



# Image Analysis

Tim B. Dyrby

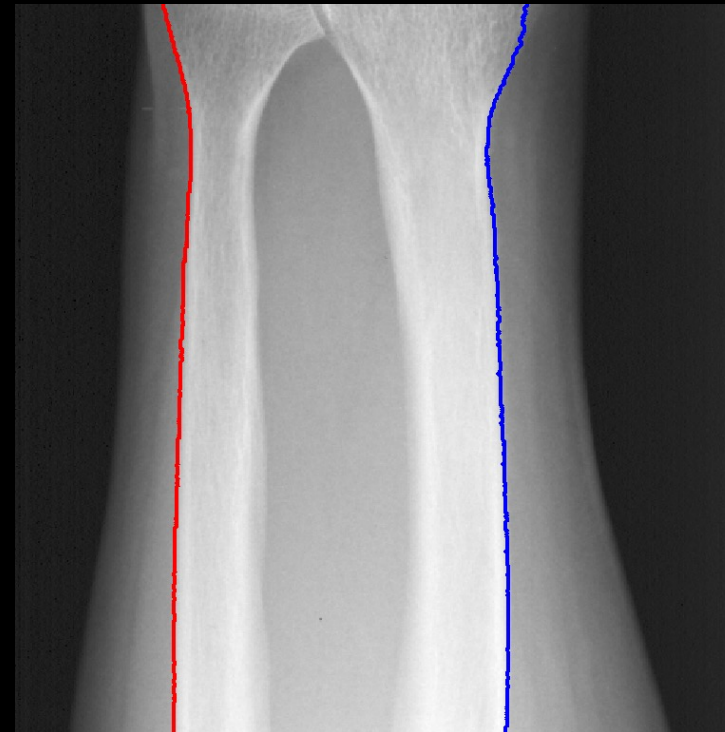
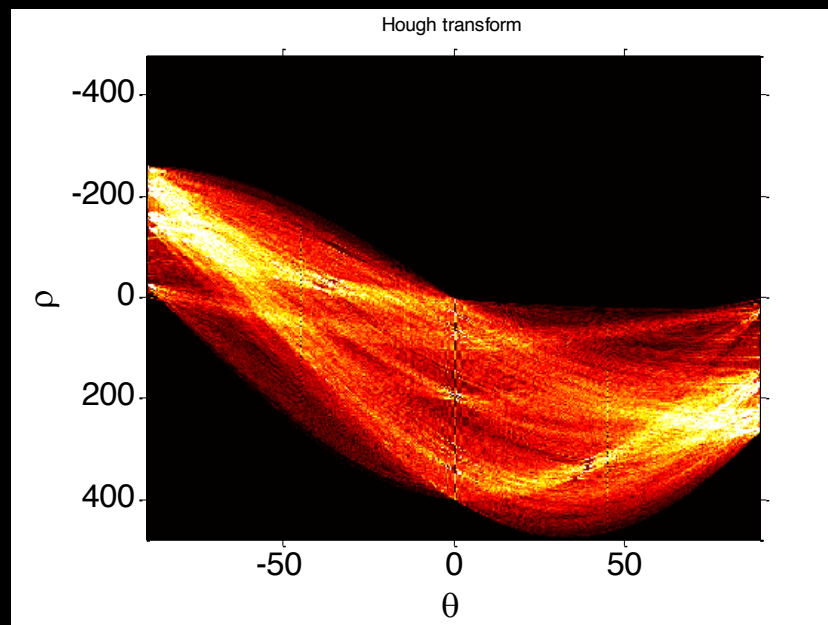
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# Lecture 8 – Hough Transformation and Path Tracing





Go to [www.menti.com](https://www.menti.com) and use the code 4006 2653

# Quiz testing How long time did it take to develop the Dijkstra

## Breaking NEWS!

One morning I was shopping in Amsterdam with my young fiancée, and tired, we sat down on the café terrace to drink a cup of coffee and I was just thinking about whether I could do this, and I then designed the algorithm for the shortest path. As I said, it was a twenty-minute invention.

— Edsger Dijkstra, in an interview with Philip L. Frana, Communications of the ACM, 2001<sup>[3]</sup>

0	0	0	0	0	0
3 years	1 year	2 weeks	1 hour	20 min	37 sec

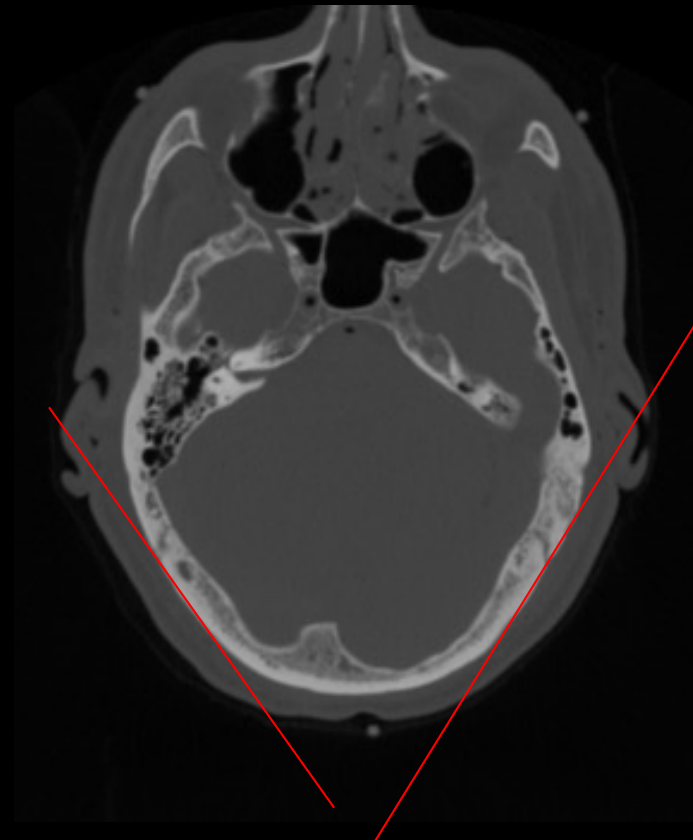
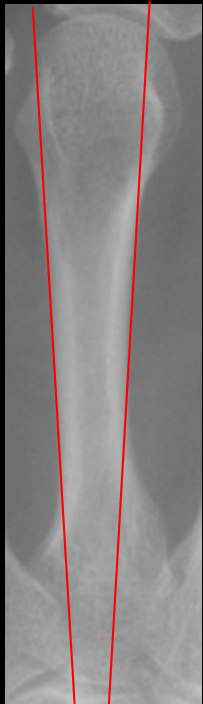


# What can you do after today?

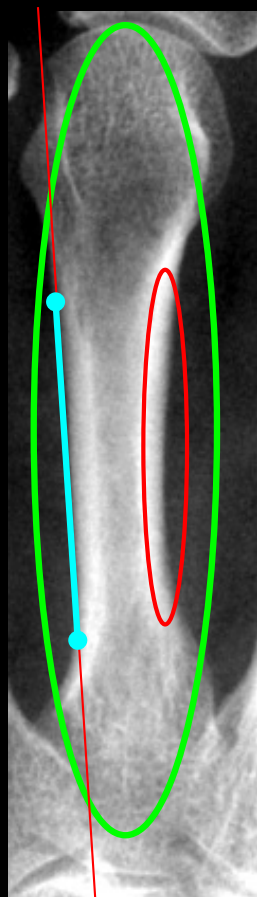
- Use the Hough transform for line detection
- Describe the slope-intercept, the general form and the normalised form of lines
- Describe the connection between lines and the Hough space
- Use edge detection to enhance images for use with the Hough transform
- Use dynamic programming to trace paths in images
- Describe how an image can be used as a graph
- Describe the fundamental properties of a cost image
- Compute the cost of path
- Compute an accumulator image for path tracing
- Compute a back tracing image for path tracing
- Choose appropriate pre-processing steps for path tracing
- Describe how circular structures can be located using path tracing

# Line Detection

- Find the lines in an image

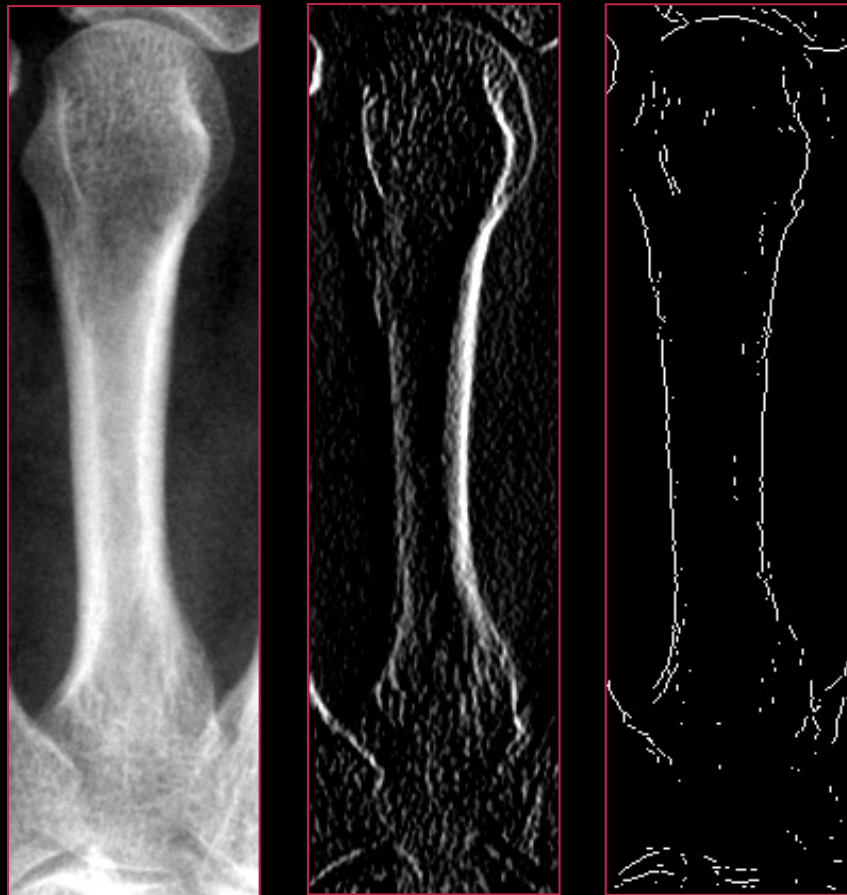


# What is a line?



- It can be the entire object
  - Large scale
- Can also be the border between an object and the background
  - Small scale
- Normally only locally defined

# Enhancing the lines



Original

Prewitt

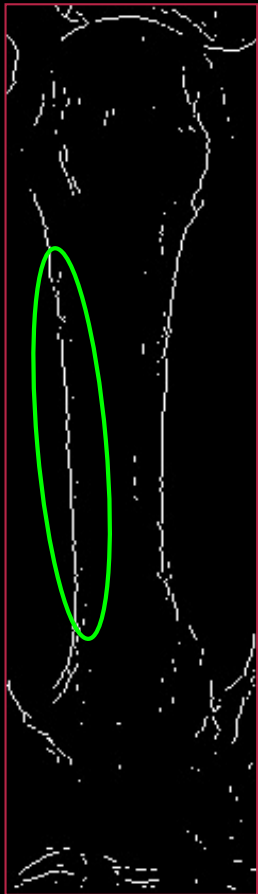
Edge

- We want to locate the borders
  - Enhance them
- Filtering (Prewitt)
- Edge detection

Prewitt:

Vertical			Horizontal		
-1	0	1	-1	-1	-1
-1	0	1	0	0	0
-1	0	1	1	1	1

## What is a line II?



- Result of the edge filter is a selection of white pixels
- Some of them define a line
  - Not a perfect straight line
  - “Linelike”
- How do we find the collection of points that define a line?



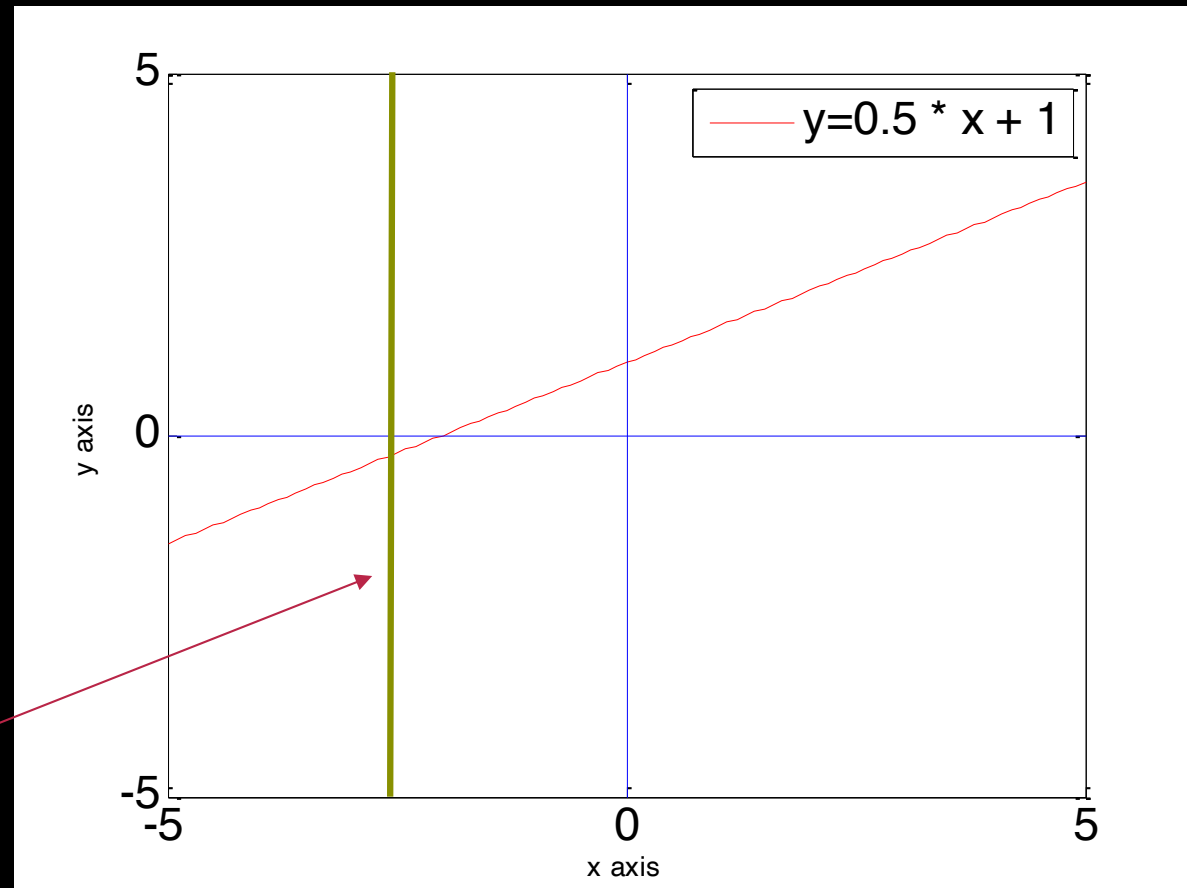
# Mathematical line definition

- The classical definition (slope-intercept form)

$$y = ax + b$$

Slope Intercept

Can not represent lines that are vertical



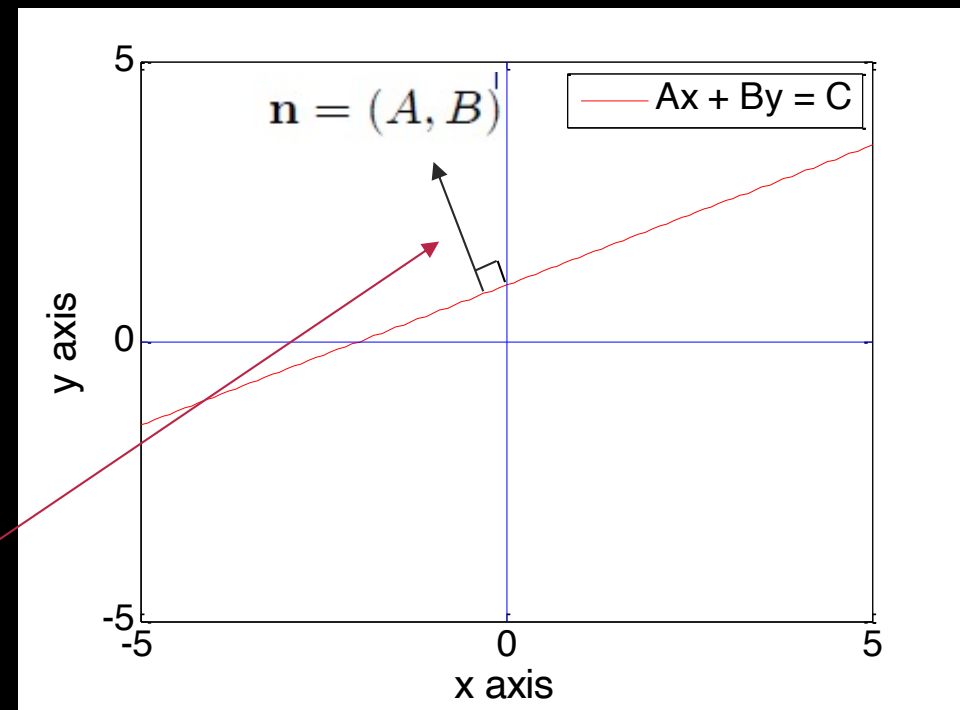
# Mathematical line definition

- General definition (the normal form)

$$Ax + By = C$$

- With

$$A^2 + B^2 = 1$$



Line normal

# Mathematical line definition

## ■ Normal form parameterisation

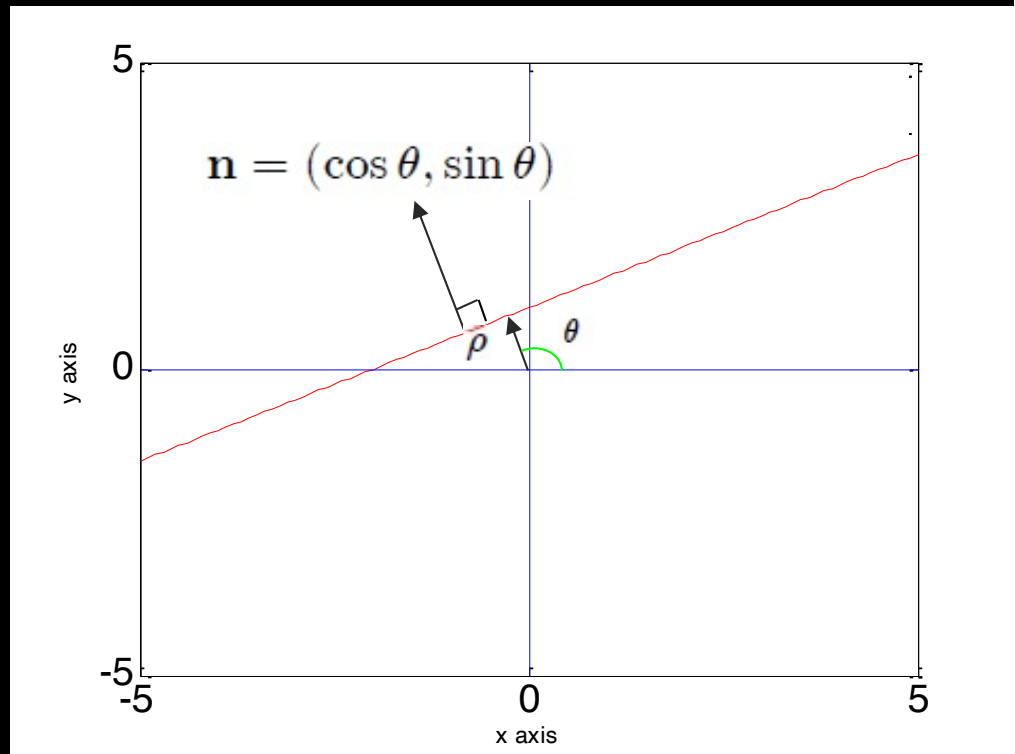
$$x \cos \theta + y \sin \theta = \rho$$

## ■ where

- $\rho$  is the distance from the origin
- $\theta$  is the angle

$$(\cos \theta)^2 + (\sin \theta)^2 = 1$$

$$A^2 + B^2 = 1$$



# Mathematical line definition

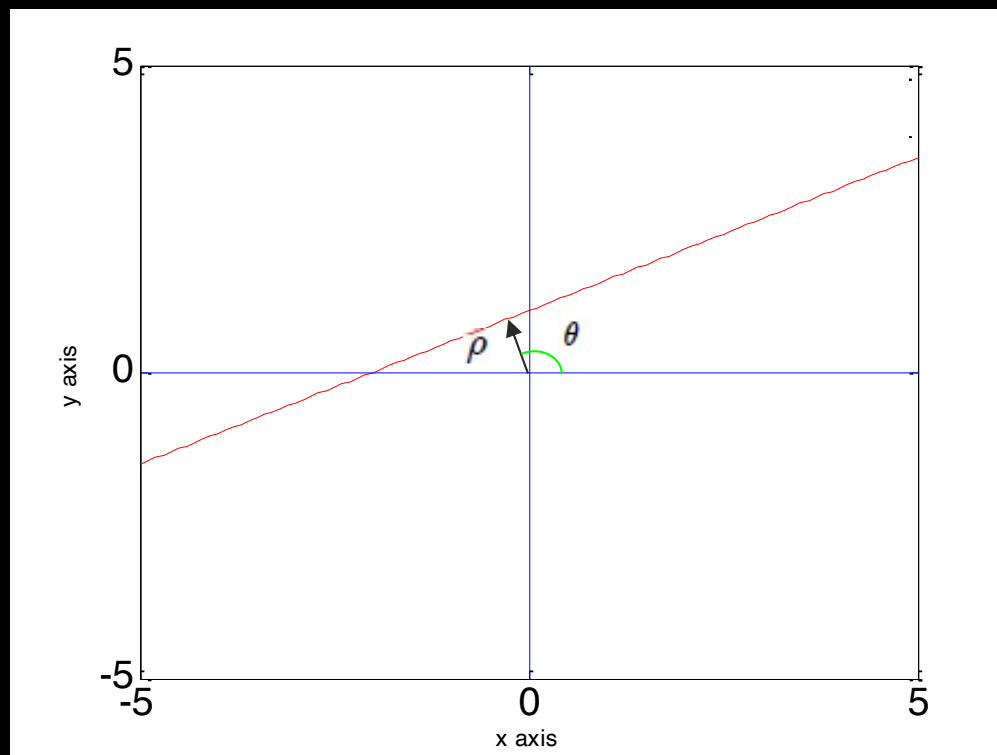
## ■ Normal form parameterisation

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

## ■ Therefore a line can be defined by two values

- $\rho$
- $\theta$

## ■ A line can therefore also be seen as a *point* in a $(\theta, \rho)$ -space



# Converting lines between definitions

## ■ From normal form to the slope-intercept form

The normal form:  $p = x \cos \theta + y \sin \theta$

The slope-intercept form:  $y = ax + b$

Start:  $p = x \cos \theta + y \sin \theta$

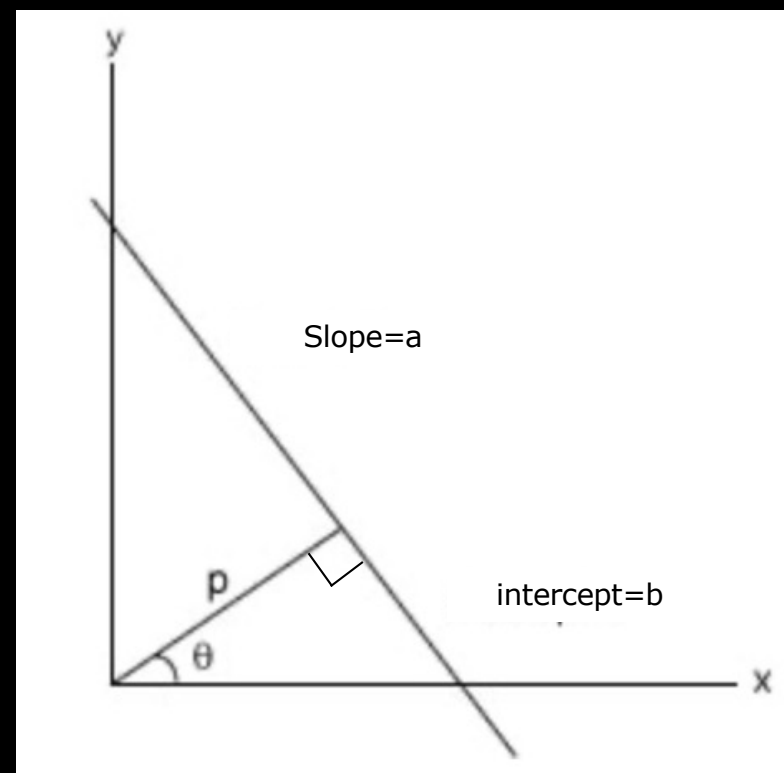
$$-x \cos \theta + p = y \sin \theta$$

$$-x \cot \theta + p \operatorname{cosec} \theta = y$$

$$y = x * (-\cot \theta) + p(\operatorname{cosec} \theta)$$

Slope=a

Intercept=b





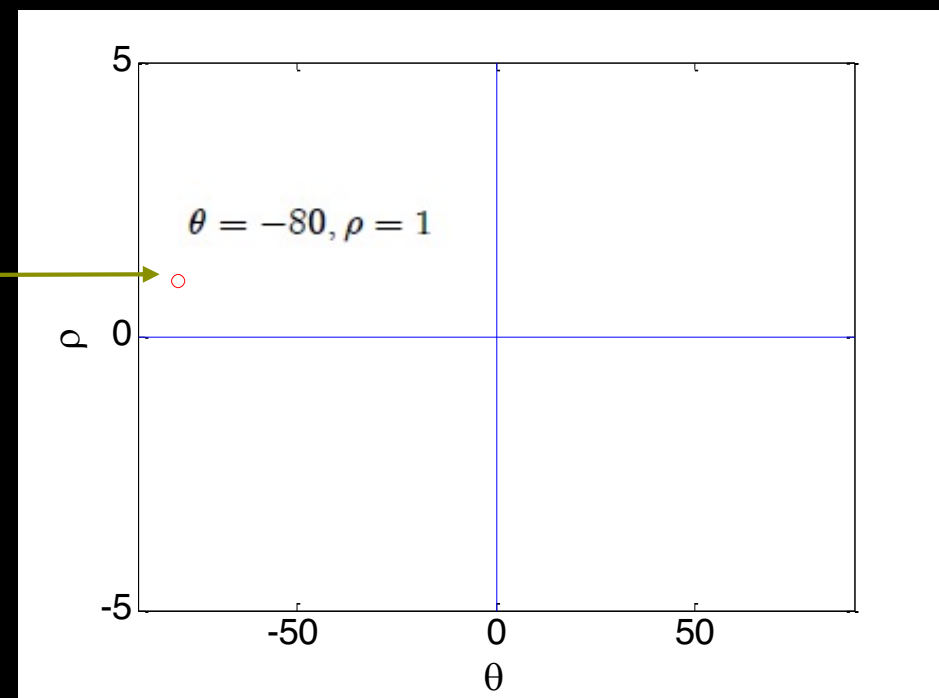
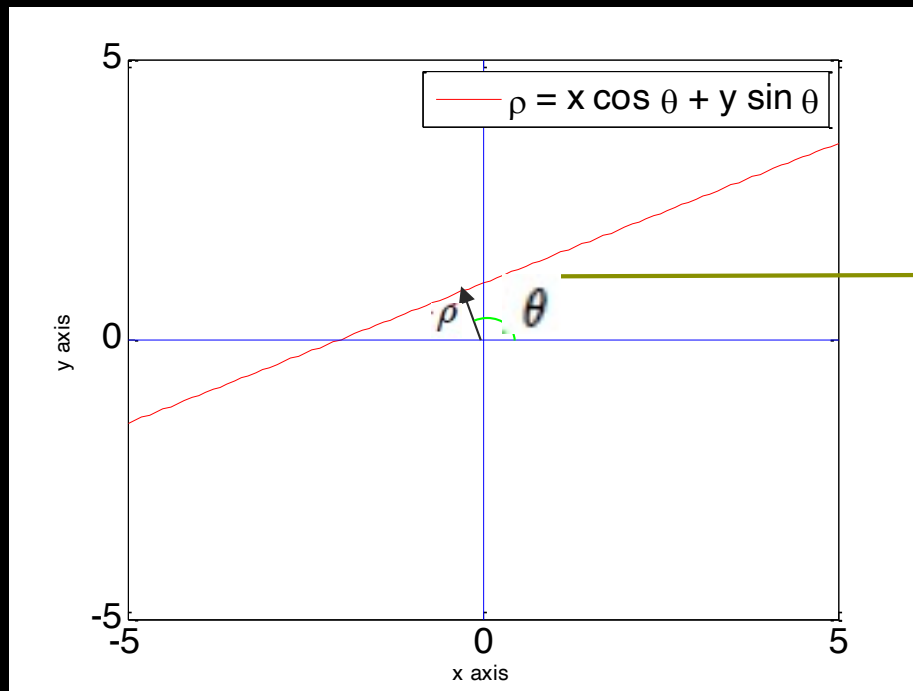
# Something about angles

$\theta \in [0^\circ, 180^\circ]$  In the course notes

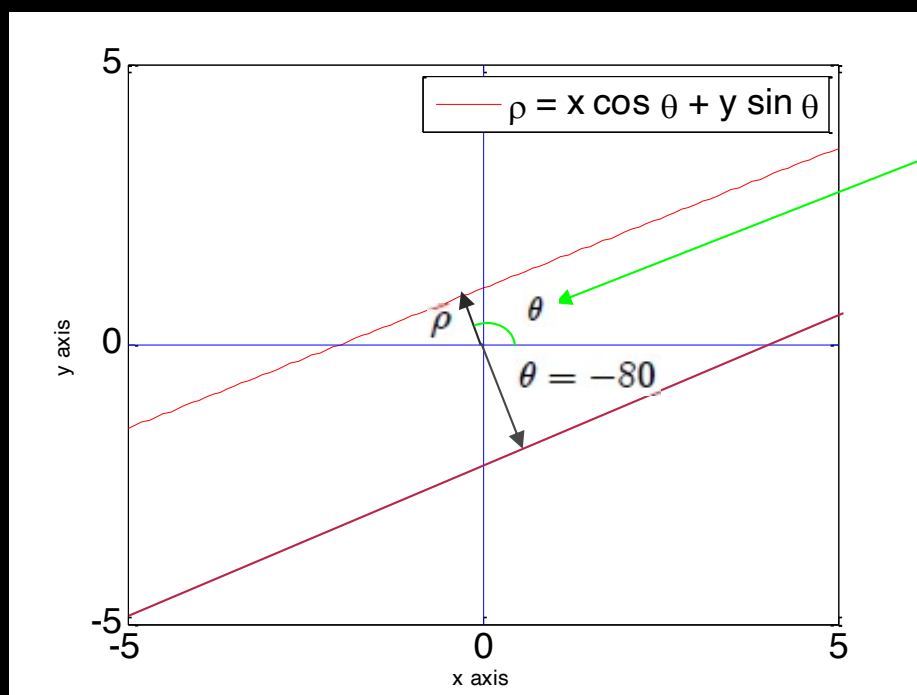
$\theta \in [-90^\circ, 90^\circ]$  In Matlab and in this presentation

# Hough Space

$$-90^\circ < \theta < 90^\circ$$



# More about angles



$$\theta = -80^\circ$$

Why?

$$\theta = 100^\circ$$

but Matlab only allows

$$-90^\circ < \theta < 90^\circ$$

look at the mirror-projection of the normal

$\rho$  is used to determine if it is the "upper" or "lower line"





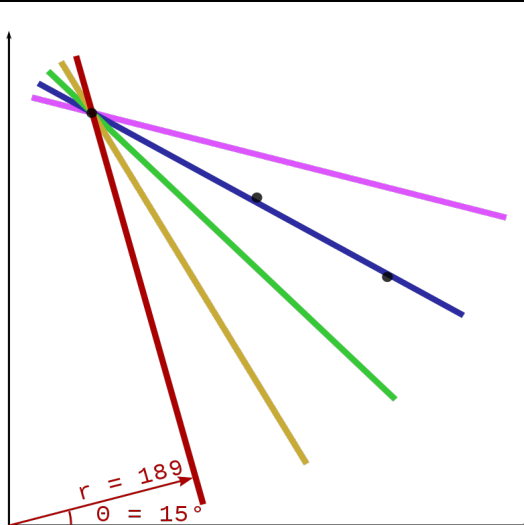
# Hough space: Let's vote for general line ...

- Basically a tool to find a line through points.

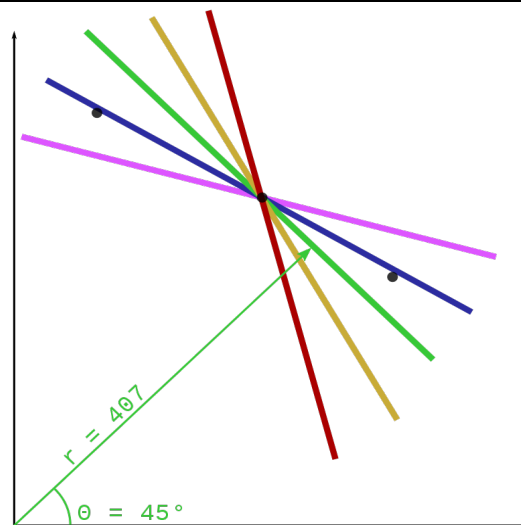
- 1) Point coordinates:  $(x, y)$
- 2) Define origin
- 3) Hough parameters space:  $(r, \theta)$
- 4) Map all possible lines through a point for different  $\theta$
- 5) "Vote" which line fit best through points

Points

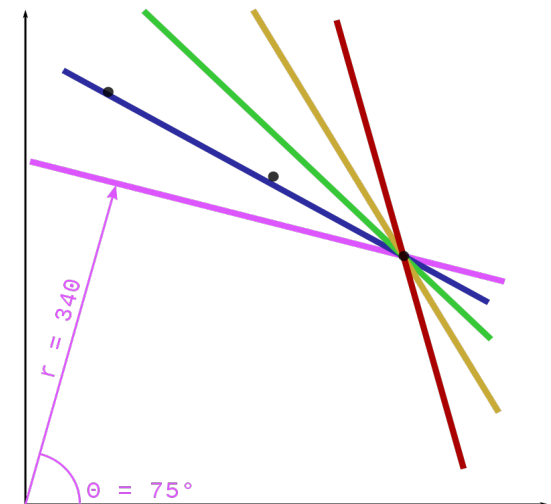
Hough space



$\theta$	$r$
15	189.0
30	282.0
45	355.7
60	407.3
75	429.4



$\theta$	$r$
15	318.5
30	376.8
45	407.3
60	409.8
75	385.3



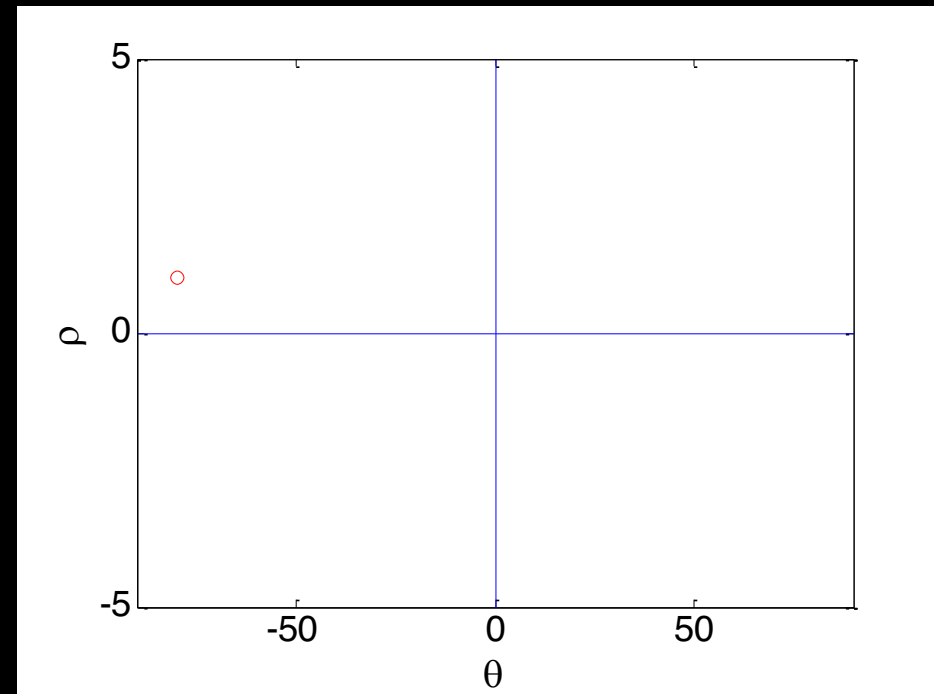
$\theta$	$r$
15	419.0
30	443.6
45	438.4
60	402.9
75	340.1



# How do we use the Hough space?

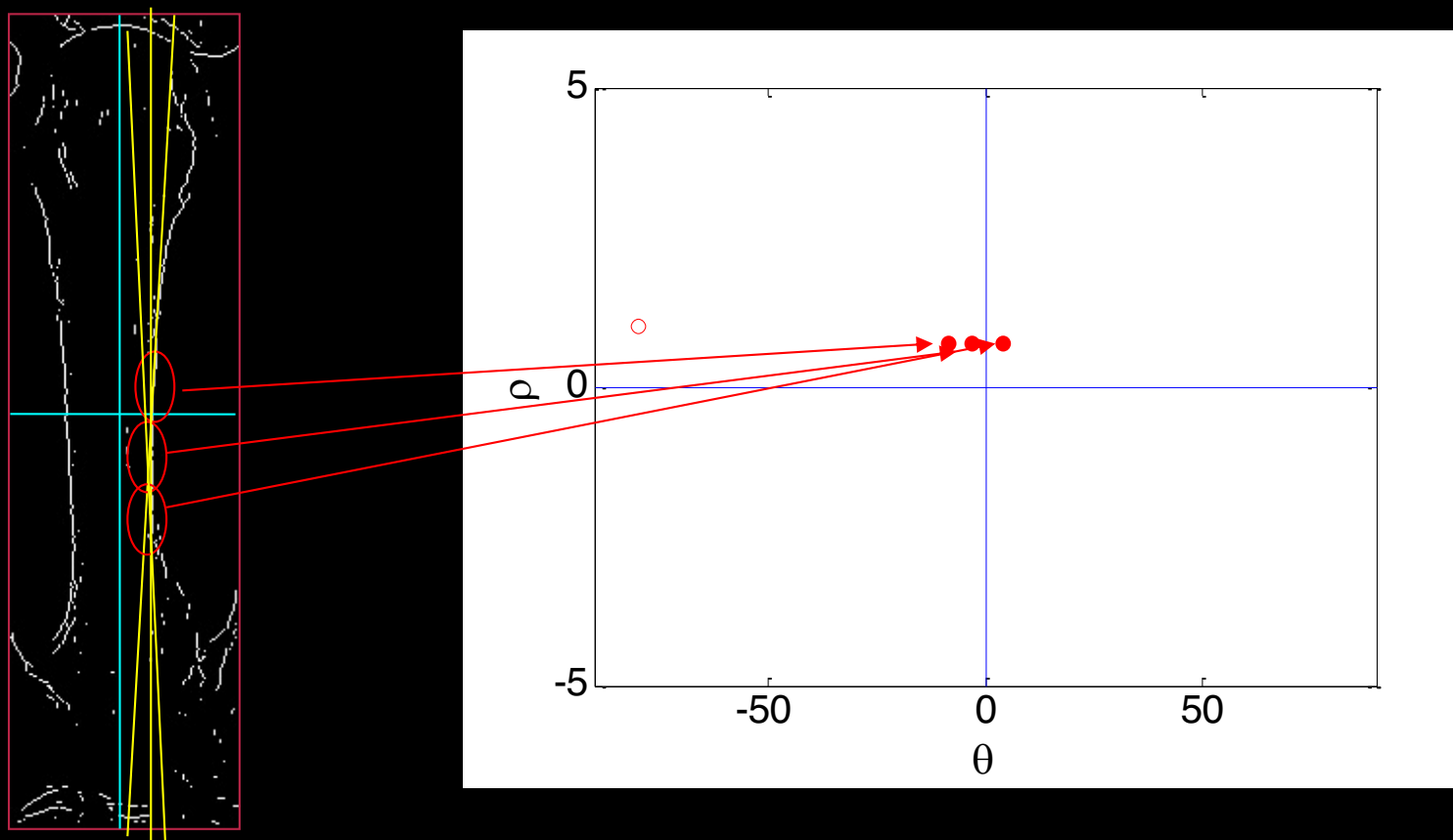


?



# How do we use the Hough space?

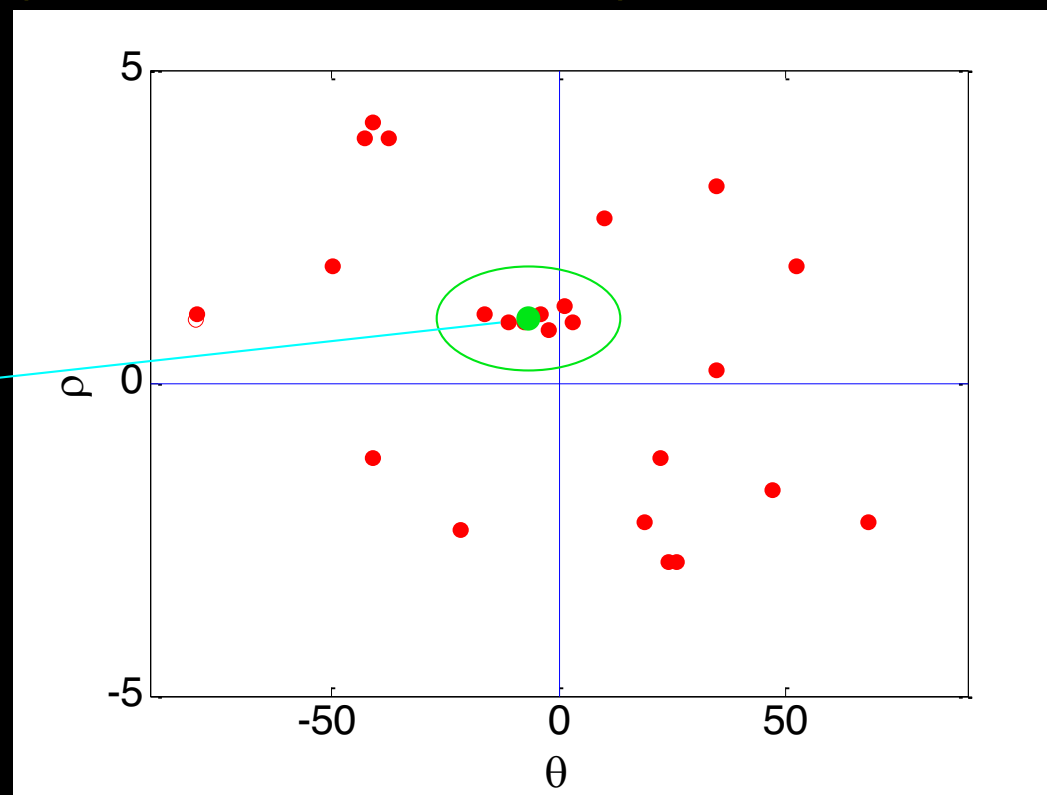
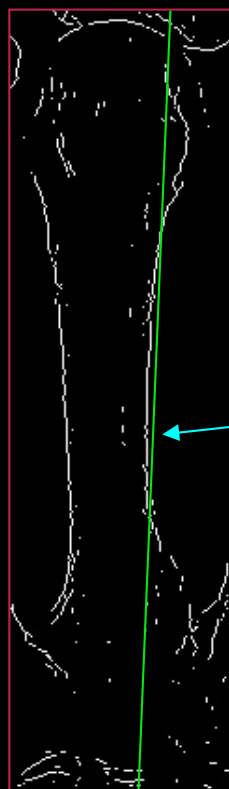
- What if every little “line-segment” was plotted in the Hough-space?



# Filled Hough-Space

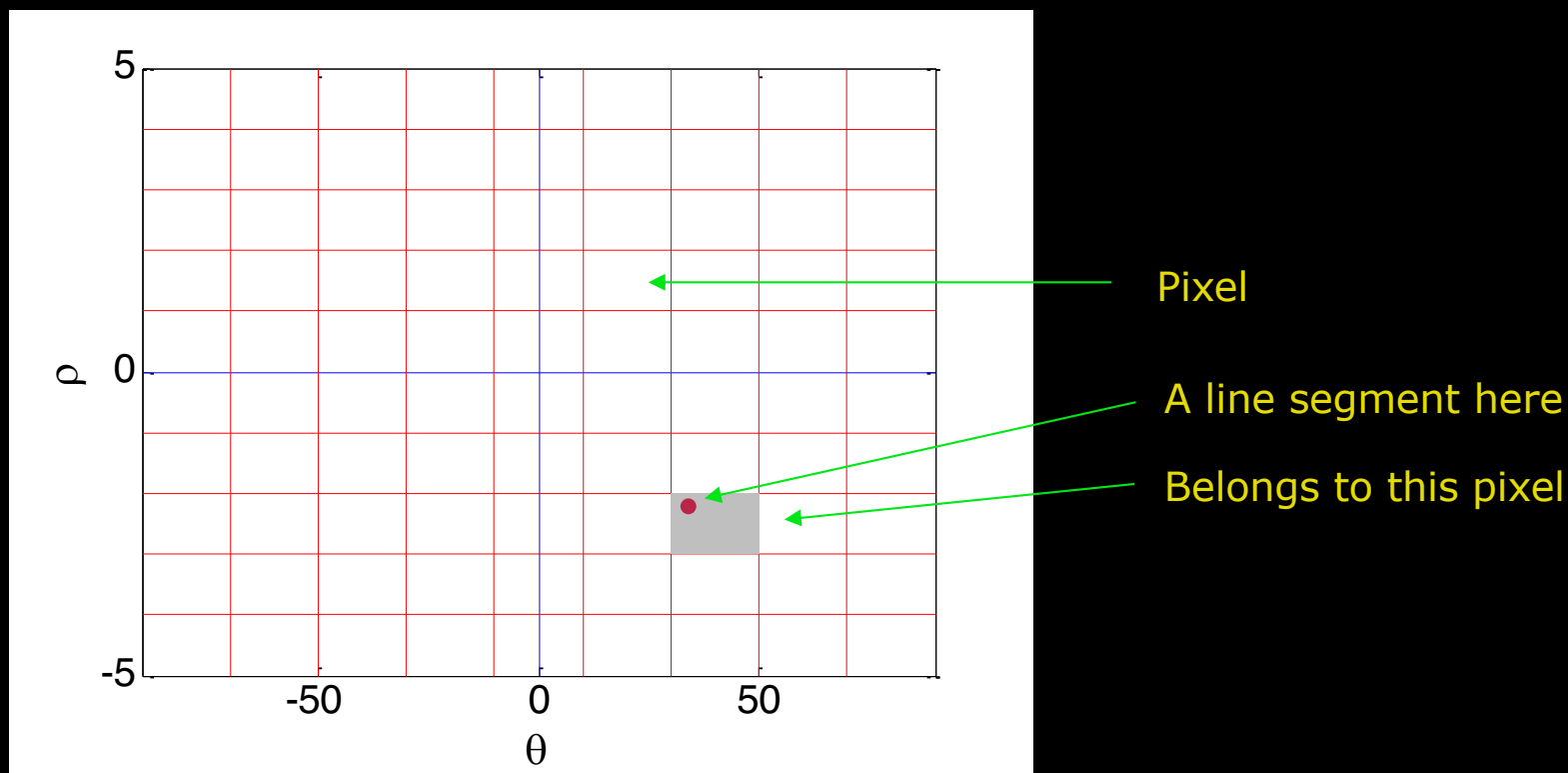
- All “line segments” in the image examined
- A “global line” can now be found as a cluster of points

In practice it is difficult to identify clusters



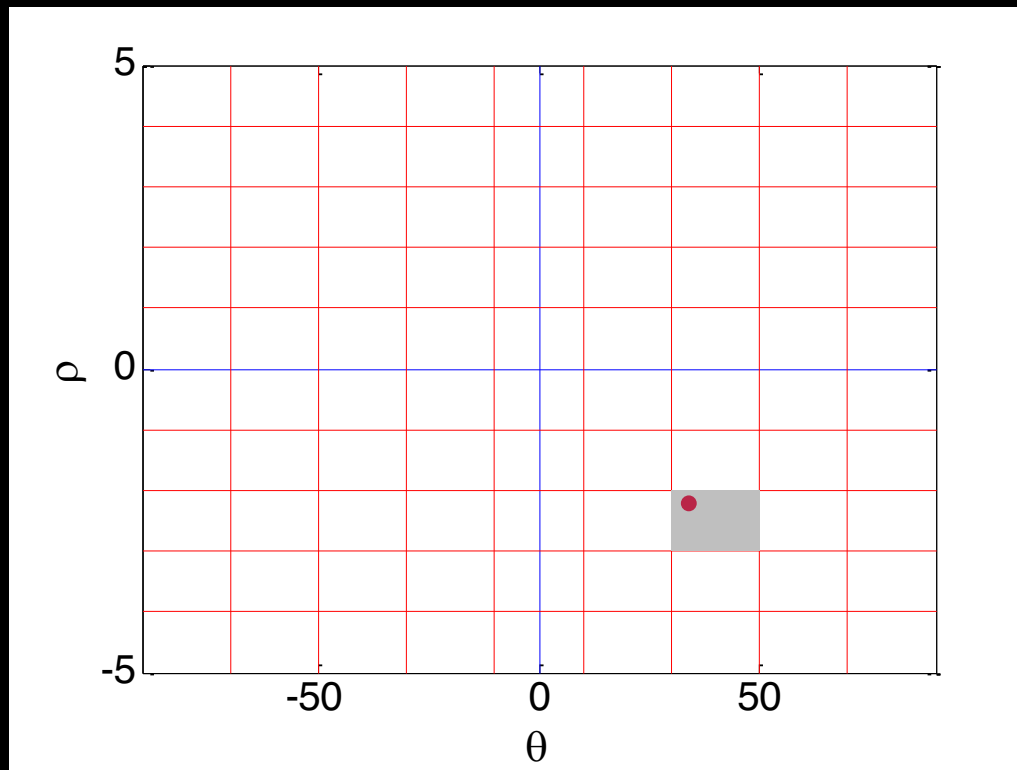
# Hough transform in practise

- Hough Space is represented as an image
- It is *quantized* – made into finite boxes



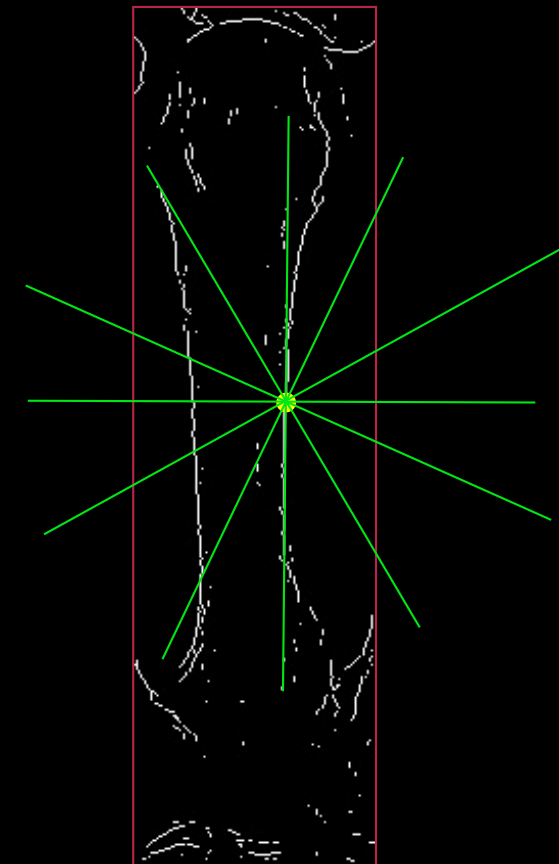
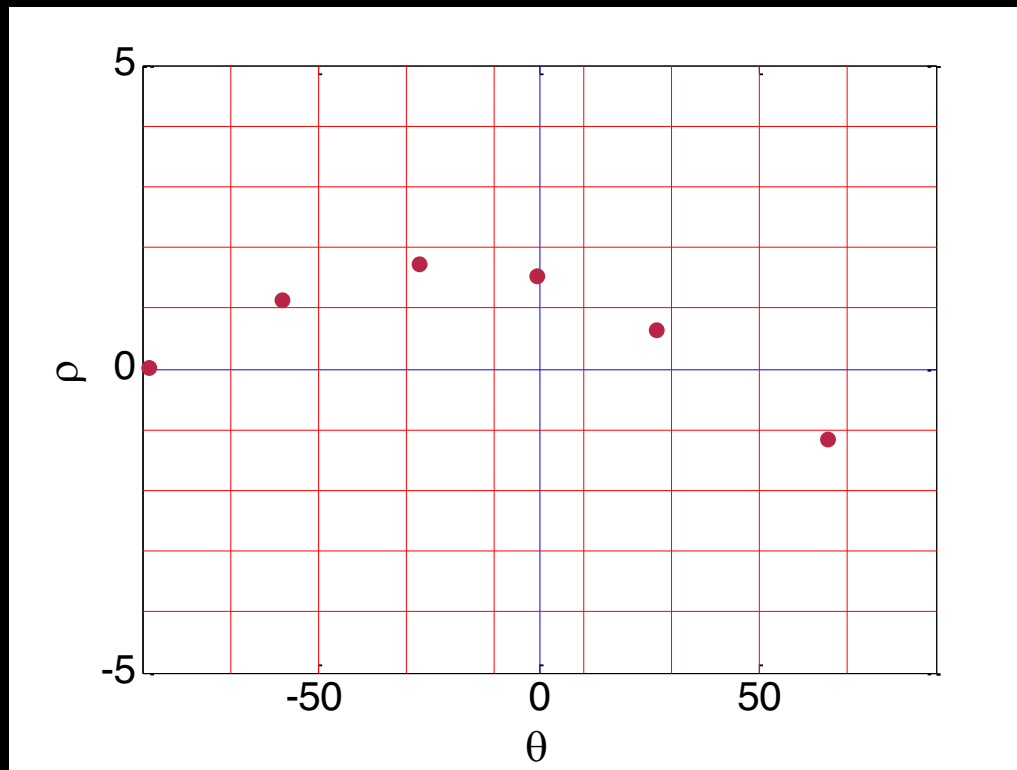
# Hough transform as a voting scheme

- The pixels in the Hough space are used to *vote* for lines.
- Each *line segment* votes by putting *one vote* in a pixel
- The pixels are also called *accumulator cells*



# Hough transform per pixel

- In practise we do not use line segments
- Each pixel in the input image votes for **all** potential lines going through it.



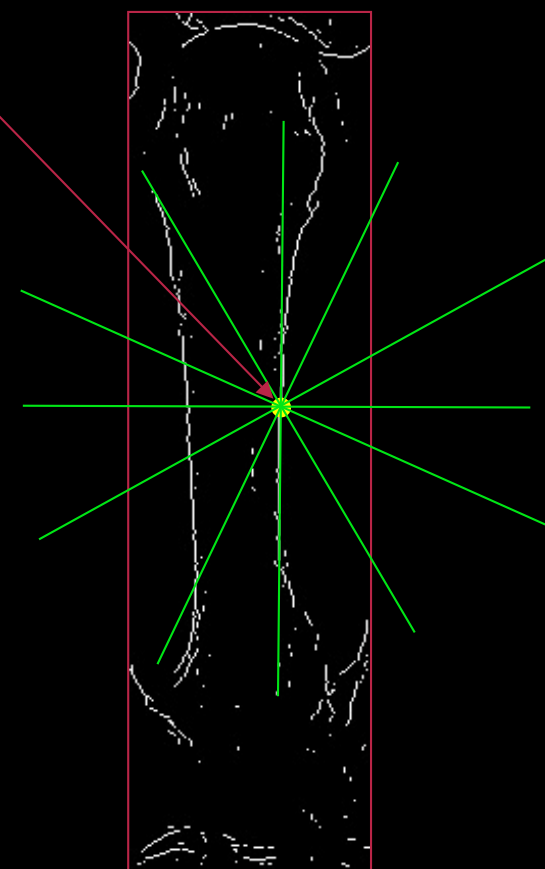
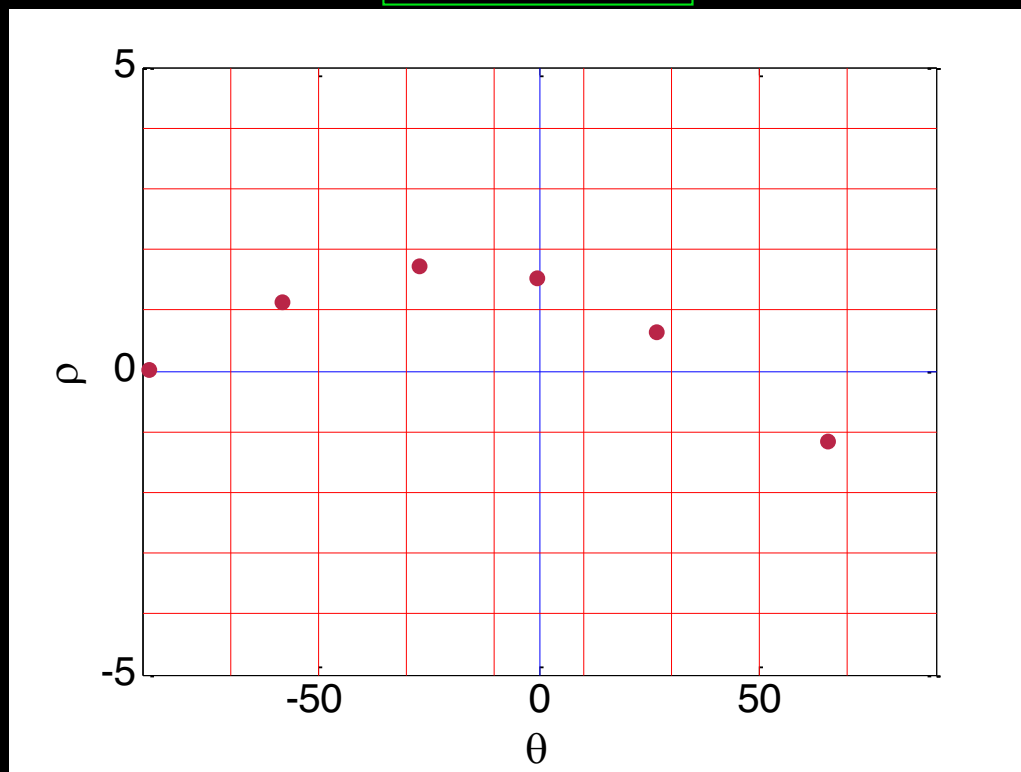
# Hough transform per pixel

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

Go through all  $\theta$  and calculate  $\rho$

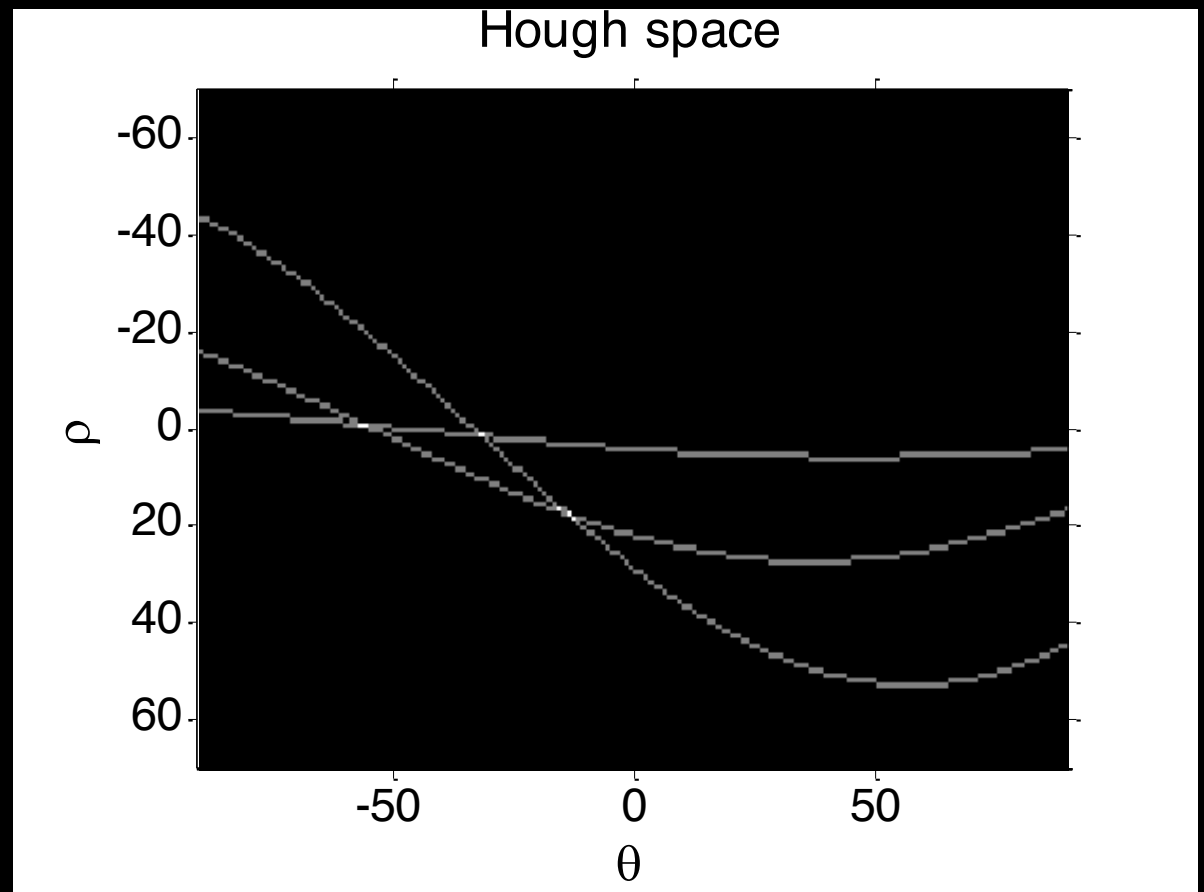
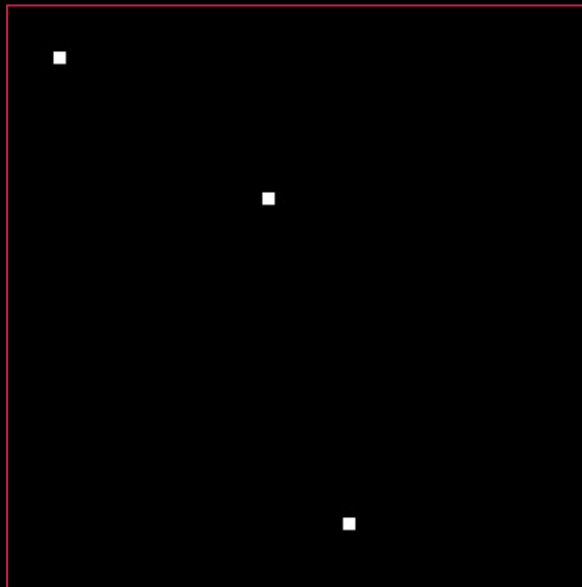
(x, y) are fixed

Sinusoid!

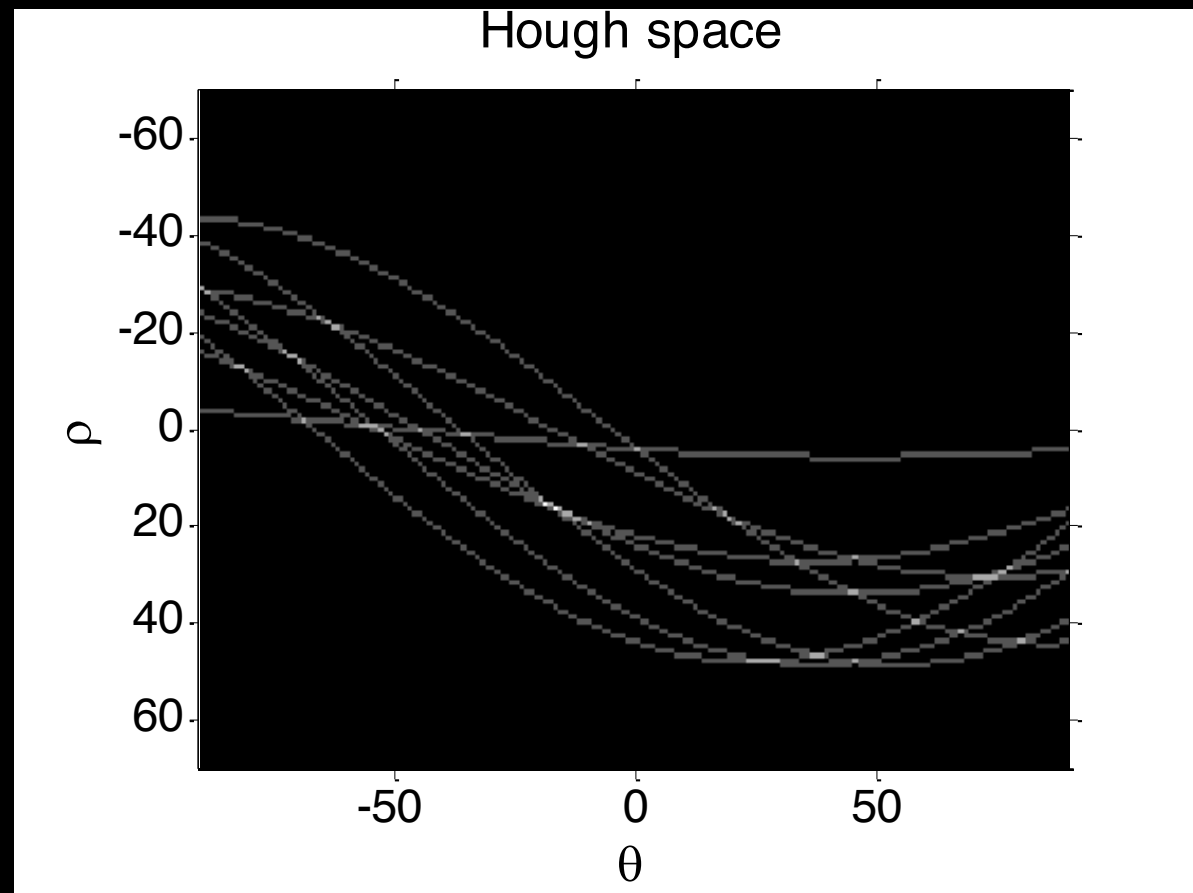
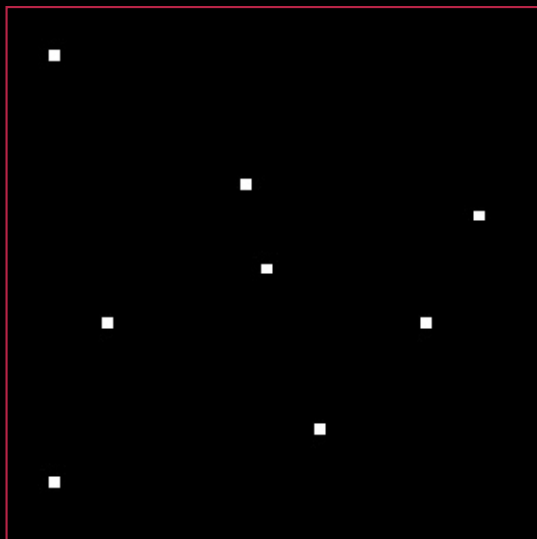




# Real Hough Transform



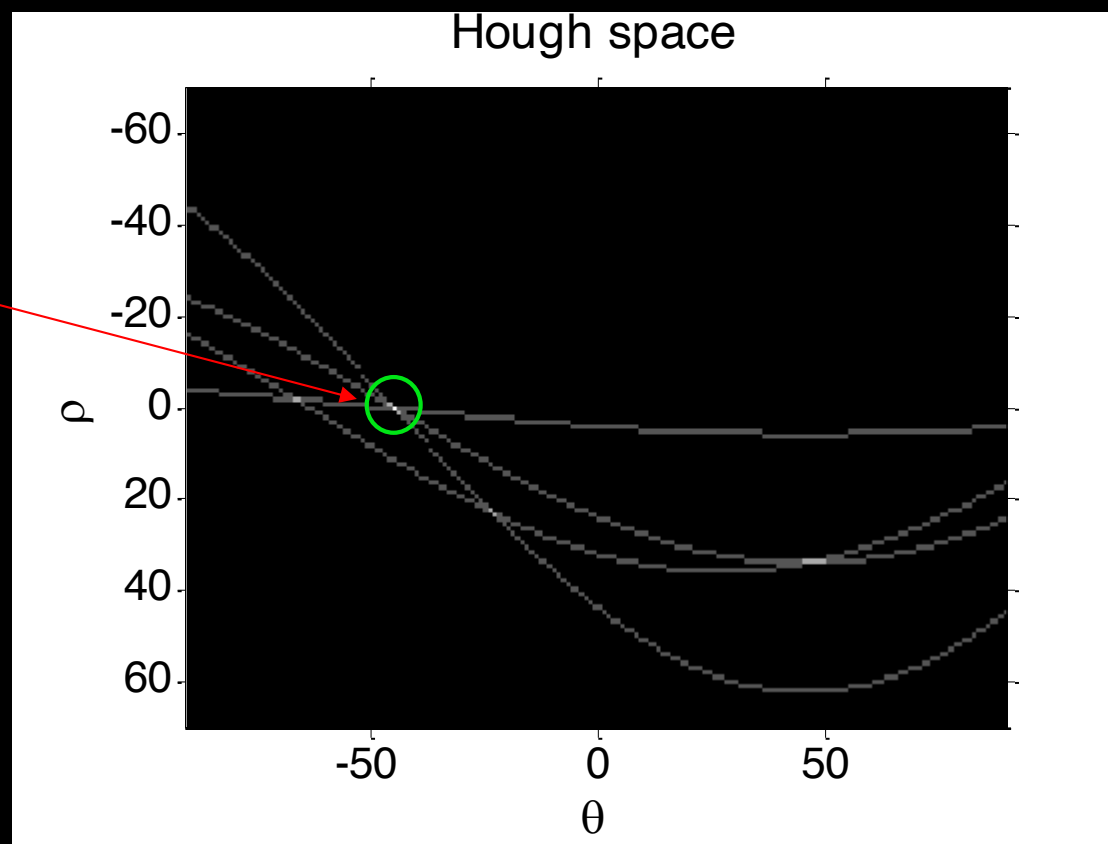
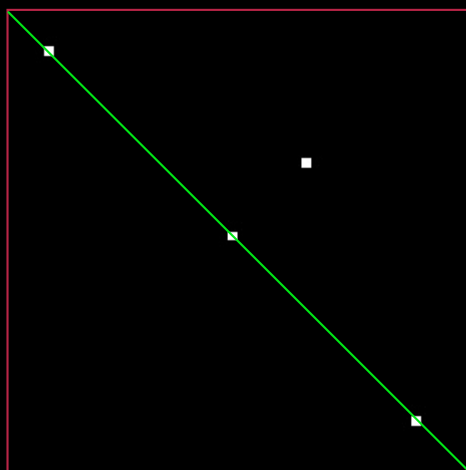
# Real Hough Transform II



# Real Hough Transform and lines

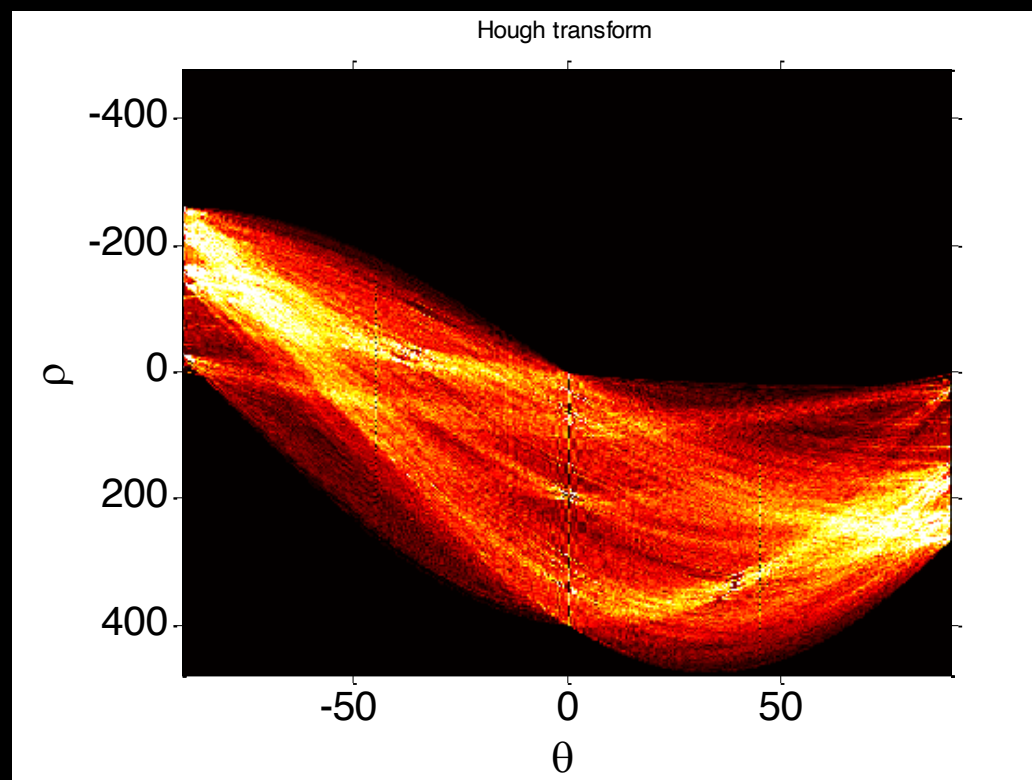
Spot the line!

A maximum where Hough pixel has value 3

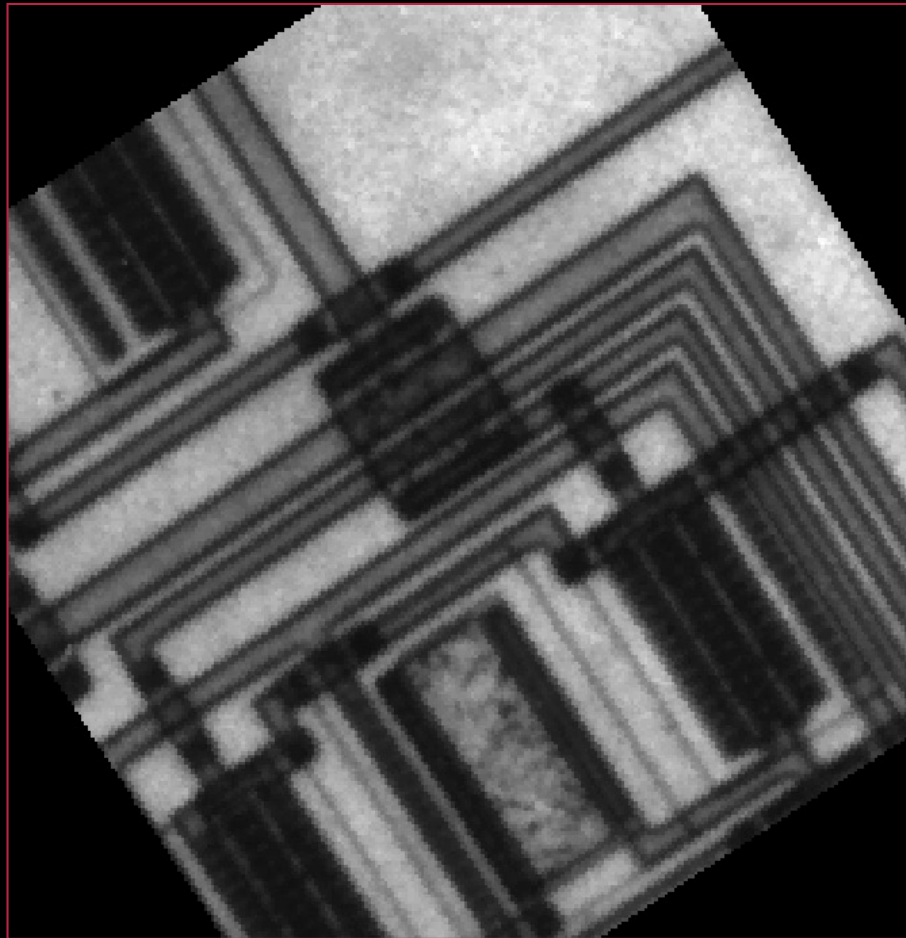


## Finding the lines in Hough space

- The lines are found in Hough space where *most pixels have voted for there being a line*
- Can be found by searching for maxima in Hough Space

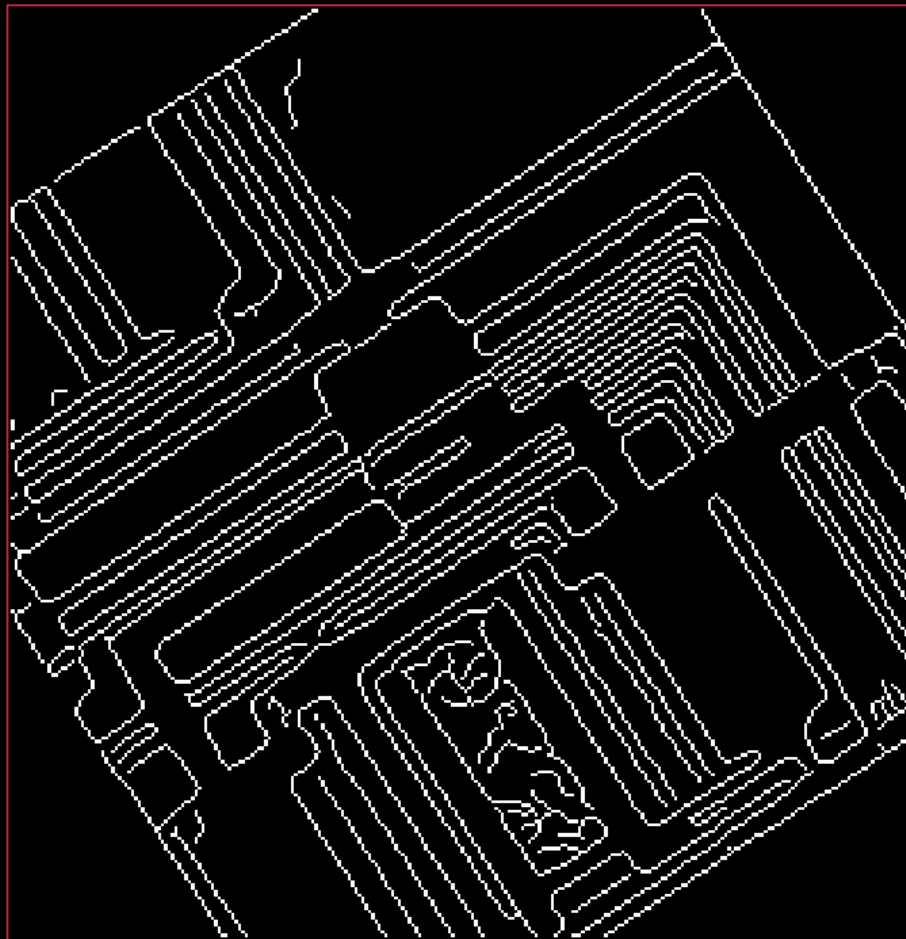


# The practical guide to the Hough Transform



- Start with an input image

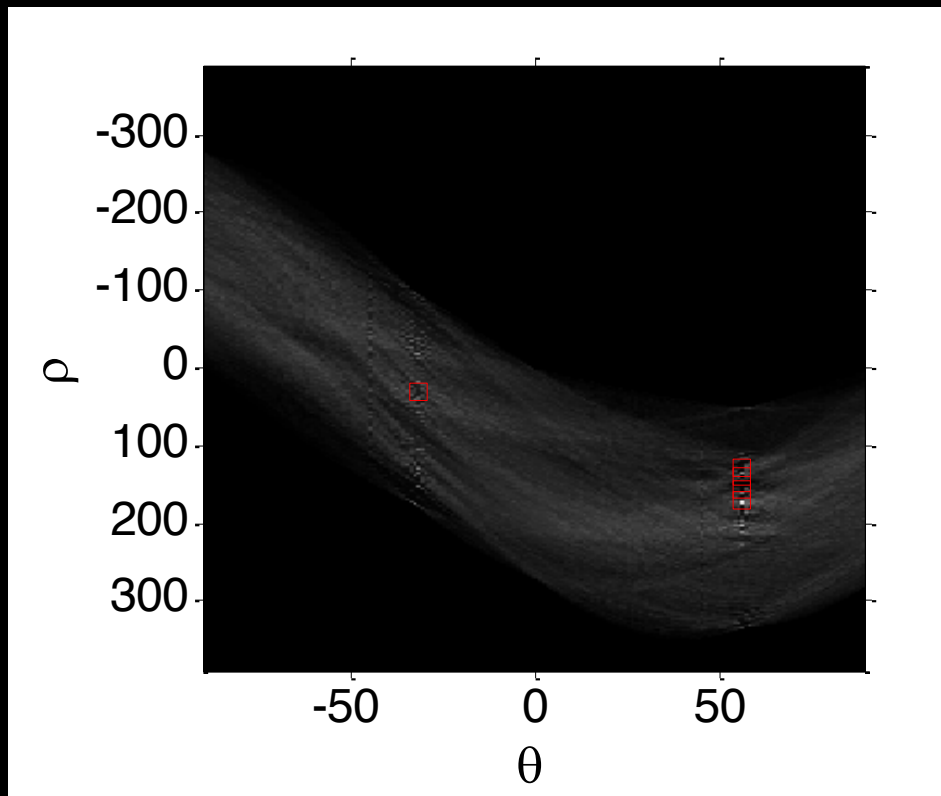
# The practical guide to the Hough Transform



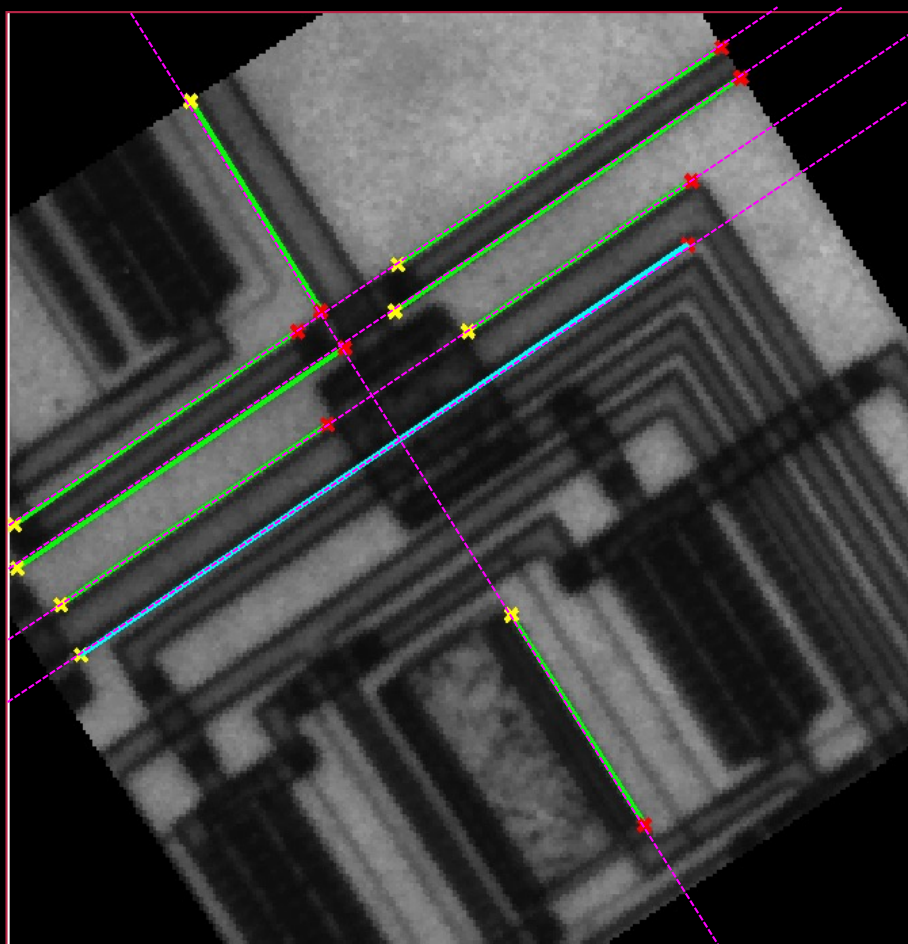
- Detect edges and create a binary image

# The practical guide to the Hough Transform

- Compute Hough transform and locate the maxima



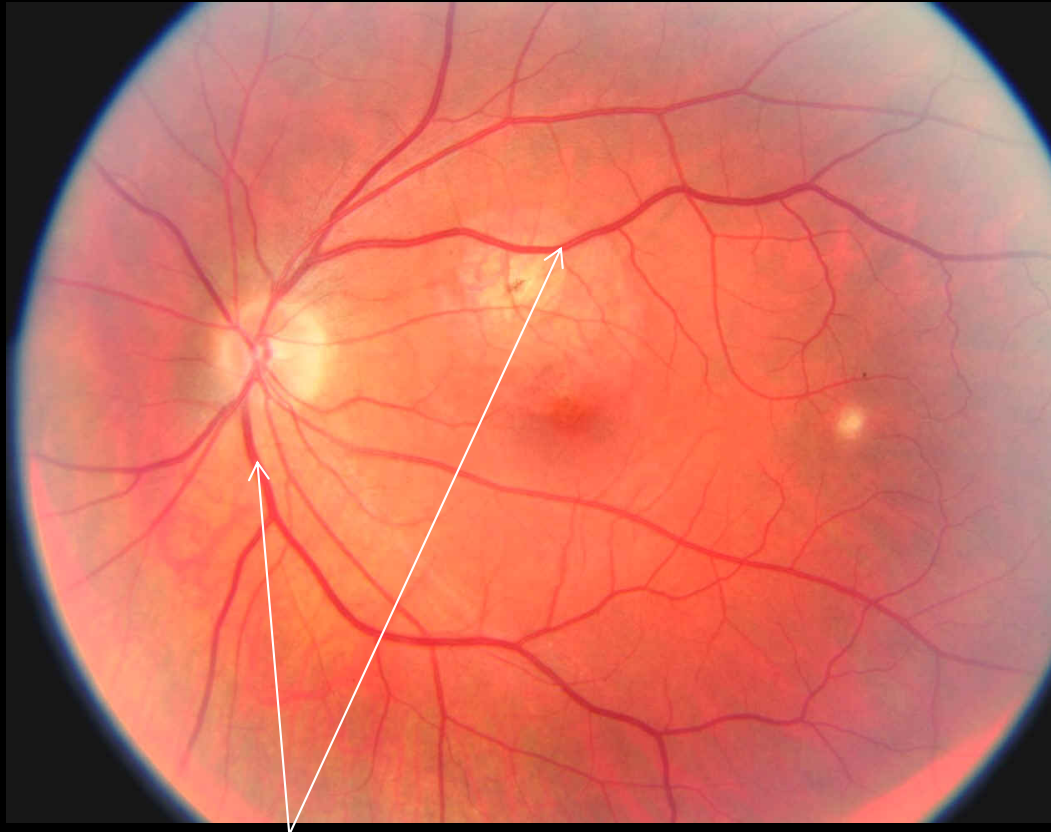
# The practical guide to the Hough Transform



- Draw the lines corresponding to the found maxima
- Here the **cyan** line is the longest



# Path Tracing

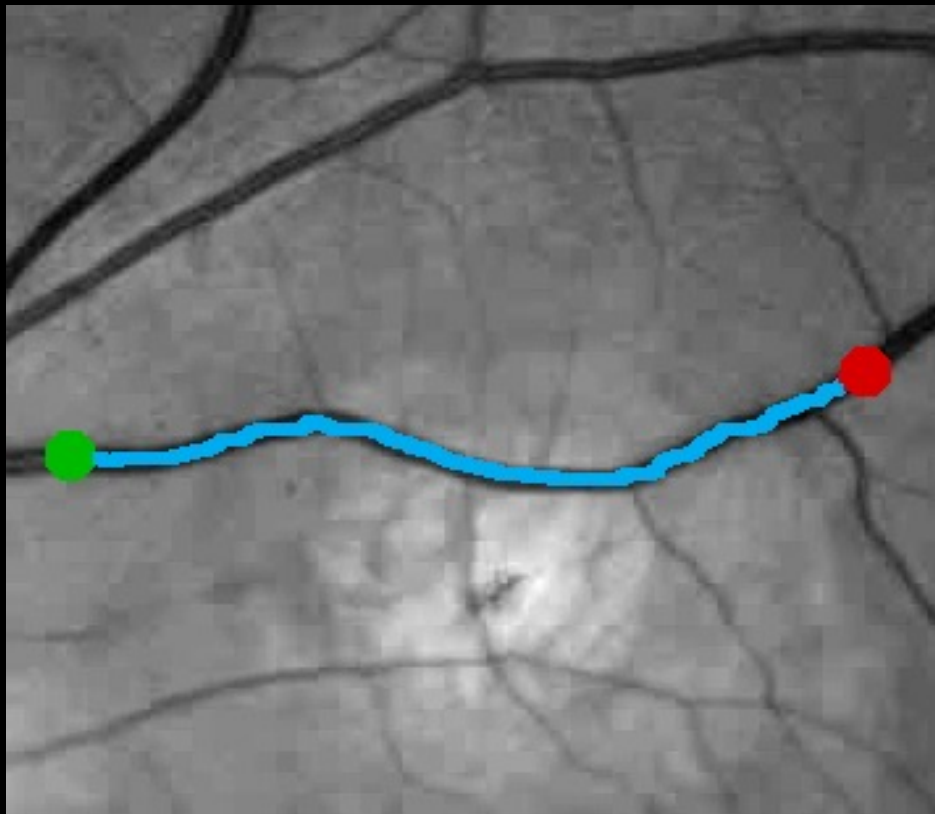


Arteries and veins

Fundus image

- The diameter as function of the distance to the optic cup tells something about the patients health
- We need to find the arteries and veins
- Path tracing is one solution

# Path tracing



- A path is defined as a curve in an image defined as *something that is different from the background*
- In this case it is a dark line
- Pre-processing can for example turn edges into dark lines.

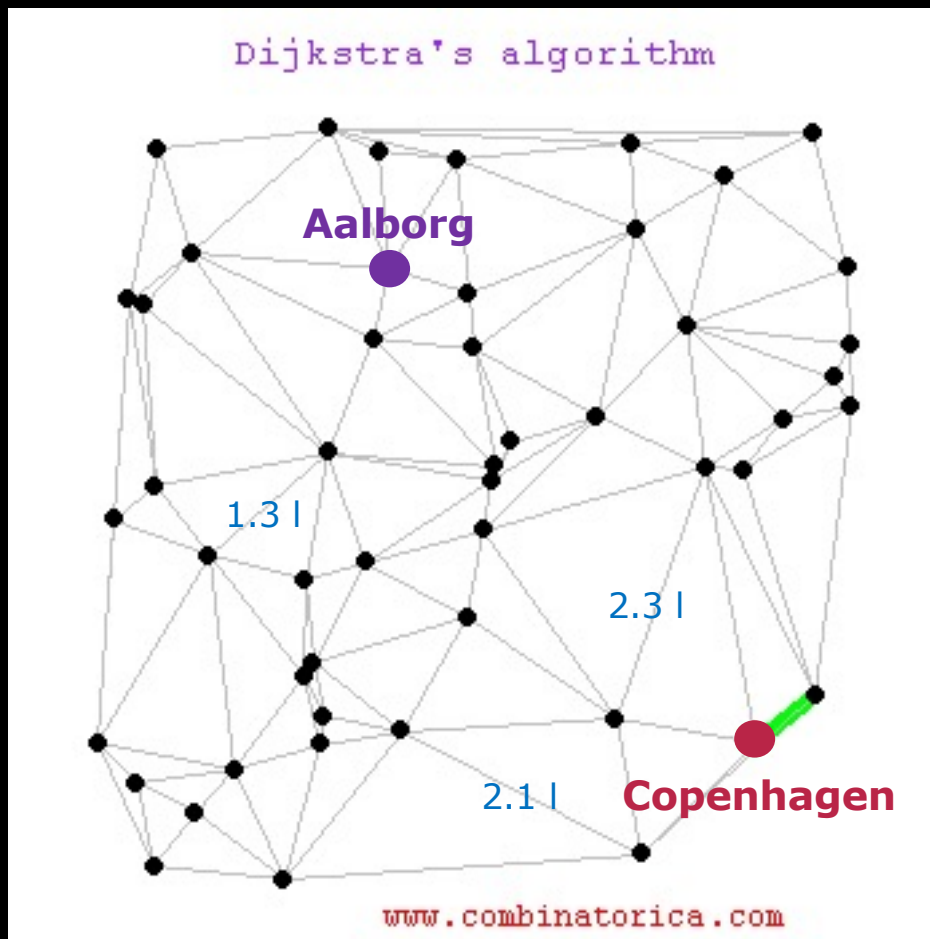
# Dynamic Programming



- Break up large problem into many small sub-problems
- A classic algorithm:
  - Dijkstra's algorithm
  - One source to all nodes shortest path
- We will look at a simplified variant

Dijkstra, E. W. (1959). "A note on two problems in connexion with graphs". Numerische Mathematik. 1: 269–271.

# Path tracing



- A GPS device uses path tracing
- Based on *graph algorithms*
  - A city is a node
  - A road is an edge. The weight of the edge is the fuel cost
- How do we come from Copenhagen to Aalborg using the least fuel?
- Dijkstra's algorithm

# Images as graphs

140	190	73	19	60
130	212	14	100	145
150	20	80	135	120
157	140	33	199	100
121	234	45	210	86

$$C(2, 3) = 14$$

- Each pixel is a node
- Pixel neighbours are connected by edges
- The edge cost ( $c(r, c)$ ) is the pixel value
- Directed graph
- Imagine a car driving on the image
- Called a *cost image*

## Simplified problem



- Track dark lines
- Path going from top to bottom
- No sharp turns – smooth
- Problem:
  - from the top to the bottom
  - Sum of pixel values should be minimal

## Simplified problem



- Pixel value at  $(r, c)$  equals the cost

$C(r, c)$

- The path  $P$  consist of pixels
- The sum of pixel values in the path

$$C_{tot} = \sum_{(r, c) \in P} C(r, c)$$

## Path cost

140	190	73	19	60
130	212	14	100	145
150	20	80	135	120
157	140	33	199	100
121	234	45	210	86

$P = [(1,3), (2,3), (3, 2), (4,3), (5,4)]$

- A path is defined as  $(r,c)$  coordinates

$$C_{tot} = \sum_{(r,c) \in \mathcal{P}} C(r,c)$$



# Quiz 1: Total cost – what is $C_{tot}$ ?

A) 167

B) 350

C) 403

D) 270

E) 345

140	190	73	19	60
130	212	14	100	145
150	20	80	135	120
157	140	33	199	100
121	234	45	210	86

$$P = [(1,3), (2,3), (3,2), (4,3), (5,4)]$$



## Path cost

140	190	73	19	60
130	212	14	100	145
150	20	80	135	120
157	140	33	199	100
121	234	45	210	86

$P = [(1,3), (2,3), (3, 2), (4,3), (5,4)]$

- This is *NOT* the optimal path
- How do we compute the path  $P$  that has minimum  $C_{tot}$ ?
- Test all possible paths?
  - No! Impossible amount of possibilities

## Quiz 2: Path Cost

A) 196

B) 154

C) 201

D) 185

E) 132

A path has been found in the image  
 $P = [(1,4), (2,4), (3,5), (4,5), (5,5), (6,4)]$ . A Matlab matrix coordinate system is used. What is the total cost of the path?

208	157	234	19	145	79
62	121	73	14	120	135
237	90	193	135	3	42
89	212	192	199	86	154
50	149	97	238	41	67
64	140	145	33	203	167

Figur 1: Grayscale billede



## Path restriction: The rules

140	190	73	19	60
130	212	14	100	145
150	20	80	135	120
157	140	33	199	100
121	234	45	210	86

- Path is only allowed to
  - Go down
  - Move one pixel left or right
- Longer jumps not allowed

# Accumulator image

140	190	73	19	60
270	285	33	119	164
420	53	113	168	239
210	193	86	312	263
314	320	131	296	354

- Keeps track of the accumulated cost for efficient paths finding
- Path ending here has cost 296



# Computing the accumulator image

140	190	73	19	60
130	212	14	100	145
150	20	80	135	120
157	140	33	199	100
121	234	45	210	86

140	190	73	19	60
0	0	0	0	0
0	0	0	0	0
0	0	0	0	0
0	0	0	0	0

Step 1: Copy first row of input image

# Computing the accumulator image

140	190	73	19	60
130	212	14	100	145
150	20	80	135	120
157	140	33	199	100
121	234	45	210	86

140	190	73	19	60
270	285	33	119	164
0	0	0	0	0
0	0	0	0	0
0	0	0	0	0

Step 2: Fill second row

$$A(r, c) = I(r, c) + \min(A(r-1, c-1), A(r-1, c), A(r-1, c+1))$$

# Computing the accumulator image

140	190	73	19	60
130	212	14	100	145
150	20	80	135	120
157	140	33	199	100
121	234	45	210	86

140	190	73	19	60
270	285	33	119	164
420	53	113	168	239
210	193	86	312	268
314	320	131	296	354

Step 3: Fill all rows by looking at the previous row

$$A(r, c) = I(r, c) + \min(A(r-1, c-1), A(r-1, c), A(r-1, c+1))$$



## Quiz 3: Accumulator Image

- A) 57
- B) 167
- C) 301
- D) 241**
- E) 145

An optimal path has been found in the image.  
What is the value of the accumulator image in the marked pixel?

117	163	74	210
223	244	171	57
132	61	110	170
241	172	17	215

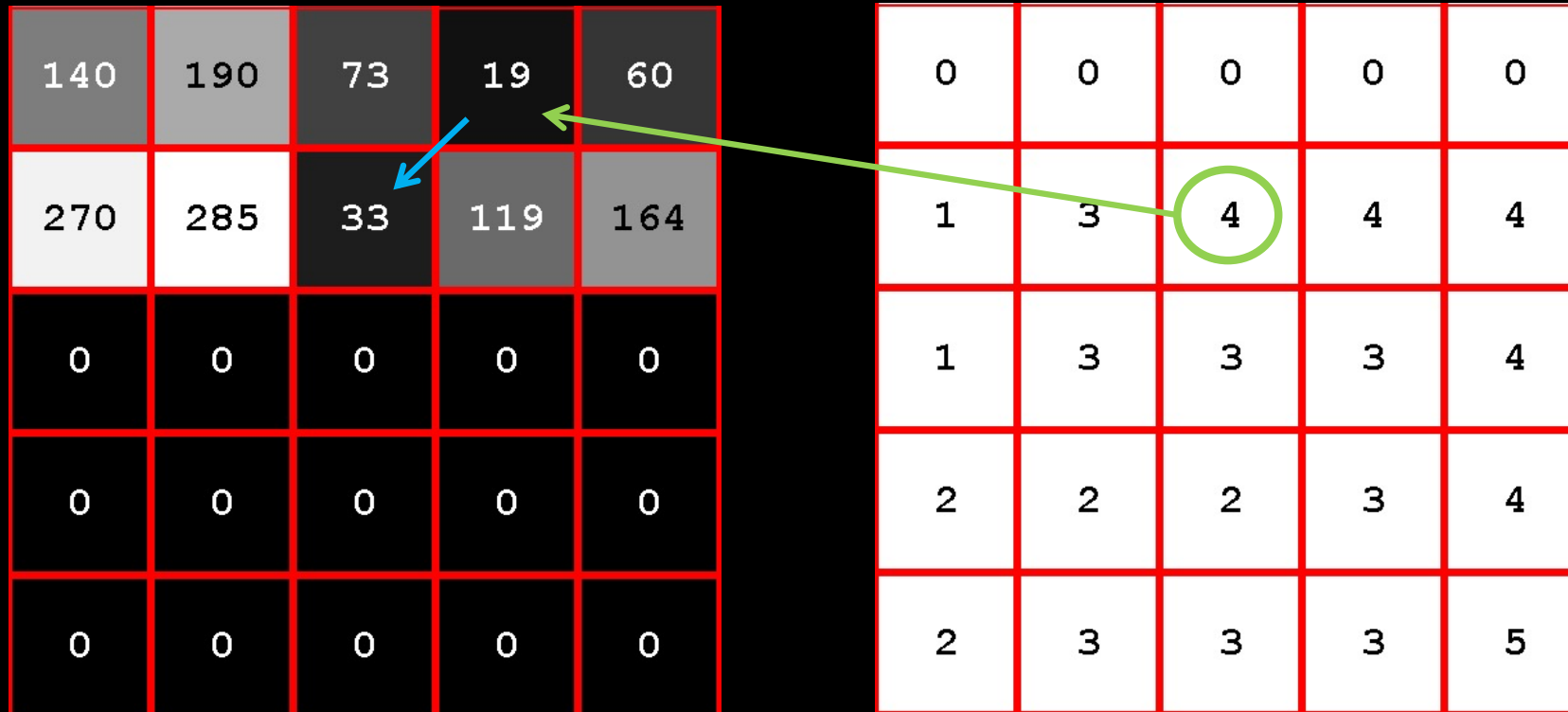
## Using the accumulator image

140	190	73	19	60
130	212	14	100	145
150	20	80	135	120
157	140	33	199	100
121	234	45	210	86

140	190	73	19	60
270	285	33	119	164
420	53	113	168	239
210	193	86	312	268
314	320	131	296	354

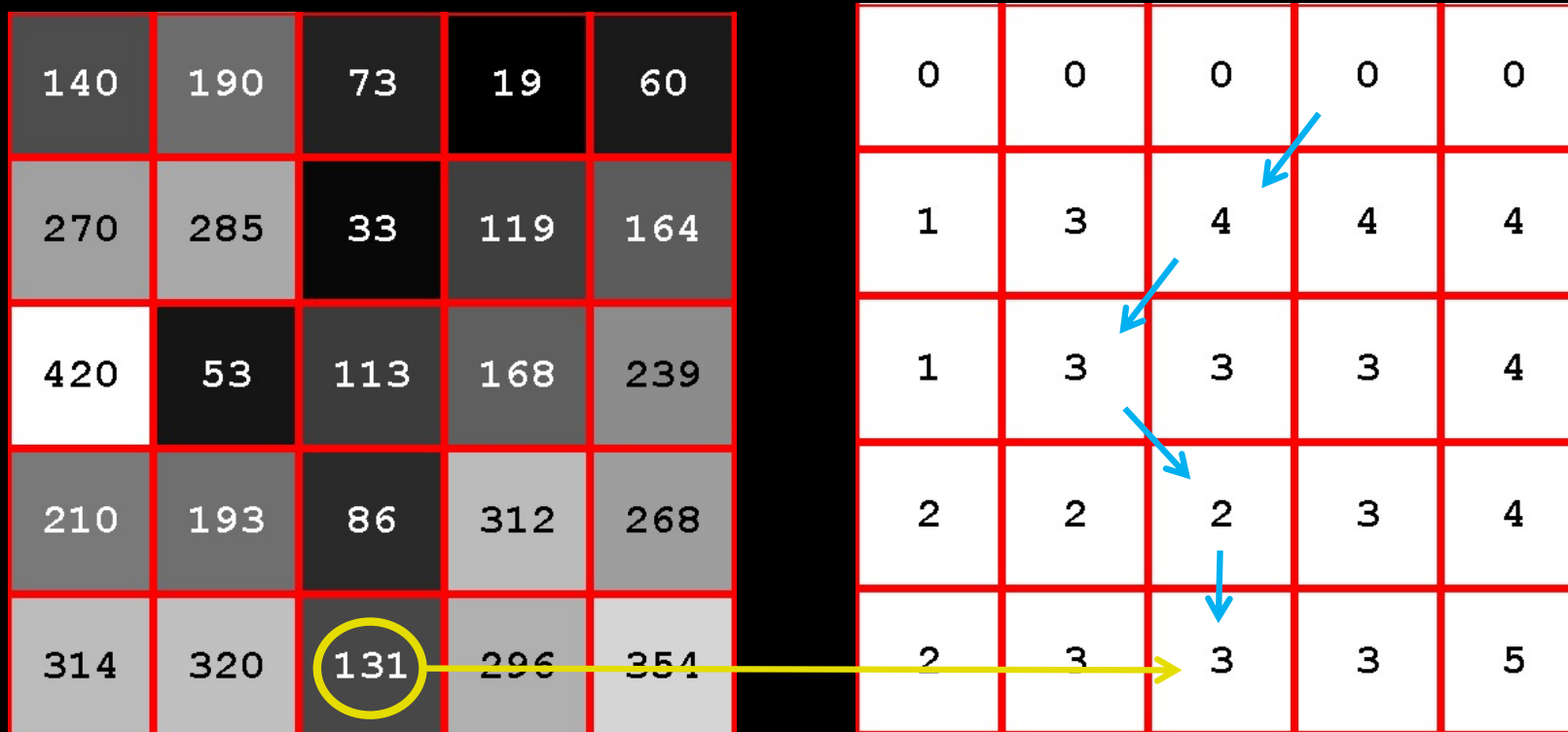
Step 4: The end of the optimal path can now be found

## The backtracing image



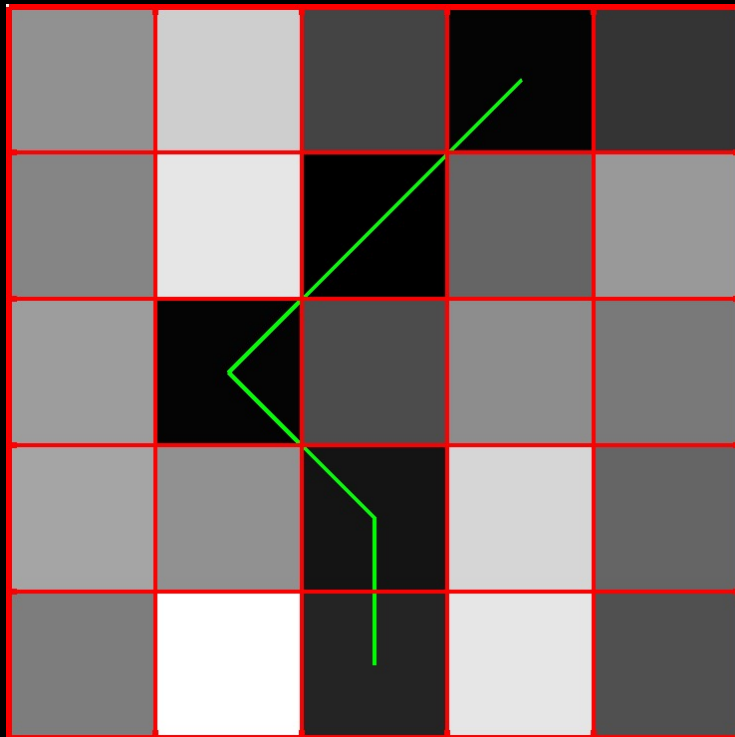
- Keeps track of where the path *came* from
- Each pixel stores the column number

## Using the backtracing image



Step 5: Trace the path in the backtracing image

# Using the backtracing image



0	0	0	0	0
1	3	4	4	4
1	3	3	3	4
2	2	2	3	4
2	3	3	3	5

A 5x5 grid of numerical values. Blue arrows indicate a path from the top-right to the bottom-left, following the values: (0,4) to (1,3) to (2,2) to (3,1) to (4,0).

## Quiz 4: Backtracing

A) 1

B) 2

C) 3

D) 4

E) 5

An optimal path has been found in an image. The backtracing image is seen below and the optimal path ends in the marked pixel. A Matlab coordinate system is used. What is the optimal path?

0	0	0	0	0
1	3	3	3	5
2	2	2	4	4
1	1	4	5	5
1	1	4	4	4

1.  $\mathcal{P} = [(1, 3), (2, 2), (3, 1), (4, 1), (5, 2)]$
2.  $\mathcal{P} = [(1, 3), (2, 2), (3, 2), (4, 2), (5, 2)]$
3.  $\mathcal{P} = [(1, 2), (2, 2), (3, 2), (4, 1), (5, 2)]$
4.  $\mathcal{P} = [(1, 3), (2, 1), (3, 1), (4, 1), (5, 2)]$
5.  $\mathcal{P} = [(1, 2), (2, 1), (3, 1), (4, 2), (5, 2)]$

# Pre-processing



- We would like to track paths that are not dark curves

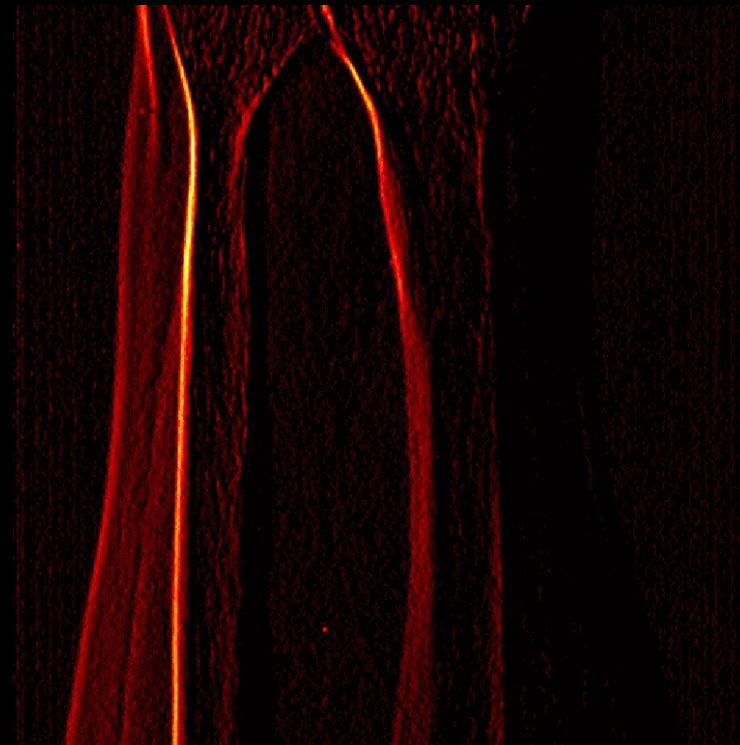
## Quiz 5 : X-ray preprocessing

- A) Gaussian smoothing
- B) 255-I
- C) Gradient filter
- D) Registration
- E) Morphological operation



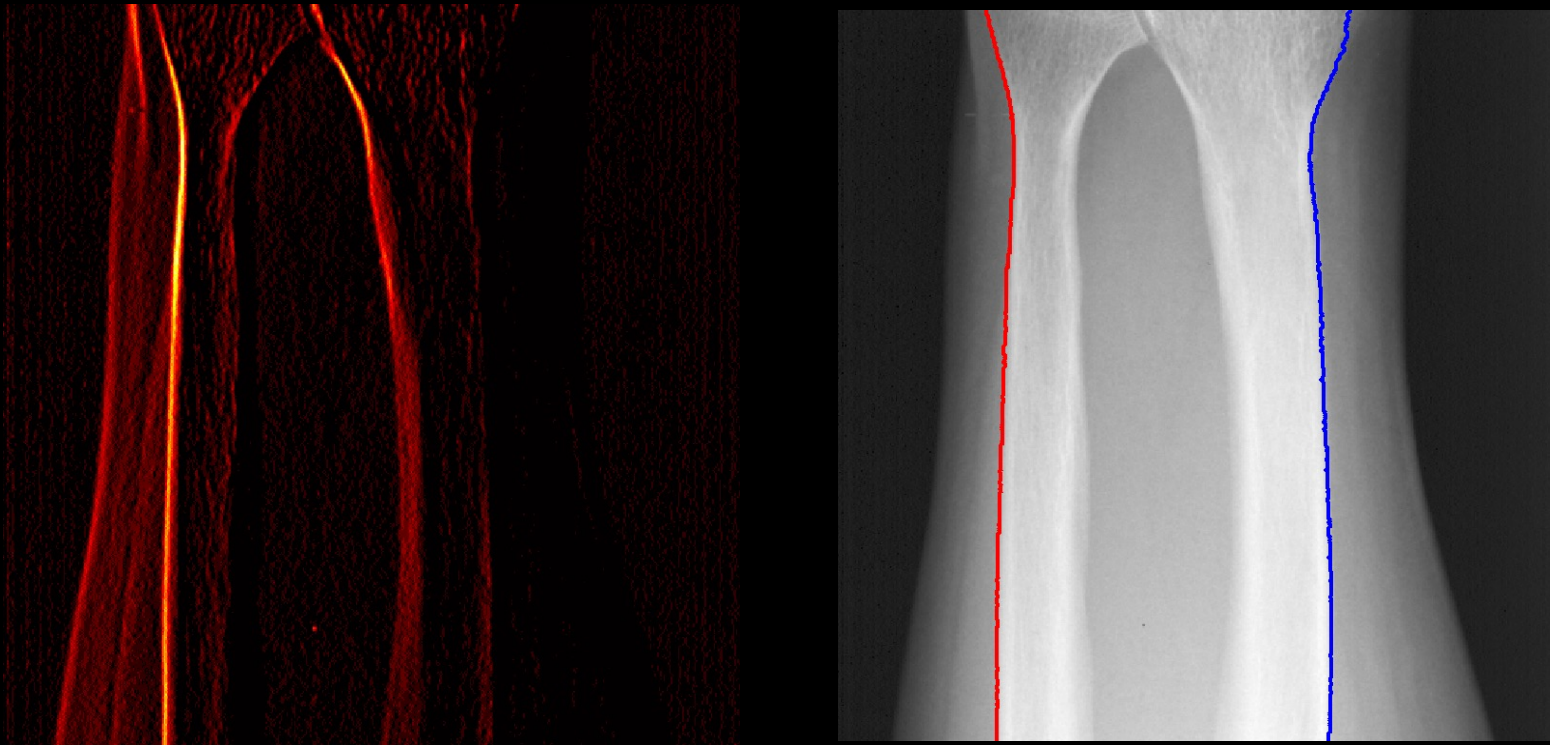


# Pre-processing



Edge filtered image  
(Gaussian smoothing followed by Prewitt)

# Path tracing on pre-processed image



Paths found on pre-processed image and  
intensity inverted



## Quiz 6: Optimal Path 2

A) 81

B) 64

C) 11

D) 73

E) 51

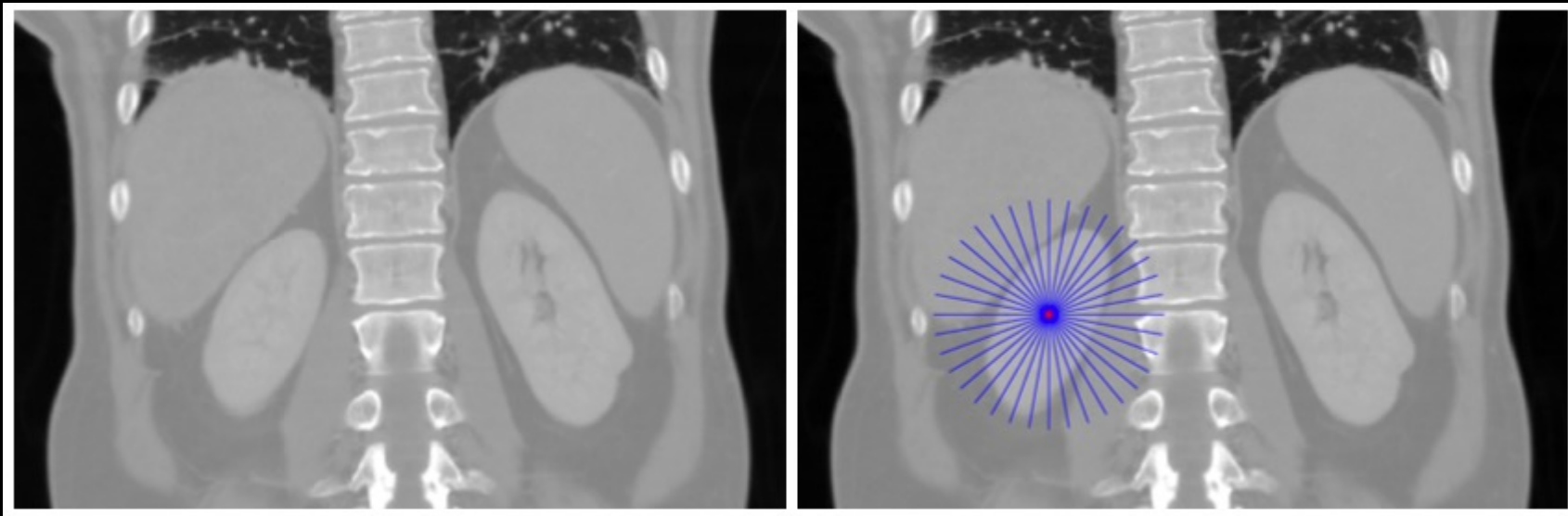
A 5 x 5 image is filled with values given the gray level run length encoding: 2, 180, 1, 15, 3, 112, 1, 8, 4, 177, 1, 20, 4, 195, 1, 12, 3, 242, 2, 25, 3, 9. After that the optimal path is found. What is the total cost?

Solution:

$$15+8+20+12+9=64$$

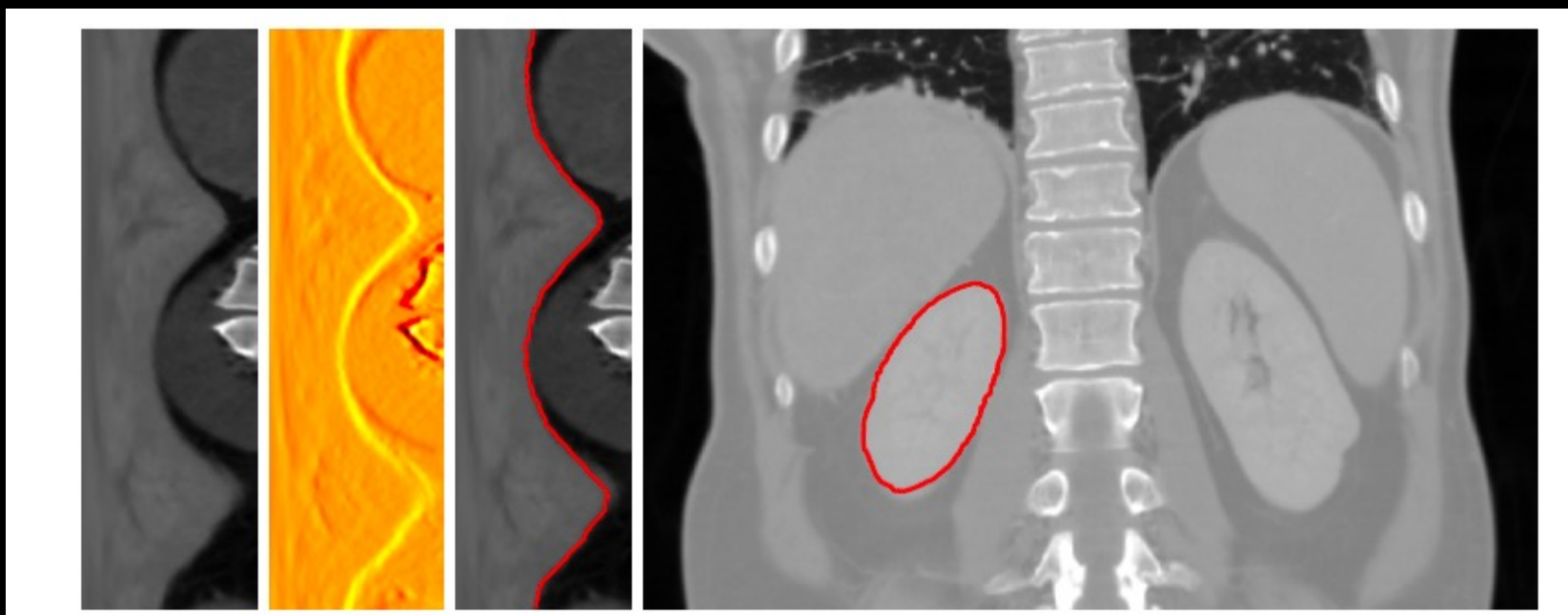
180	180	15	112	112
112	8	177	177	177
177	20	195	195	195
195	12	242	242	242
25	25	9	9	9

# Locating Circular Structures



- Define origin inside structure
- Send out spokes

# Locating Circular Structures



- Each spoke is a line in a new image (surface- layer detection)
- Prewitt
- Dijkstra's algorithm
- Map back the spokes into image



# What did you learn today?

- Use the Hough transform for line detection
- Describe the slope-intercept, the general form and the normalised form of lines
- Describe the connection between lines and the Hough space
- Use edge detection to enhance images for use with the Hough transform
- Use dynamic programming to trace paths in images
- Describe how an image can be used as a graph
- Describe the fundamental properties of a cost image
- Compute the cost of path
- Compute an accumulator image for path tracing
- Compute a back tracing image for path tracing
- Choose appropriate pre-processing steps for path tracing
- Describe how circular structures can be located using path tracing



# Lecture 9 – Statistical models of shape and appearance

