

Computer Vision

Hough Transform
Morphological
Operations

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What we have learnt

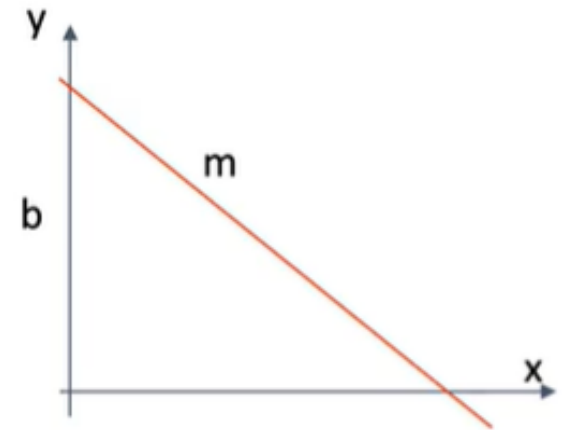
- Gradient Detection
 - Sobel Filter
 - Prewitt
 - Laplacian Filter
- Edge detection
 - Noise reduction
 - Detection of edge points
- Counters

Content

- Hough Transform
- Morphological operations

Hough Transform

- Hough transform is a popular technique to detect straight lines
- Normally a straight line is defined as:
$$y = mx + b$$
- Where m is the slope and b is the intercept

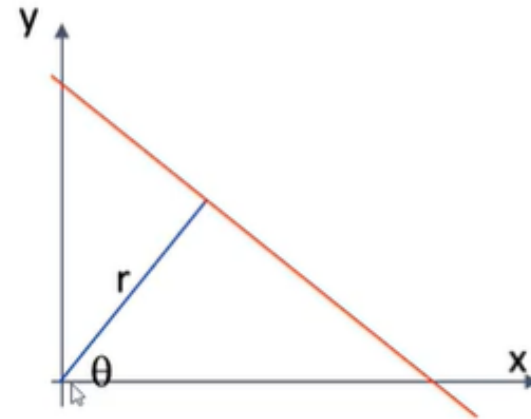


Hough Transform

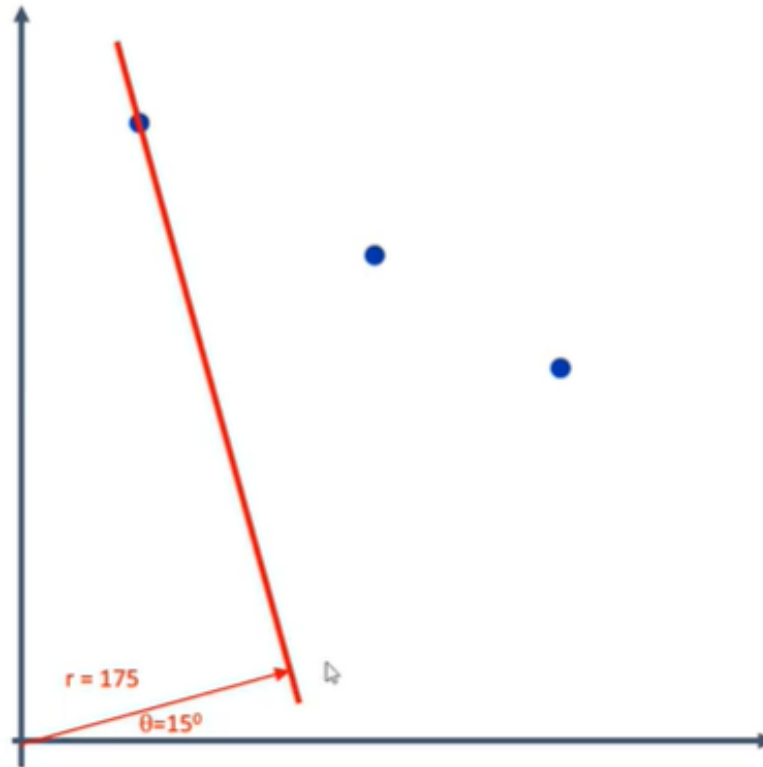
- The line can also be represented as:

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

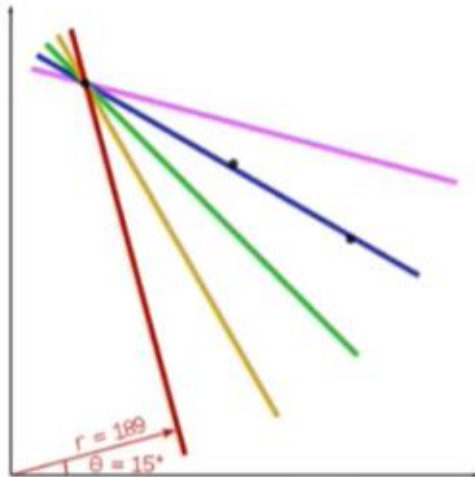
- Where r is the distance from the origin to the closest point on the straight line
- (r, θ) corresponds to the Hough space representation of a line



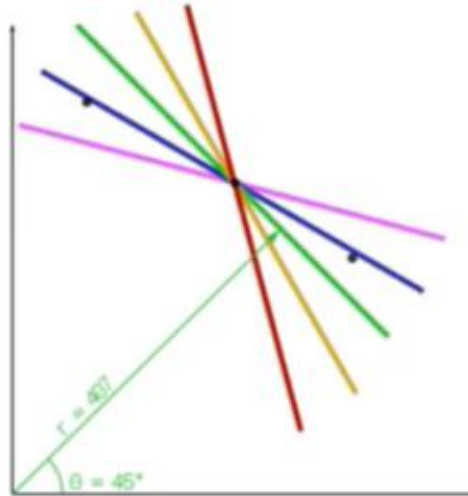
Hough Transform



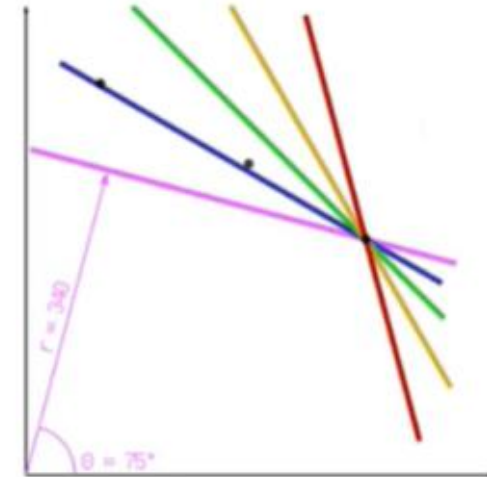
Hough Transform



θ	r
15	189.0
30	282.0
45	355.7
60	407.3
75	429.4

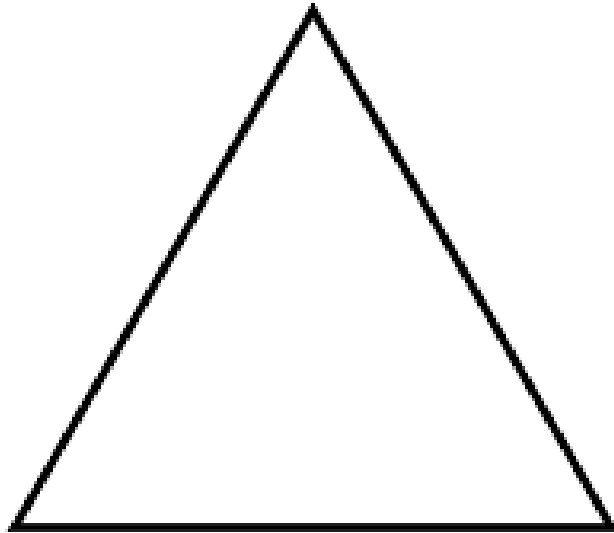


θ	r
15	318.5
30	376.8
45	407.3
60	409.8
75	385.3



θ	r
15	419.0
30	443.6
45	438.4
60	402.9
75	340.1

Hough Transform



- Using the image triangle.png apply:
 - Gray conversion
 - Gaussian blur
 - Canny to binarize and detect edges
 - Detect lines with Hough transform using `HoughLinesP()` function. This function returns the coordinates of the points

`HoughLinesP(image, r, theta, threshold)`

- Draw in the original image in blue colour the detected lines

Hough Transform

```
import cv2
import numpy as np

image = cv2.imread("triangle.png")
gray = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
blurred = cv2.GaussianBlur(gray, (5, 5), 0)
cv2.imshow("Blurred image", blurred)
edged = cv2.Canny(blurred, 50, 150)
cv2.imshow("Edges", edged)

# Detect points that form a line
max_slider = 50
lines = cv2.HoughLinesP(edged, 1, np.pi/180, max_slider)
for line in lines:
    x1, y1, x2, y2 = line[0]
    cv2.line(image, (x1, y1), (x2, y2), (255, 0, 0), 1)
# Show result
cv2.imshow("Result Image", image)
cv2.waitKey(0)
```

Hough Transform: Use case



Morphological Operations

- Simple operations applied to binary or grayscale images
- Morphological operations are applied to shapes and structures inside of images
 - Increase the size of objects
 - Decrease the size of objects
 - Close gaps or open

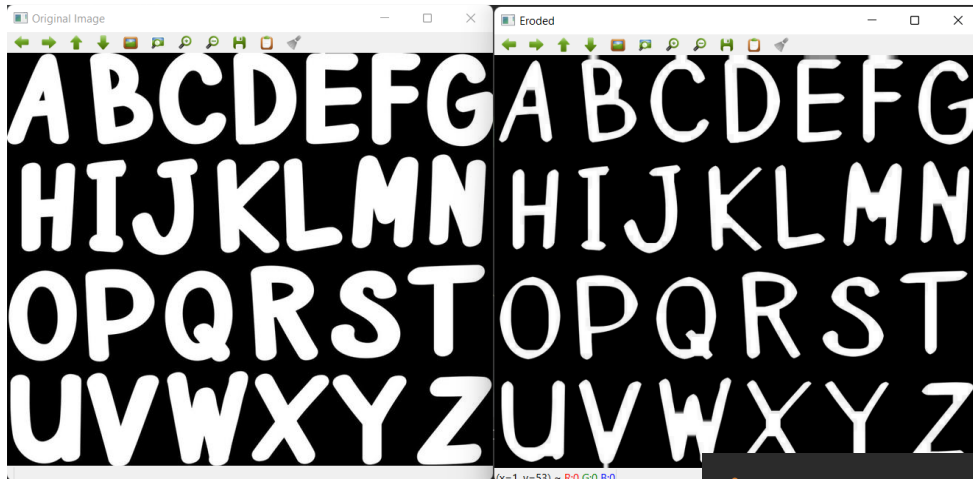
Morphological Operations

- Erosion
- Dilatation
- Opening
- Closing

Morphological Operations: Erosion

- Import the necessary packages
- Read the image
- Binarize the image.
- As it is advised to keep the foreground in white, invert operation on the binarized image to make the foreground as white.
- We are defining a 5×5 kernel filled with ones
- Then we can make use of Opencv `erode()` function to erode the boundaries of the image.
- All the pixels near boundary will be discarded depending upon the size of kernel.

Morphological Operations: Erosion



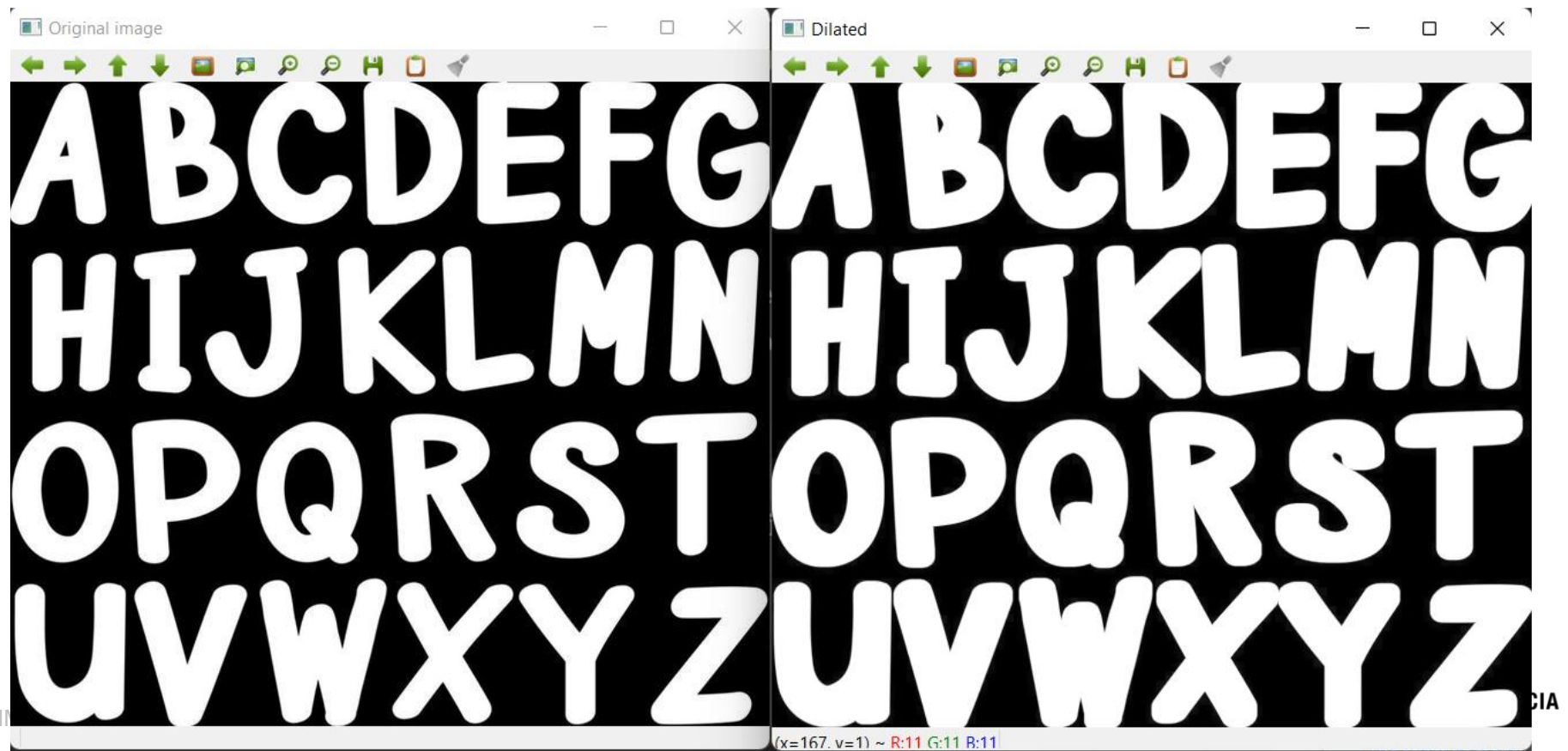
```
import cv2
import numpy as np

image = cv2.imread("morpho.jpg")
invert = cv2.bitwise_not(image)
# define the kernel
kernel = np.ones((3, 3), np.uint8)

erosion = cv2.erode(invert.copy(), kernel,
iterations=1)
cv2.imshow("Eroded", erosion)
cv2.waitKey(0)
```

Morphological Operations: Dilatation

```
dilatation = cv2.dilate(invert.copy(), kernel, iterations=1)
```



Morphological Operations:opening

- Erosion followed by dilatation
- Useful to remove noise

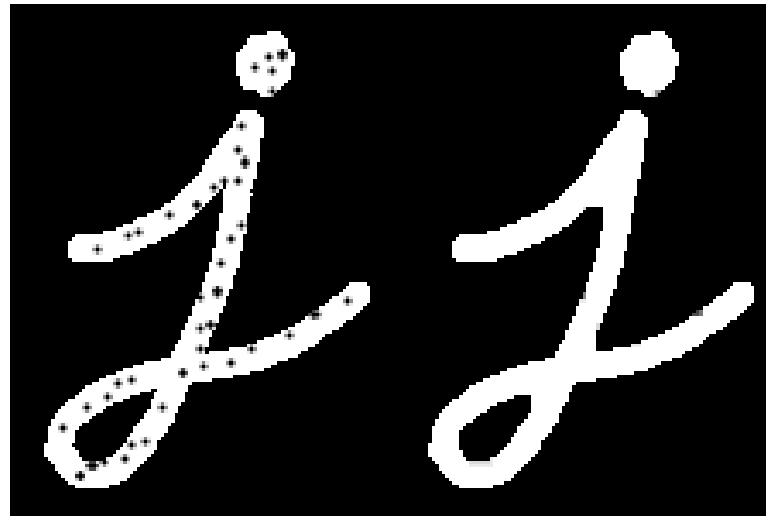
```
opening = cv2.morphologyEx(invert.copy(), cv2.MORPH_OPEN, kernel)
```



Morphological Operations:closing

- Dilatation followed by erosion
- Useful in closing small holes inside the foreground objects, or small black points on the object.

```
closing = cv2.morphologyEx(invert.copy(), cv2.MORPH_CLOSE, kernel)
```



Morphological Operations

- Morphological gradient: difference between dilatation and erosion

```
gradient = cv2.morphologyEx(invert.copy(), cv2.MORPH_GRADIENT, kernel)
```

- Top Hat: difference between the input image and the opening

```
tophat = cv2.morphologyEx(invert.copy(), cv2.MORPH_TOPHAT, kernel)
```

- Black hat: difference between the input image and the closing

```
blackhat = cv2.morphologyEx(invert.copy(), cv2.MORPH_BLACKHAT, kernel)
```

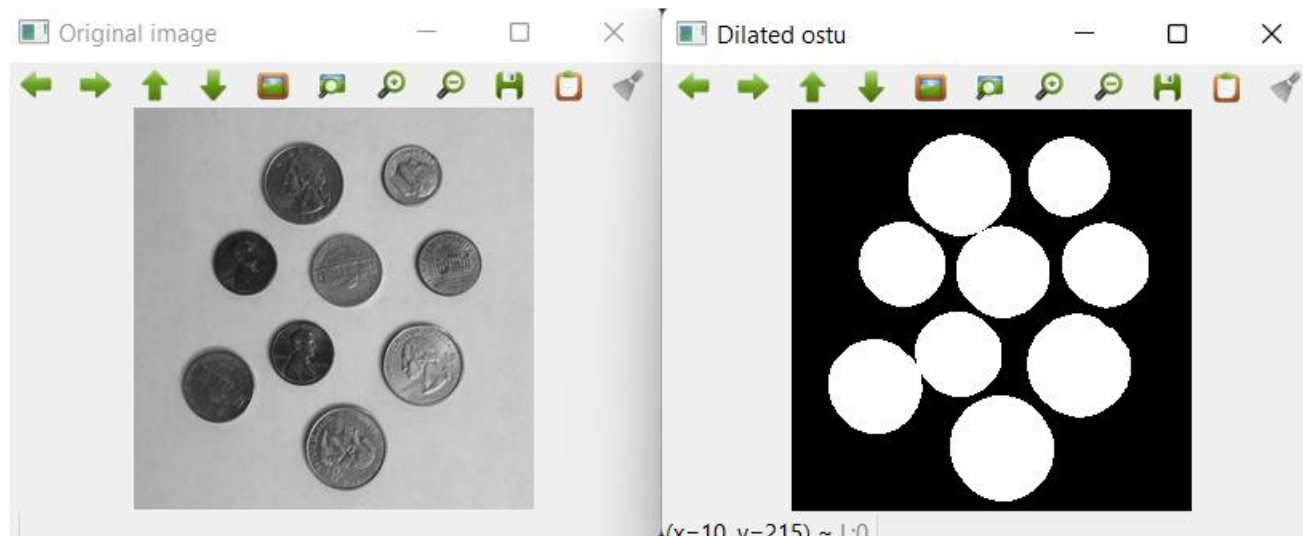
Morphological Operations: Structuring Element

- The previous examples use rectangular kernels with the help of numpy
- Some use cases may need elliptical/circular shaped kernels.
- OpenCV has a function, `cv.getStructuringElement()`. You just pass the shape and size of the kernel, you get the desired kernel.

```
kernel = cv2.getStructuringElement(cv2.MORPH_ELLIPSE, (13,13))
```

Morphological Operations: Structuring Element

- Increase the coins using morphological operations with an ellipse kernel



Morphological Operations: Use Case



Let's play with images!





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