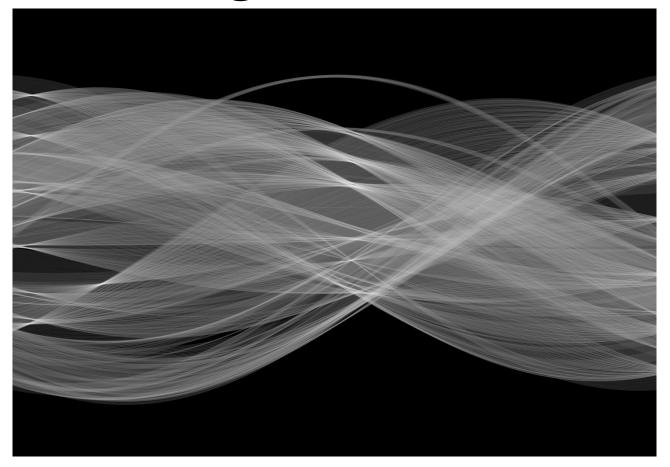
Hough transform

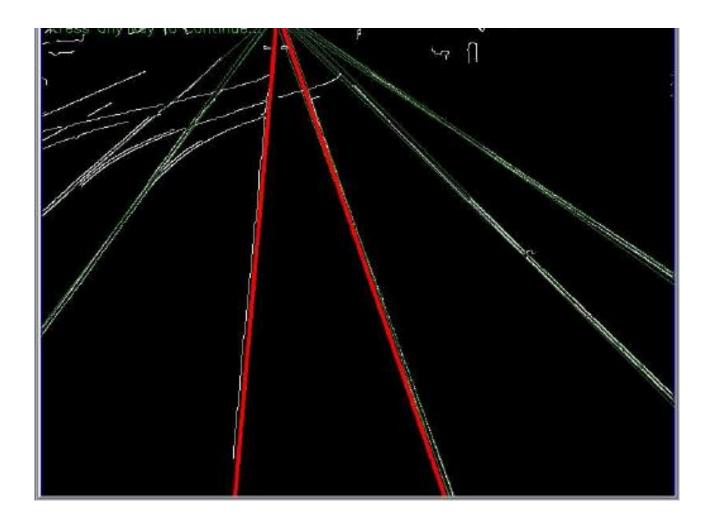


Yoni Chechik

www.AlisMath.com

Problem 3: fitting multiple lines

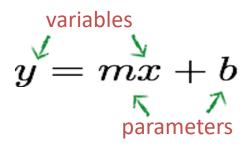
- (This class is a follow-up of the last curve fitting class).
- What if the initial data has multiple lines needed to detect. What can we do?

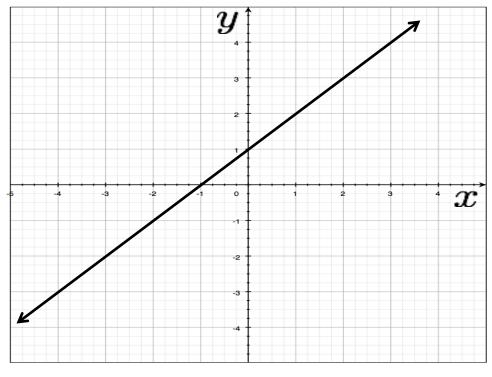


TOC

- Hough transform
 - -(m,b) parameter space
 - $-\left(
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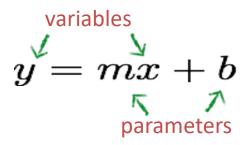
- Line function: y = mx + b
 - Usually we are given (m, b) constants, and the variables are (x, y).
- In a regression problem we are given (x, y), and the unknowns we wish to find are the best fit for (m, b).
 - Let's look at (m, b) as our variables.

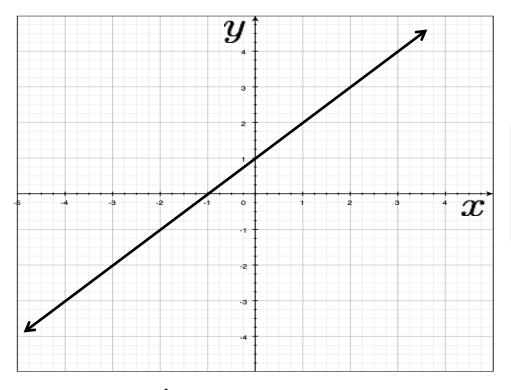




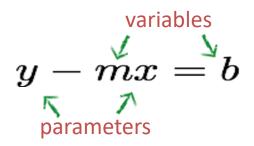
Let's look at (m,b) as our variables

Image space





a line becomes a point



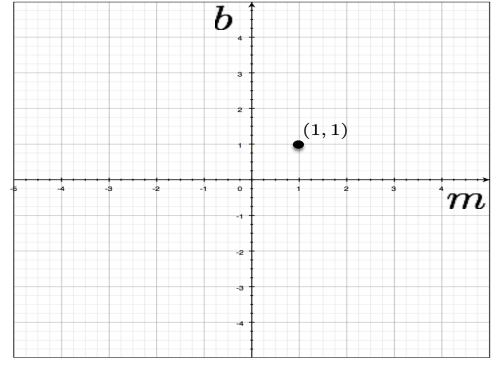
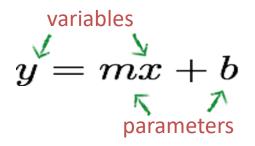


Image space

Parameter space



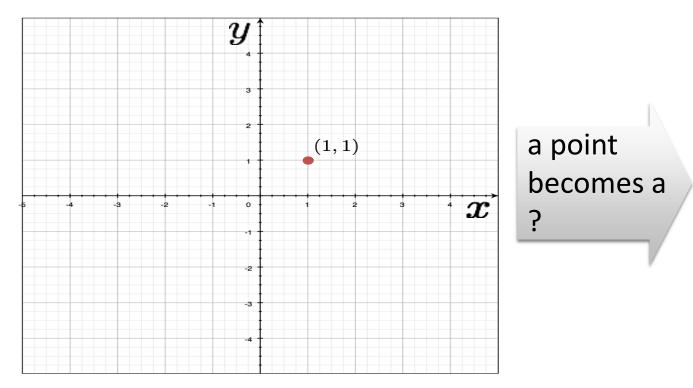
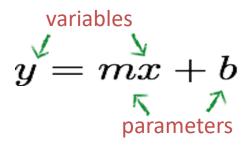
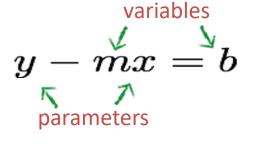
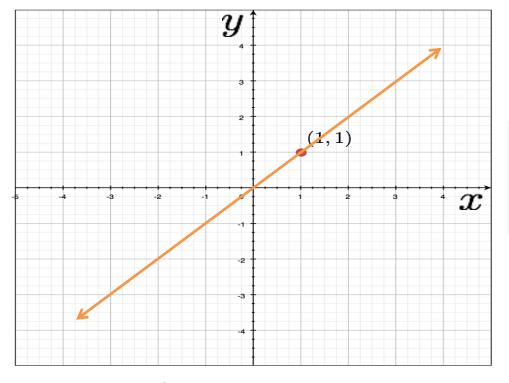


Image space







a point becomes a ?

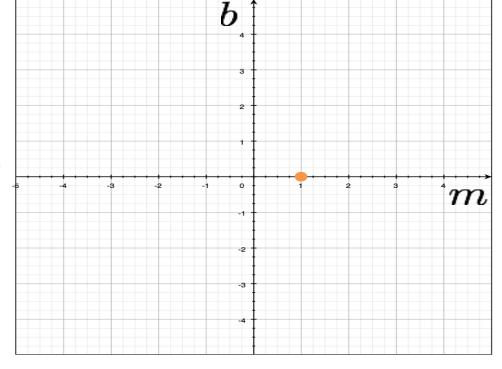
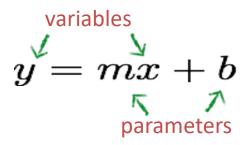
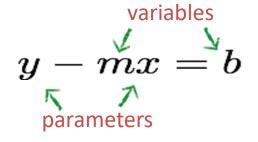
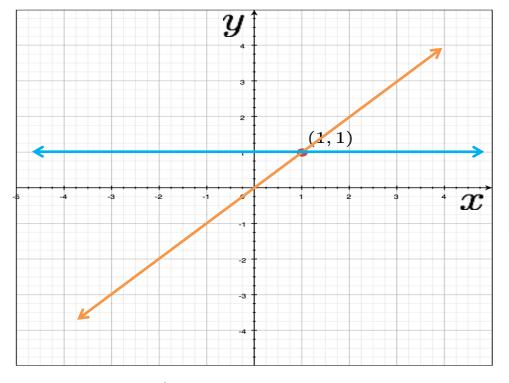


Image space

Parameter space







a point becomes a ?

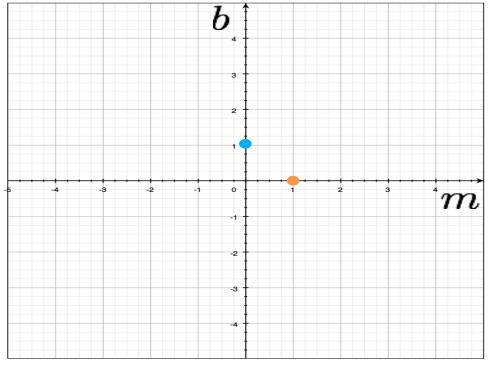
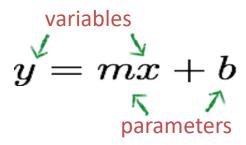
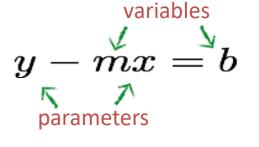
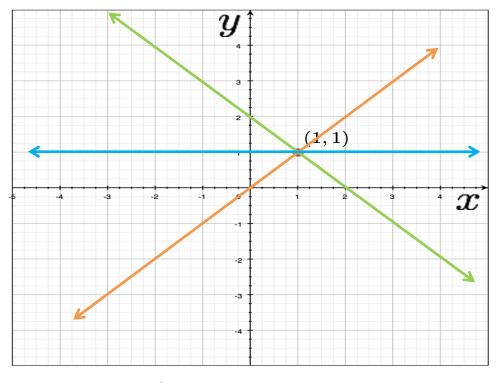


Image space

Parameter space







a point becomes a ?

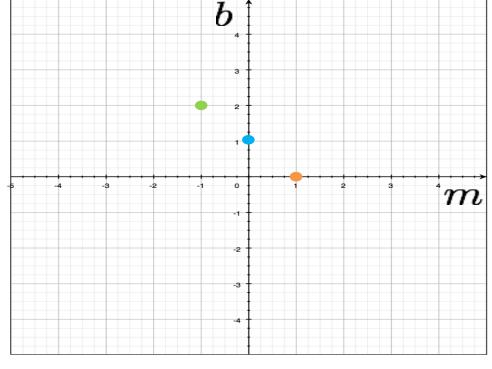
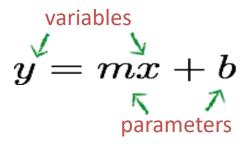
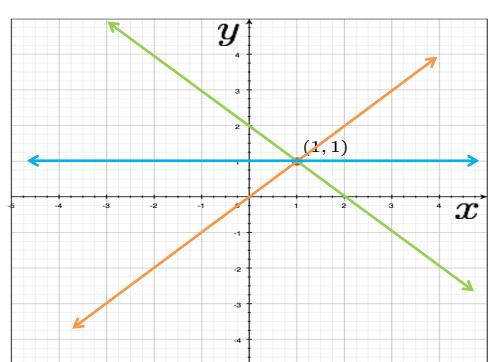


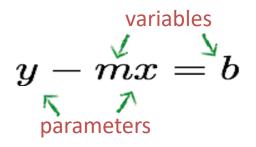
Image space

Parameter space





a point becomes a <u>line</u>



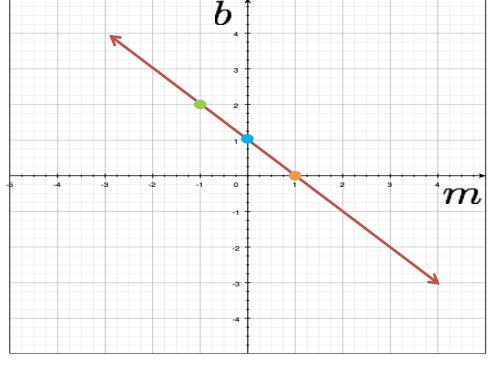
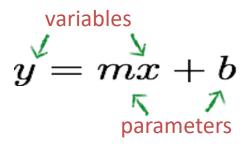
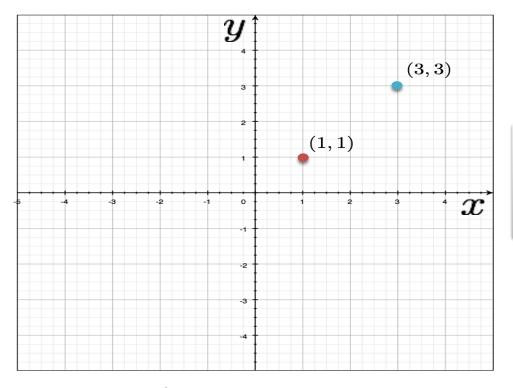


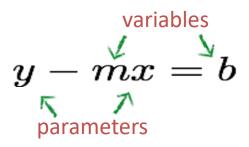
Image space

Parameter space





two points become ?



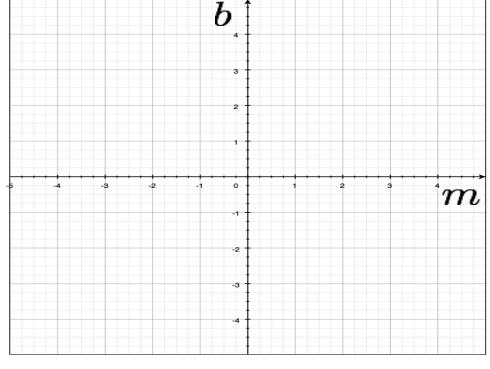
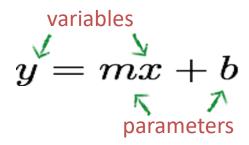
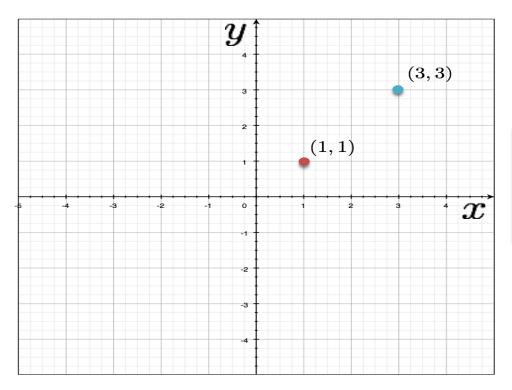


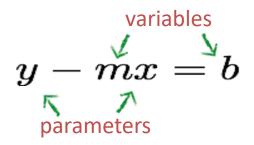
Image space

Parameter space





two points become **Two lines**



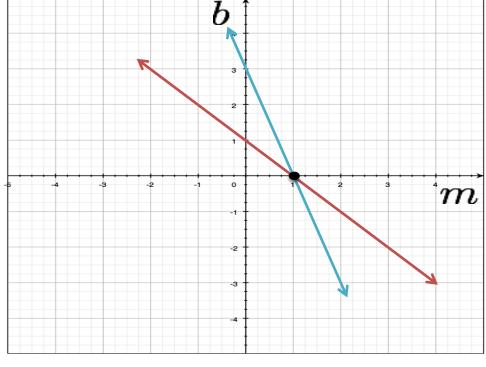
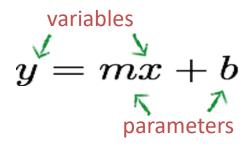


Image space

Parameter space



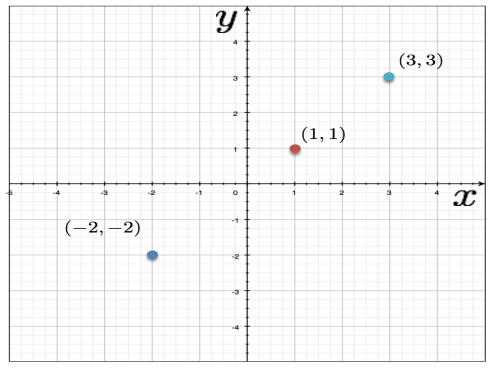
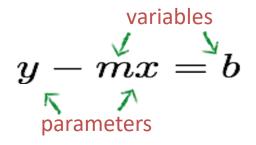
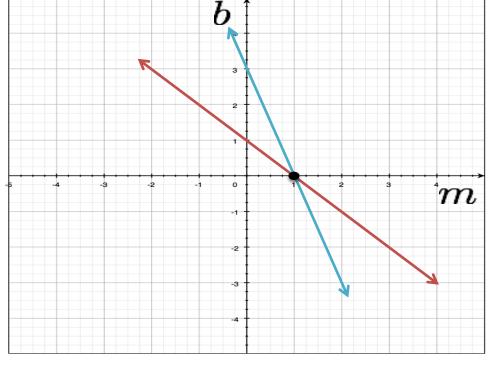


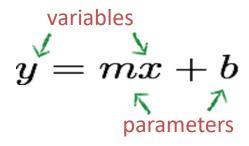
Image space

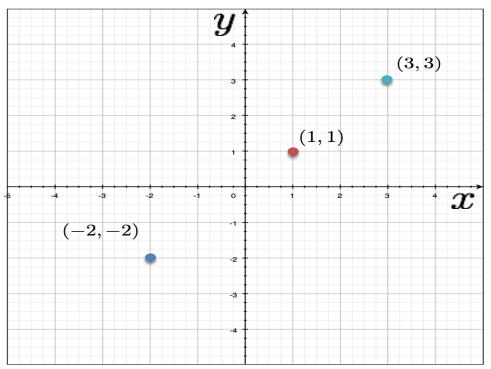
three points become



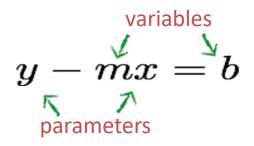


Parameter space





three points become Three lines



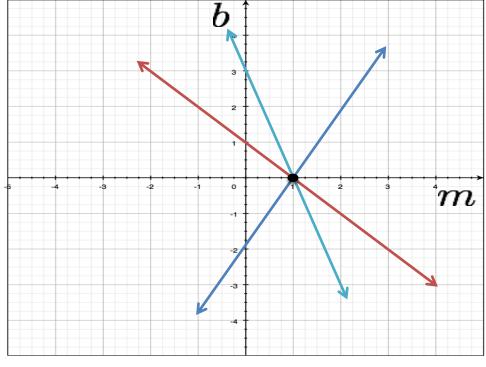
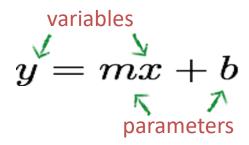
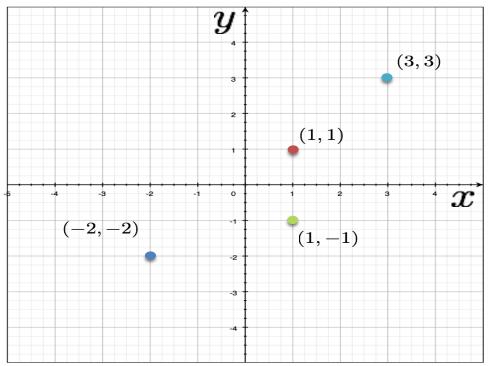


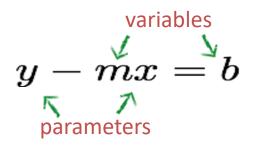
Image space

Parameter space





four points become



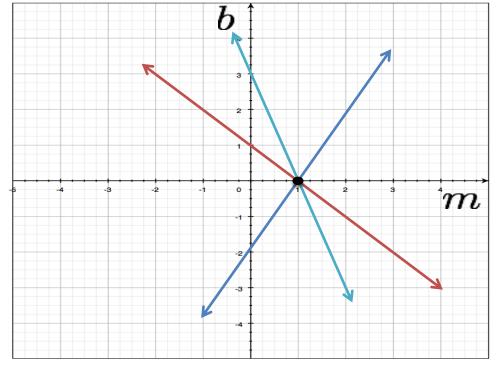
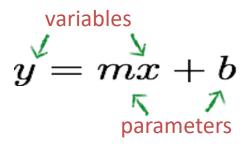
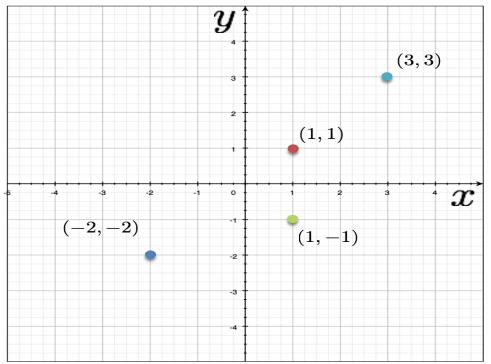


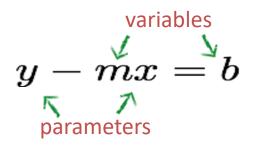
Image space

Parameter space





four points become Four lines



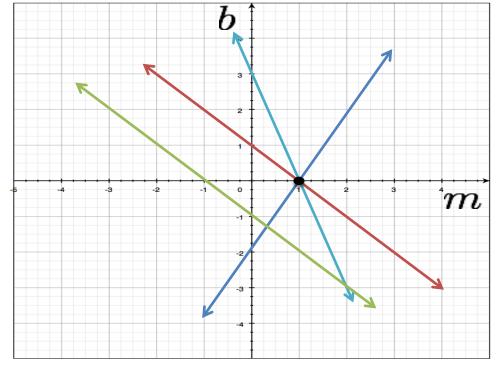


Image space

Parameter space

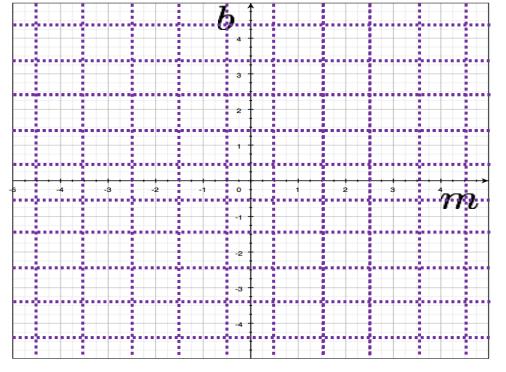
How can we find lines in dataset using the parameter space?

How can we find lines in dataset using the parameter space?

1. Quantize the output parameter space to user defined bins- we will call this table the accumulator table.



Image space



Parameter space

- How can we find lines in dataset using the parameter space?
 - 2. For each point in image space- find corresponding line in parameter space and increment +1 the intersecting bins.

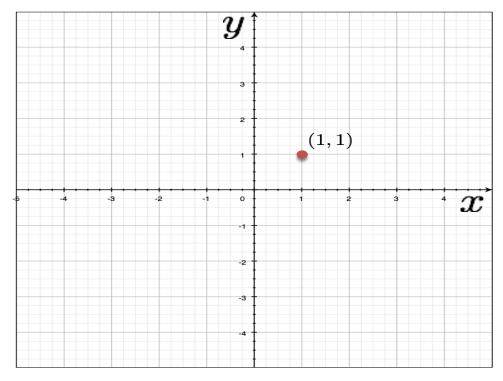
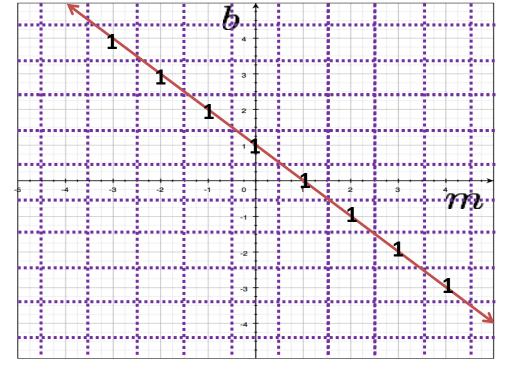


Image space



Parameter space

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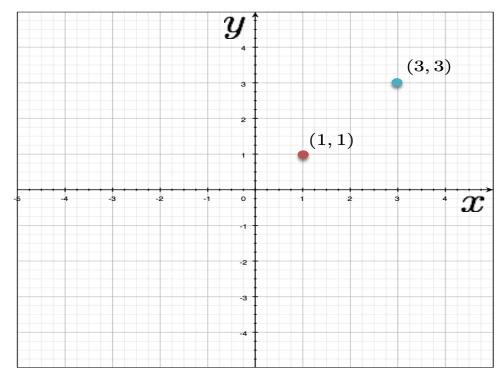
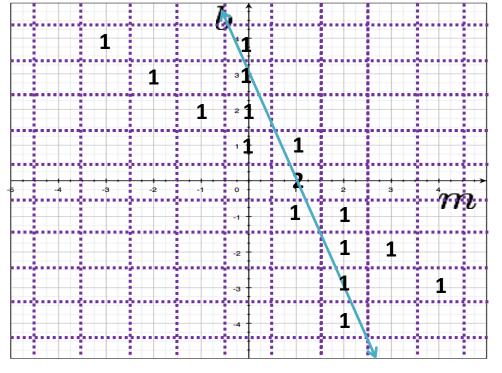


Image space



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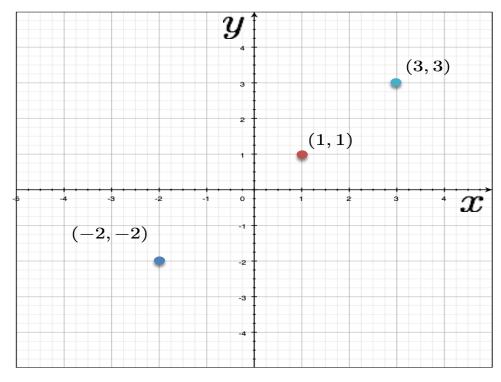
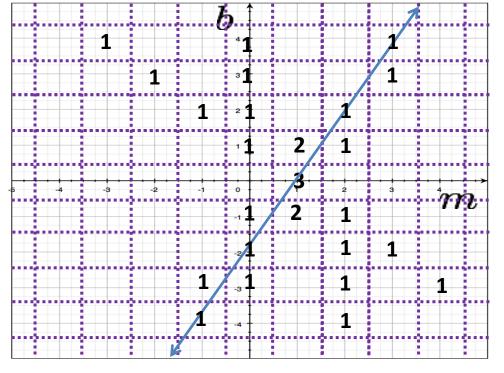


Image space



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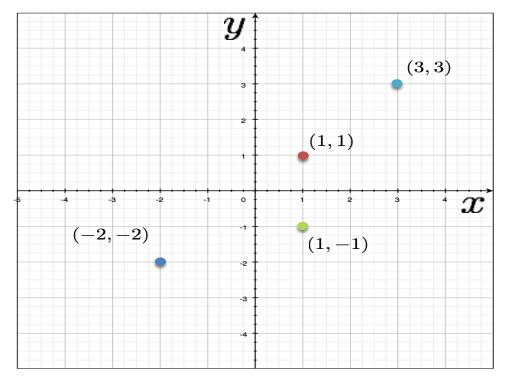
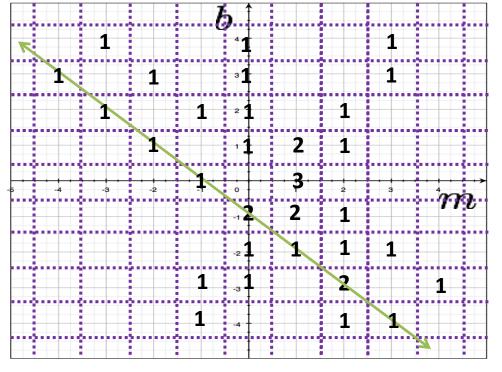


Image space



Parameter space

How can we find lines in dataset using the parameter space?

3. Threshold the accumulator table result by some TH and get the corresponding line parameters.

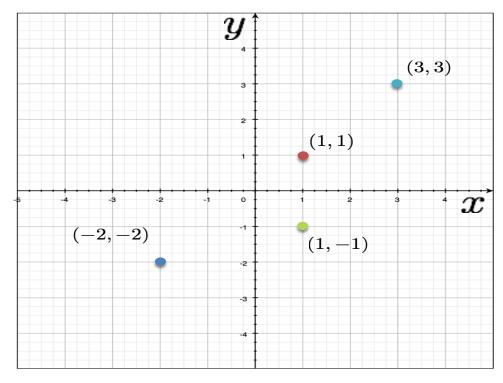
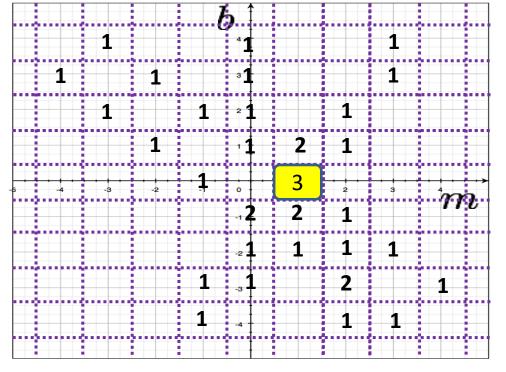


Image space



Parameter space

Hough transform

```
Build accumulator table.
For each point in image space:
    find corresponding line in parameter space and
    increment +1 the intersecting bins.
Threshold the accumulator table result by some TH and get
the corresponding line parameters.
```

Hough transform- pros & cons

Pros:

- Can detect multiple lines in image space.
- can be extended to detect different parameterized curves (e.g.: circles, ellipsoids), and even un-parameterized curves (generalized hough transform [out of scope]- similar to template matching).

• Cons:

- For the shown (m, b) parameter space, can't detect vertical lines. Why?
- Susceptive to noise. Why?
- Computationally costly.

- Vertical (or near vertical) lines have a big slope: $m \to \infty$. This causes the accumulator table to be very big in m direction.
- A solution is to give a **different parameterization to lines**:

$$y = mx + b$$

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$$\xrightarrow{x=0} y = b$$

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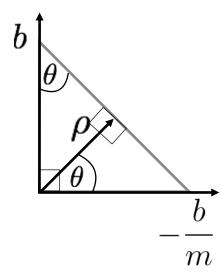
$$\xrightarrow{y=0} x = -\frac{b}{m}$$

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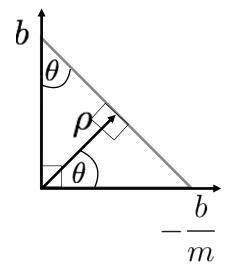


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$$\frac{y = mx + b}{\xrightarrow{x=0}} y = b$$

$$\xrightarrow{y=0} x = -\frac{b}{m}$$

$$\frac{-b}{b} = -\frac{1}{m} = tg\theta$$



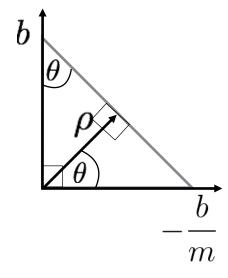
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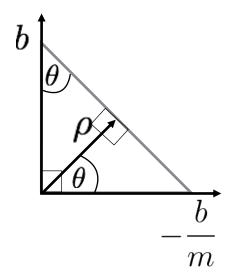
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$$\frac{\rho}{b} = \sin\theta \to b = \frac{\rho}{\sin\theta}$$



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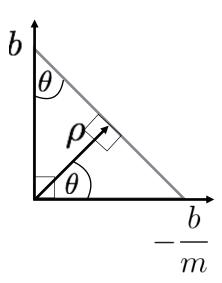
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$$y = mx + b \to y = -\frac{cos\theta}{sin\theta}x + \frac{\rho}{sin\theta}$$



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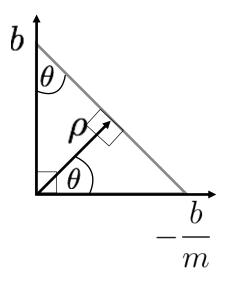
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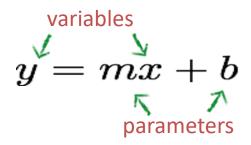
$$y = mx + b \to y = -\frac{\cos\theta}{\sin\theta}x + \frac{\rho}{\sin\theta} \to x\cos\theta + y\sin\theta = \rho$$

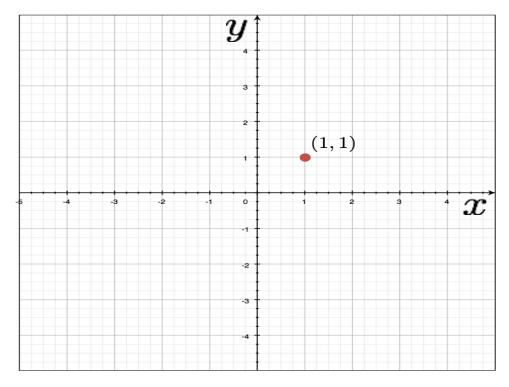


$$y = mx + b \rightarrow y = -\frac{\cos\theta}{\sin\theta}x + \frac{\rho}{\sin\theta} \rightarrow x\cos\theta + y\sin\theta = \rho$$

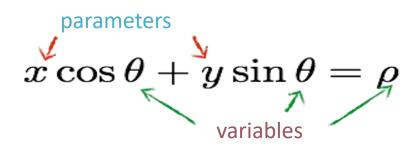
TOC

- Hough transform
 - -(m,b) parameter space
 - $-\left(oldsymbol{
 ho},oldsymbol{ heta}
 ight)$ parameter space





a point becomes a wave



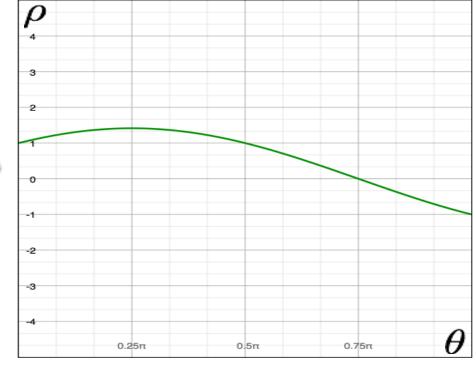


Image space

Parameter space

variables
$$y = mx + b$$
 parameters

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

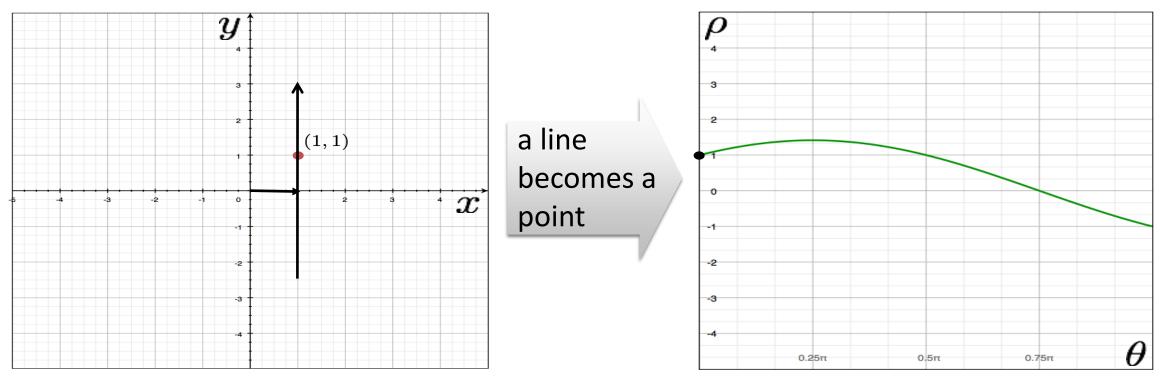


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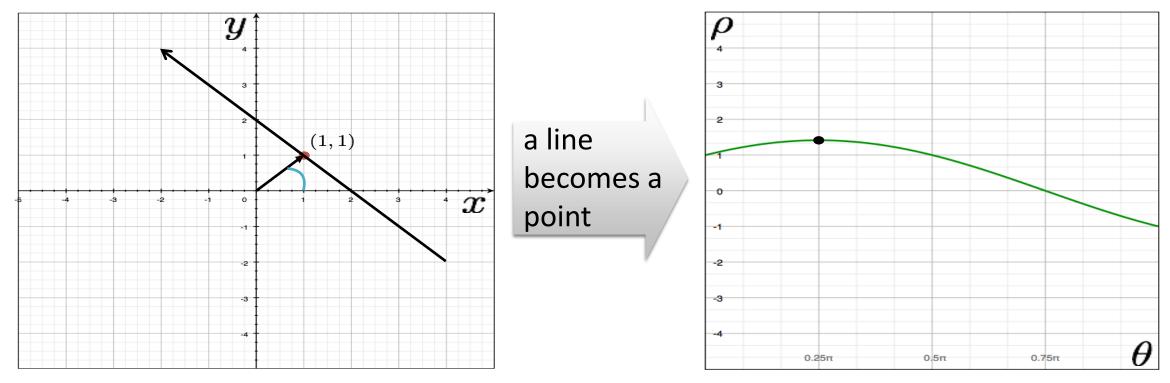
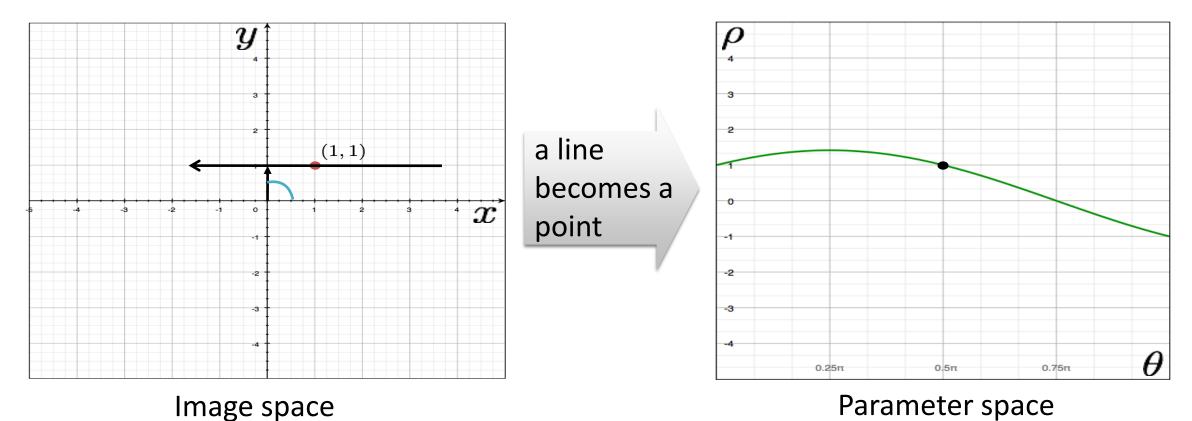


Image space

Parameter space

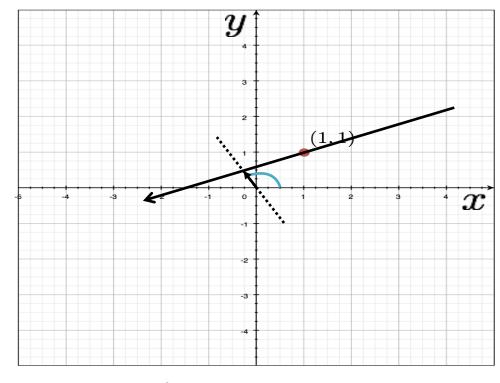
variables
$$y=mx+b$$
 parameters

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



variables
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 parameters

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



a line becomes a point

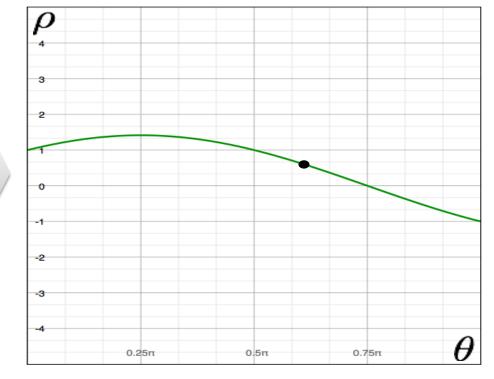
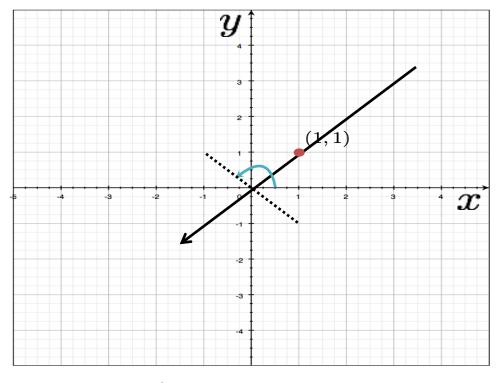


Image space

Parameter space

variables
$$y = mx + b$$
 $\sqrt{}$ parameters

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



a line becomes a point

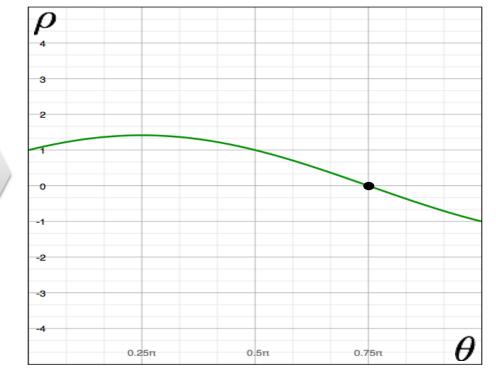
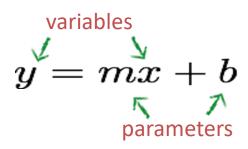
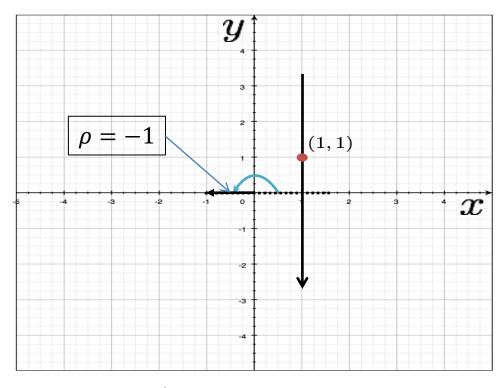


Image space

Parameter space



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



a line becomes a point

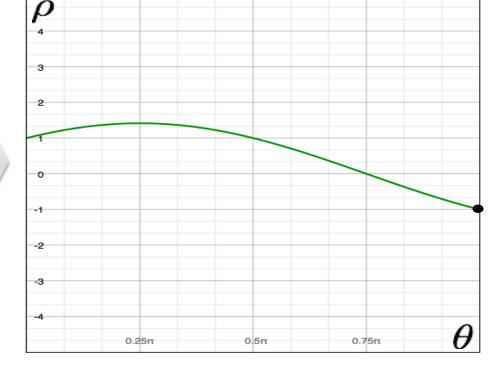
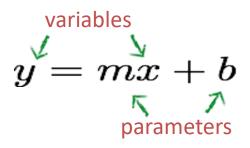


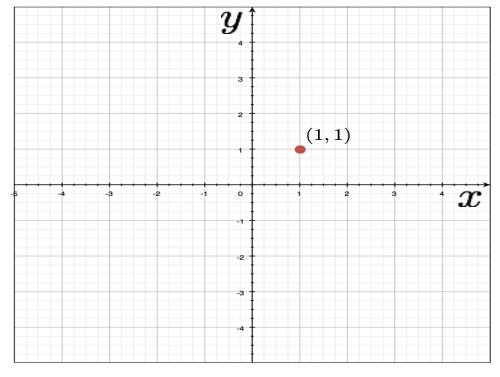
Image space

Parameter space

Notes on (ρ, θ) parameter space

- We only care about $0 \le \theta < \pi$, otherwise it's symmetric.
- As we so earlier, for some $\theta \to sign(\rho) = -1$. this is acceptable since the derivation earlier was right only for the first quadrant.





a point becomes a wave

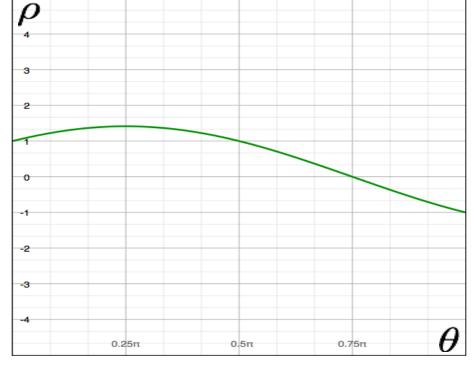
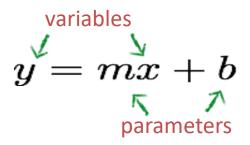


Image space

Parameter space



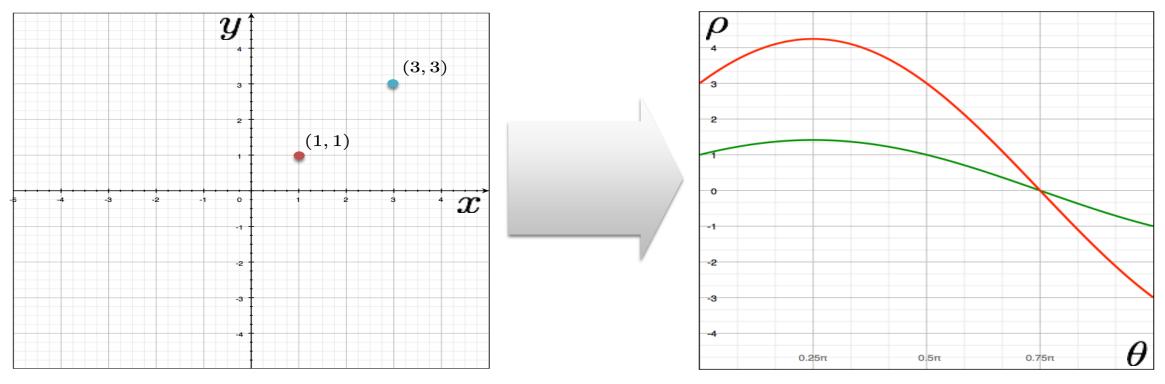
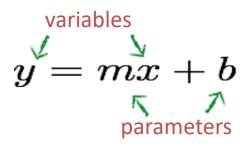


Image space

Parameter space



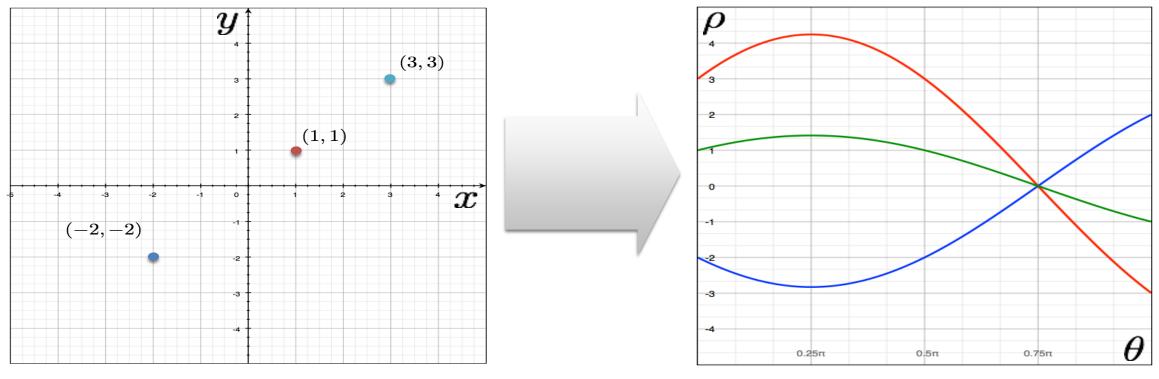
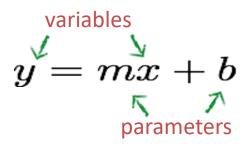


Image space

Parameter space



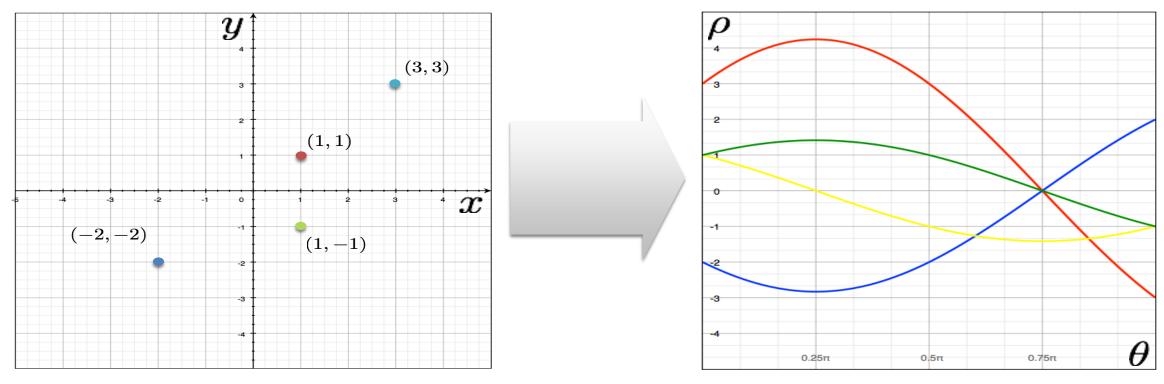


Image space

Parameter space

Hough transform algorithm stays the same!

```
Build accumulator table.
For each point in image space:
   find corresponding line in parameter space and
   increment +1 the intersecting bins.
Threshold the accumulator table result by some TH and get
the corresponding line parameters.
```

Hough transform- pros & cons

Pros:

- Can detect multiple lines in image space.
- can be extended to detect different parameterized curves (e.g.: circles, ellipsoids), and even un-parameterized curves (generalized hough transform [out of scope]- similar to template matching that will be covered later in course).

• Cons:

- For the shown (m, b) parameter space, can't detect vertical lines. Why?
- Susceptive to noise. Why?
- Computationally costly.

Hough transform noise

- In case that the discovered edge is noisy, the binning process in the accumulation matrix can be problematic.
- This is a known problem of Hough transform... some ways to get better results is to try:
 - Different bin size (different step size for $R \& \theta$).
 - Smooth the accumulation matrix before thresholding.