# CV Practice Class Lecture 3

2017-07-05

FIRA 인공지능 에이전트 과정 SNUVL Lab

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# **Update Practice Git Repo.**

- Git pull
  - At the <u>fira-2017-cv</u> root folder, get the last commits via <u>git pull</u>

```
$ git pull
```

In case you get error such as

```
error: cannot pull with rebase: You have unstaged changes error: please commit or stash them.
```

First, you should enter your name and email address(for the first time),

```
$ git config --global user.name "YOUR NAME"
$ git config --global user.email "YOUR@EMAIL.com"
```

Then you should <u>stash</u> the changes, <u>pull</u> last commits, and <u>stash pop</u>.

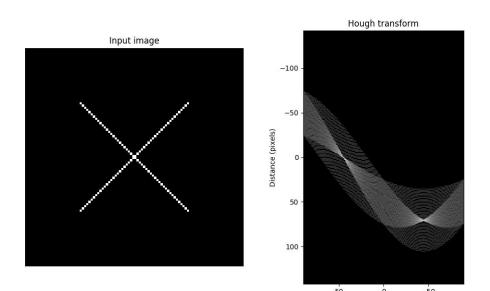
```
$ git stash
$ git pull
$ git stash pop
```

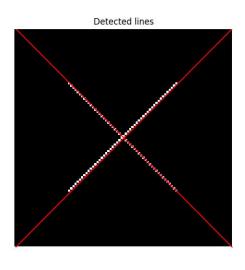
Introduction
Hough Line Transform
Hough Circle Transform

#### Hough Transform

- A feature extraction technique used to find imperfect instances of objects.
- Transformation? It convert an image to some parameter space by a voting process, and find objects by finding maxima in that sapce.
- What parameters? Any parameters that can represent a certain class of shapes(of interest; lines, circles, ...)
- How? For each pixels, vote for the parameters of which the pixel can be part
  of the shapes.

Angles (degrees)



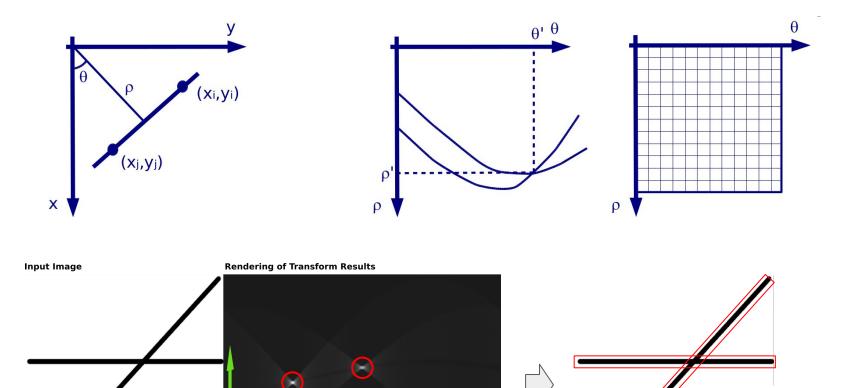


Hough Line Transform

Distance from Centre

Angle

 $\circ$  Voting process for Hough line transform( $\varrho$ (rho) and  $\theta$ (theta))



Hough line transform

#### Inputs

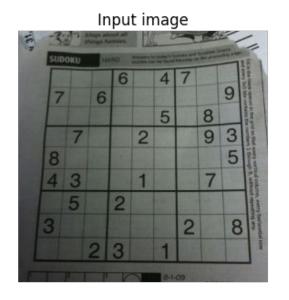
- image: A 8-bit, single-channel input image. Usually the output of edge detection to find lines as edges
- $\circ$  **rho**: The histogram bin size in rho( $\rho$ ) dimension
- $\circ$  theta: The histogram bin size in theta( $\theta$ ) dimension
- threshold: The threshold of accumulator for detecting a line
- lines: The output lines, if you have allocated beforehand.
- o srn: A divisor for the rho, for the multi-scale Hough transform
- stn: A divisor for the theta, for the multi-scale Hough transform
- min\_theta: Minimum angle to check for lines
- max\_theta: Maximum angle to check for lines

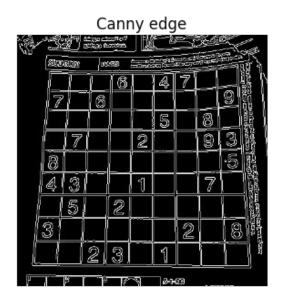
#### Output

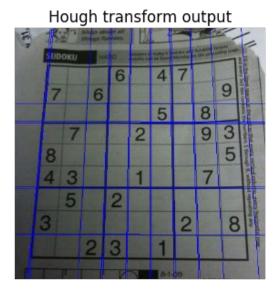
lines: Array of line parameters(rho, theta)

Hough line transform

```
img = cv2.imread('sudoku.jpg', cv2.IMREAD_COLOR)
canny_edges = cv2.Canny(img, 50, 100)
lines = cv2.HoughLines(canny_edges, 1, np.pi/180, 200)
```







 Note that the lines are drawn via cv2.line() function given the parameters of lines(rho, theta).

Hough circle transform

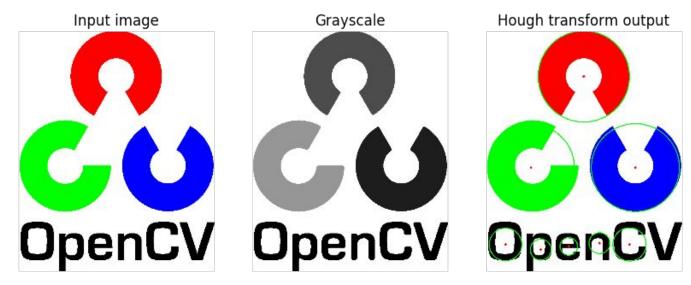
#### Input

- image: A 8-bit, single-channel input image
- method: The method to perform Hough circle transform, Currently, only cv2.HOUGH\_GRADIENT is available.
- o dp: Inverse ratio of the accumulator resolution to the image resolution.
- o minDist: Minimum distance between the centers of circles.
- o circles: The output circle, if you have allocated beforehand.
- param1: First parameter. If method is cv2.HOUGH\_GRADIENT, it is the higher threshold for canny edge detector(the lower one is twice smaller).
- o param2: Second parameter. If method is cv2.HOUGH\_GRADIENT, it is the accumulator threshold for the circle centers at the detection stage.
- o minRadius: Minimum circle radius / maxRadius: Maximum circle radius

#### Output

circles: Array of circle parameters(center\_x, center\_y, radius)

Hough circle transform



- Note that since the only method used is Hough Gradient Method, so the canny edge detection is performed inside cv2. Hough Circles() function.
- It is preferred to blur the input image before to reduce the noise.

- Practice: Coin Removal
  - Given image of coins on a table, detect the coins and remove them
  - cv2.HoughCircles() for detecting circles
  - o cv2.circle() for filling the circle with the background color





# Let's Check the Code 1\_hough\_transform.ipynb

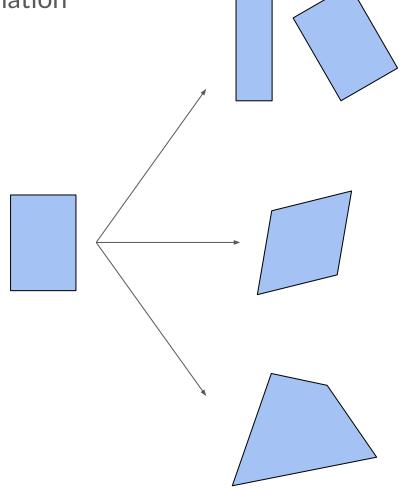
Introduction
Applying Transformation
Getting Transformation Matrix

Similarity Perspective Transformation

Scale / Translation / Rotation

Affine Transformation

- Perspective Transformation
  - Homograpy



Scale

$$\left(\begin{array}{c} x' \\ y' \end{array}\right) = \left(\begin{array}{cc} S_x & 0 & 0 \\ 0 & S_y & 1 \end{array}\right) \left(\begin{array}{c} x \\ y \\ 1 \end{array}\right)$$

Translation

$$\left(\begin{array}{c} x' \\ y' \end{array}\right) = \left(\begin{array}{ccc} 1 & 0 & T_x \\ 0 & 1 & T_y \end{array}\right) \left(\begin{array}{c} x \\ y \\ 1 \end{array}\right)$$

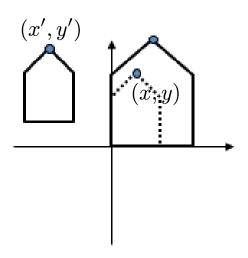


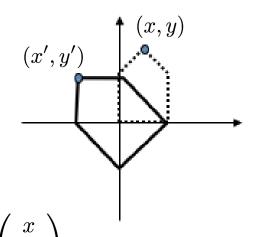
With no scale, translation

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$$

With scale, translation(rotation center on (Cx, Cy))

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} S\cos\theta & -S\sin\theta & C_x(1-\cos\theta) + C_y\sin\theta \\ S\sin\theta & S\cos\theta & C_x\sin\theta + C_y(1-\cos\theta) \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$$

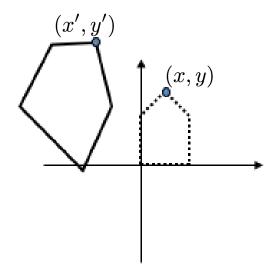




Affine Transformation

$$\left(\begin{array}{c} x' \\ y' \end{array}\right) = \left(\begin{array}{ccc} a & b & T_x \\ c & d & T_y \end{array}\right) \left(\begin{array}{c} x \\ y \\ 1 \end{array}\right)$$

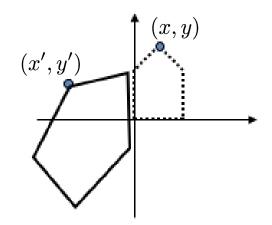
6 free parameters



Perspective Transformation

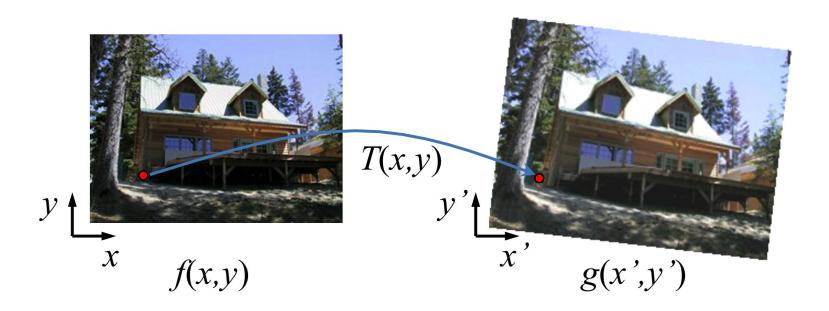
$$\begin{pmatrix} u' \\ v' \\ w' \end{pmatrix} = \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} \quad x' = u'/w' \\ y' = v'/w'$$

8 free parameters



# **Image Warping**

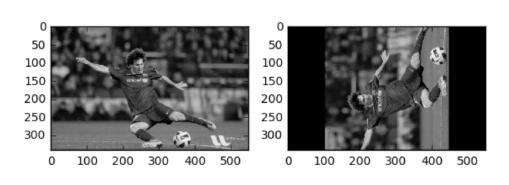
- Image Warping
  - Given a coordinate transform T and a source image f(x,y), finding a transformed image which satisfies g(x',y') = f(T(x,y)).

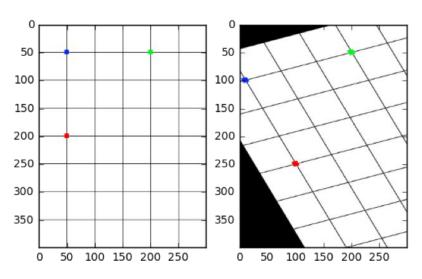


Apply Affine Transformation

```
pts2 = cv2.transform(pts1, M, ...) # points
dst = cv2.warpAffine(src, M, dsize, dst, flags=cv2.CV_INTER_LINEAR+cv2
.CV_WARP_FILL_OUTLIERS, borderMode=cv2.BORDER_CONSTANT, borderValue=0)
# image
```

- M: Affine matrix(2×3)
- flags: Interpolation methods and optional flag cv2.WARP\_INVERSE\_MAP
- borderMode: Pixel extrapolation method



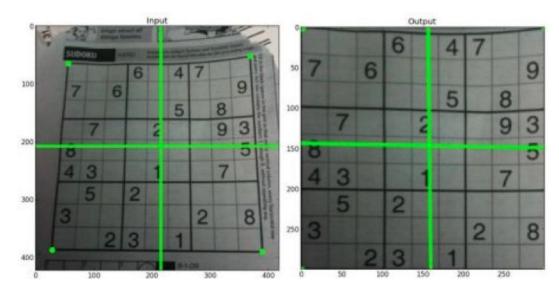


Apply Perspective Transformation

```
pts2 = cv2.perspectiveTransform(pts1, M) # points

dst = cv2.warpPerspective(src, M, dsize, dst, flags=cv2.CV_INTER_LINEA
R+cv2.CV_WARP_FILL_OUTLIERS, borderMode=cv2.BORDER_CONSTANT, borderVal
ue=0) # image
```

- M: Homography matrix(3×3 with M[3,3]==1)
- flags: Interpolation methods and optional flag cv2.WARP\_INVERSE\_MAP
- borderMode: Pixel extrapolation method



- Apply Perspective Transformation
  - borderMode: Pixel extrapolation method

Original Image



borderMode= cv2.BORDER\_WRAP



borderMode= cv2.BORDER CONSTANT



borderMode= cv2.BORDER\_REFLECT



Get Transformation Matrix

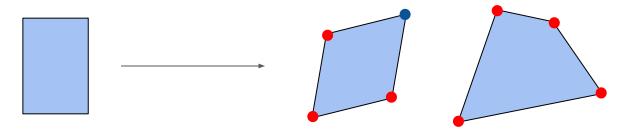
#### M = cv2.getRotationMatrix2D(center, angle, scale)

- center: Center of the rotation in the source image.
- angle: Rotation angle in degrees.
- scale: Isotropic scale factor.

```
M = cv2.getAffineTransform(src, dst)
M = cv2.getPerspectiveTransform(src, dst)
```

- src: Coordinates of triangle/quadrangle vertices in the source image.
- dst: Coordinates of the corresponding triangle/quadrangle vertices in the destination image.

(src, dst are  $N\times 2(N: 3 \text{ for affine or } 4 \text{ for perspectice})$  NumPy matrices)



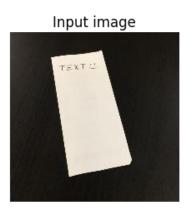
Practice #1: Track Advertisement



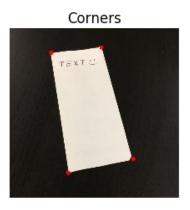




Practice #2: Cam Scanner









# Let's Check the Code 2\_geometric\_transform.ipynb