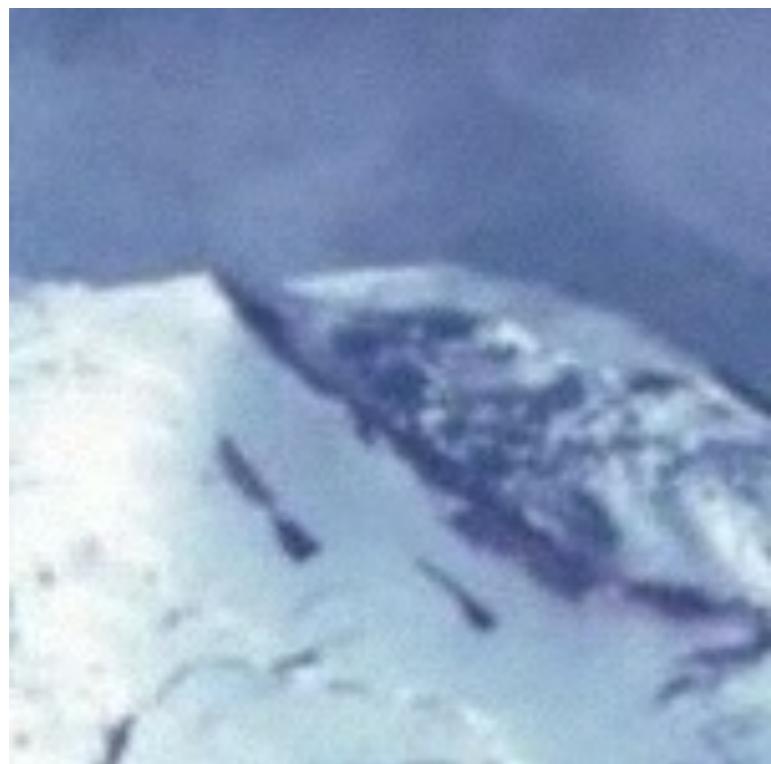


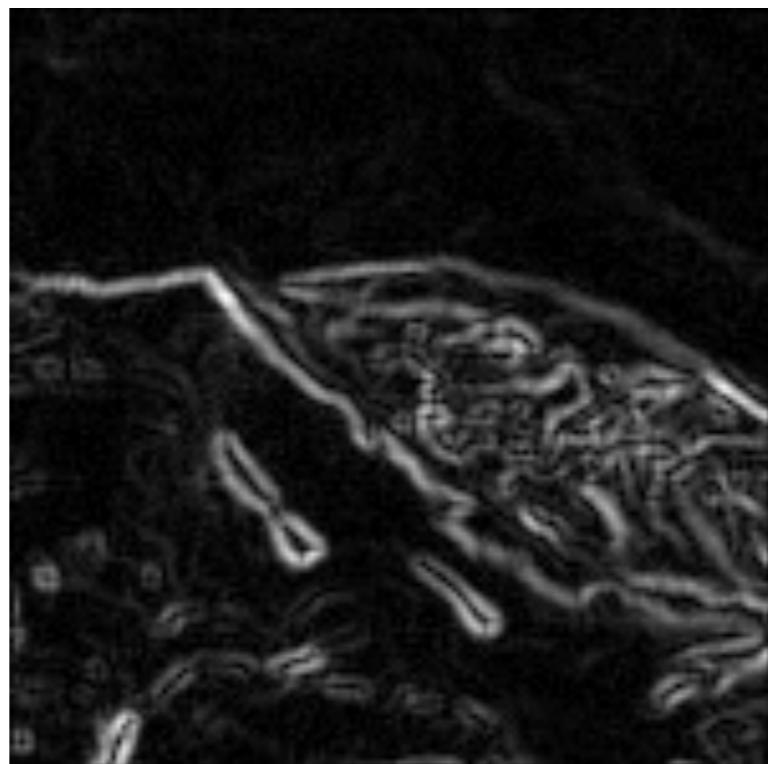
Hough Transform

16-385 Computer Vision

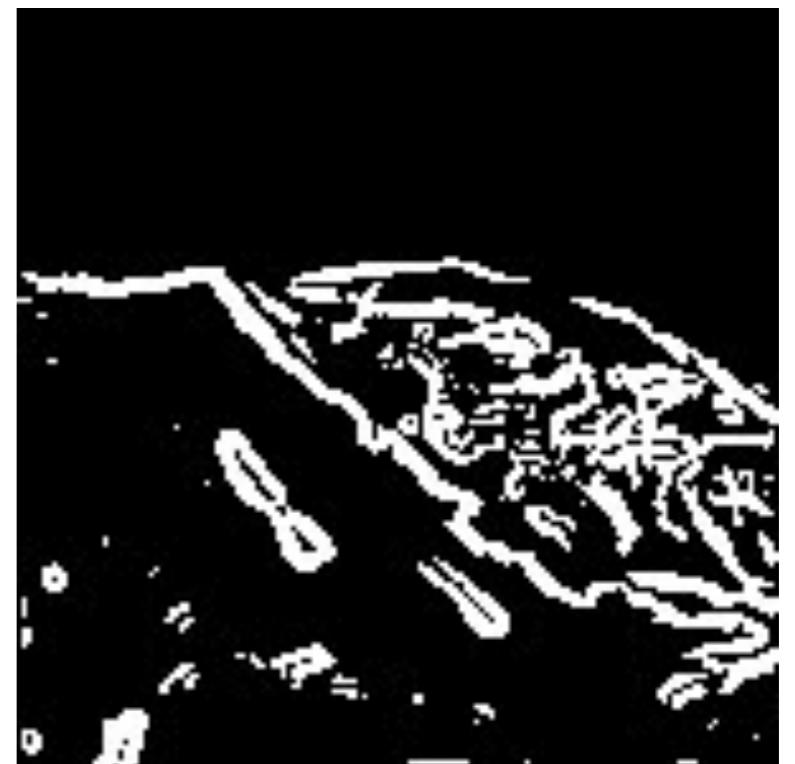
motivation



Original image



Edge detection



Thresholding

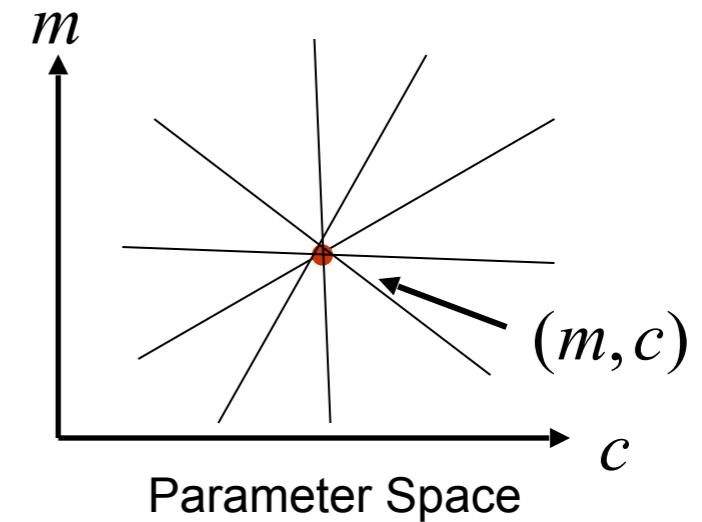
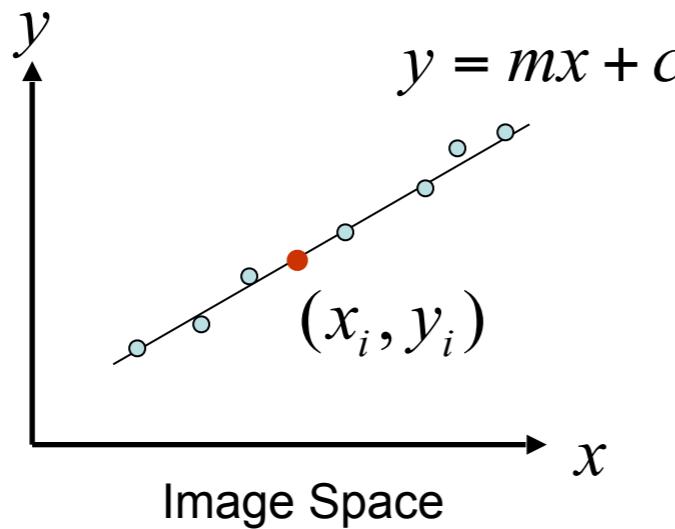
How do we find image boundaries (lines)?

Hough Transform

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

Voting in parameter space

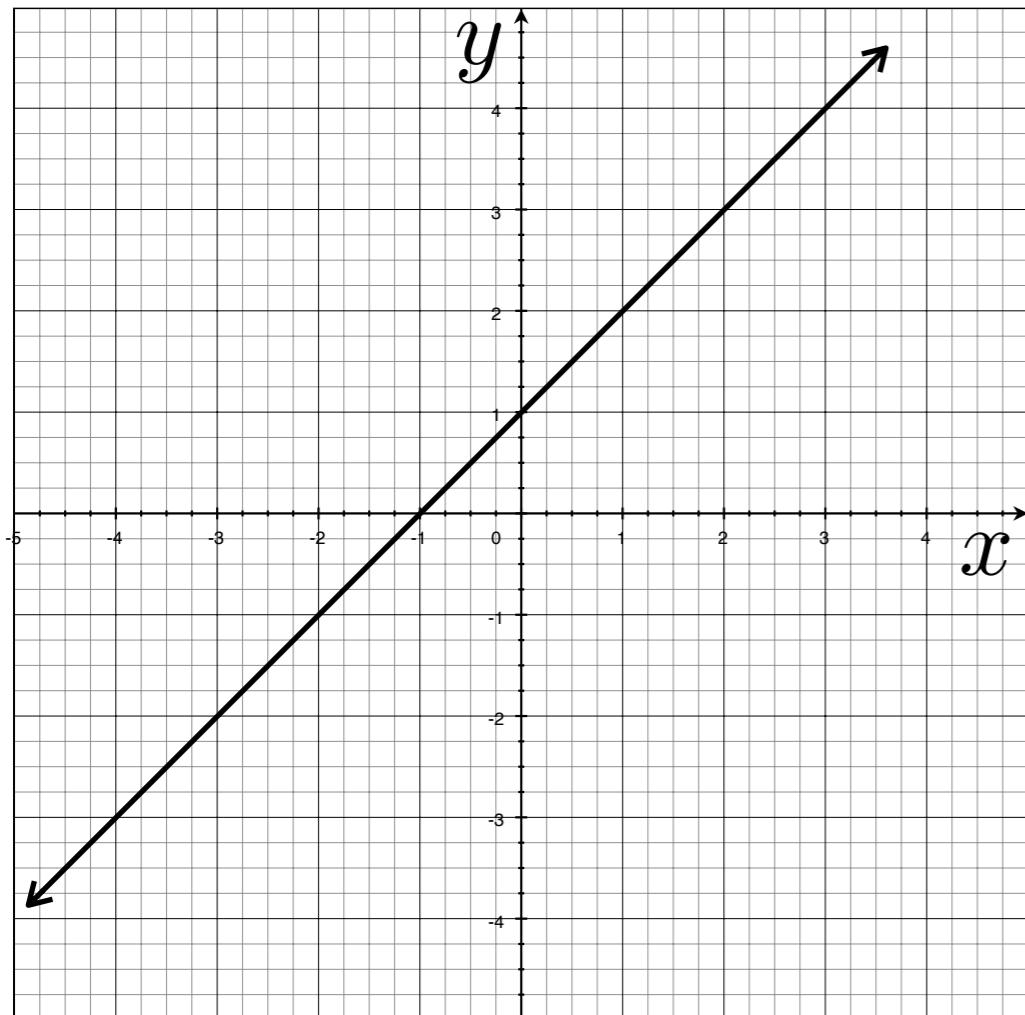
Example: 'line detection'



Each point votes for the line it belongs to

Image and parameter space

variables
 $y = mx + b$
parameters



a line
becomes
a point

variables
 $y - mx = b$
parameters

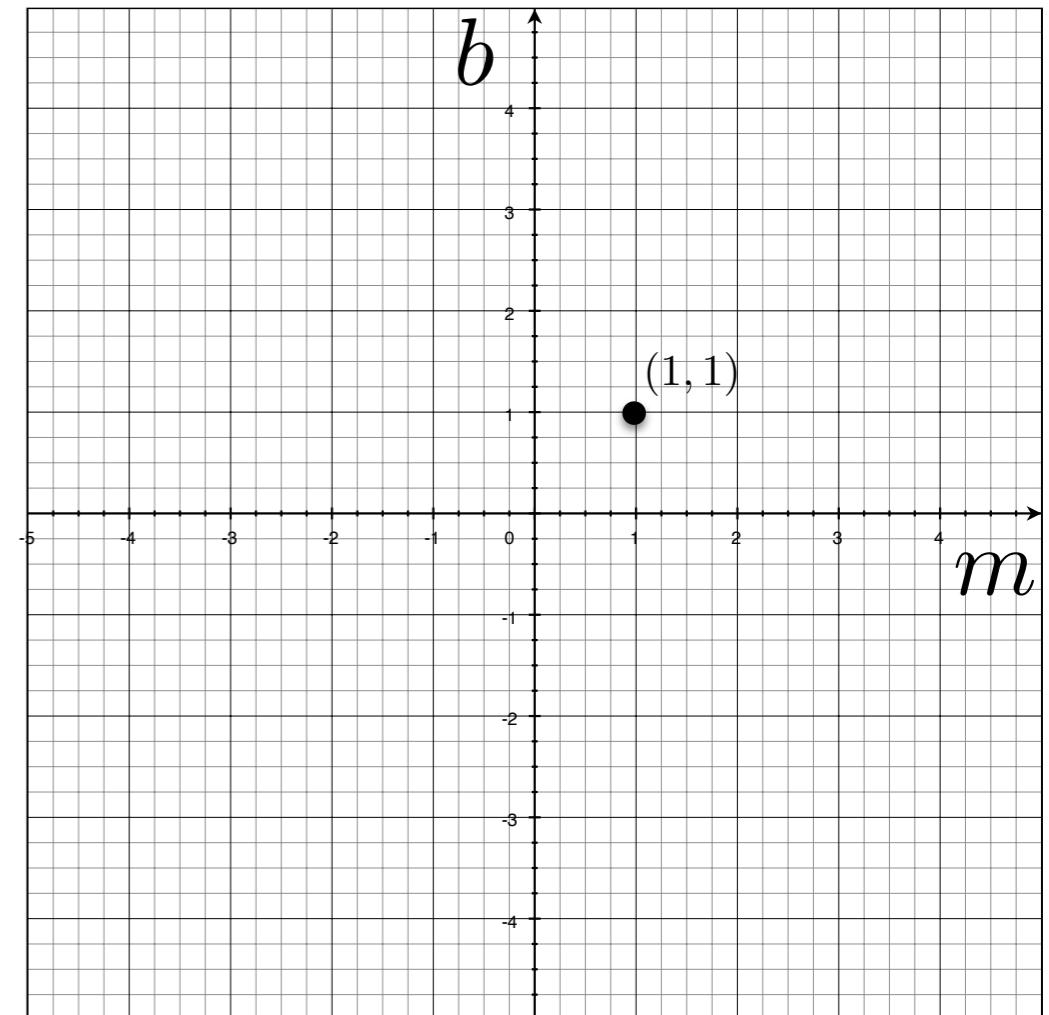
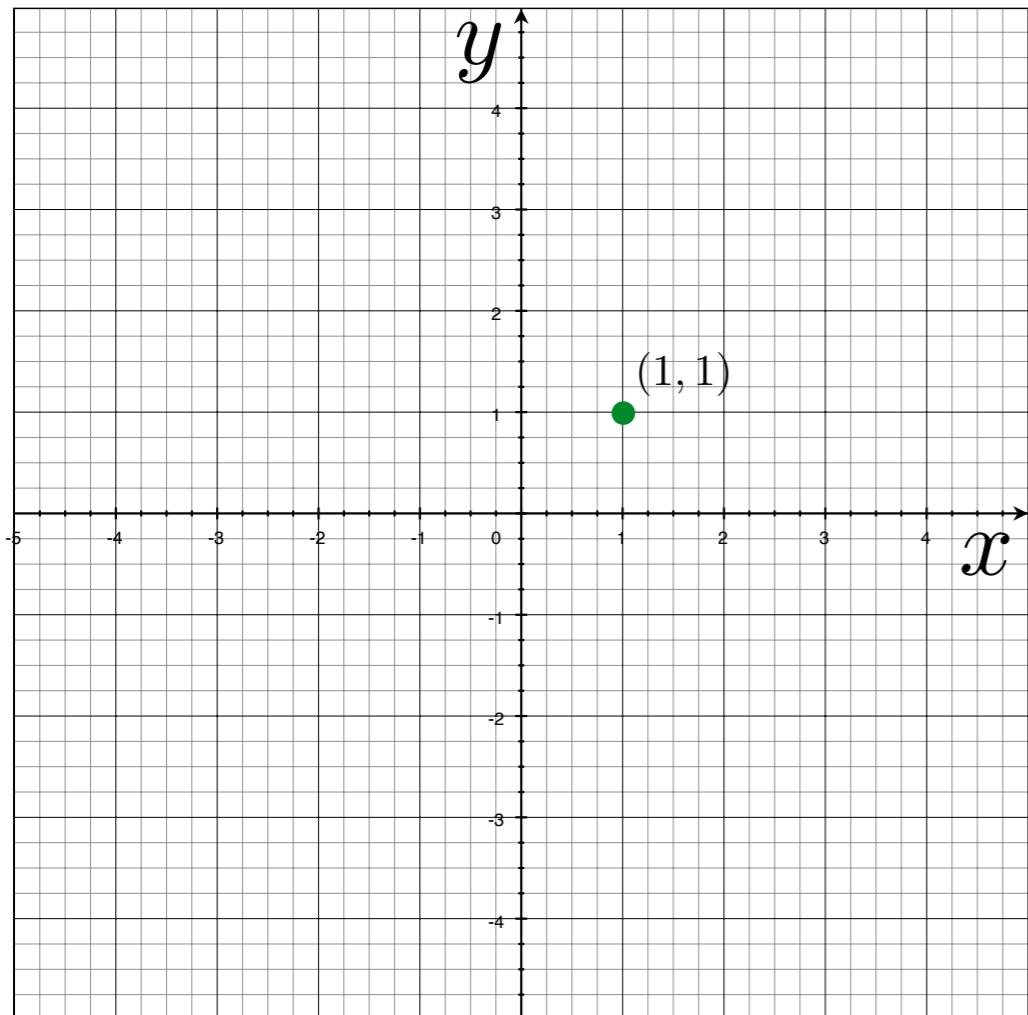


Image space

Parameter space

Image and parameter space

variables
 $y = mx + b$
parameters



a point becomes
a line

variables
 $y - mx = b$
parameters

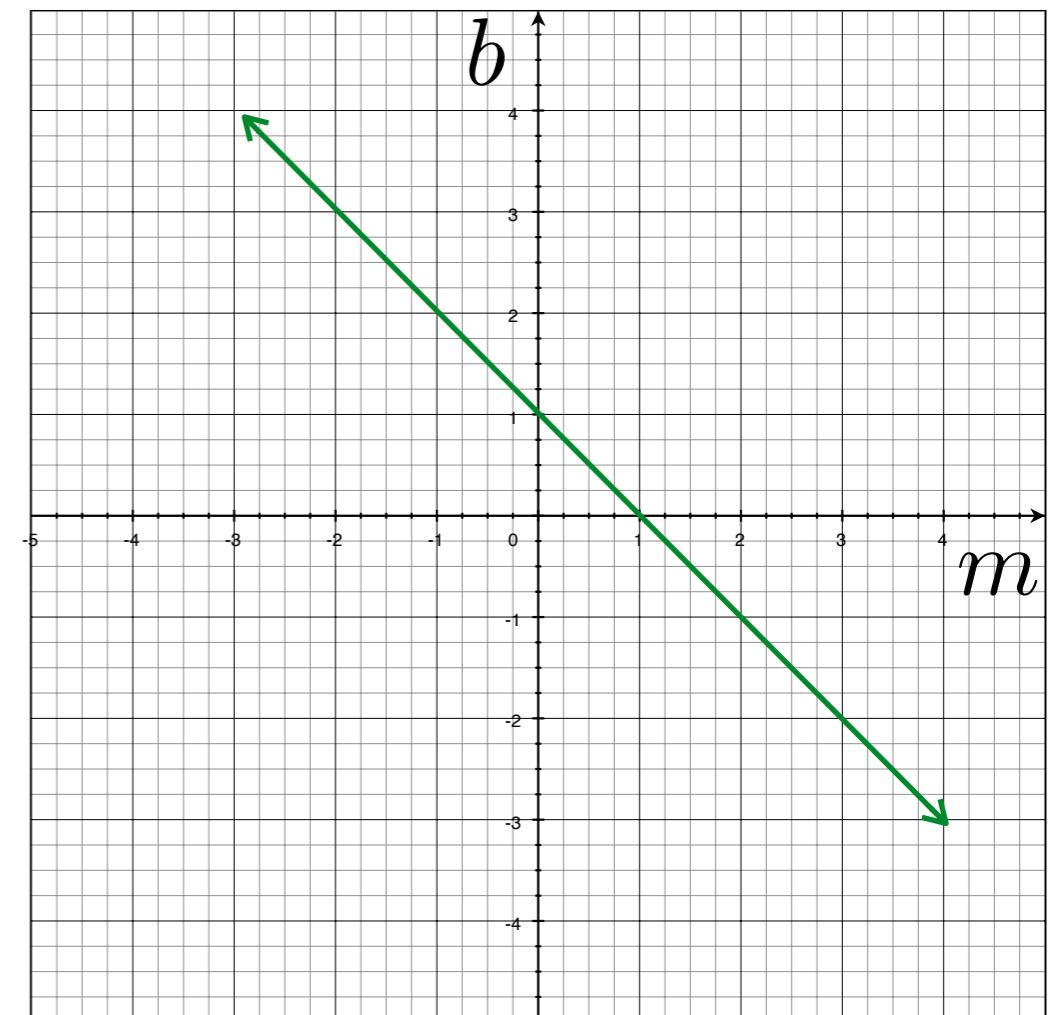
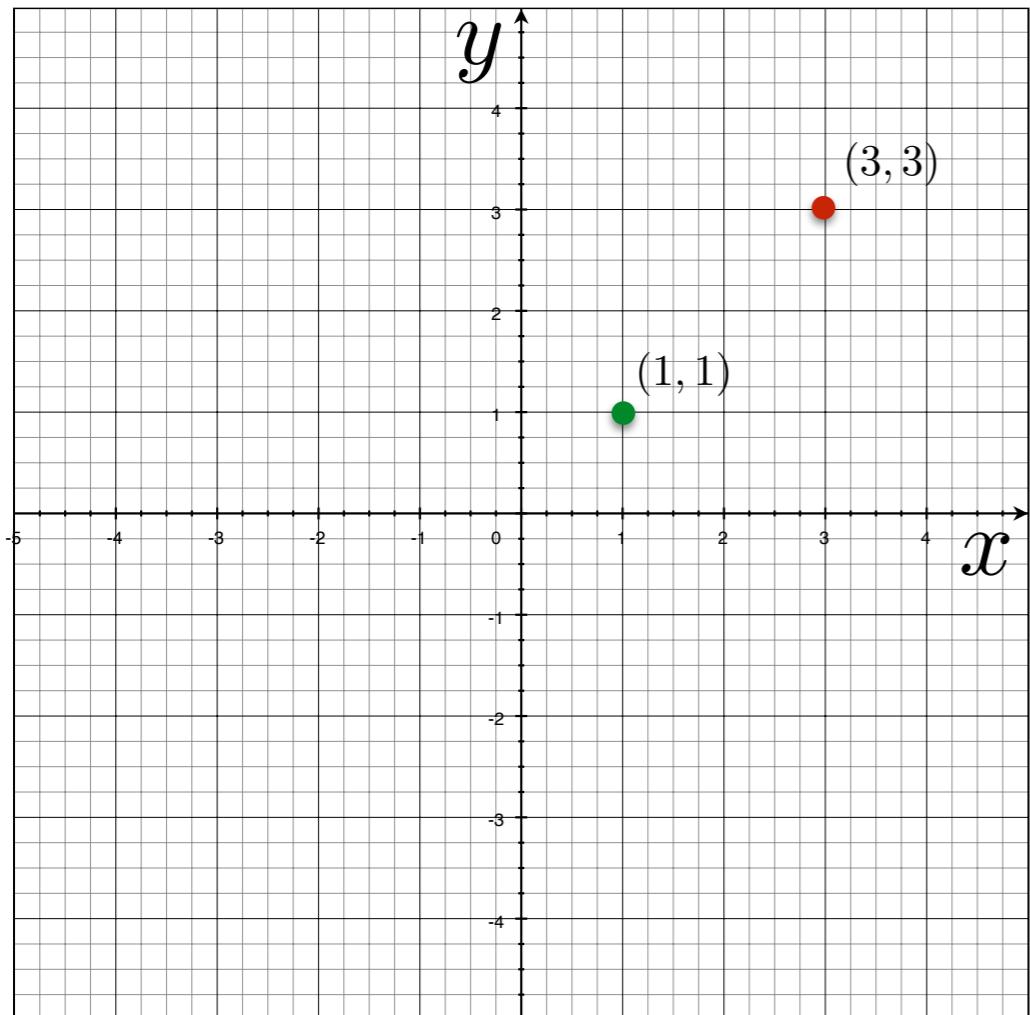


Image space

Parameter space

Image and parameter space

variables
 $y = mx + b$
parameters



two points
become
?

variables
 $y - mx = b$
parameters

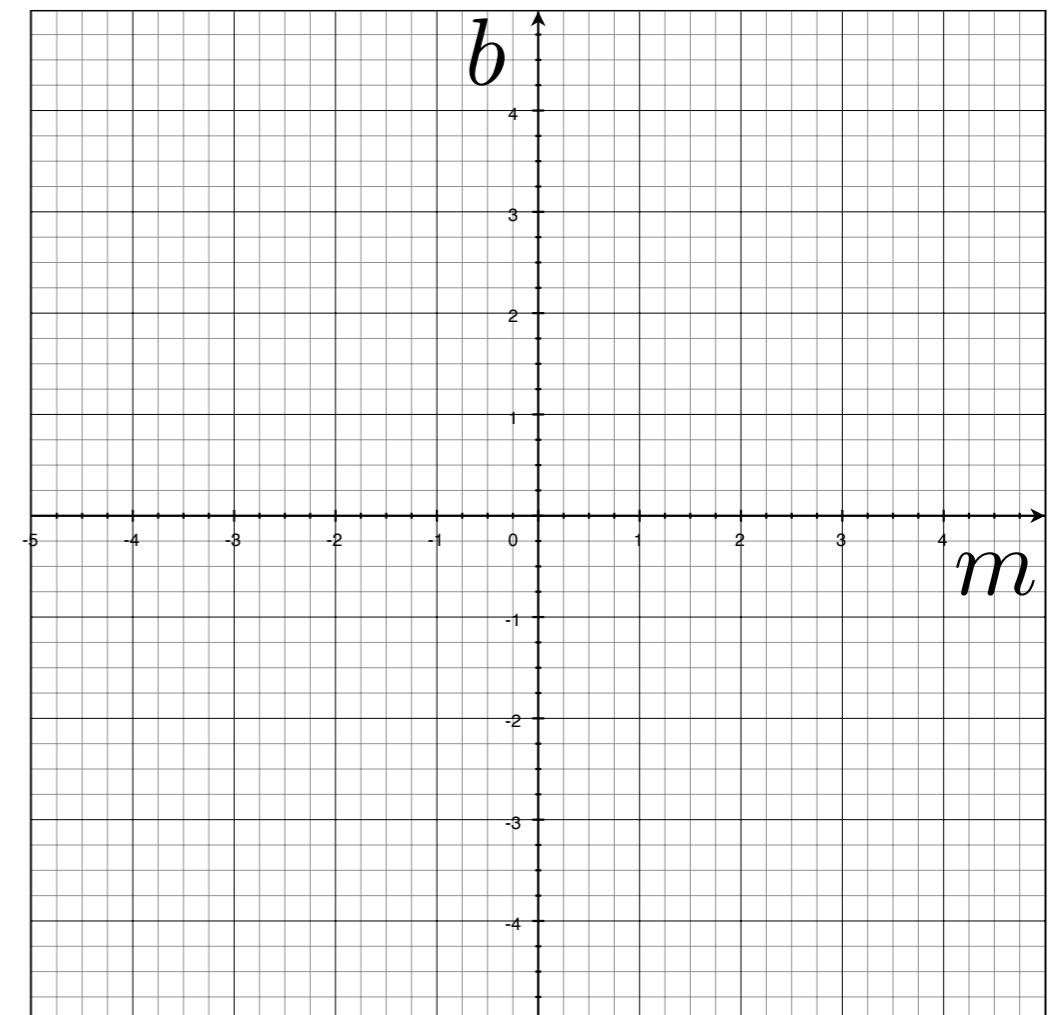
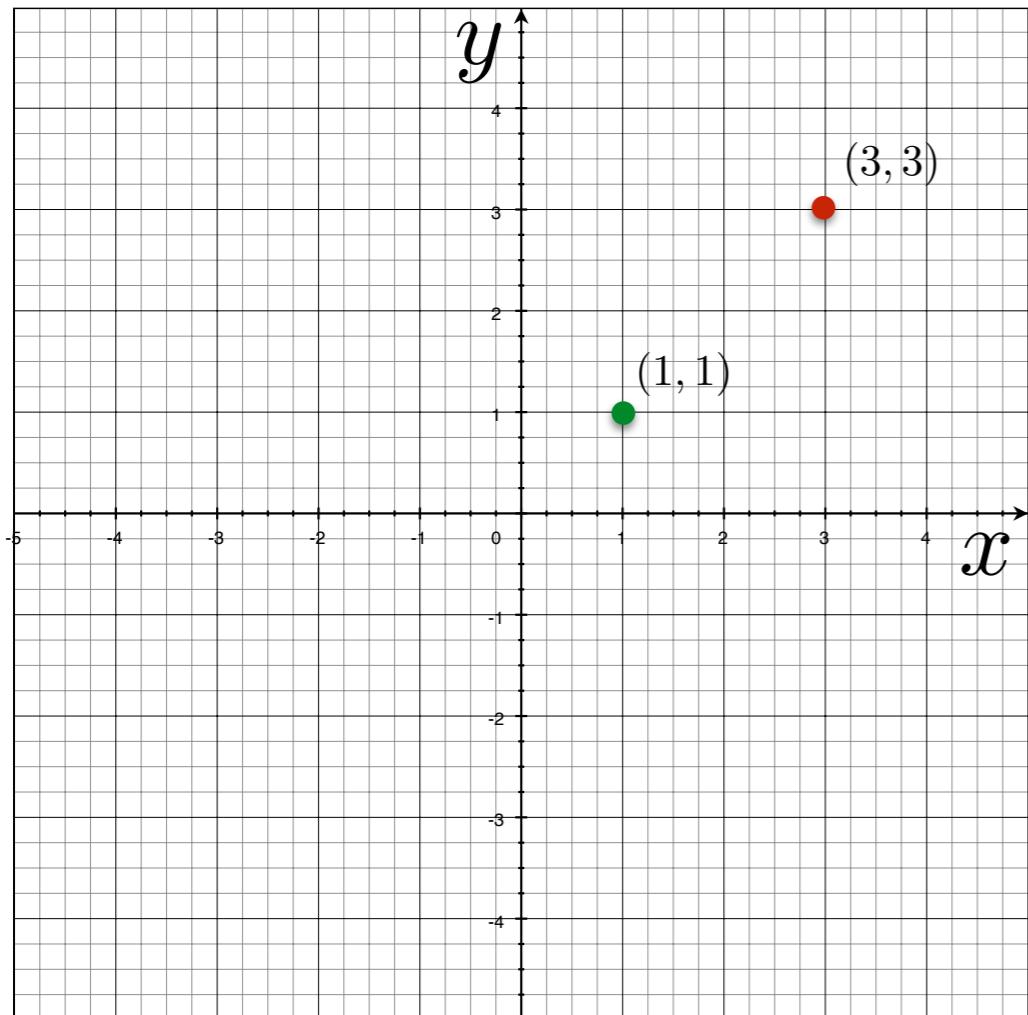


Image space

Parameter space

Image and parameter space

variables
 $y = mx + b$
parameters



two points
become
?

variables
 $y - mx = b$
parameters

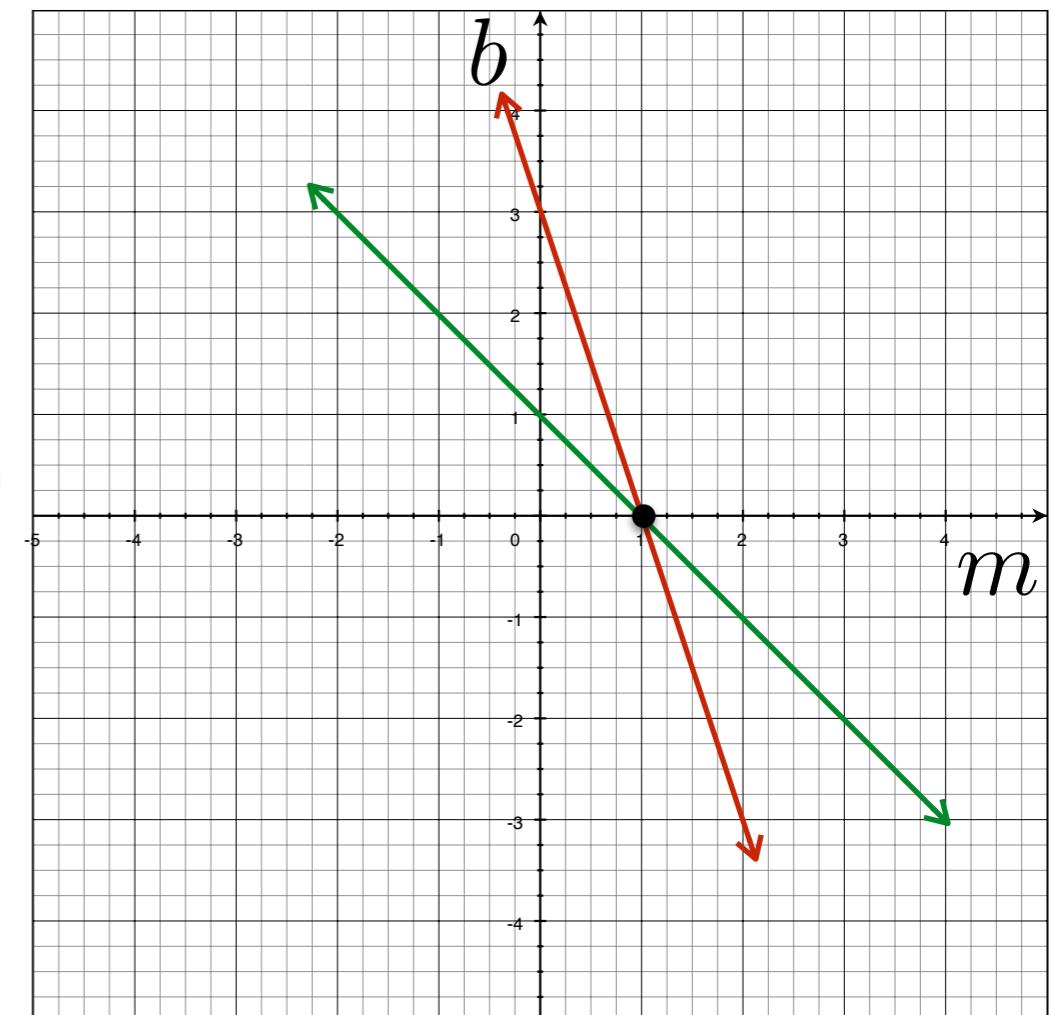
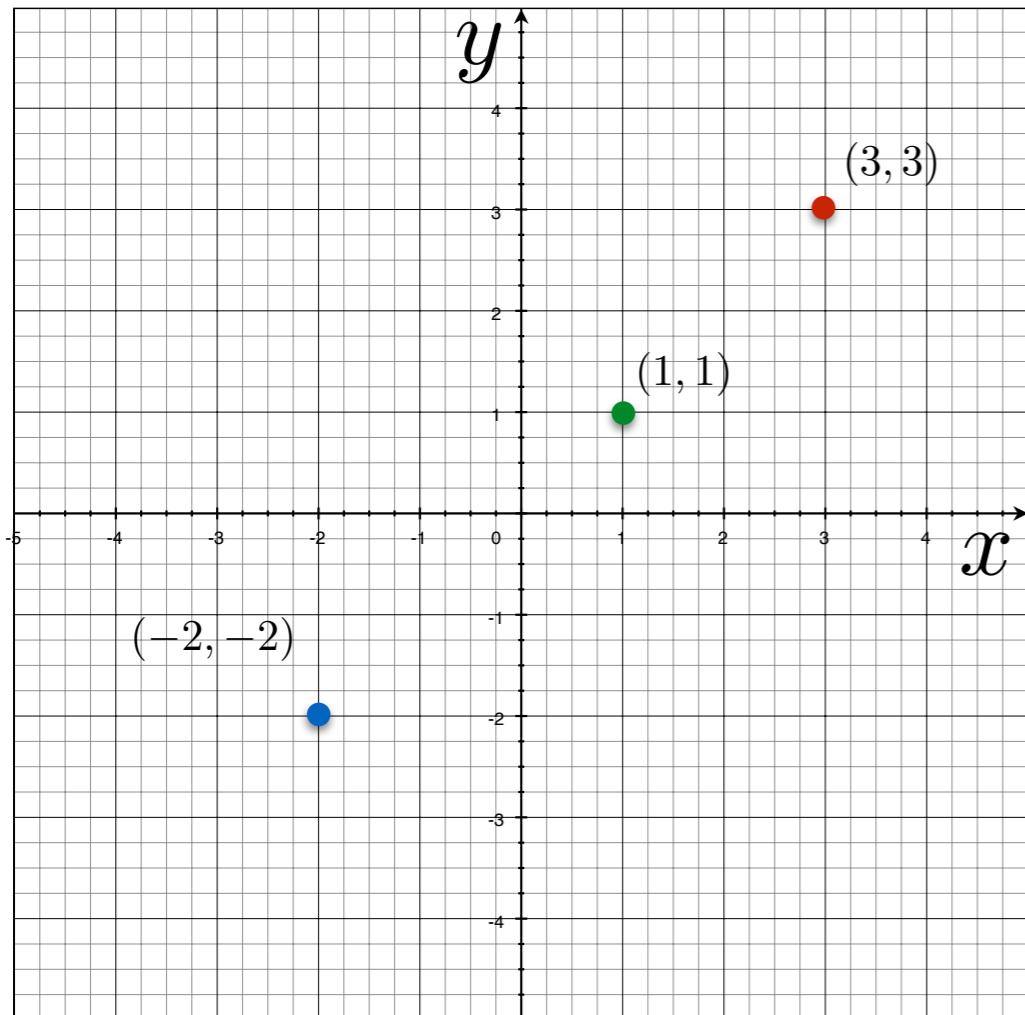


Image space

Parameter space

Image and parameter space

variables
 $y = mx + b$
parameters



three points
become
?

variables
 $y - mx = b$
parameters

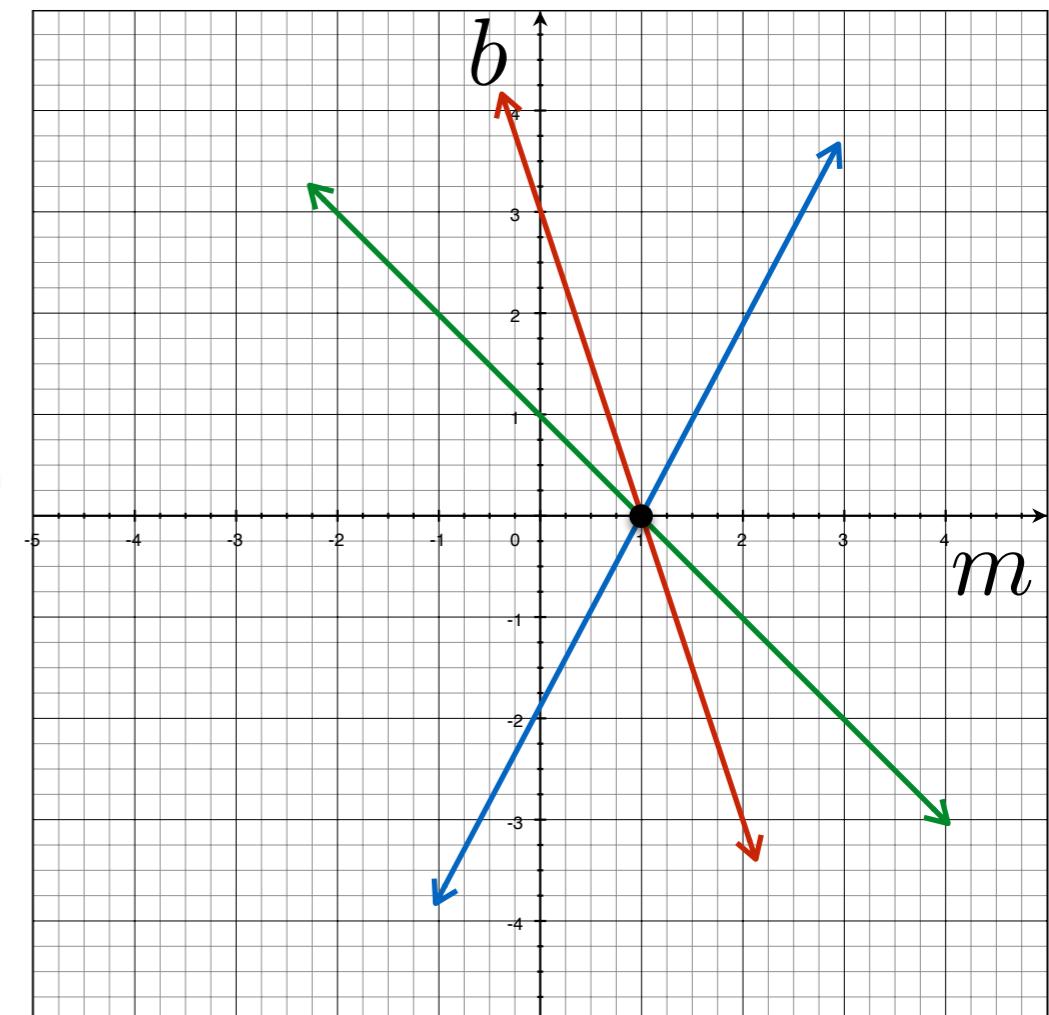
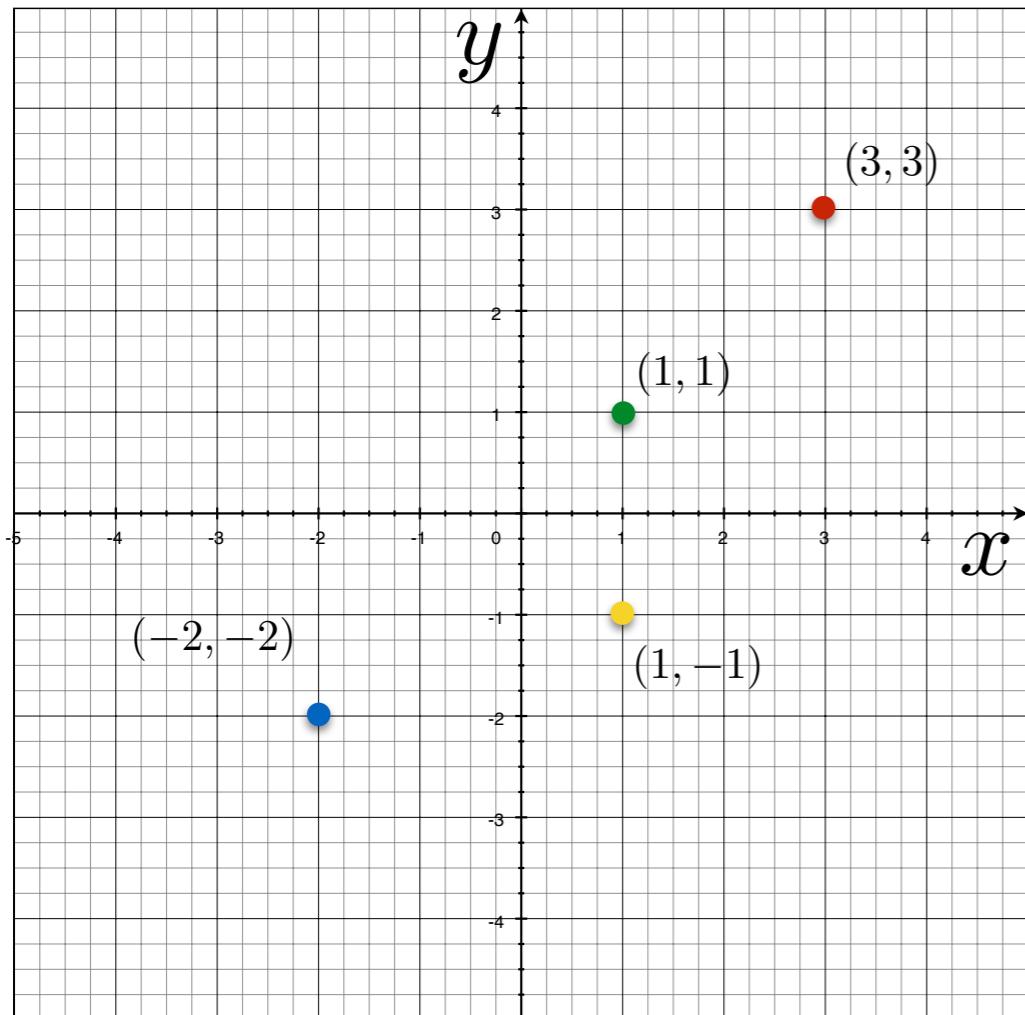


Image space

Parameter space

Image and parameter space

variables
 $y = mx + b$
parameters



four points
become
?

variables
 $y - mx = b$
parameters

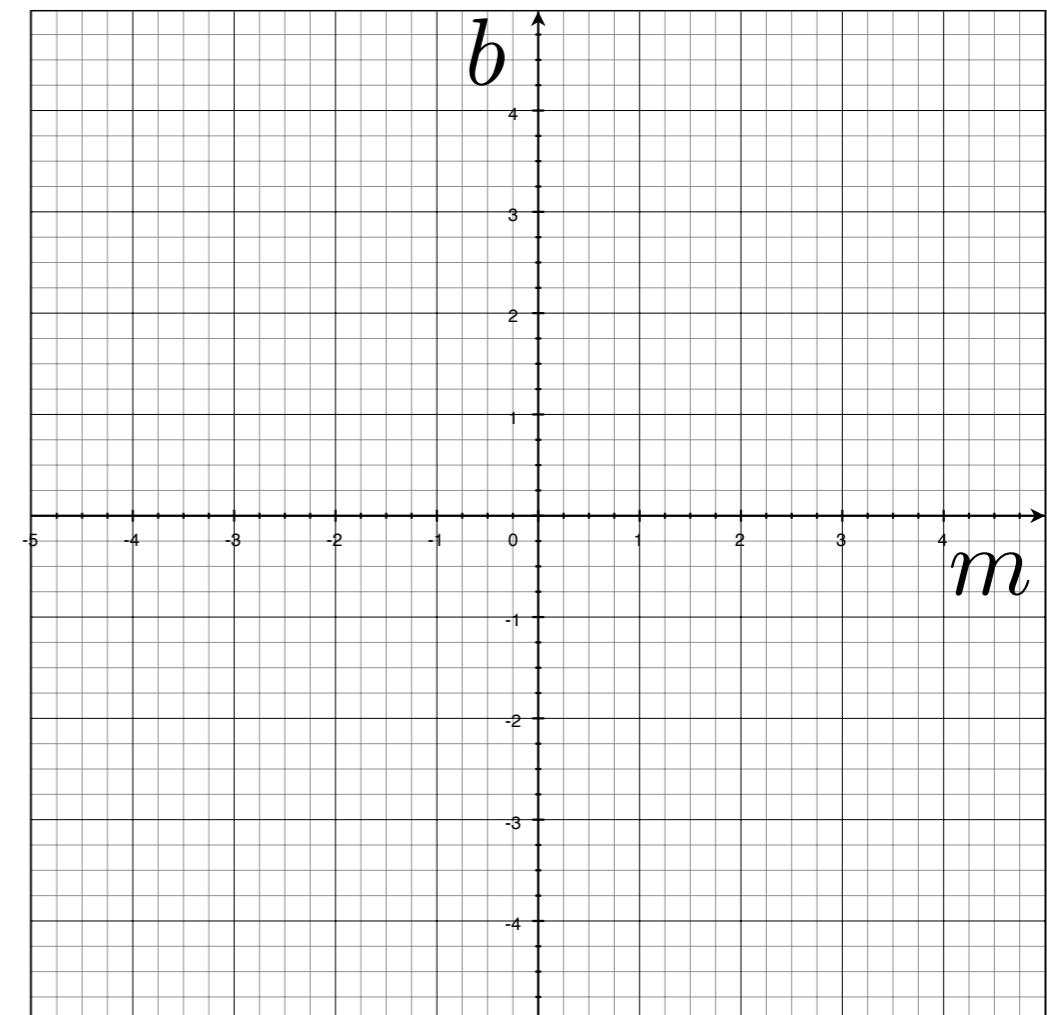
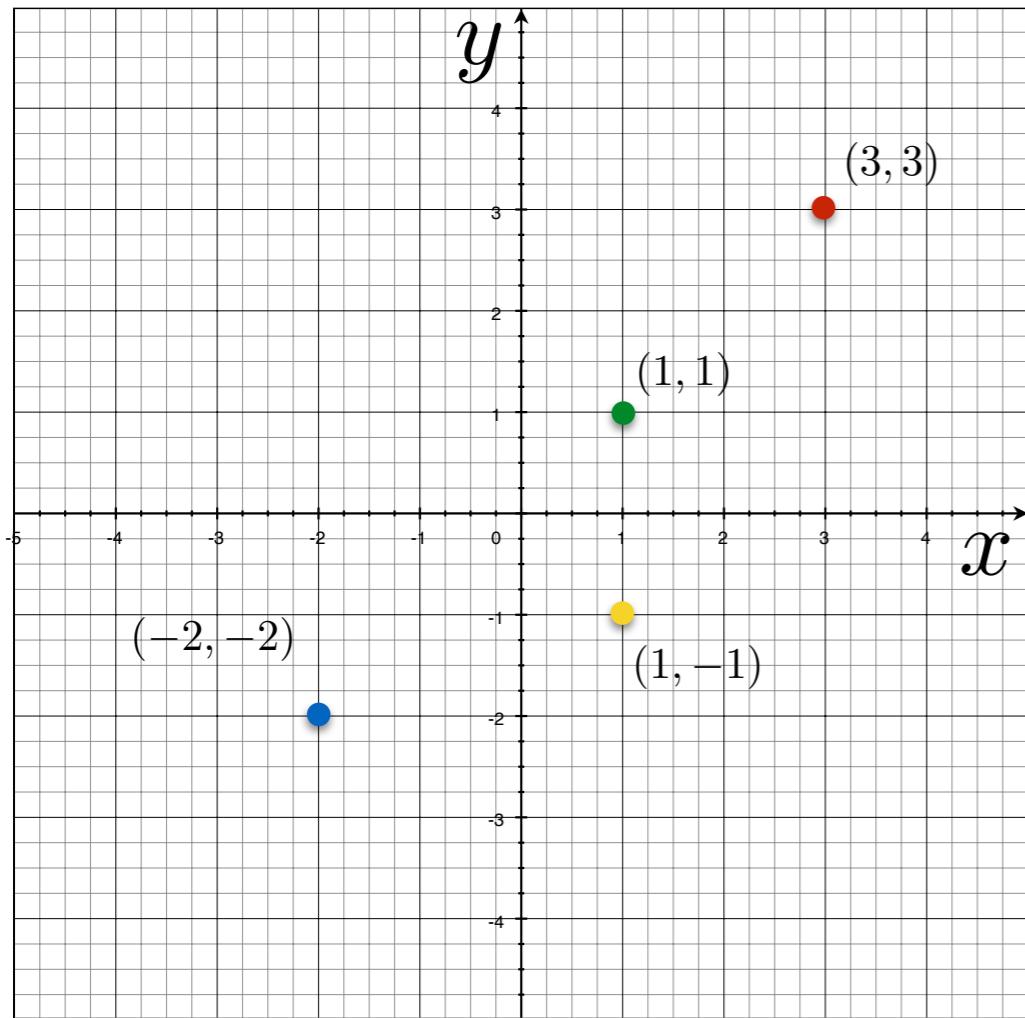


Image space

Parameter space

Image and parameter space

variables
 $y = mx + b$
parameters



four points
become
?

variables
 $y - mx = b$
parameters

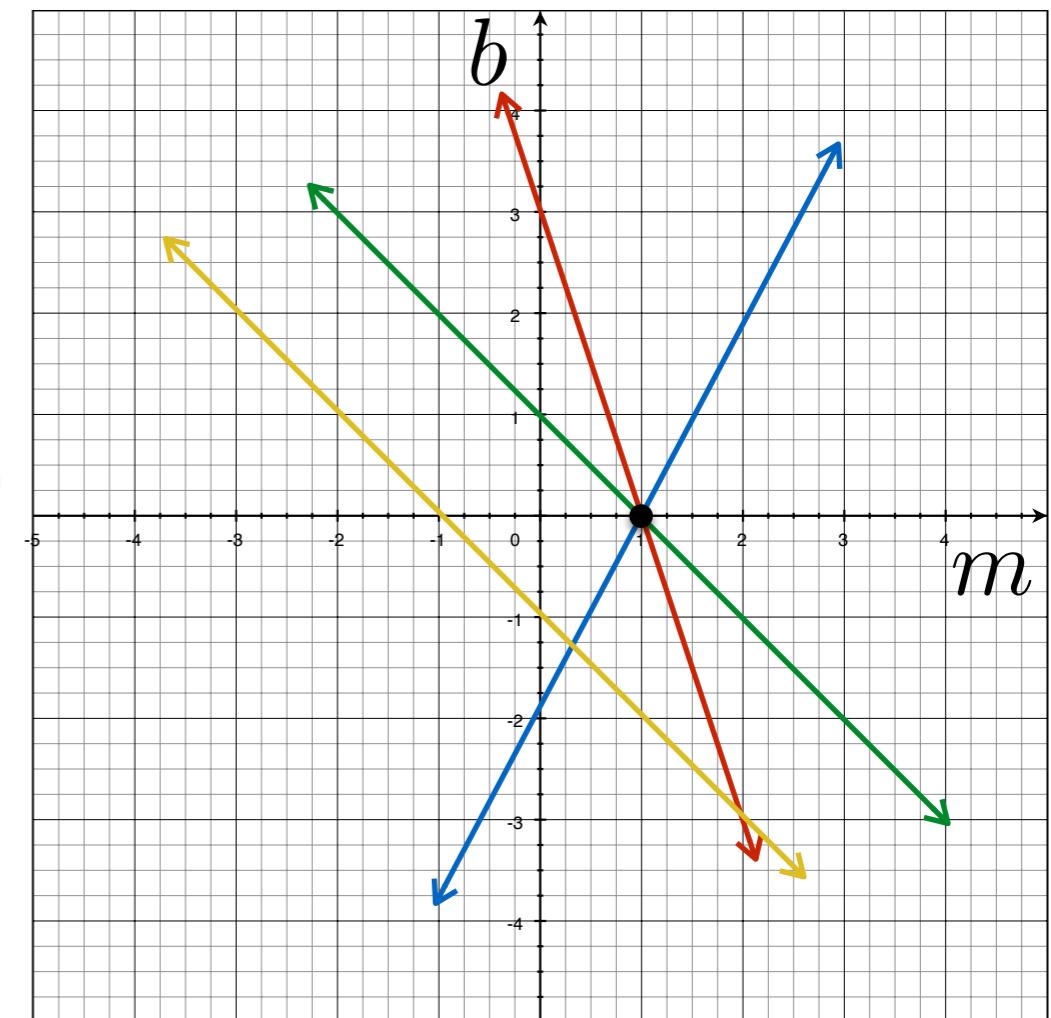


Image space

Parameter space

How would you find the best fitting line?

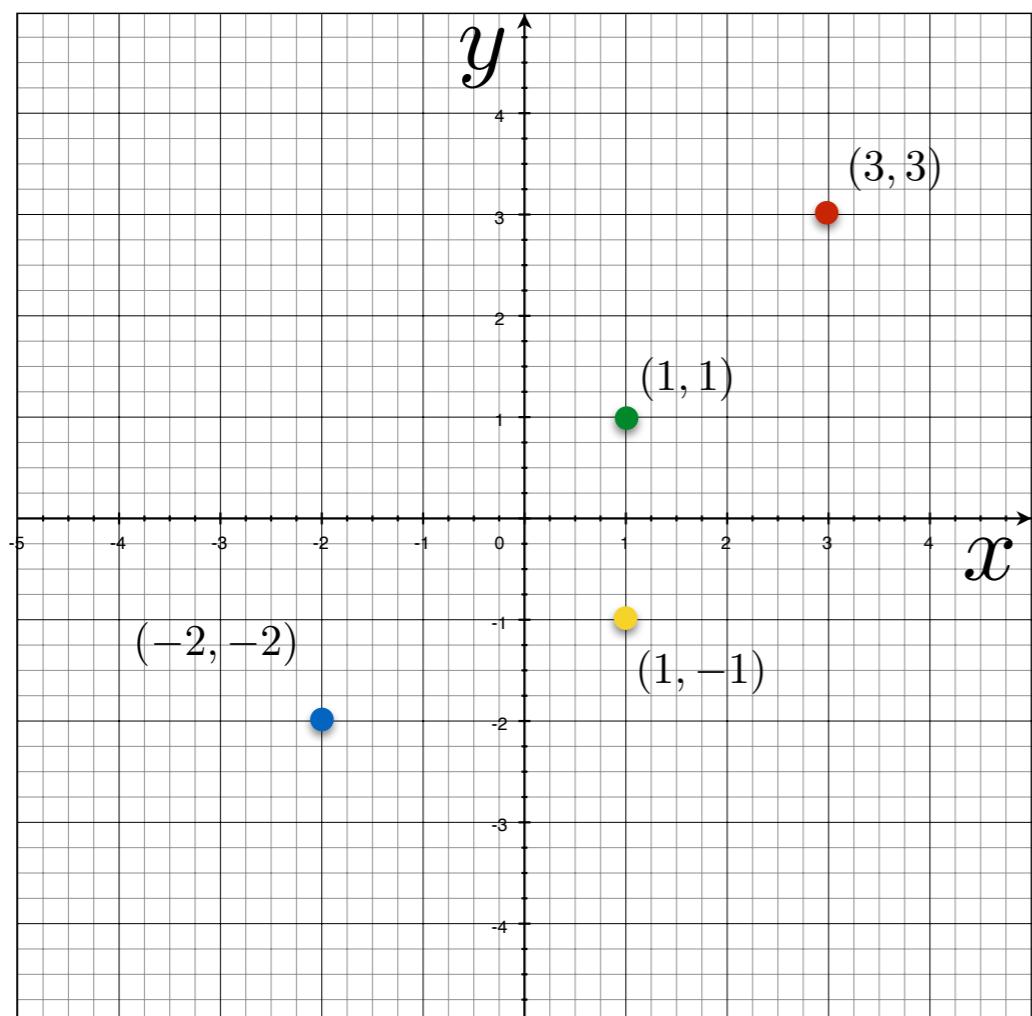
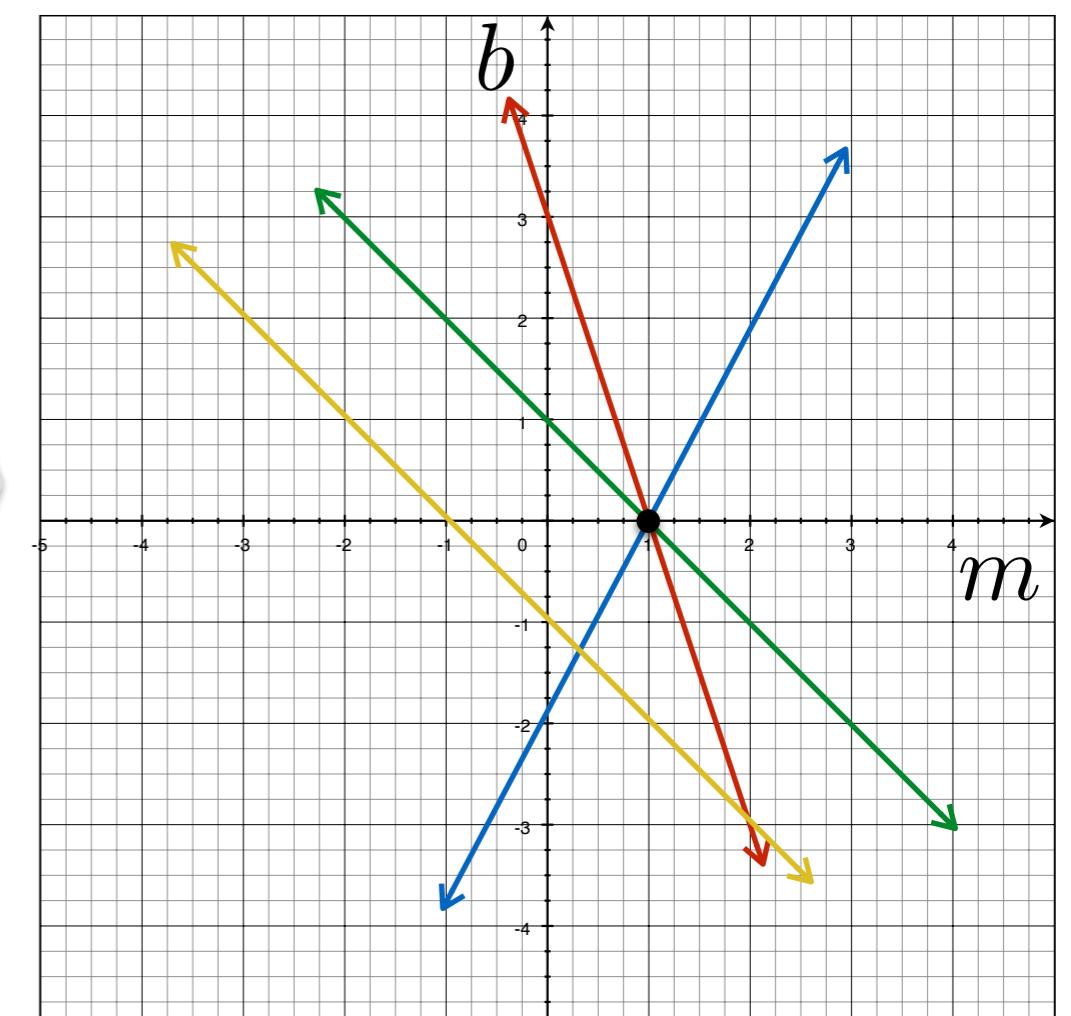


Image space



Parameter space

Is your method robust to measurement noise?

Is your 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 $A(m, c) = 0 \quad \forall m, c$

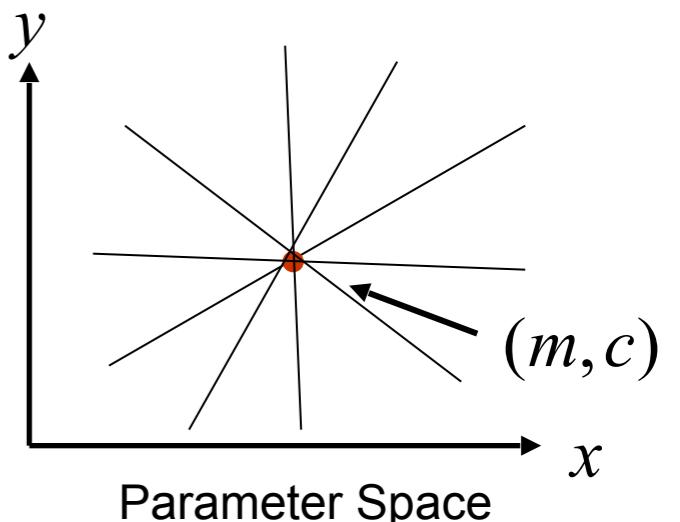
4. For each image edge (x_i, y_i)

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 $A(m, c)$



Parameter Space

$$A(m,c)$$

Problems with parameterization

What's wrong with the parameterization (m, c) ?

How big does the accumulator need to be?

$$A(m,c)$$

Problems with parameterization

What's wrong with the parameterization (m, c) ?

How big does the accumulator need to be?

$$A(m,c)$$

The space of m is huge! The space of c is huge!

Problems with parameterization

What's wrong with the parameterization (m, c) ?

How big does the accumulator need to be?

$$A(m,c)$$

The space of m is huge! The space of c is huge!

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

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)

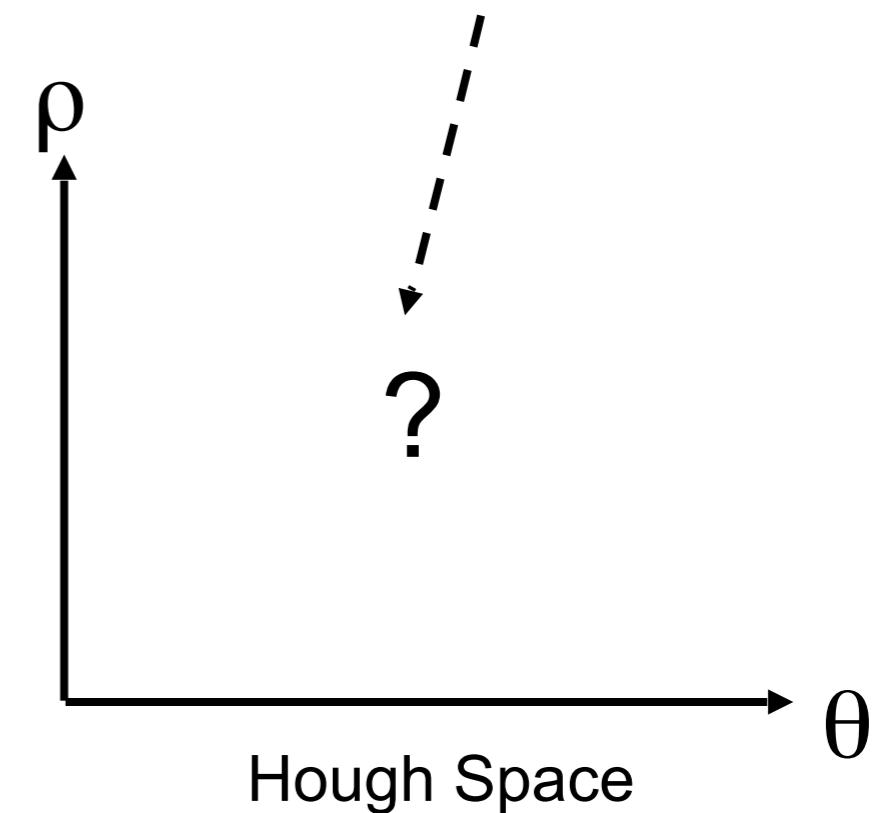
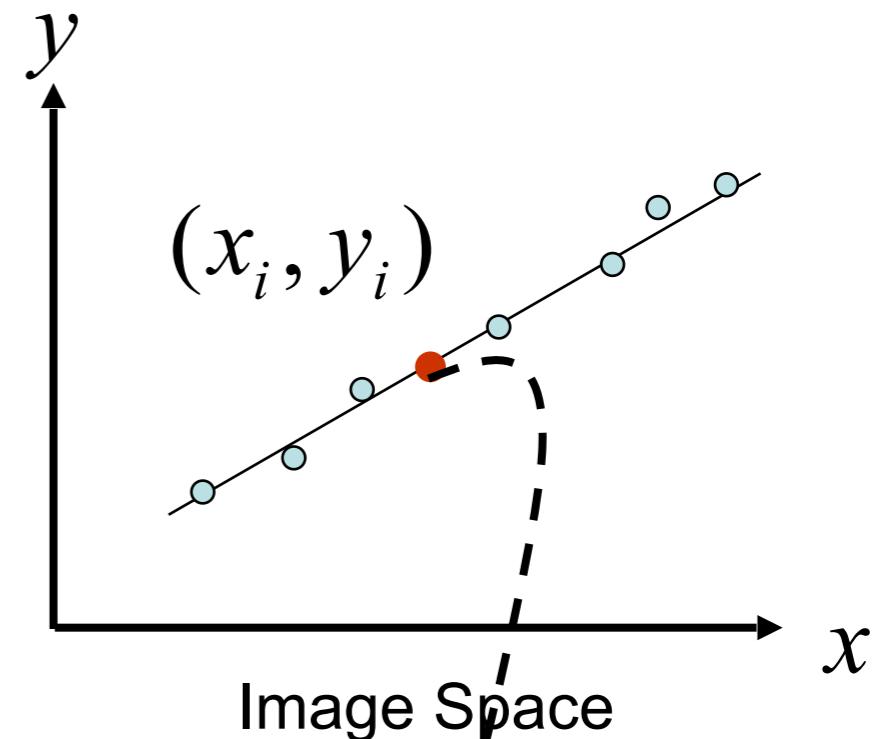
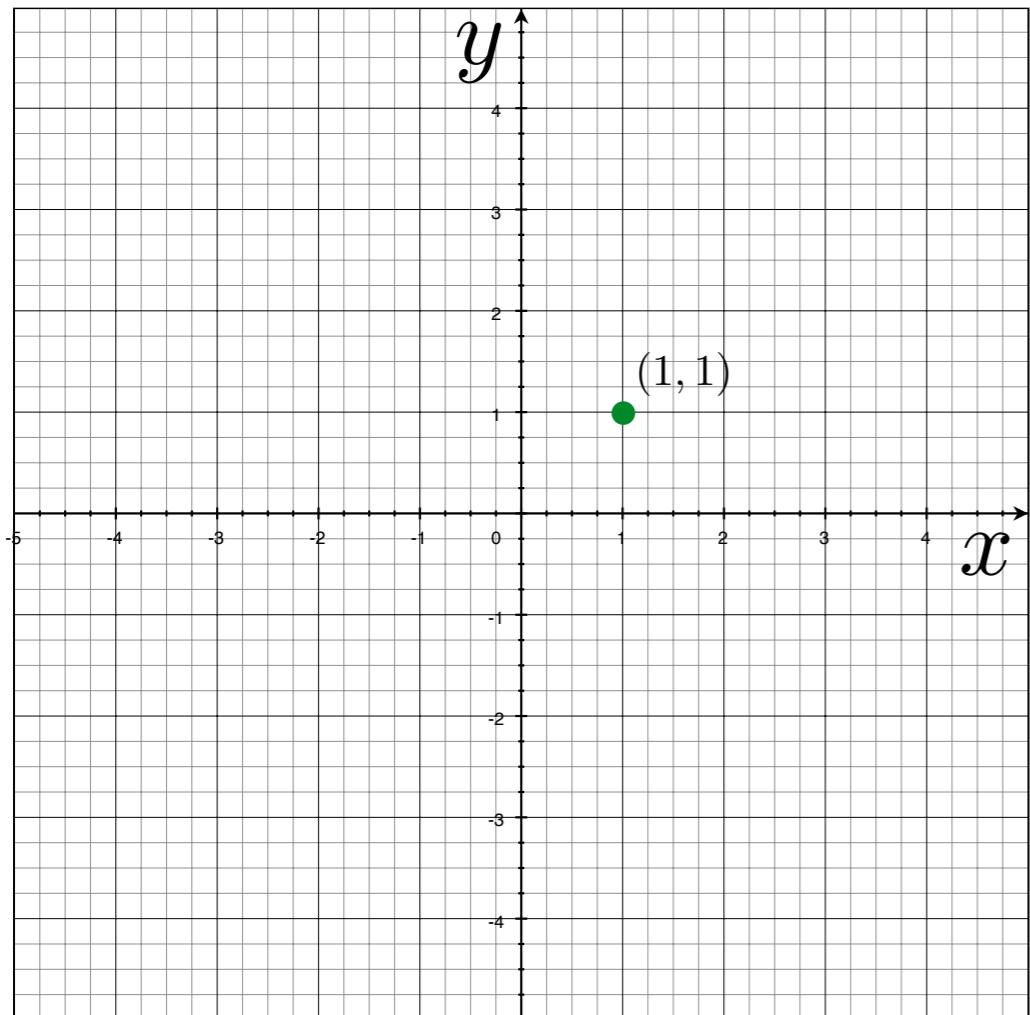


Image and parameter space

variables
 $y = mx + b$
parameters



a point becomes a wave

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

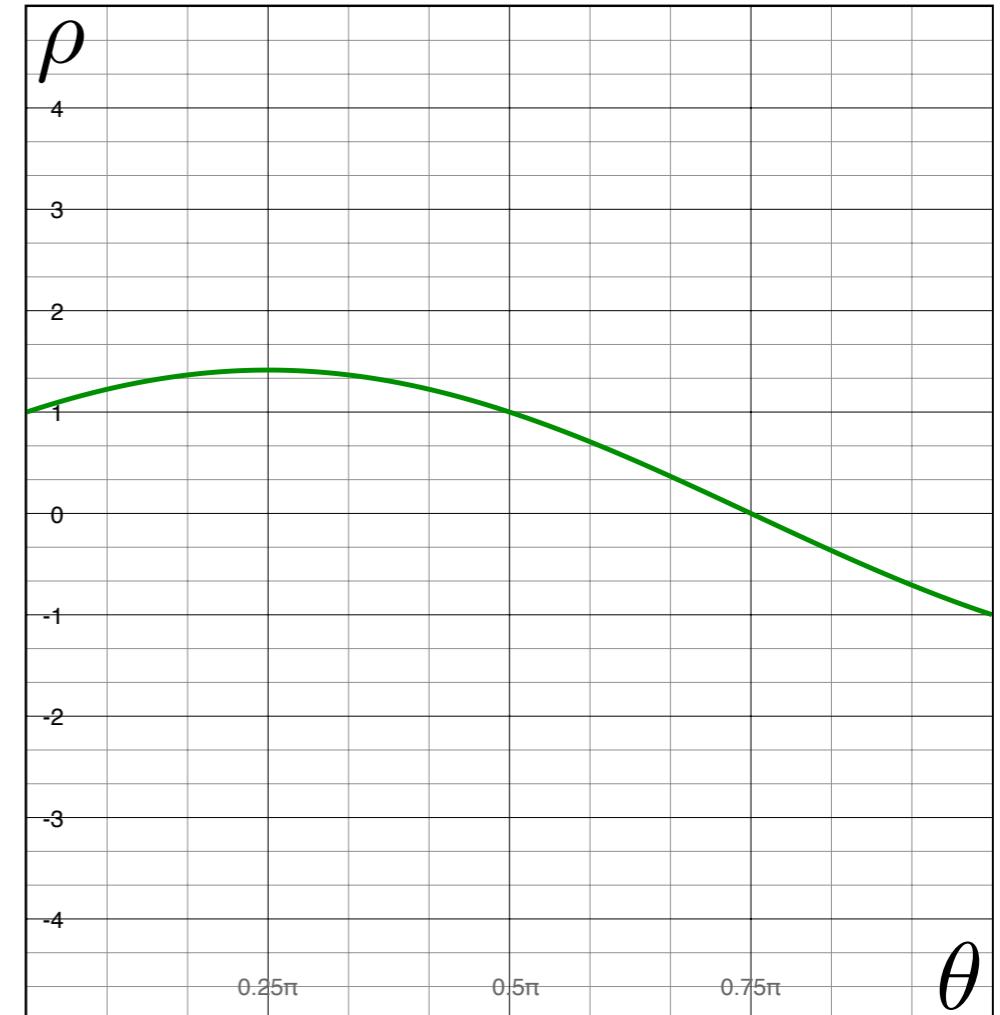
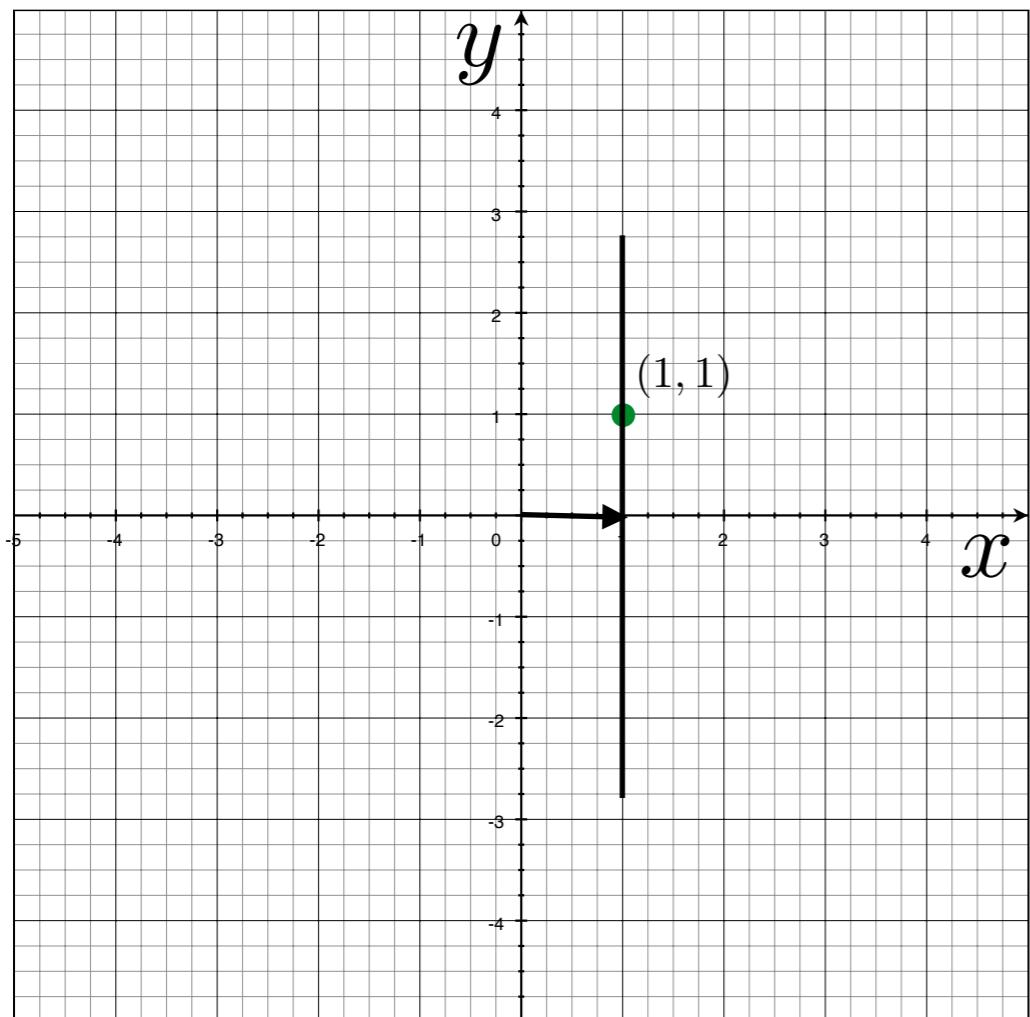


Image space

Parameter space

Image and parameter space

variables
 $y = mx + b$
parameters



a line
becomes
a point

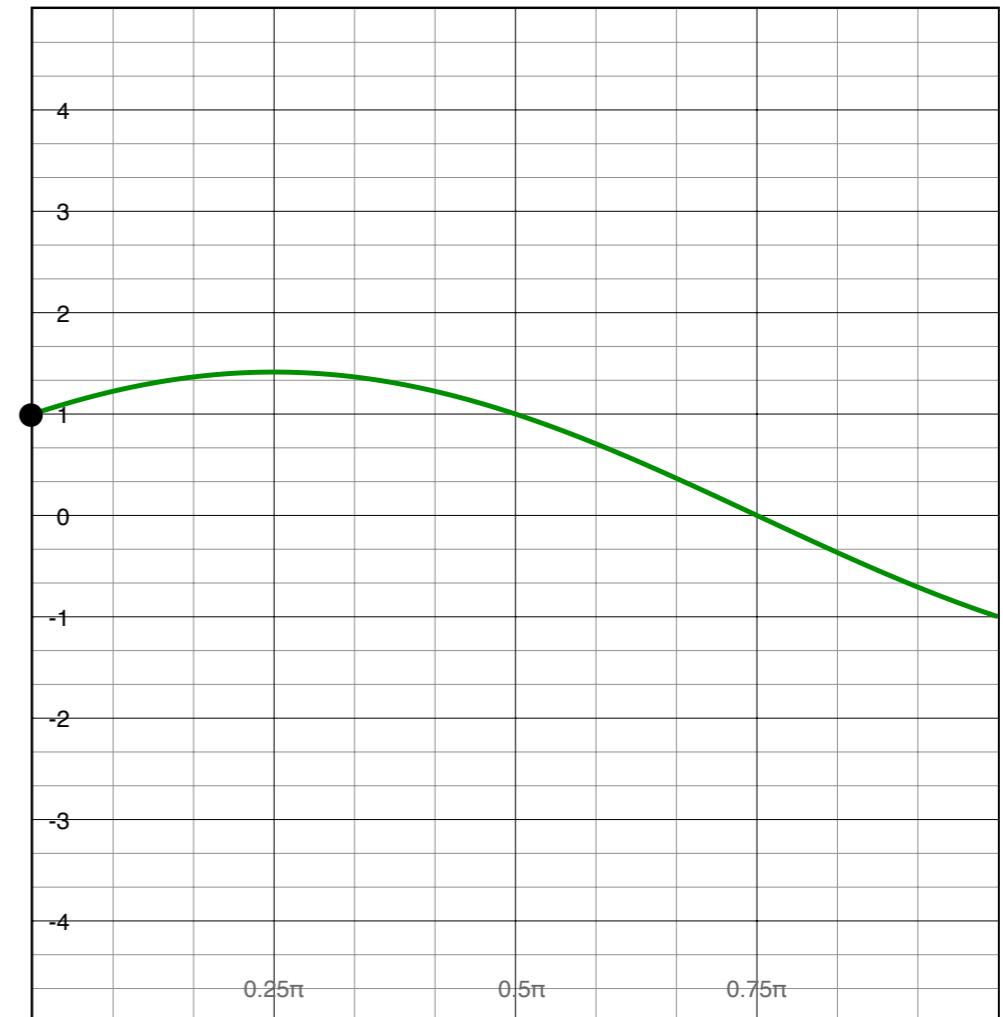


Image space

Parameter space

Image and parameter space

variables
 $y = mx + b$
parameters

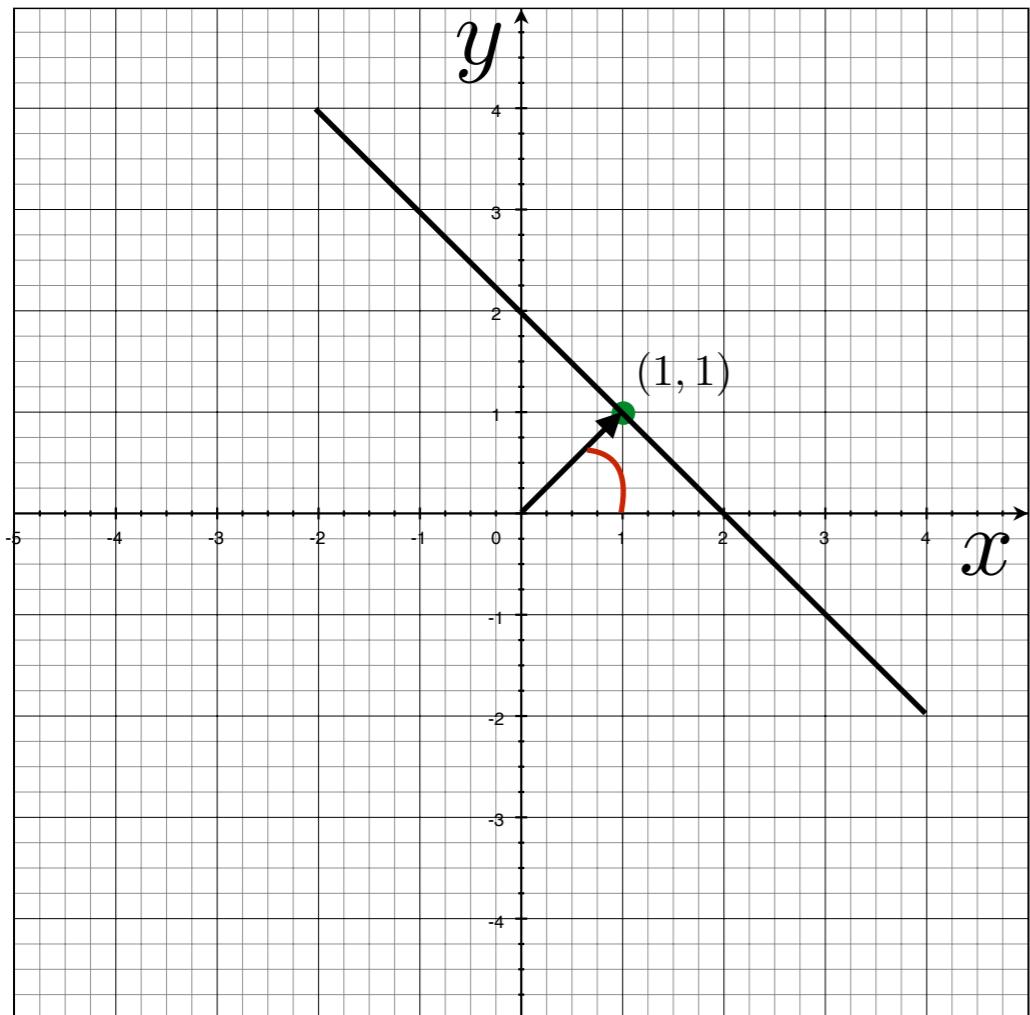
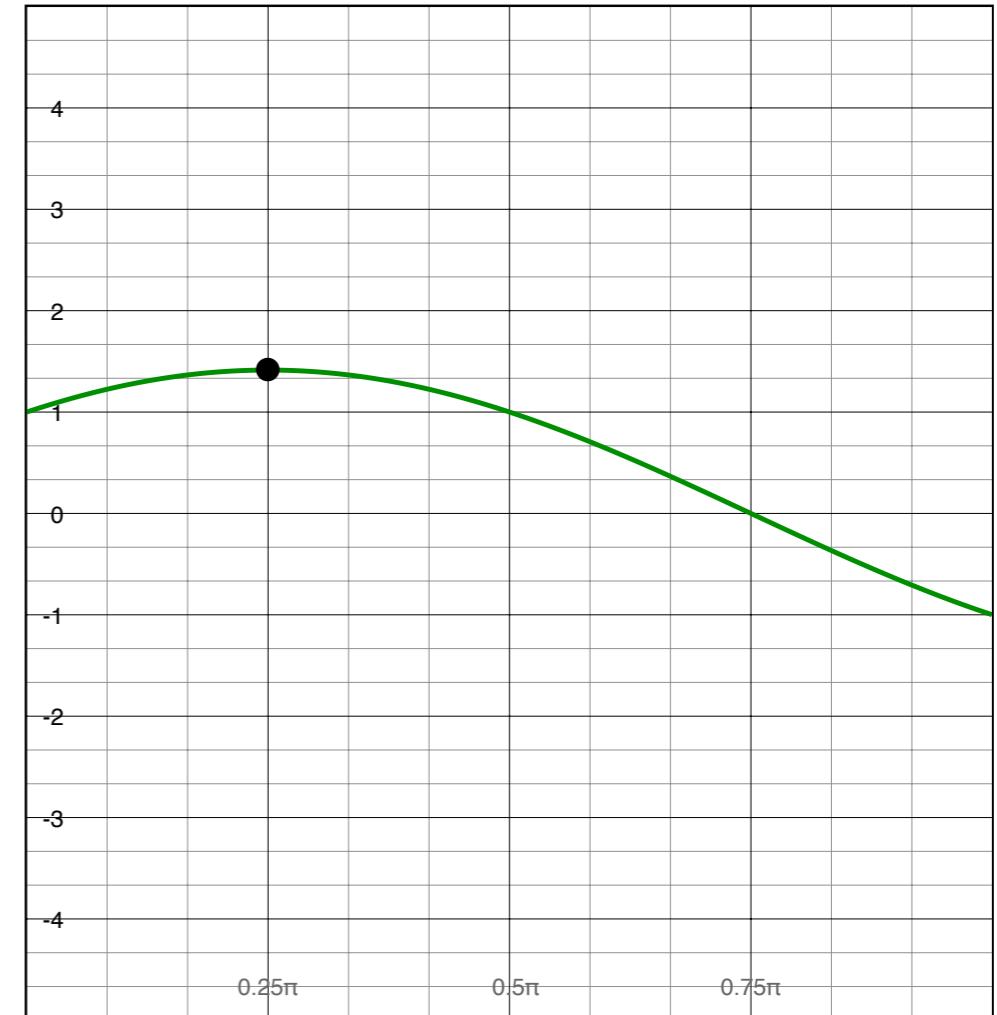


Image space

a line becomes a point



Parameter space

Image and parameter space

variables
 $y = mx + b$
parameters

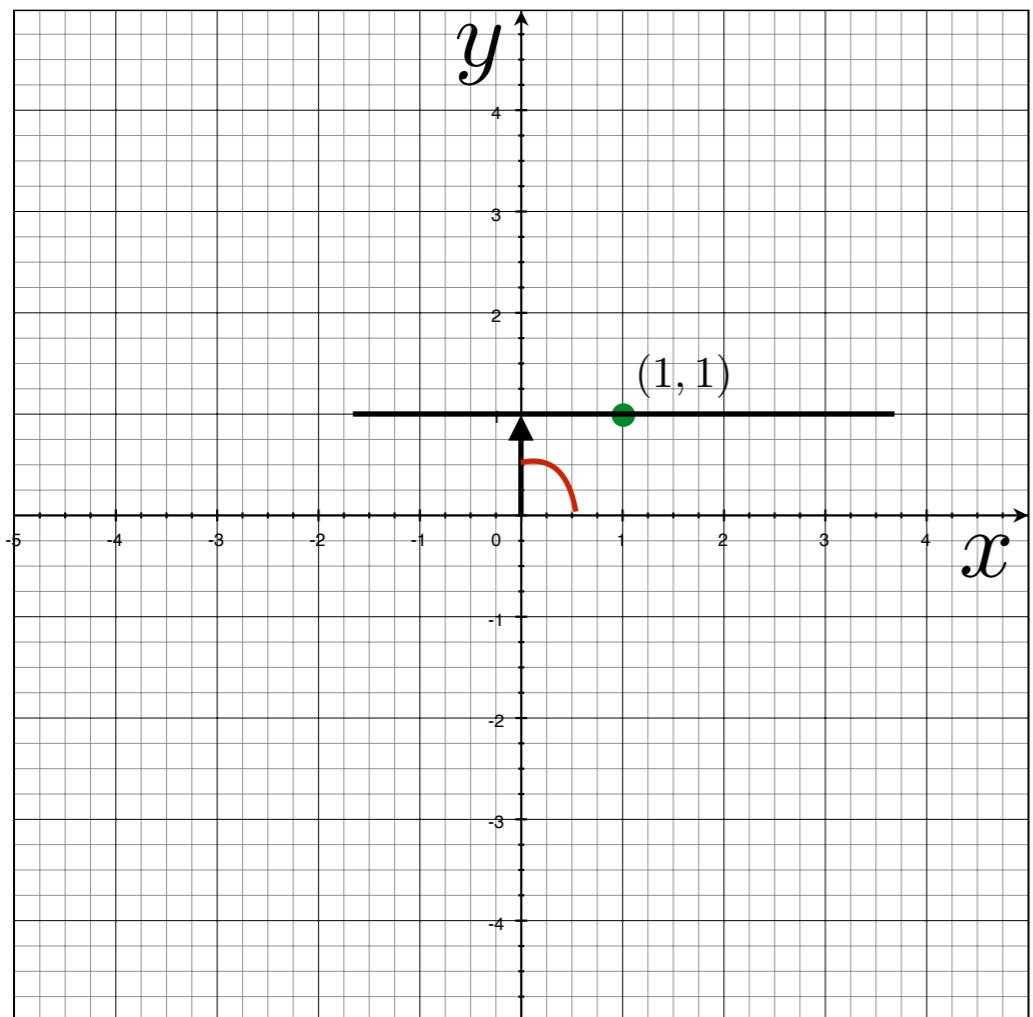
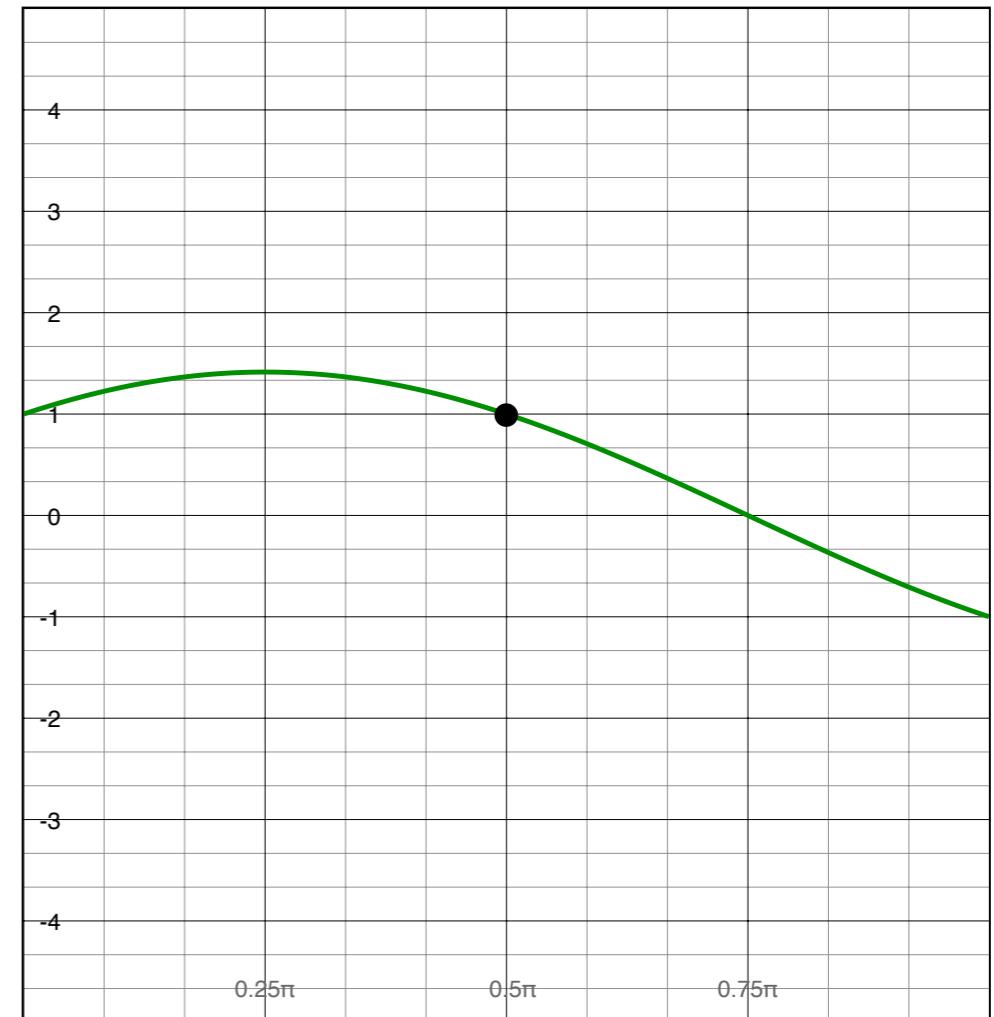


Image space

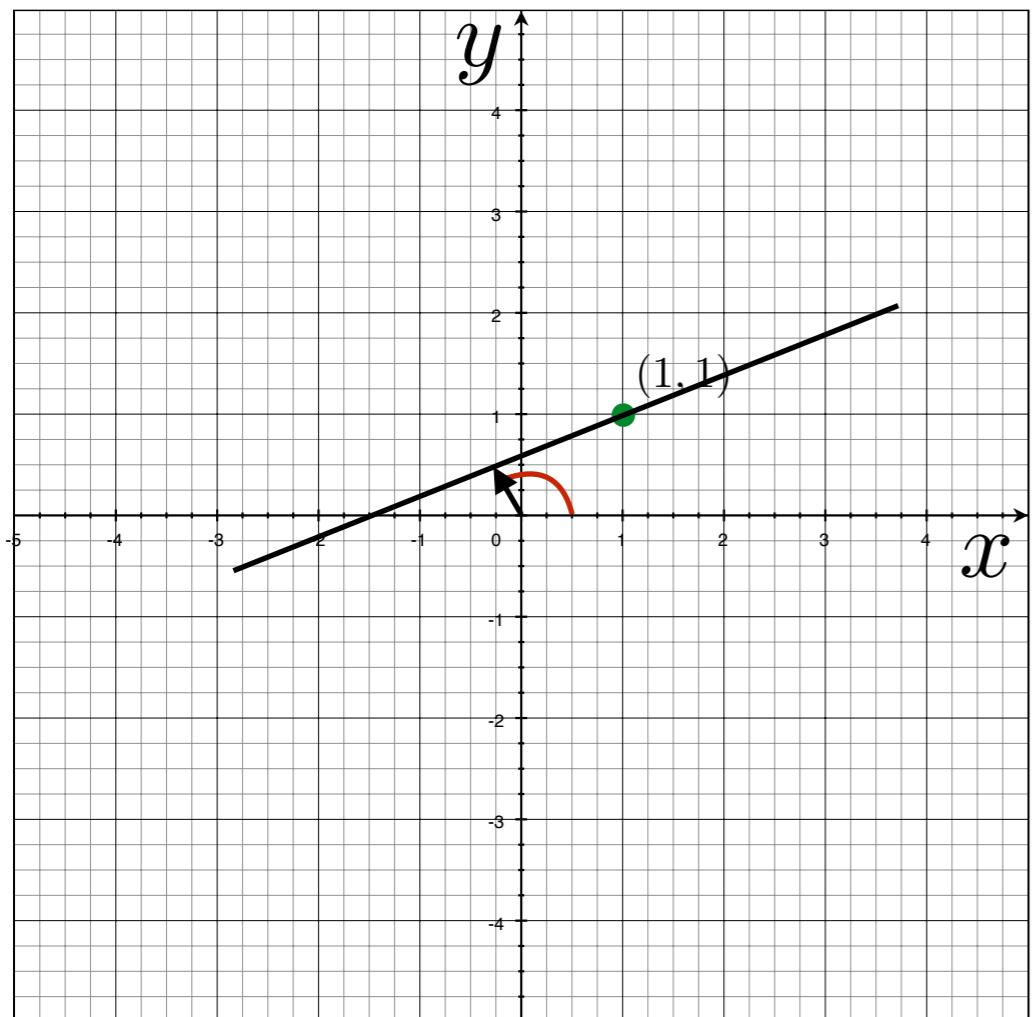
a line becomes a point



Parameter space

Image and parameter space

variables
 $y = mx + b$
parameters



a line
becomes
a point

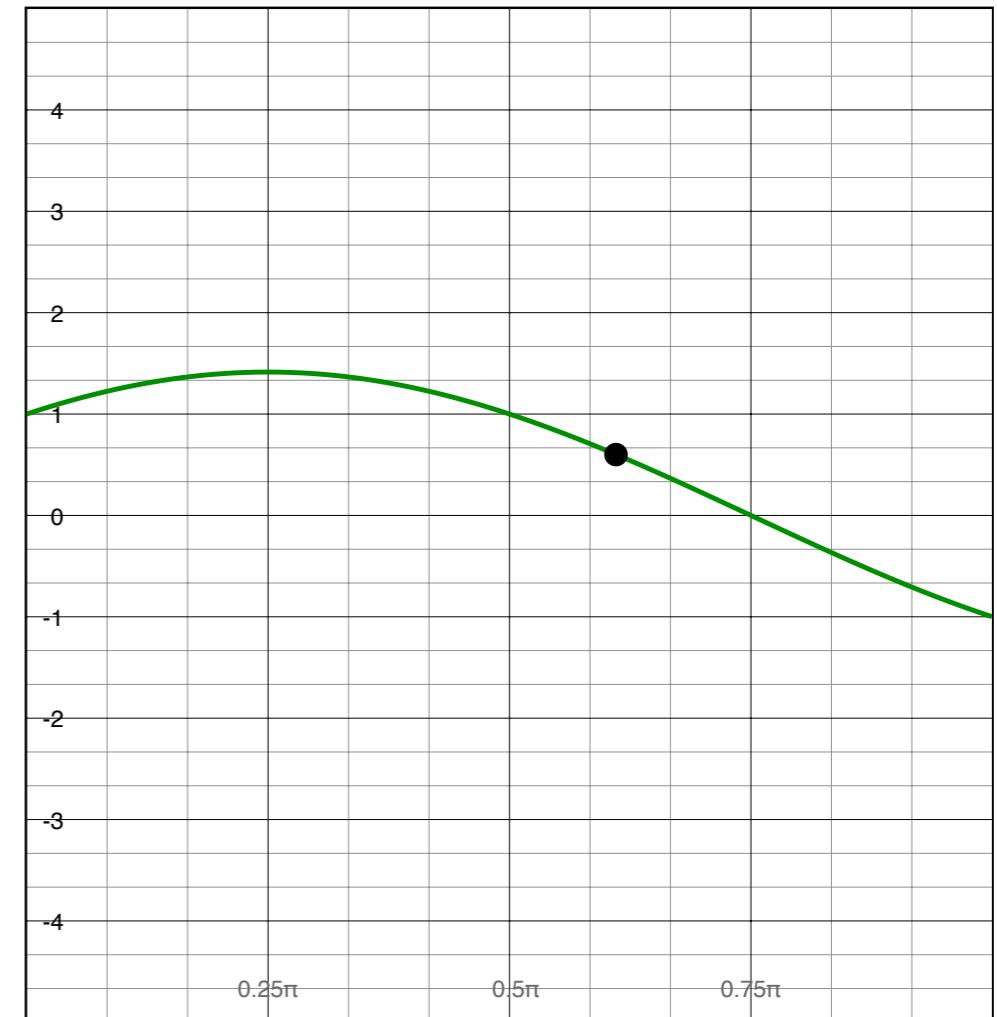
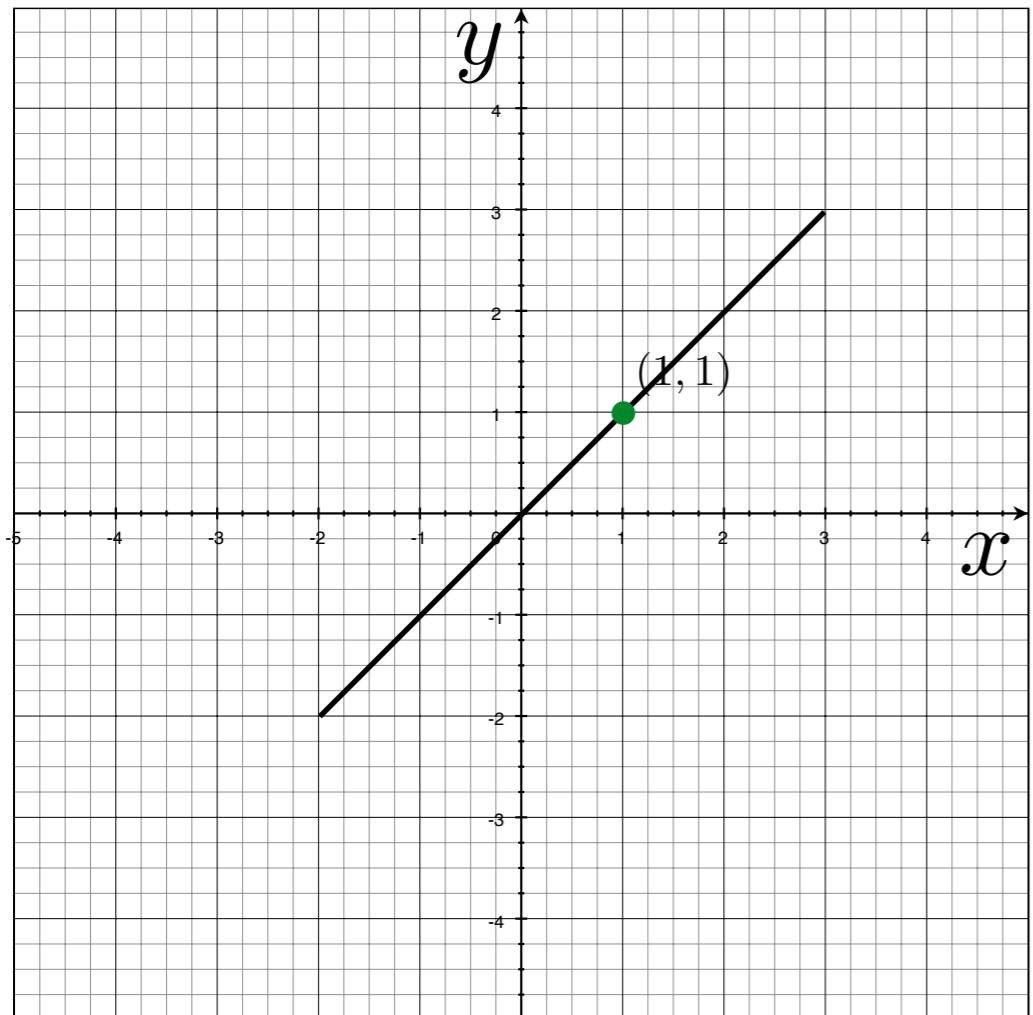


Image space

Parameter space

Image and parameter space

variables
 $y = mx + b$
parameters



a line
becomes
a point

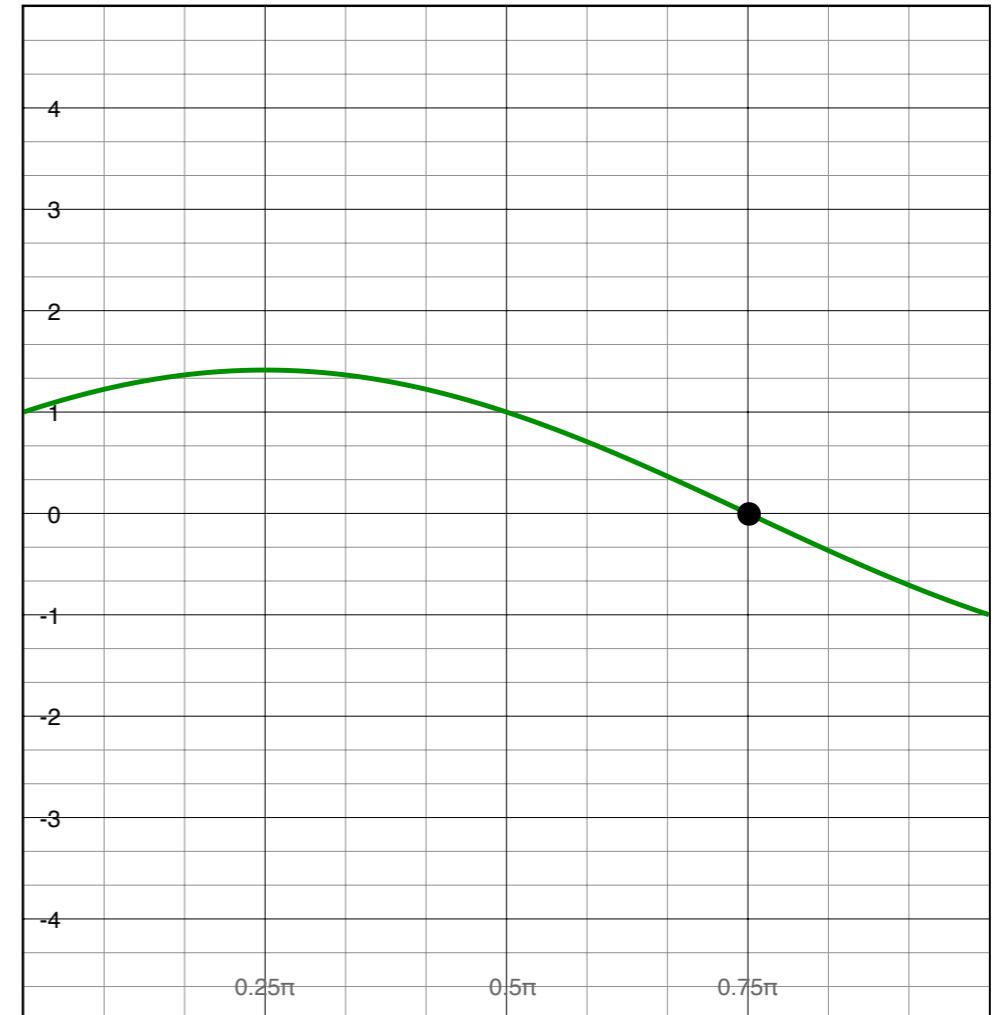
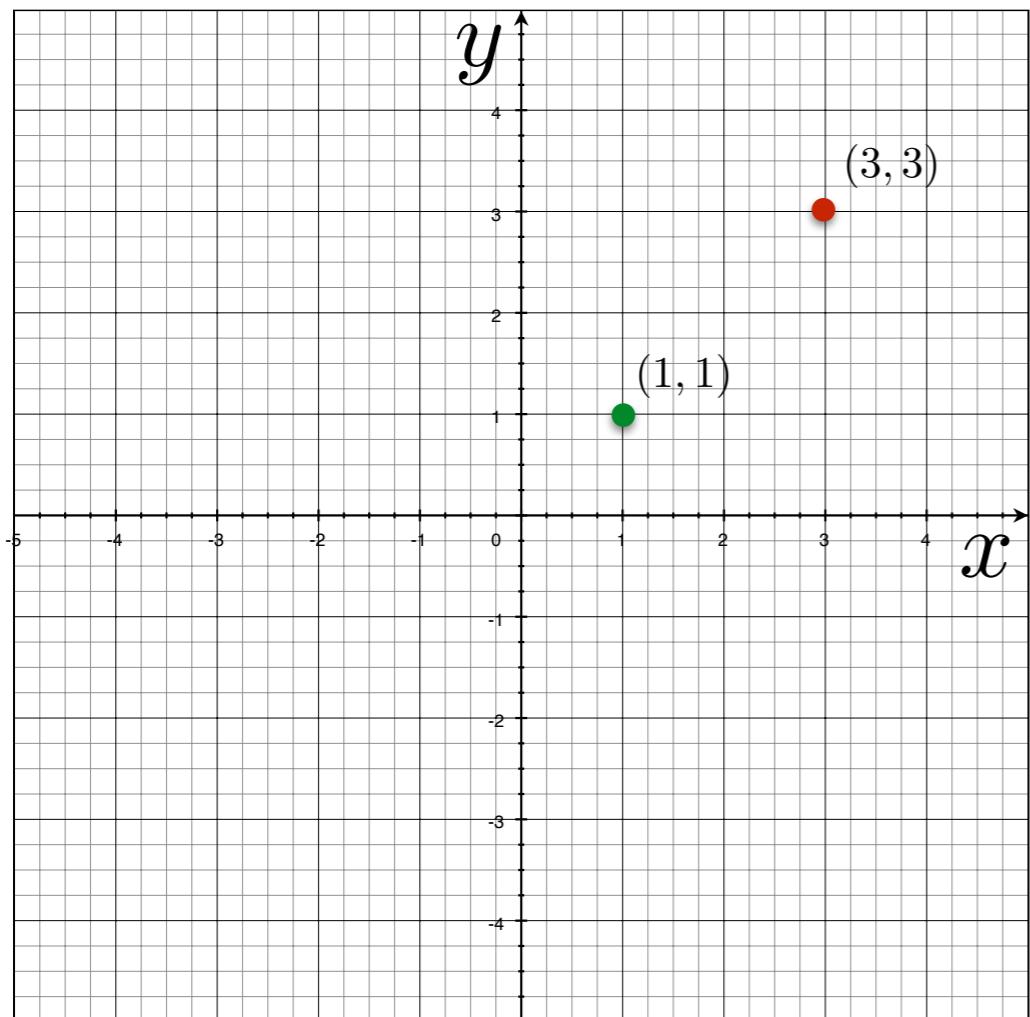


Image space

Parameter space

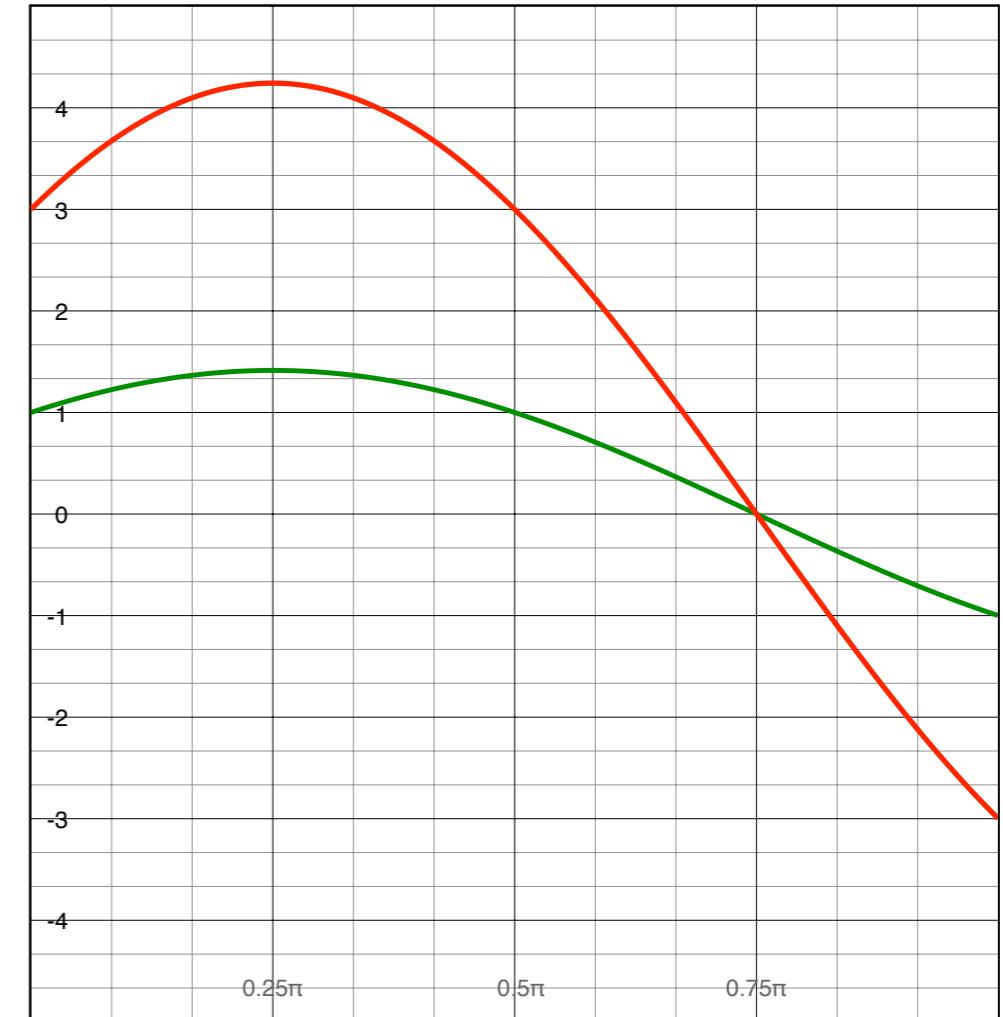
Image and parameter space

variables
 $y = mx + b$
parameters



two points
become
?

Image space



Parameter space

Image and parameter space

$$y = mx + b$$

variables
parameters

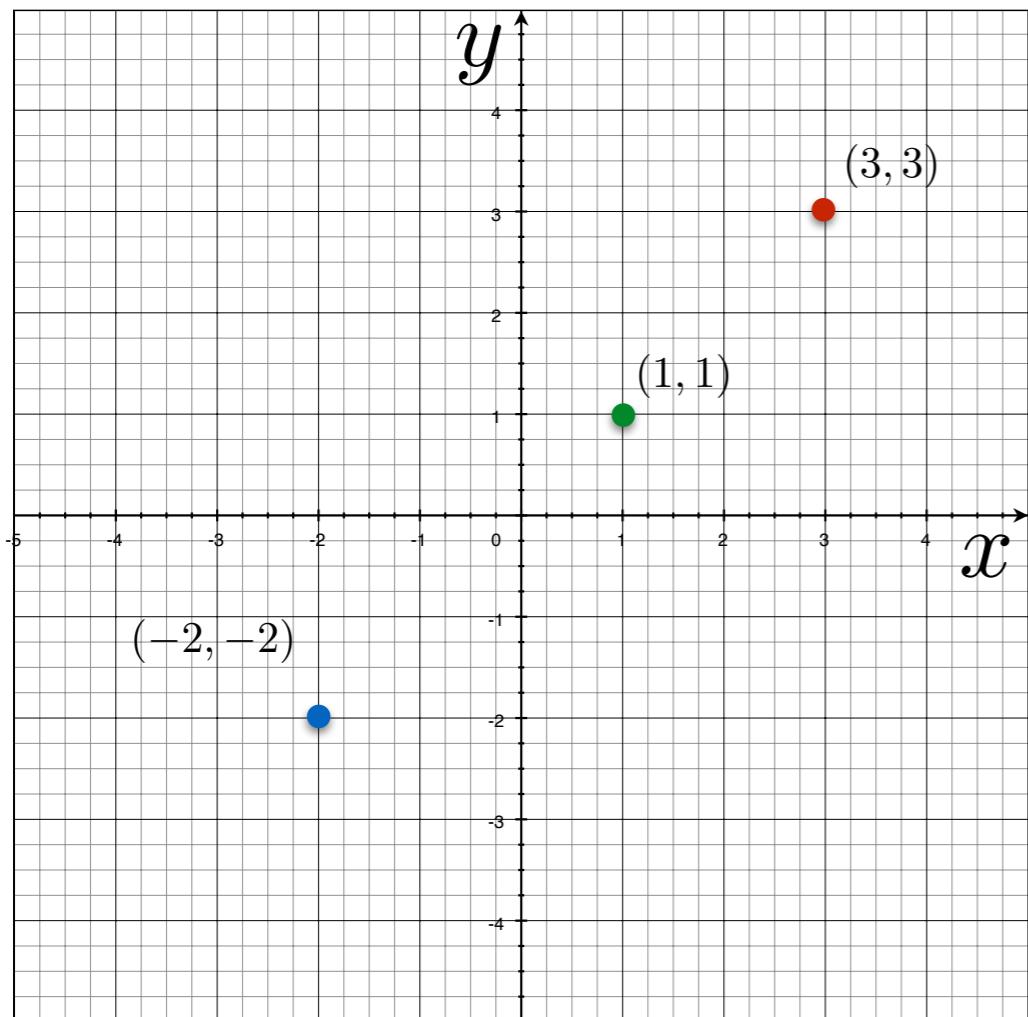
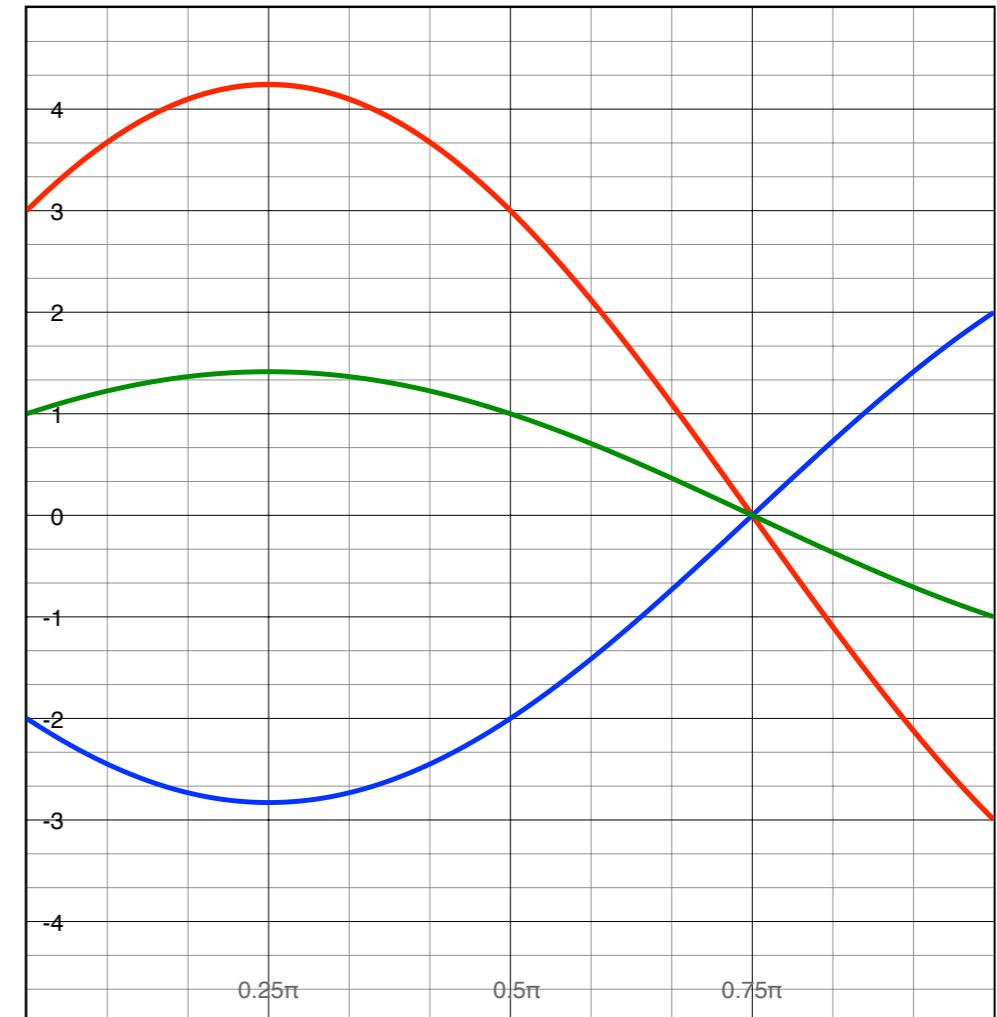


Image space

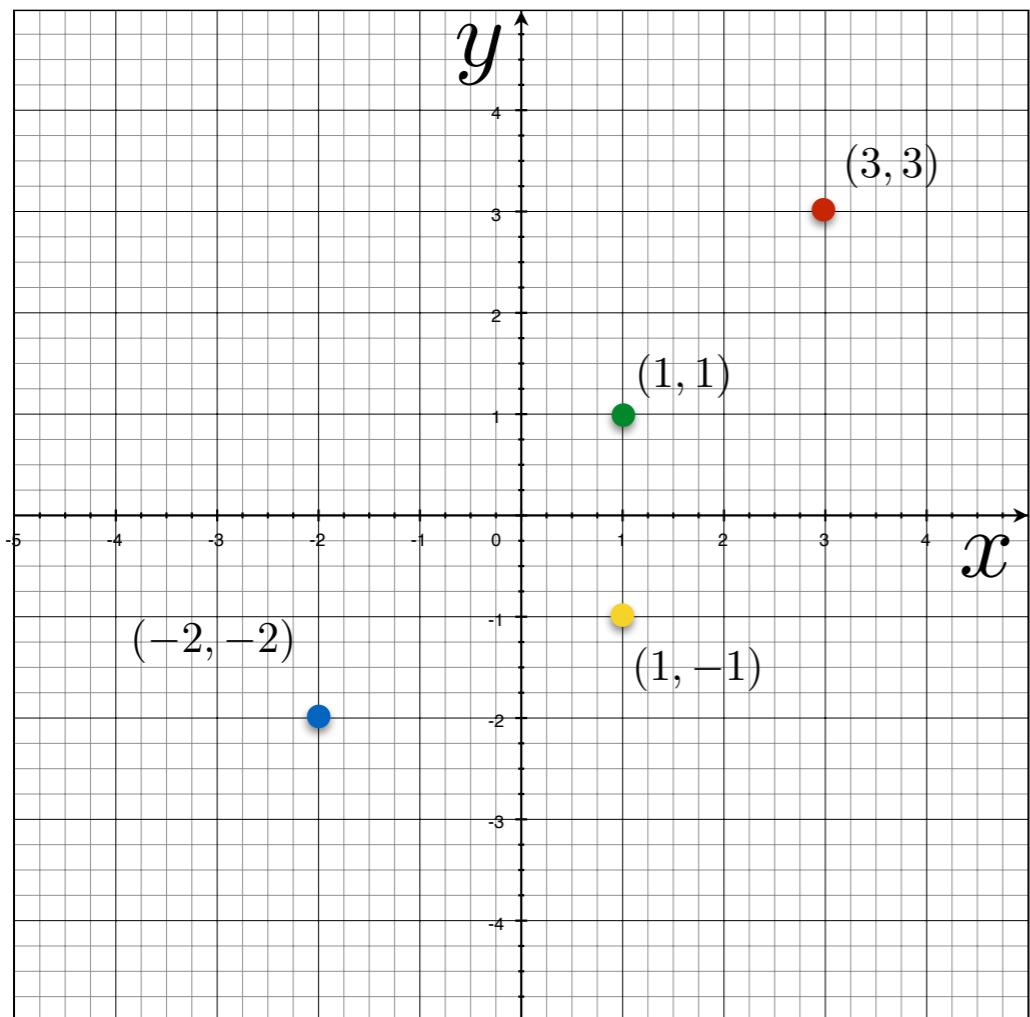
three points
become
?



Parameter space

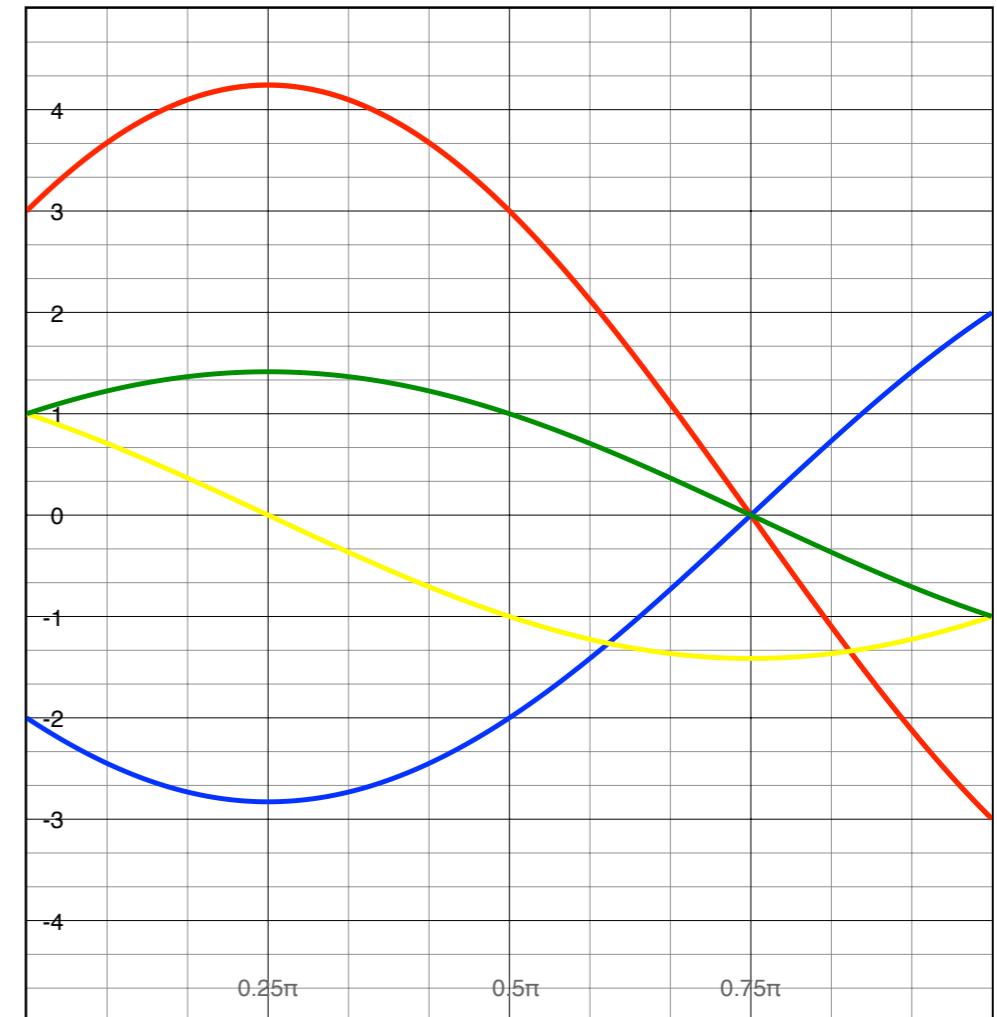
Image and parameter space

variables
 $y = mx + b$
parameters



four points
become
?

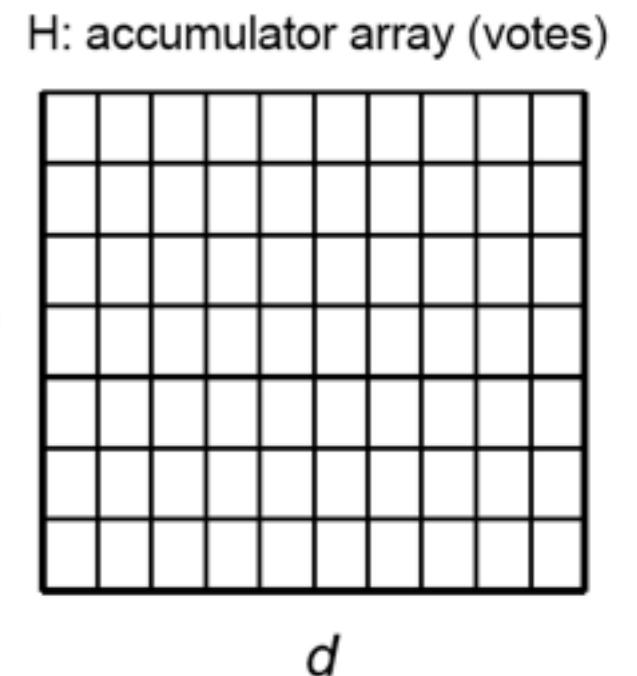
Image space



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



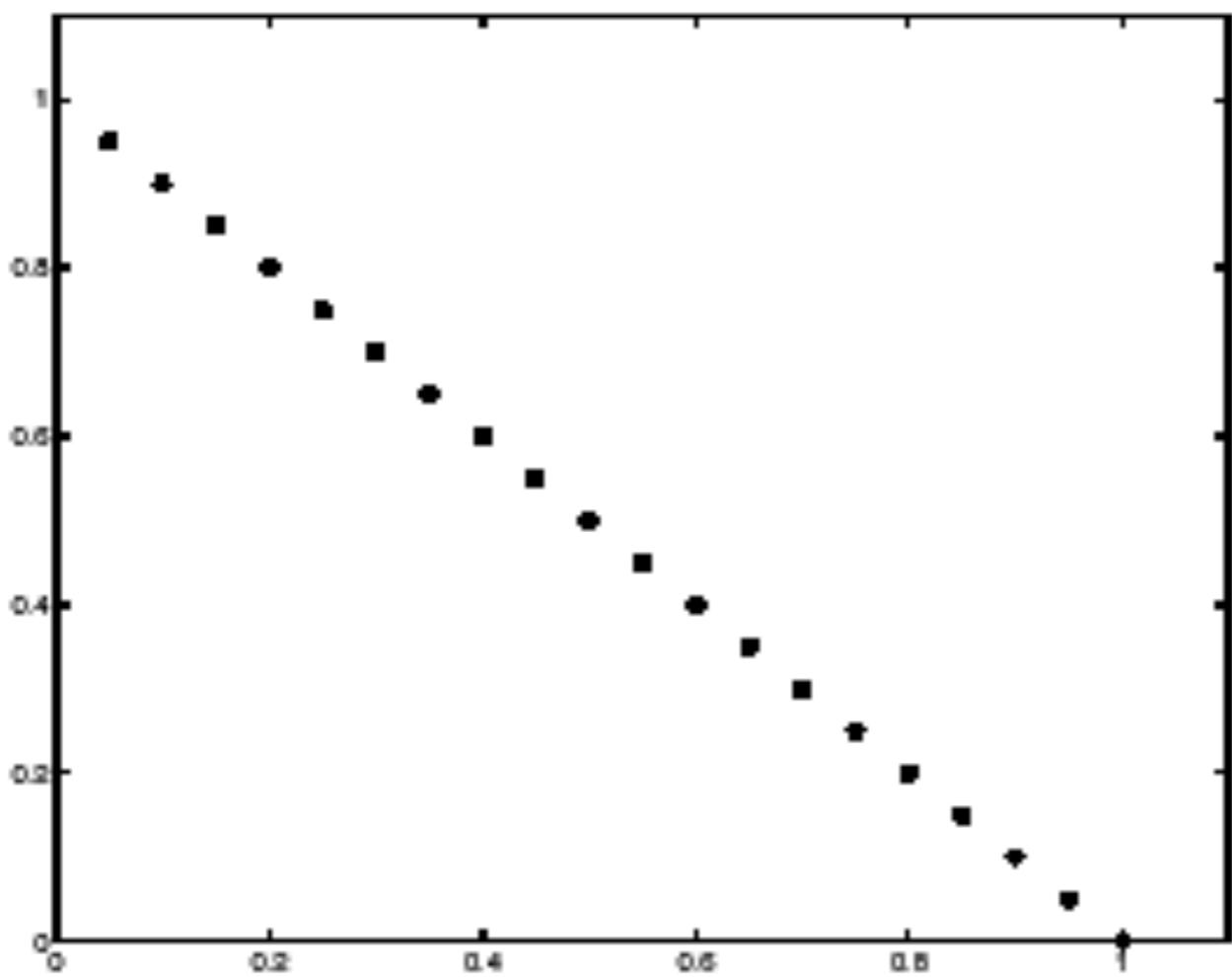
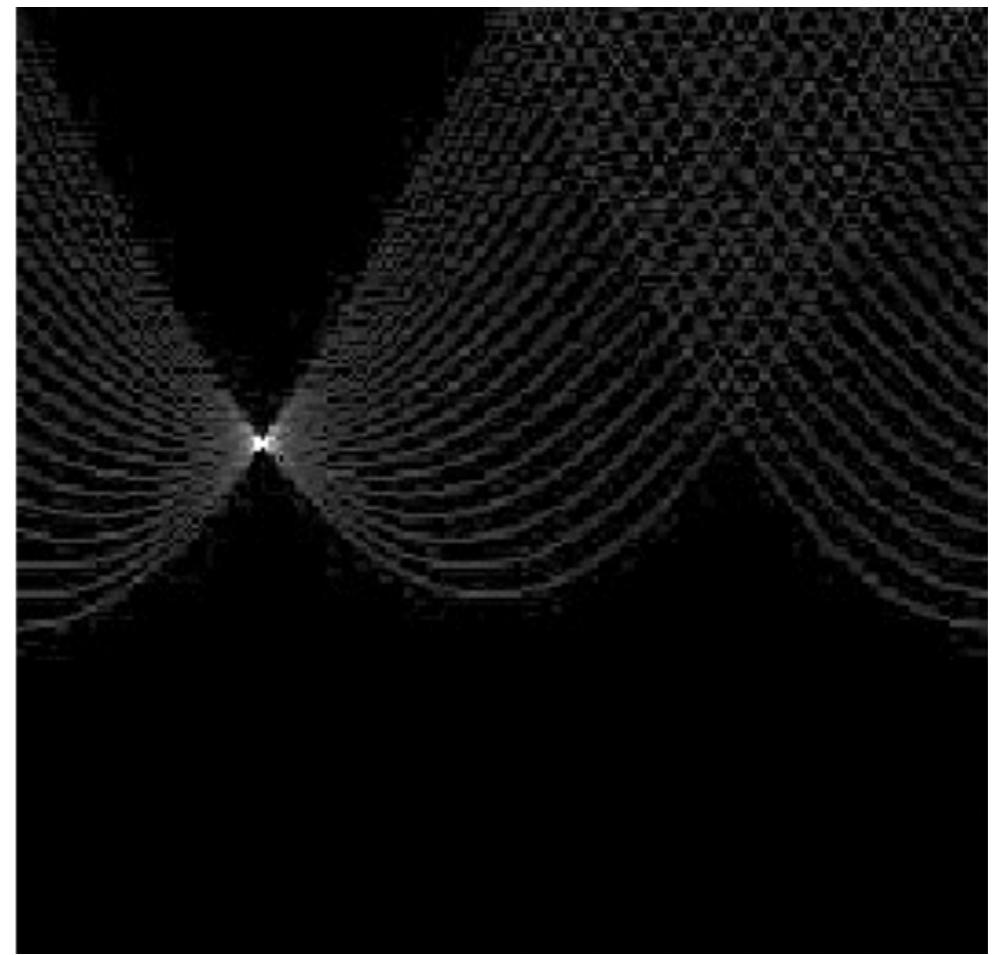


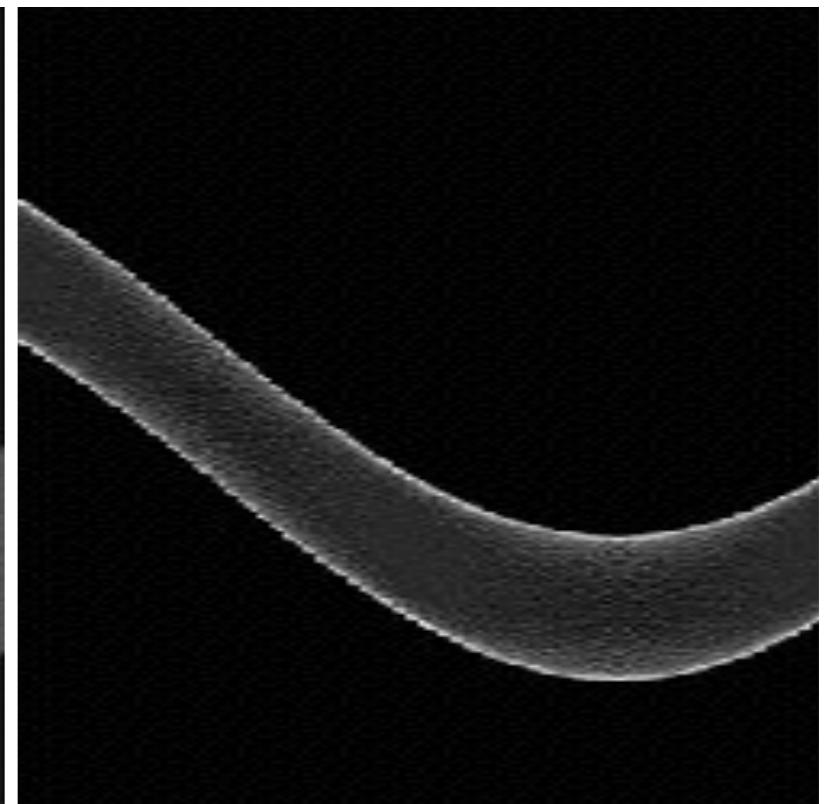
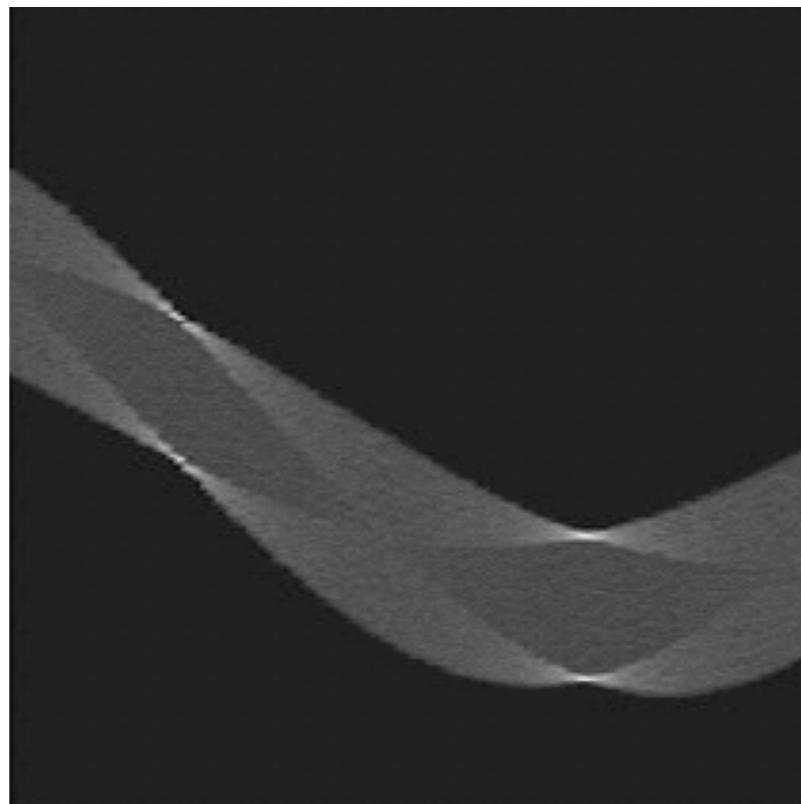
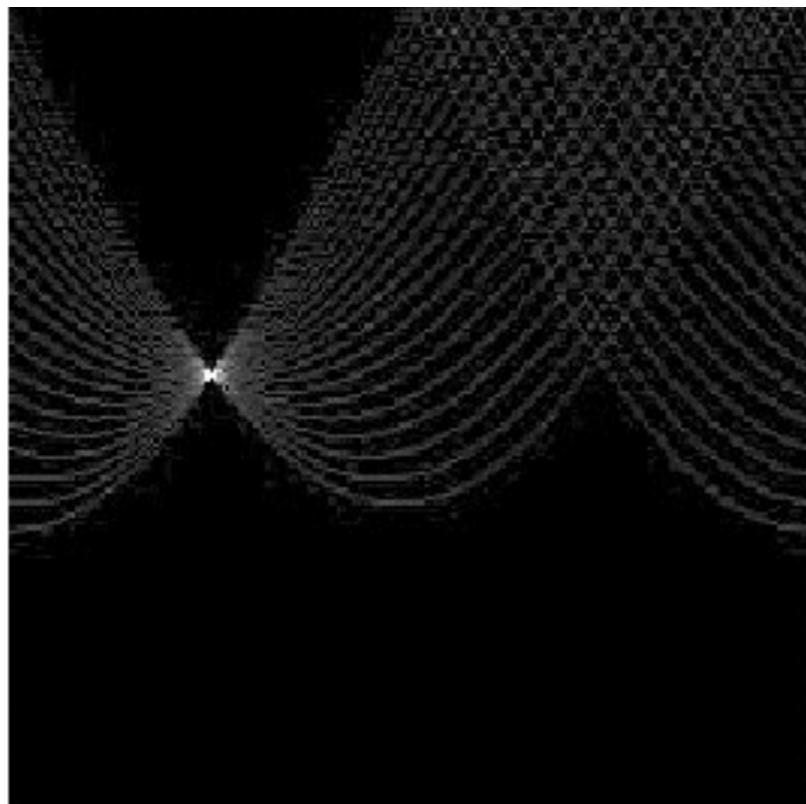
Image space



Votes

Basic shapes

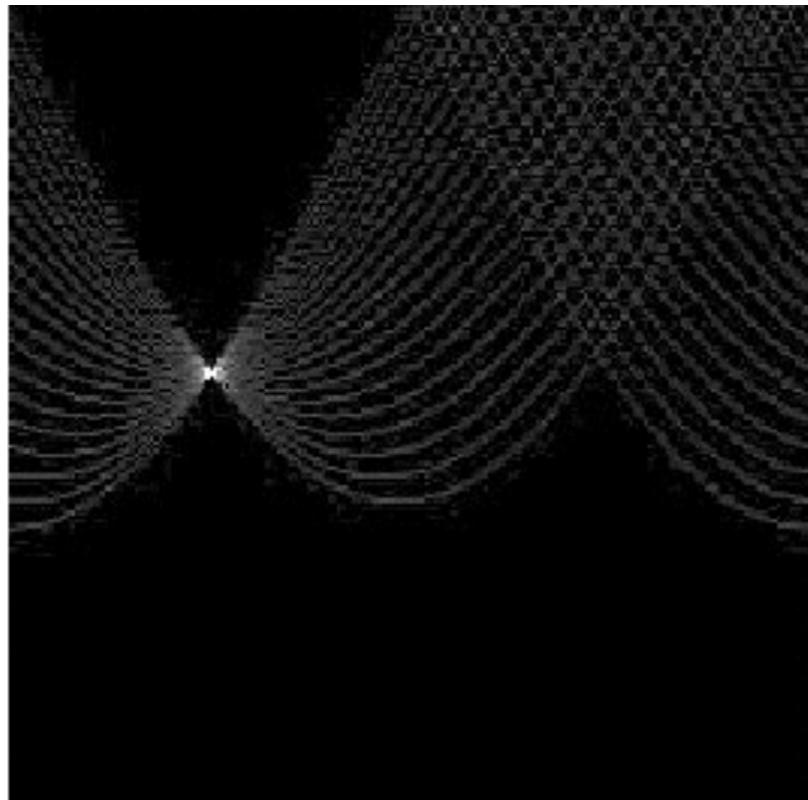
(in polar parameter space)



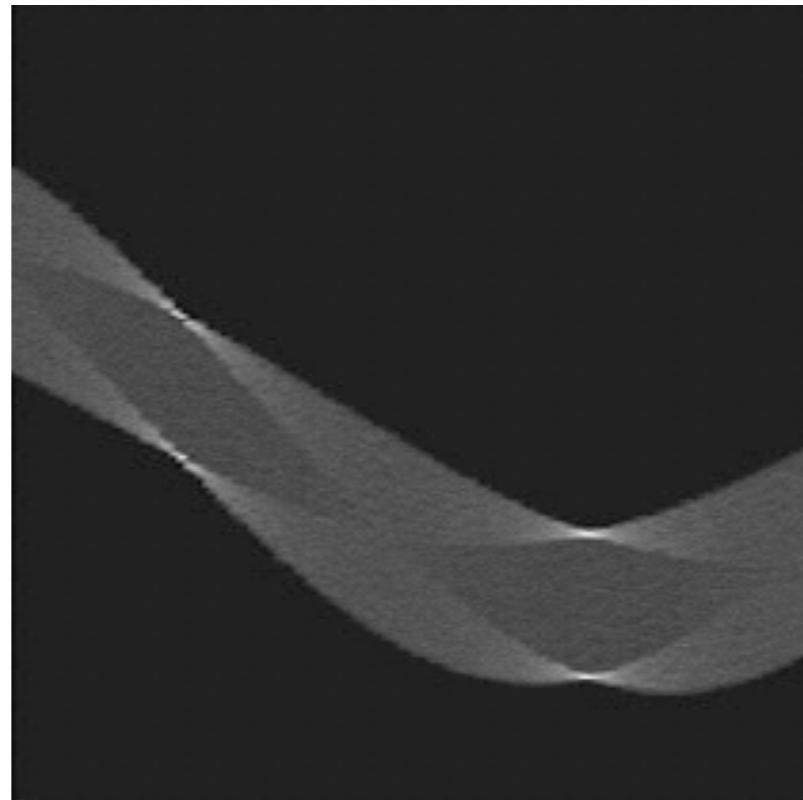
can you guess the shape?

Basic shapes

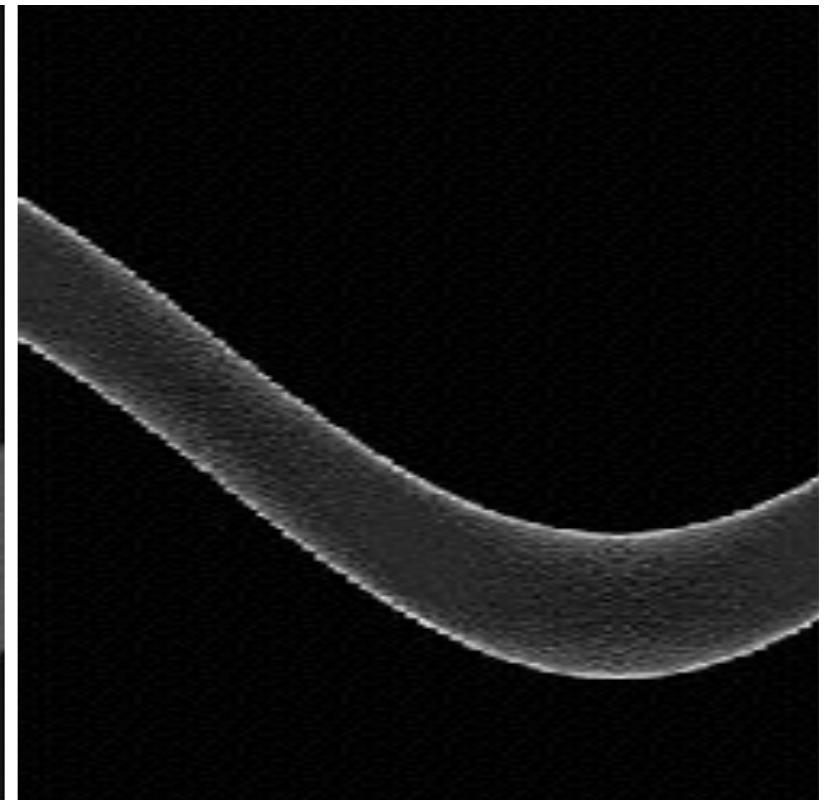
(in polar parameter space)



line

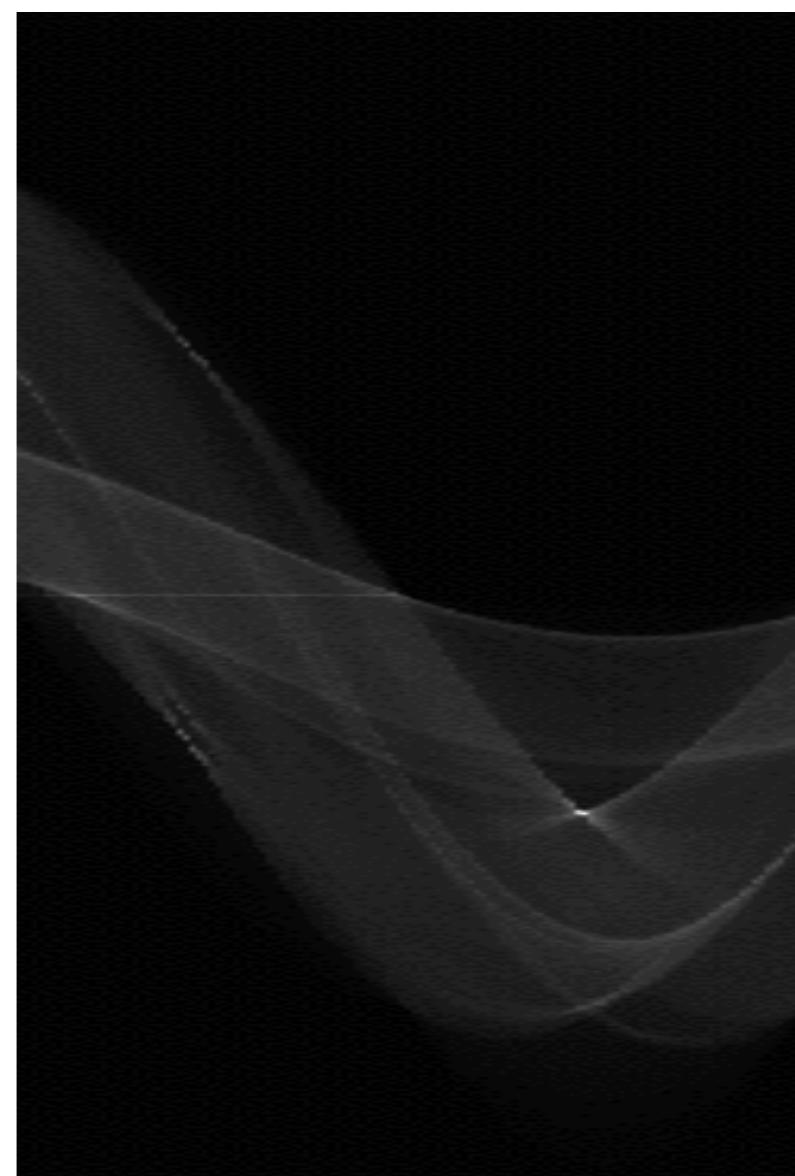
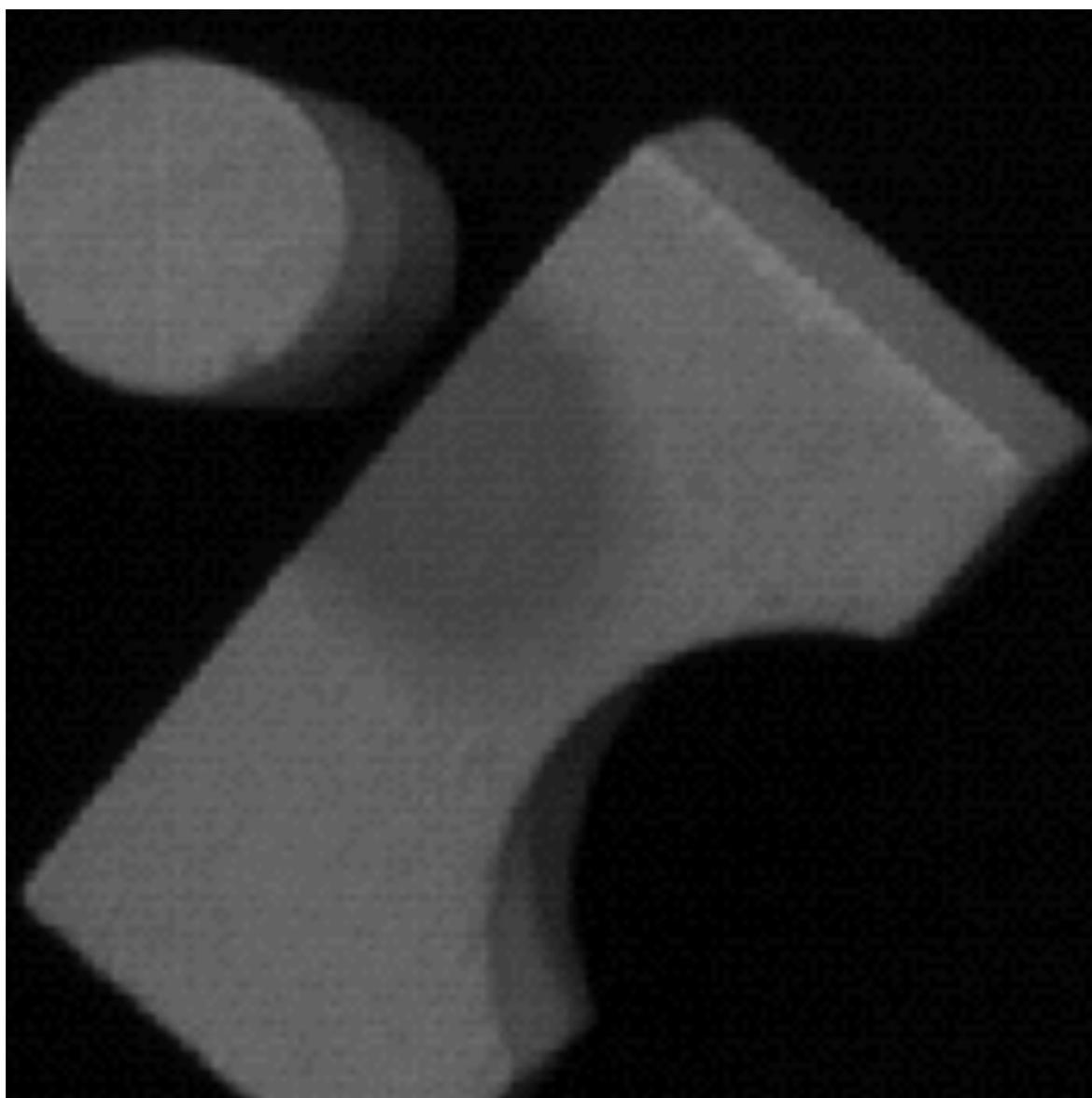


rectangle

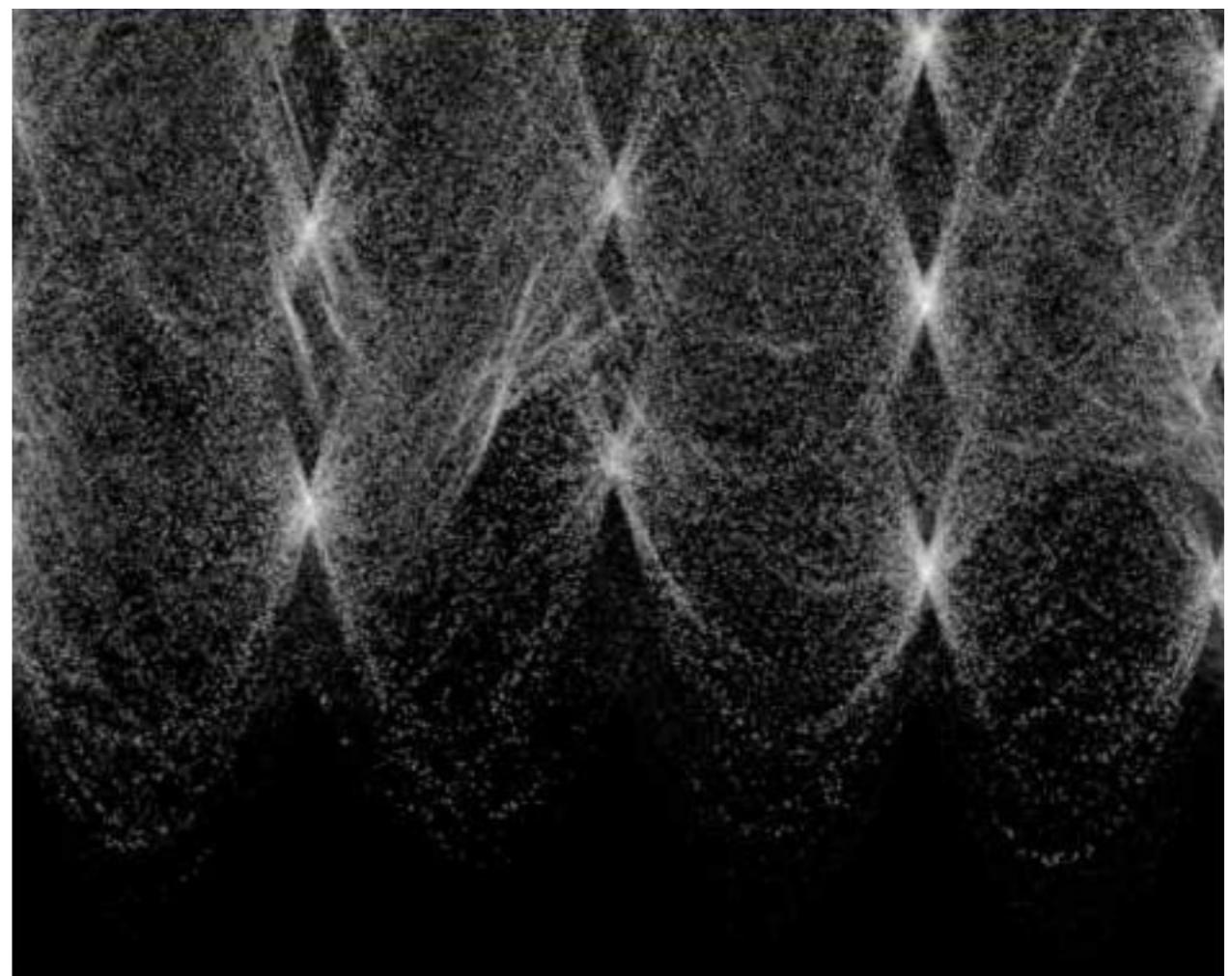


circle

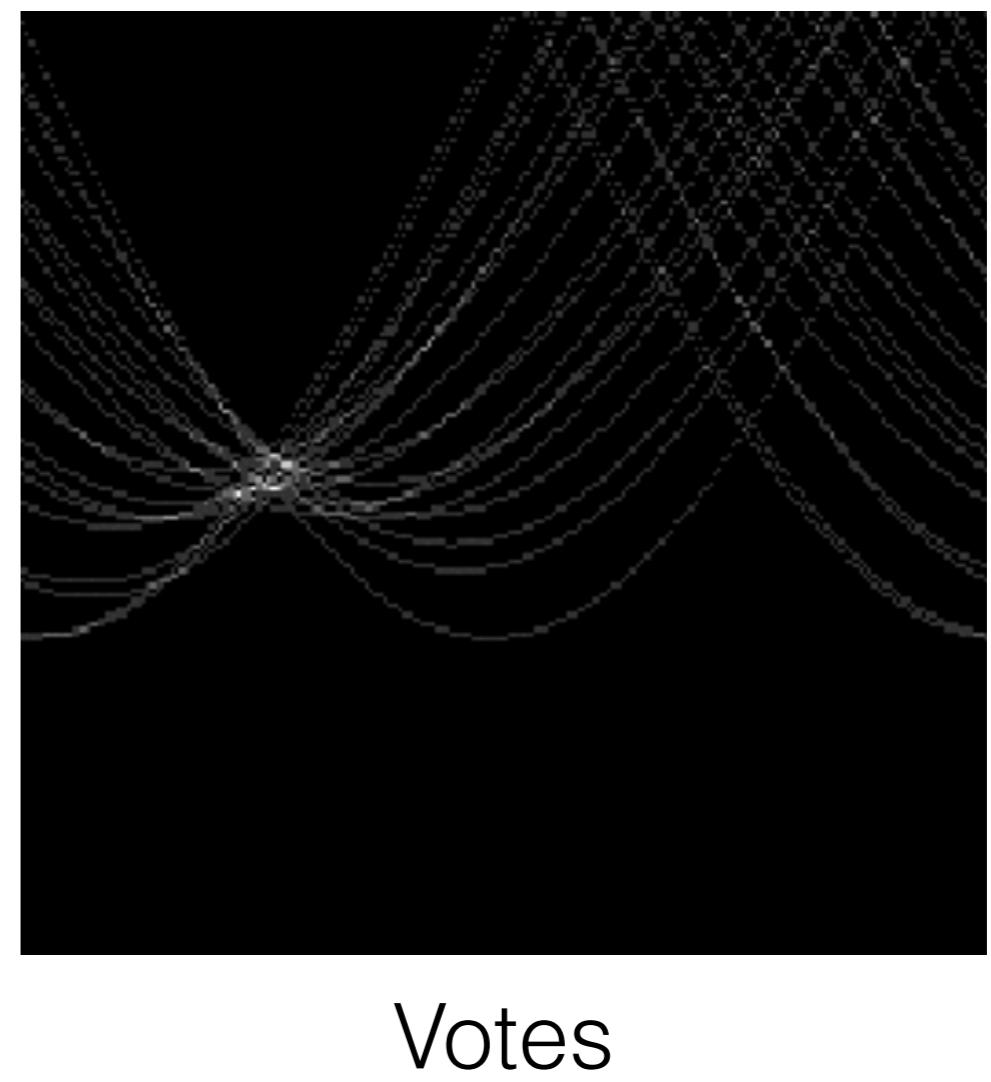
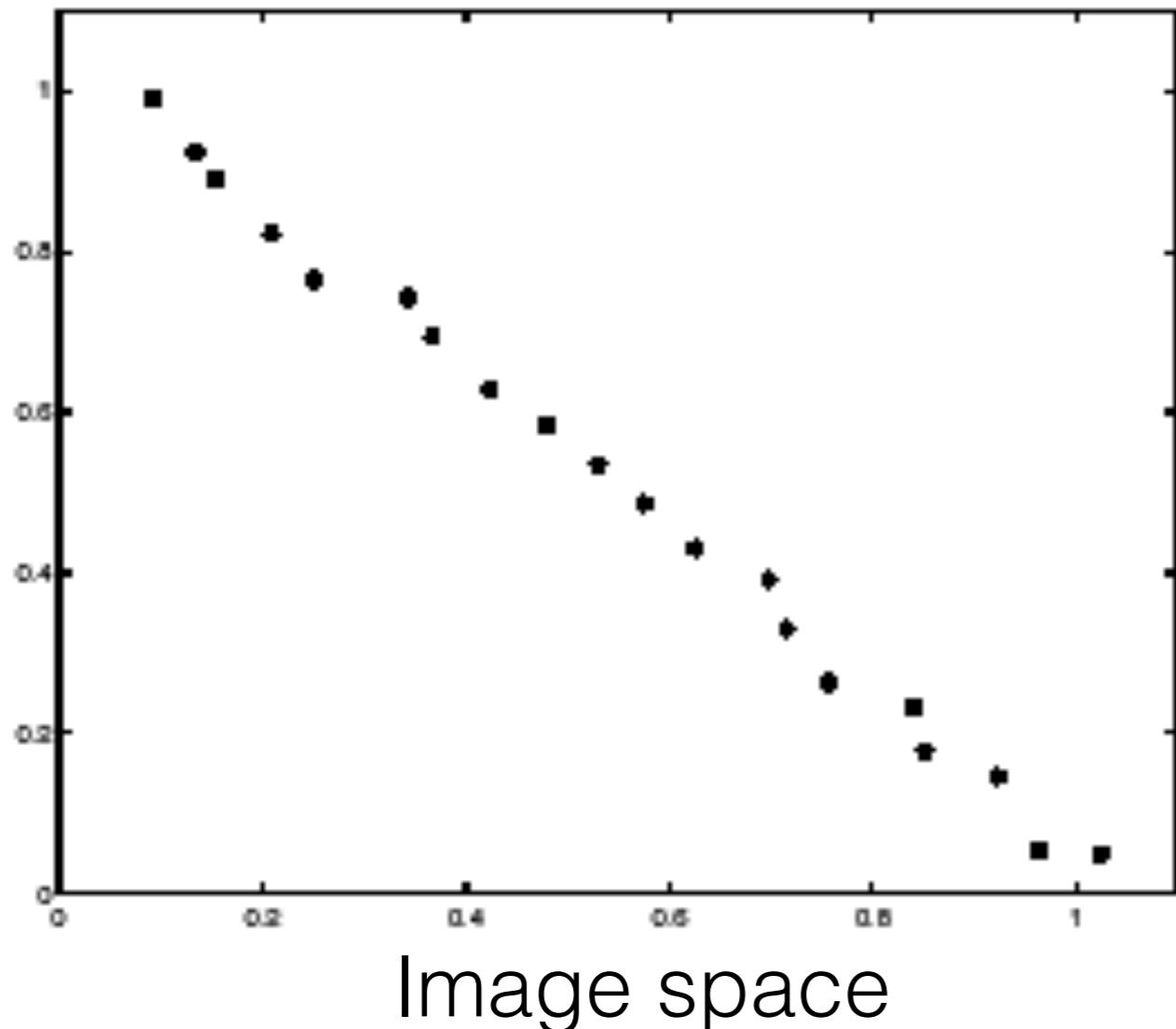
Basic Shapes



More complex image



In practice, measurements are noisy...



Too much noise ...

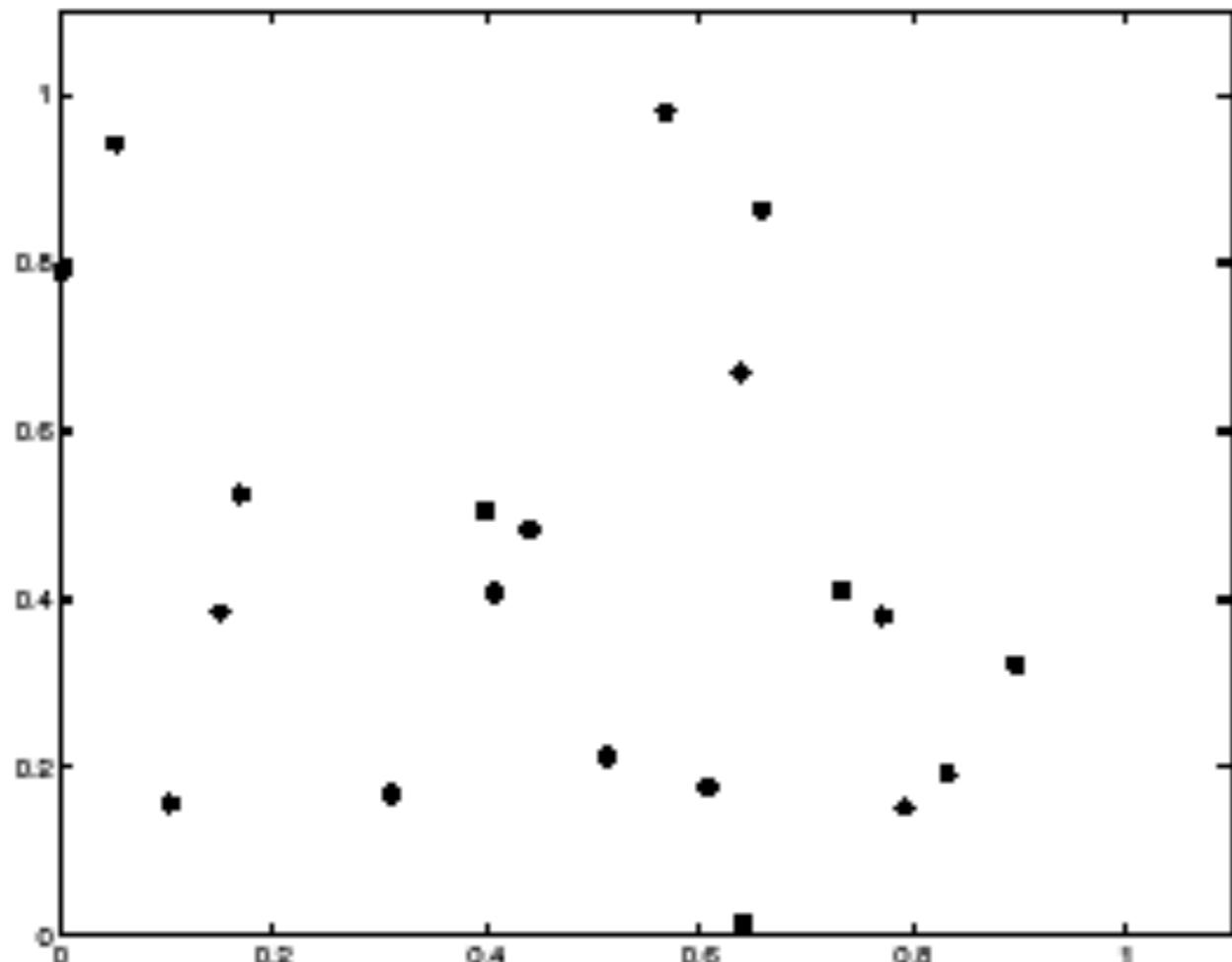
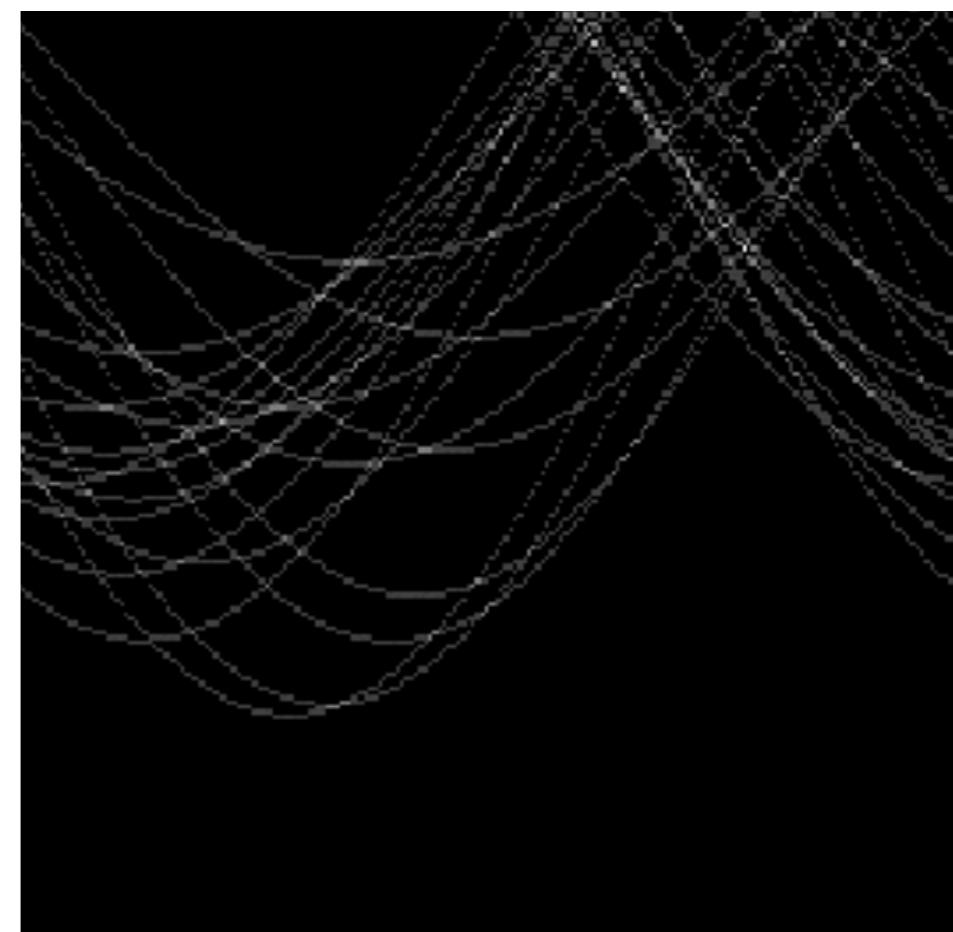


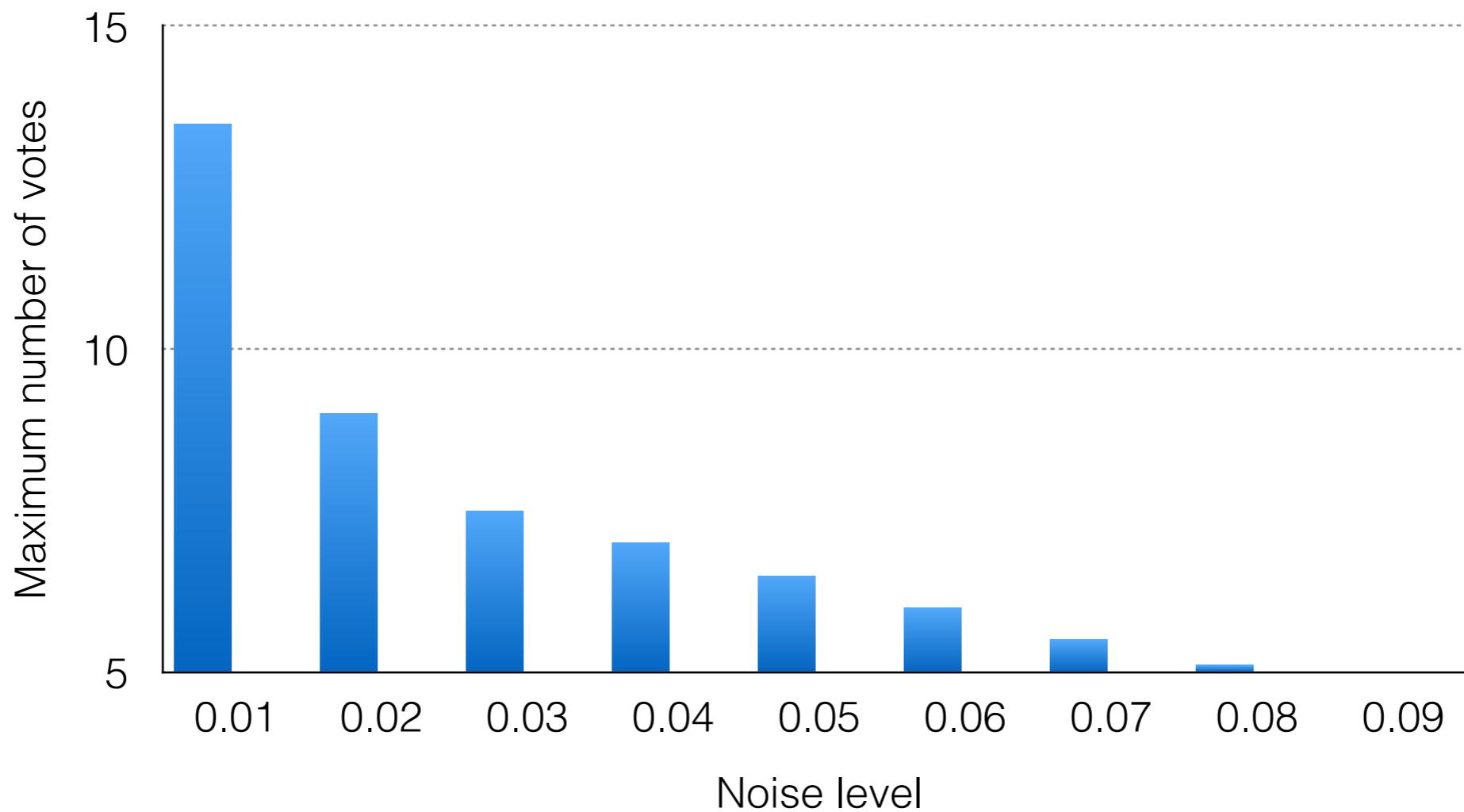
Image space



Votes

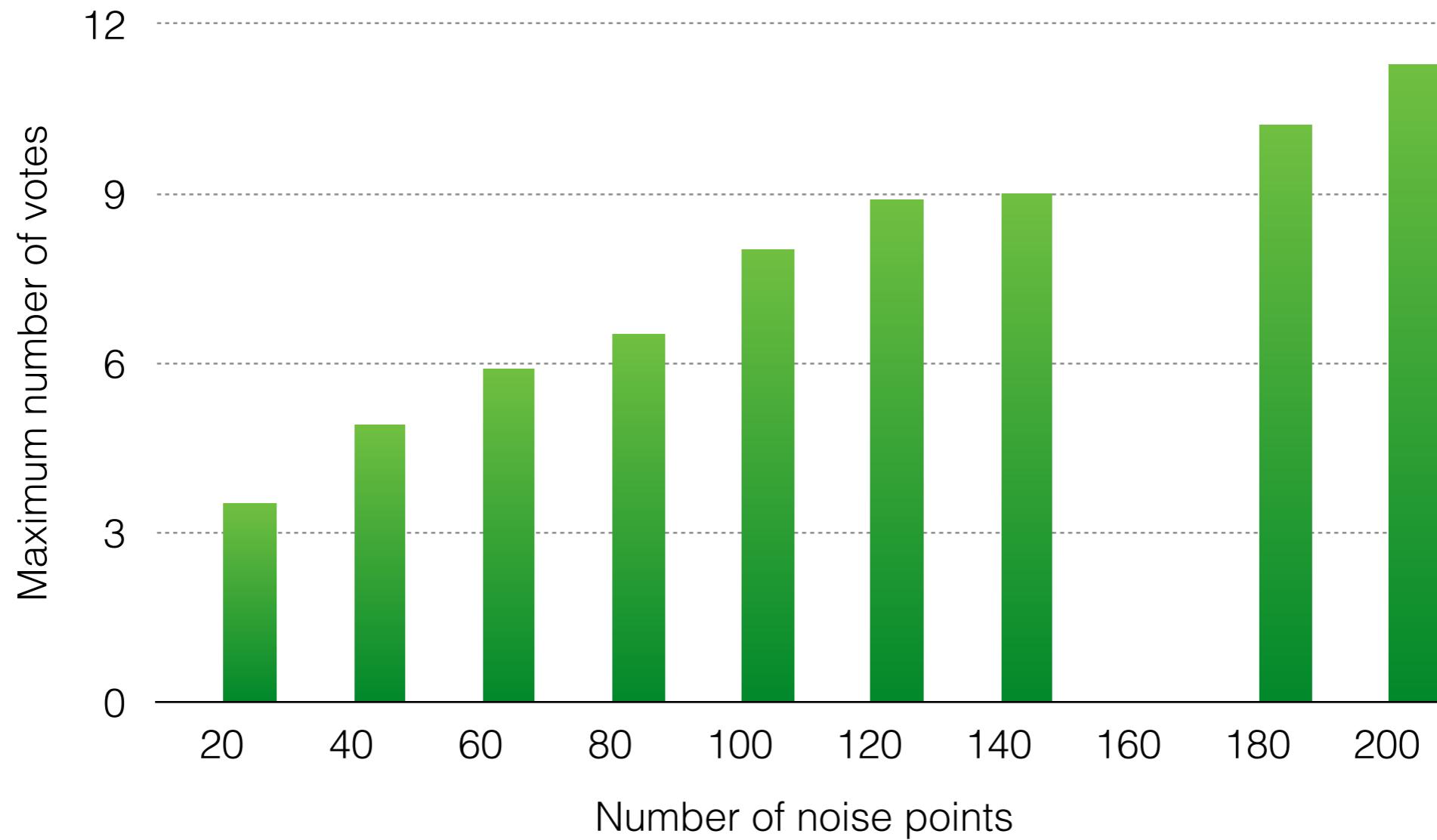
Effects of noise level

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



More noise, less votes (in the right bin)

Effect of noise points

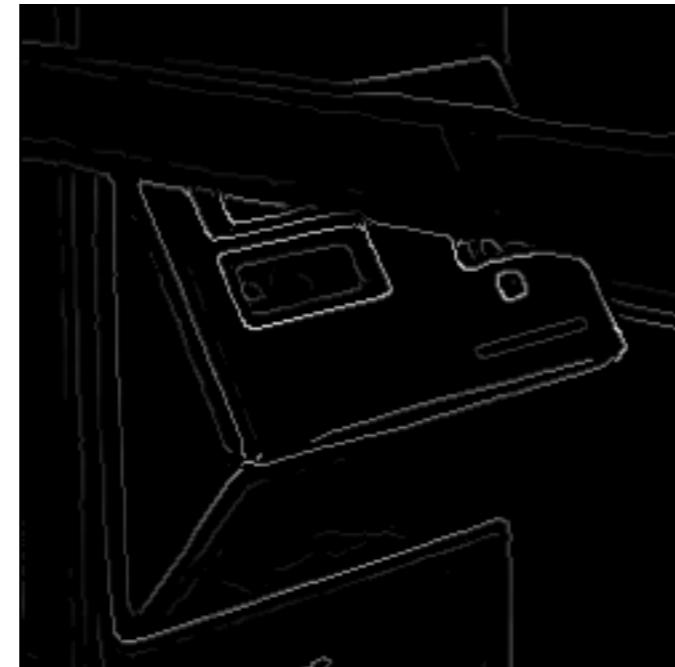


More noise, more votes (in the wrong bin)

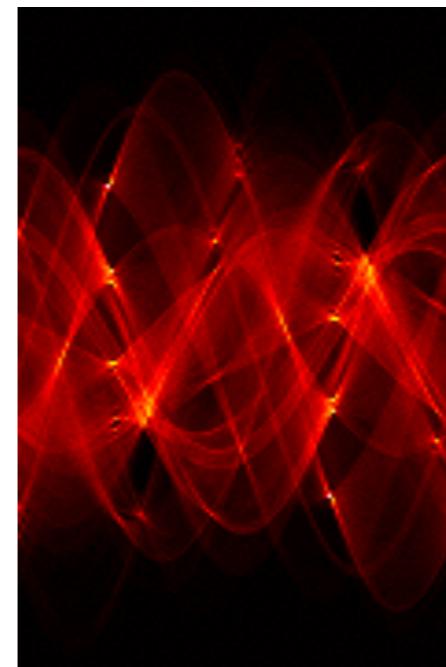
Real-world example



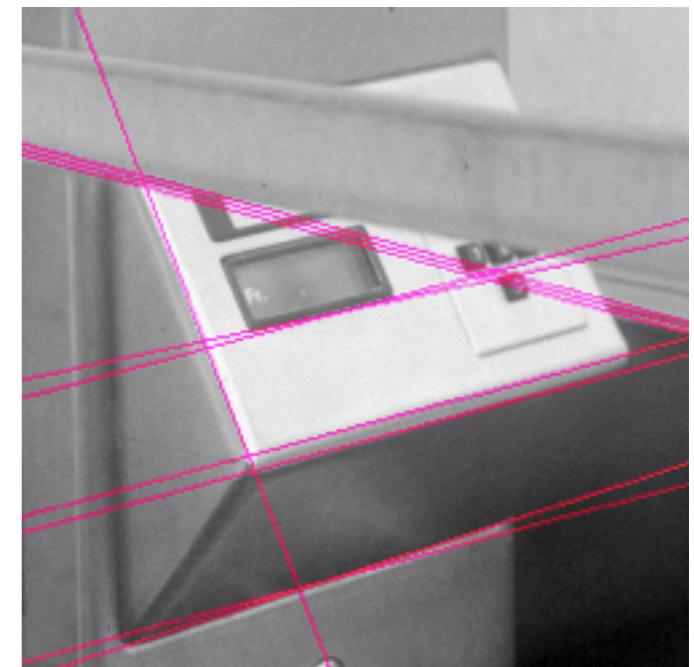
Original



Edges



parameter space



Hough Lines