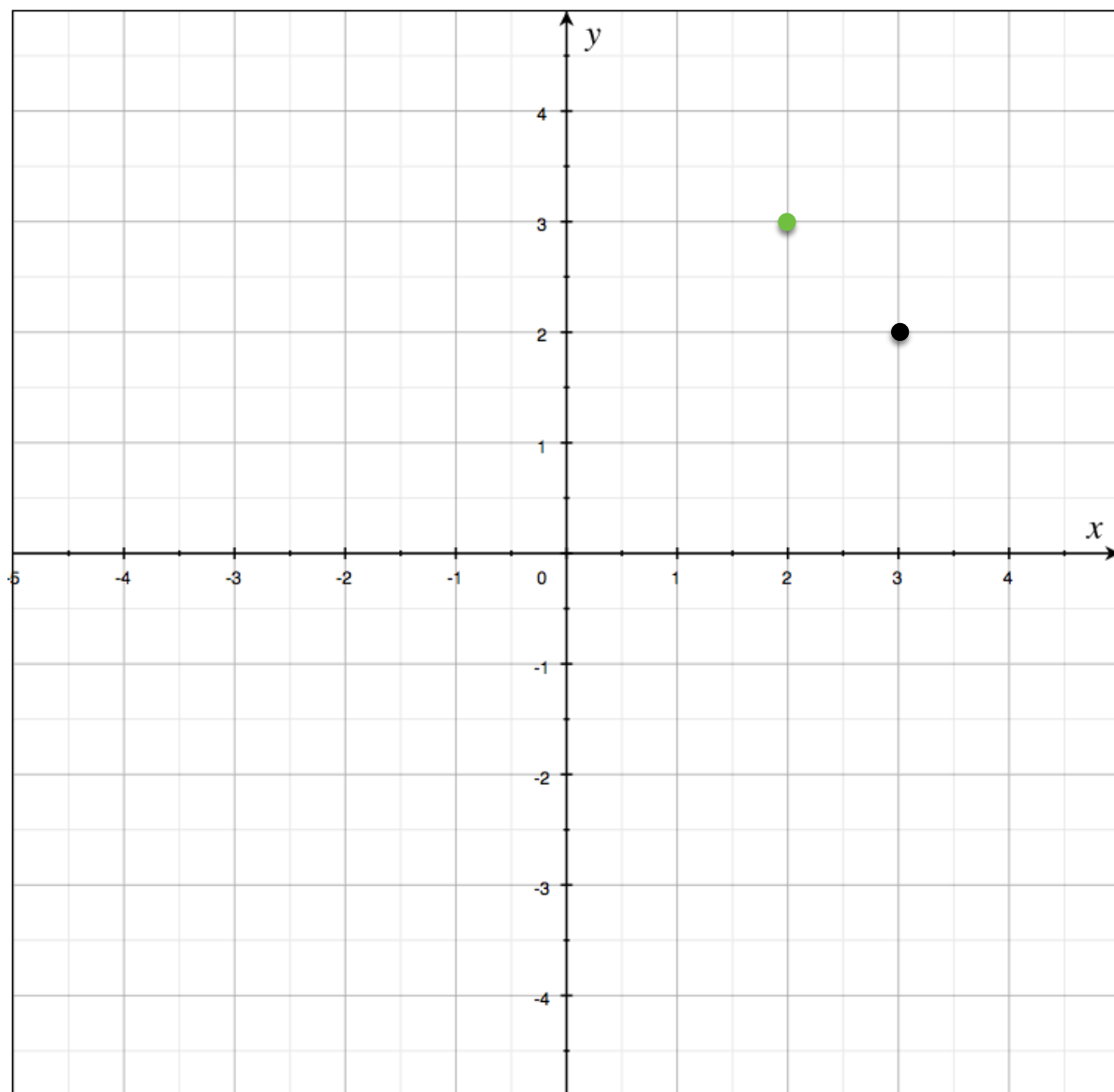
variables



parameters

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

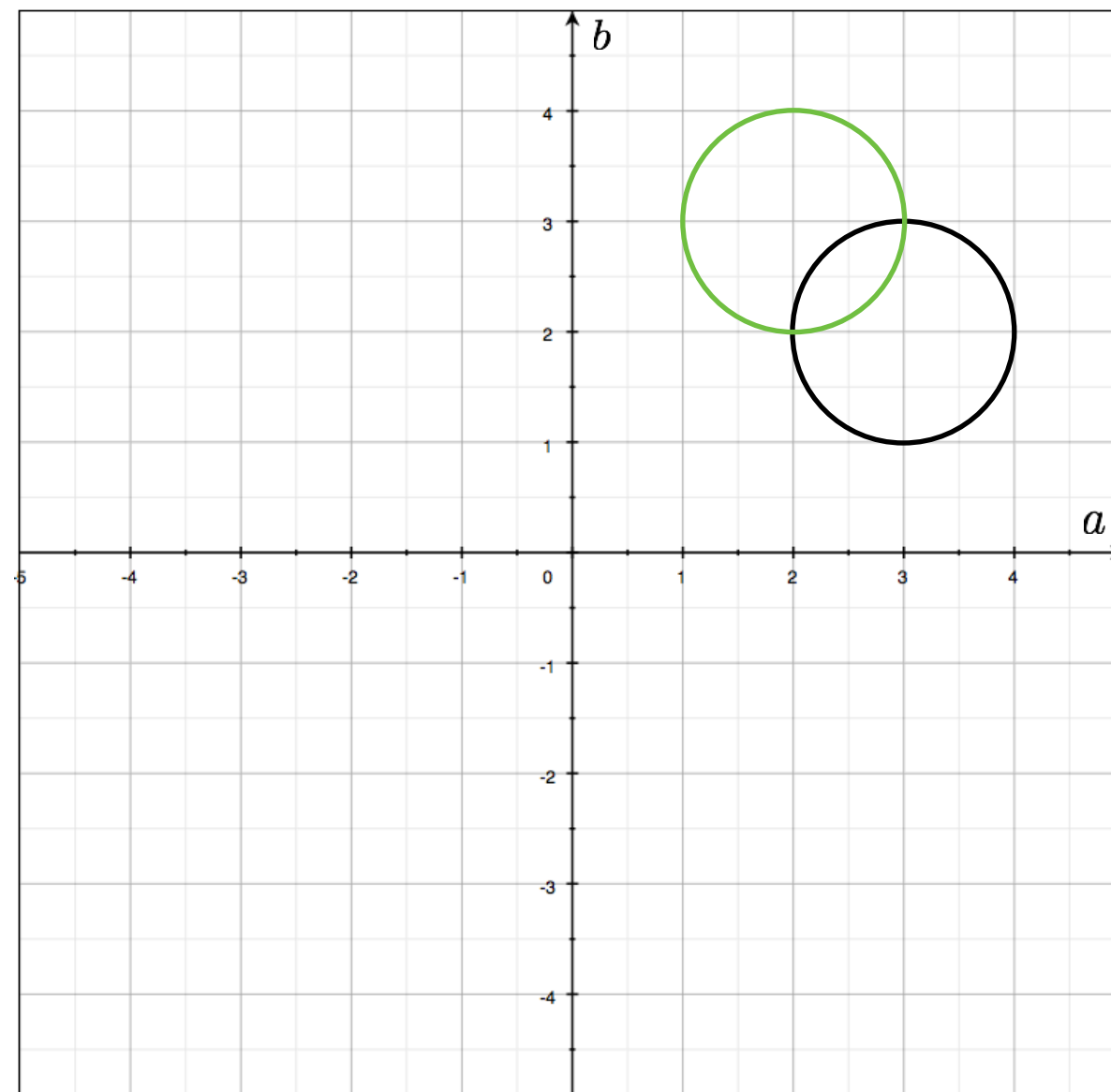
variables



parameters

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

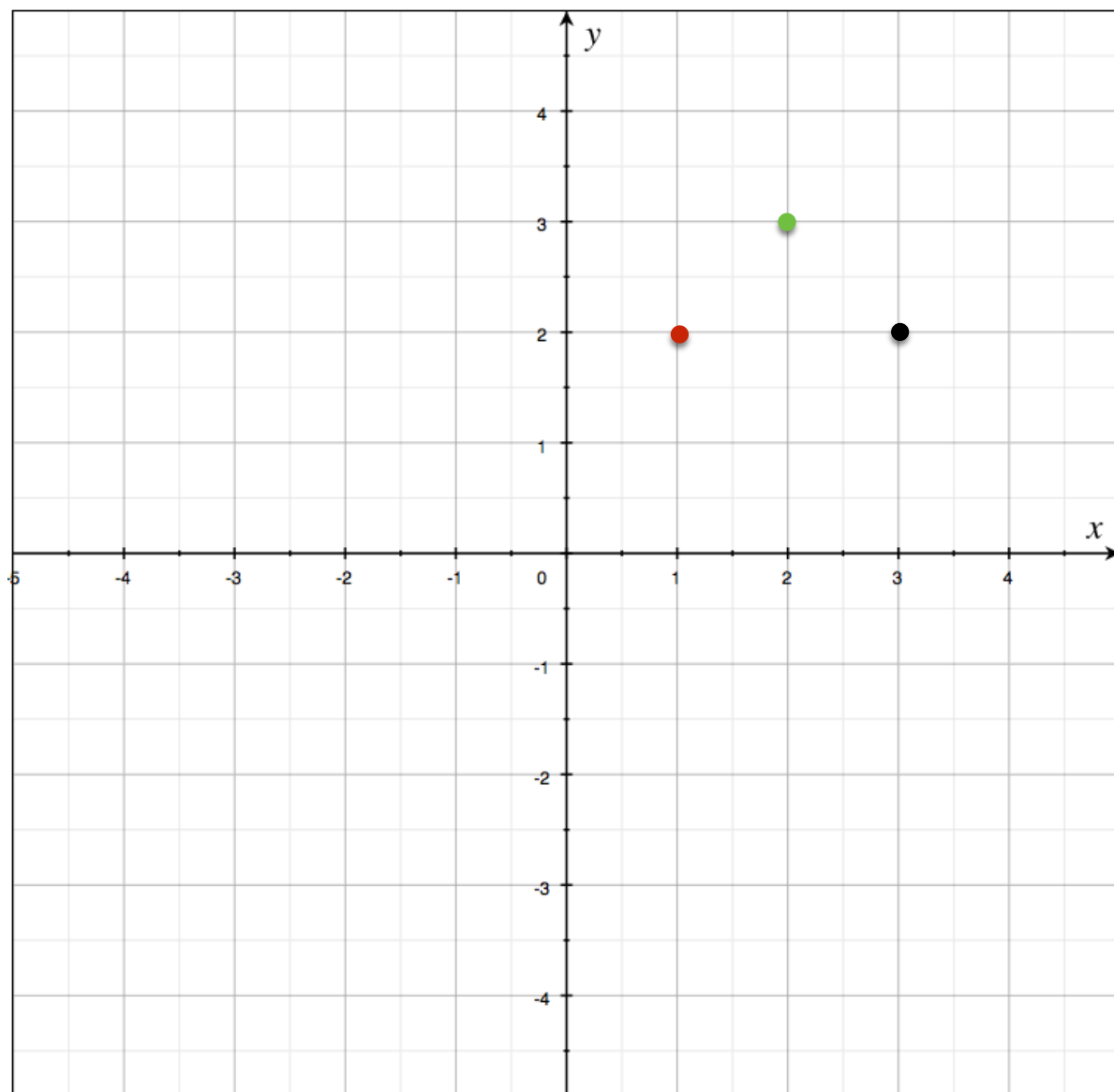
variables



parameters

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

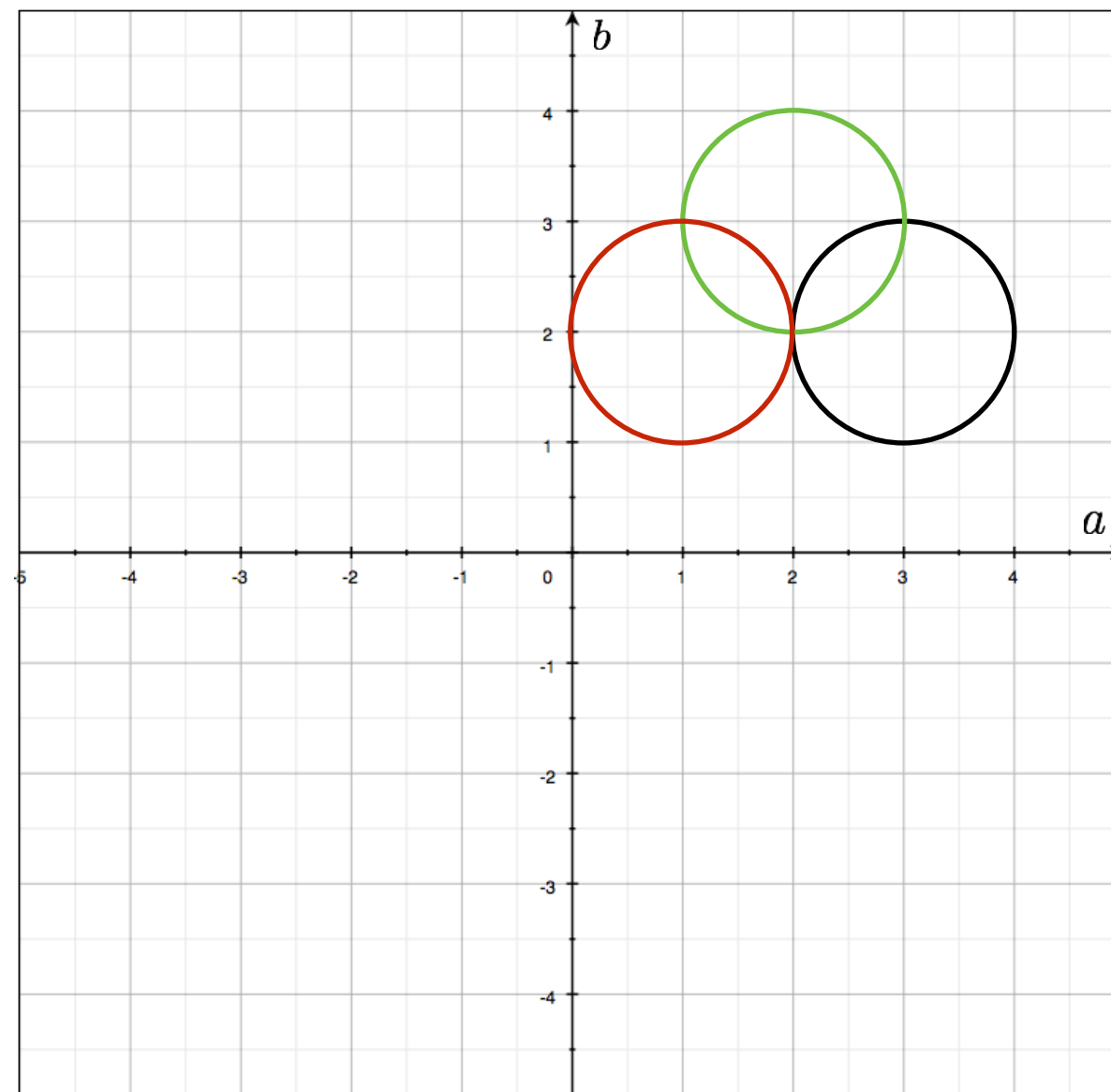
variables



parameters

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

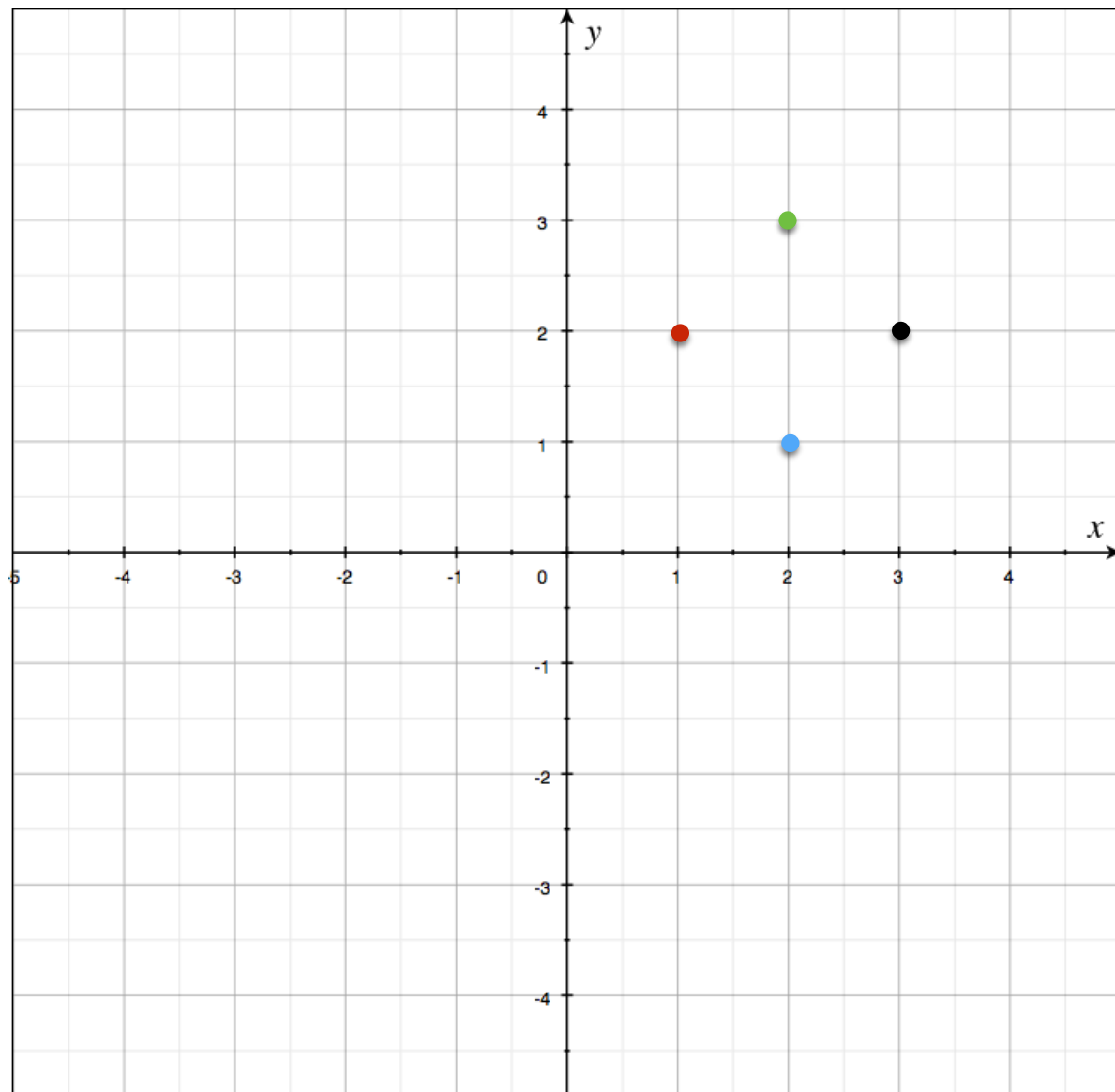
variables



parameters

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

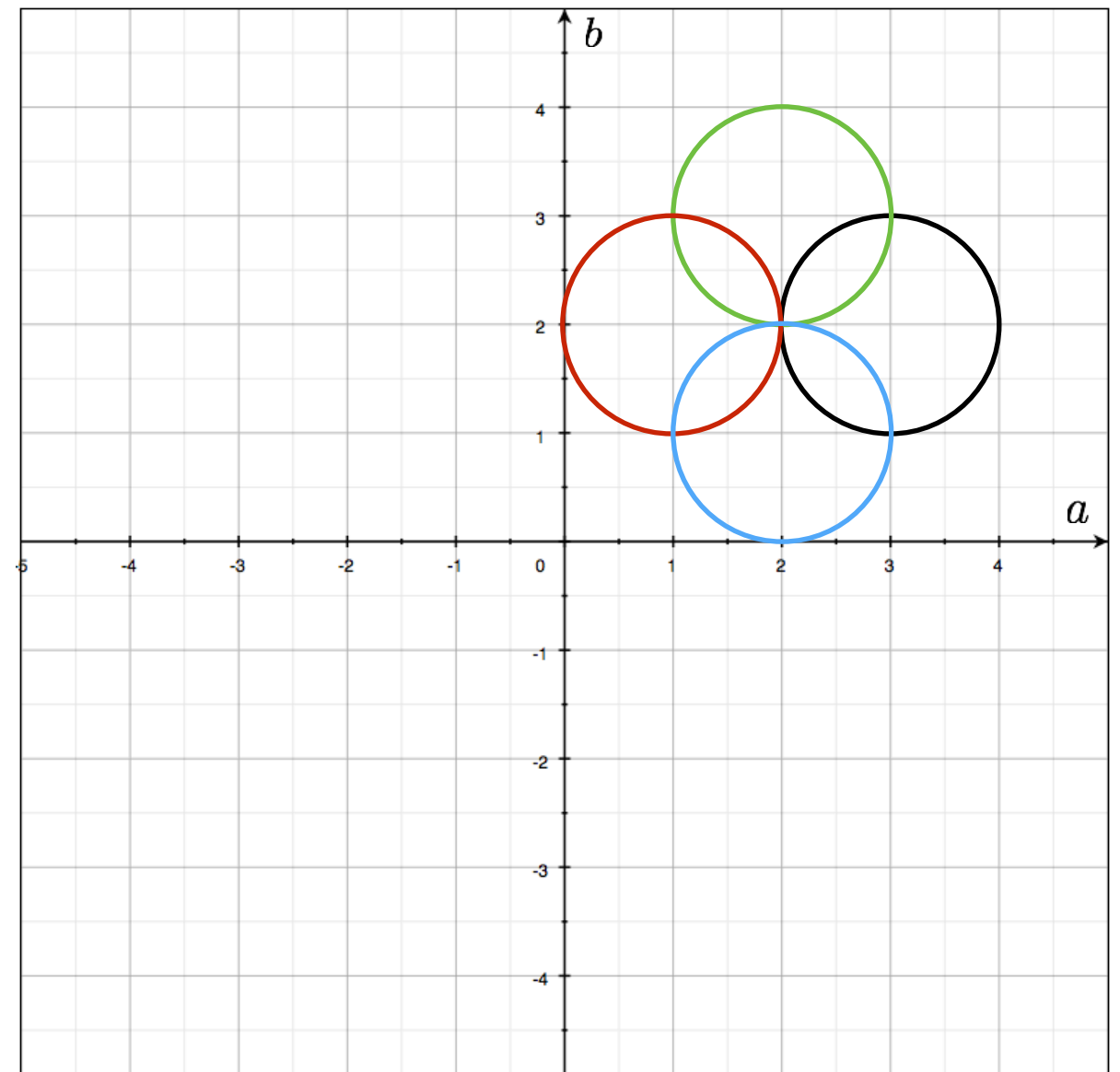
variables



parameters

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

variables



The Hough transform ...

Deals with occlusion well?



Detects multiple instances?



Robust to noise?

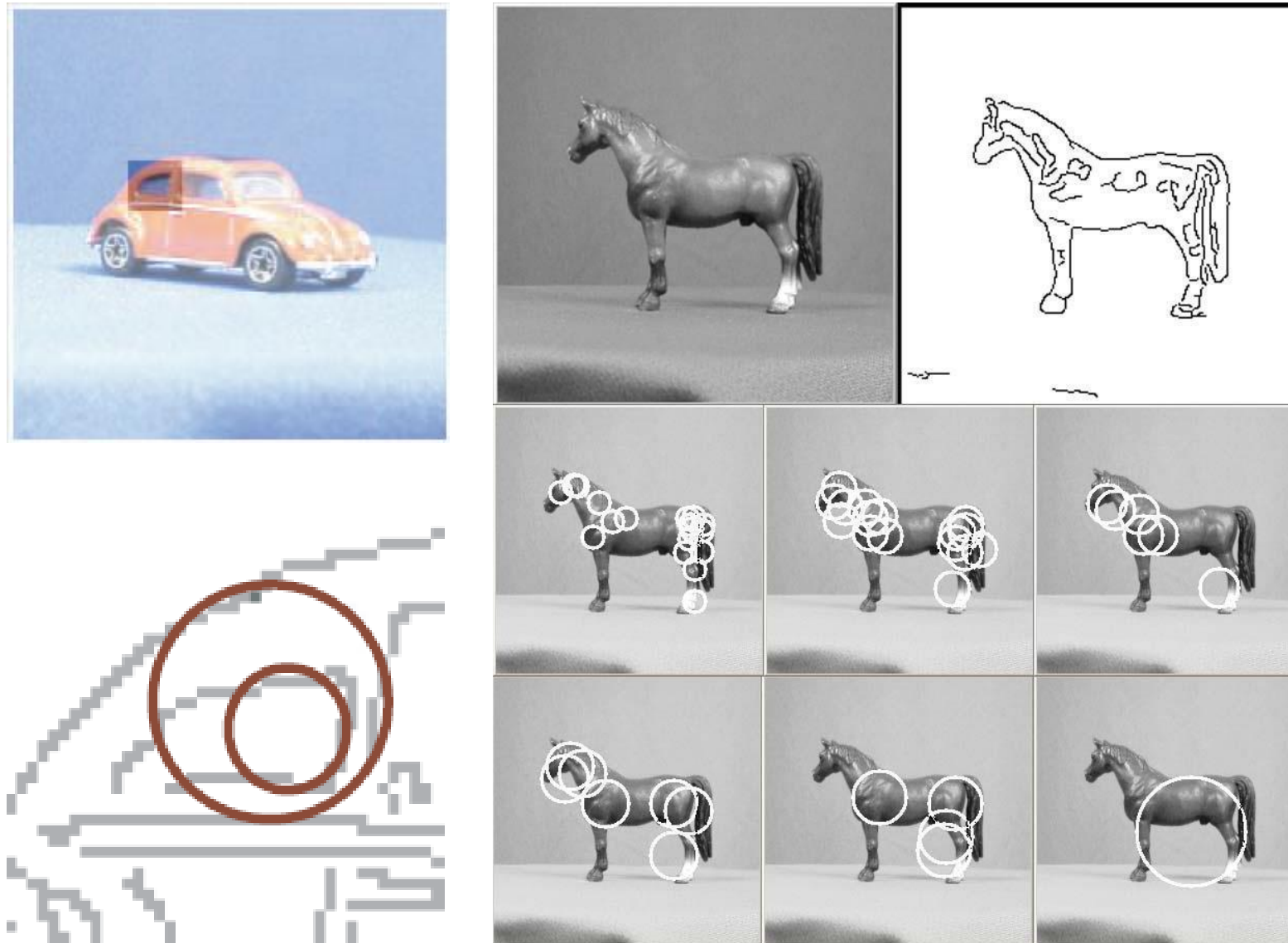


Good computational complexity?

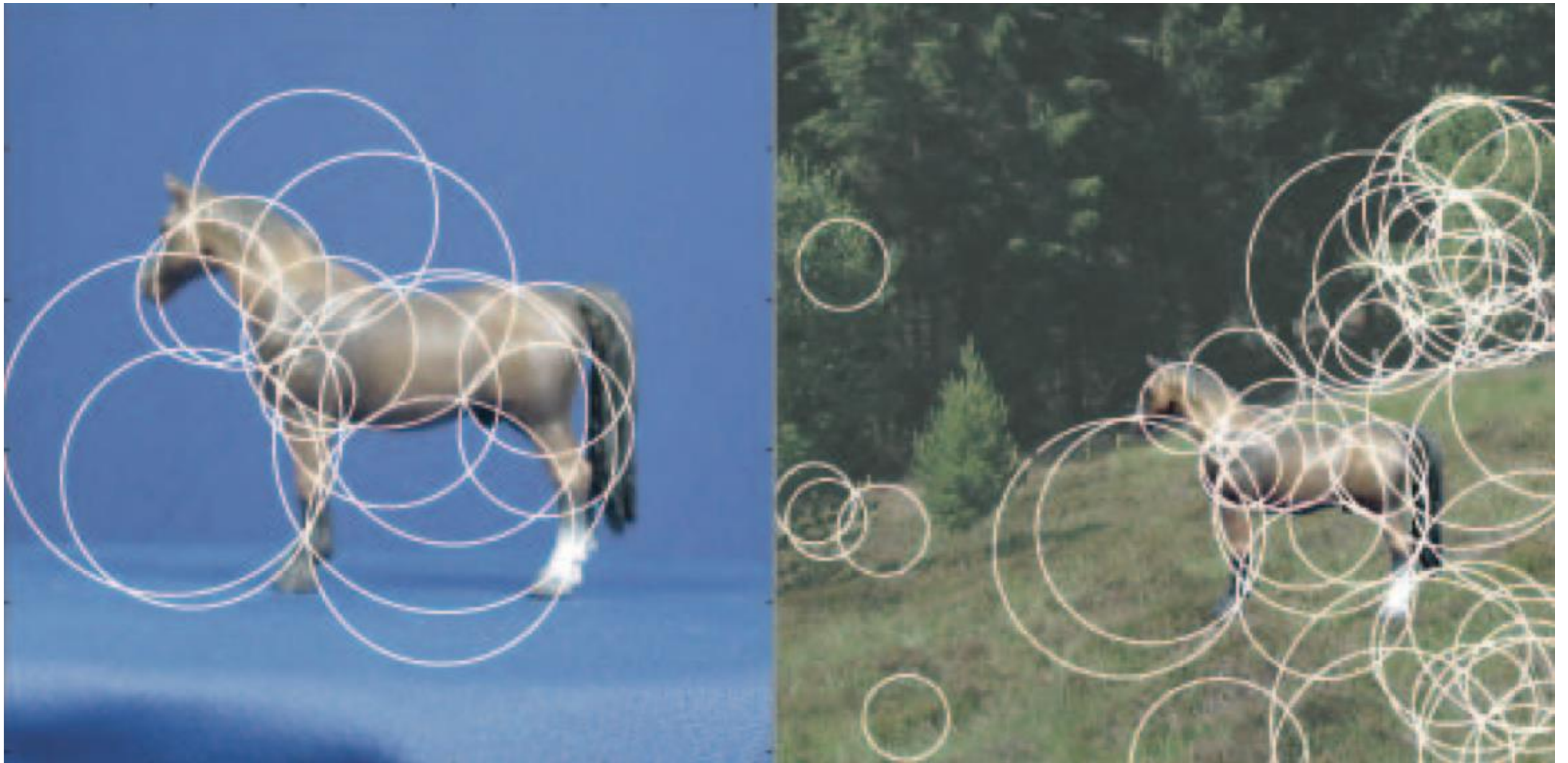


Application of Hough transforms

Detecting shape features



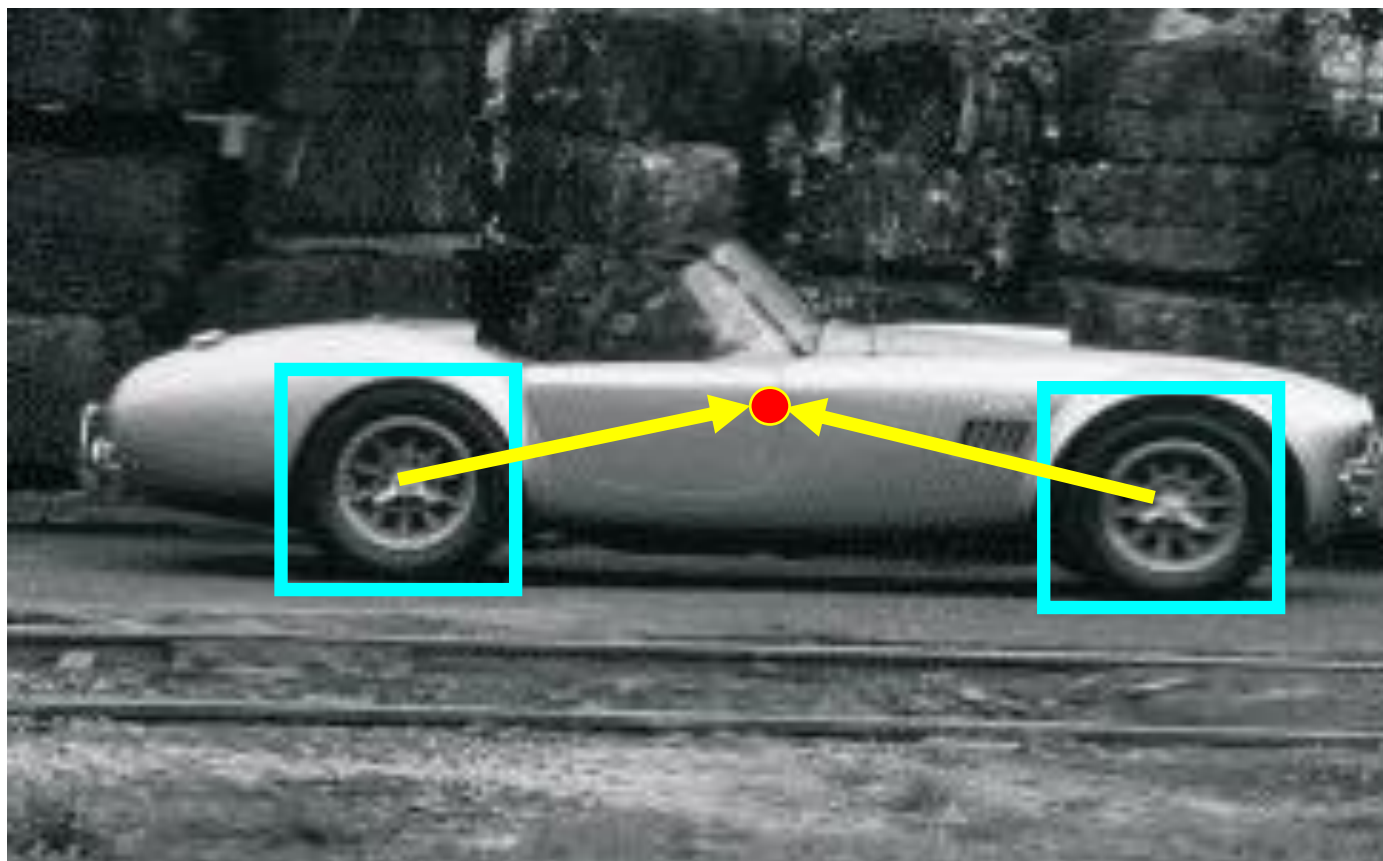
F. Jurie and C. Schmid, Scale-invariant shape features for recognition of object categories, CVPR 2004



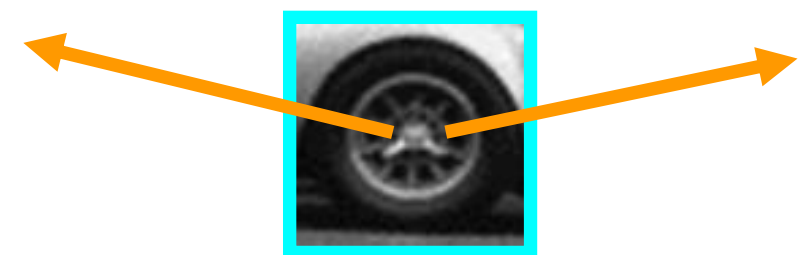
Robustness to scale and clutter

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



References

Basic reading:

- Szeliski textbook, Sections 4.2, 4.3.