

영상 데이터의 처리와 정보추출

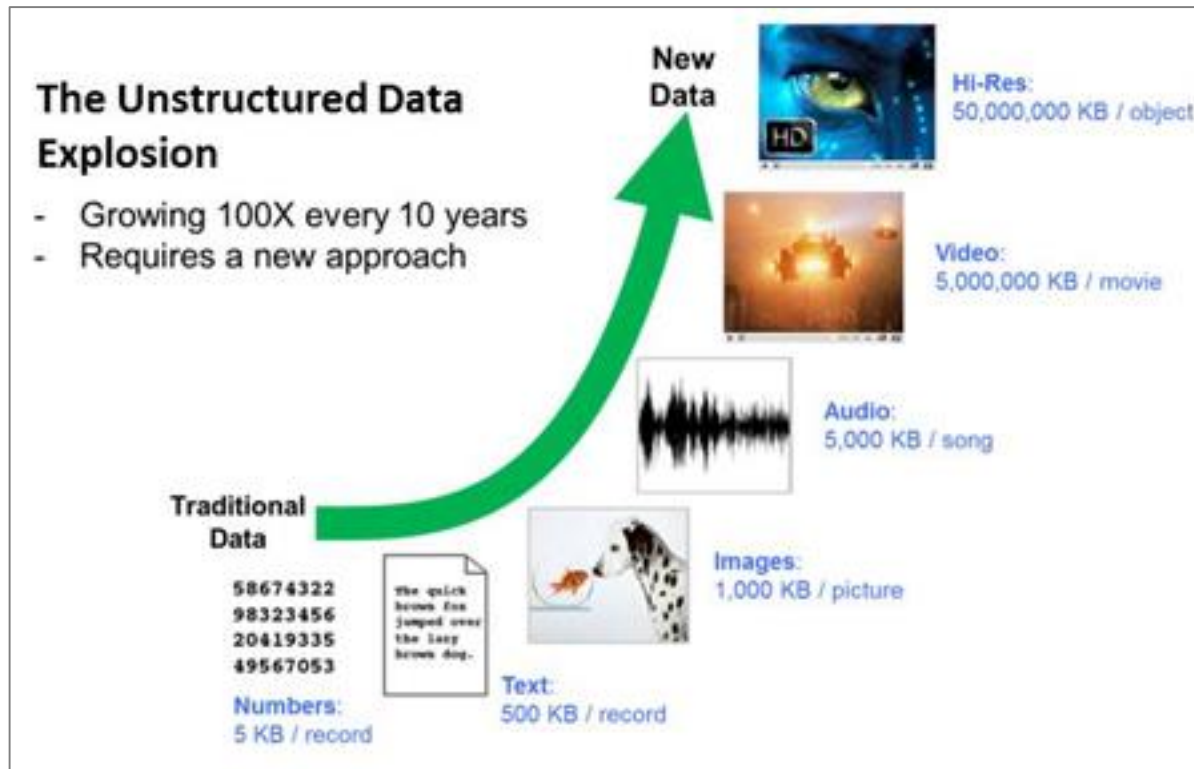
이동윤

LG CNS

Advanced Analytics Center

빅데이터에서의 영상 데이터

- 영상 데이터를 정형 데이터처럼 다룰 필요성 ↑



강연자 소개

- 컴퓨터비전, 패턴인식 전공 (석사)

- 산업계 경력 11년차
 - 삼성전자 반도체연구소
 - 컴퓨터비전, 빅 데이터 (CAD)
 - LG CNS 고급분석 센터
 - 컴퓨터비전, 고급분석/예측



※ iPad 1, iPhone 5GS 두 모델의 개발, 제조에 참여했으며
본 강의의 내용+a가 그에 적용되었음

- 페이스북 컴퓨터비전/
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www.facebook.com/groups/cvprml
(> 2,200명)

영상 데이터로부터의 정보추출

- 컴퓨터비전

※ 영상으로부터 '정보'를 추출하는 것을 목적으로 삼아 온 공학/과학 분과
("영상만을 보고 이해하는 컴퓨터의 실현")



어떤 종류의 장면인가?

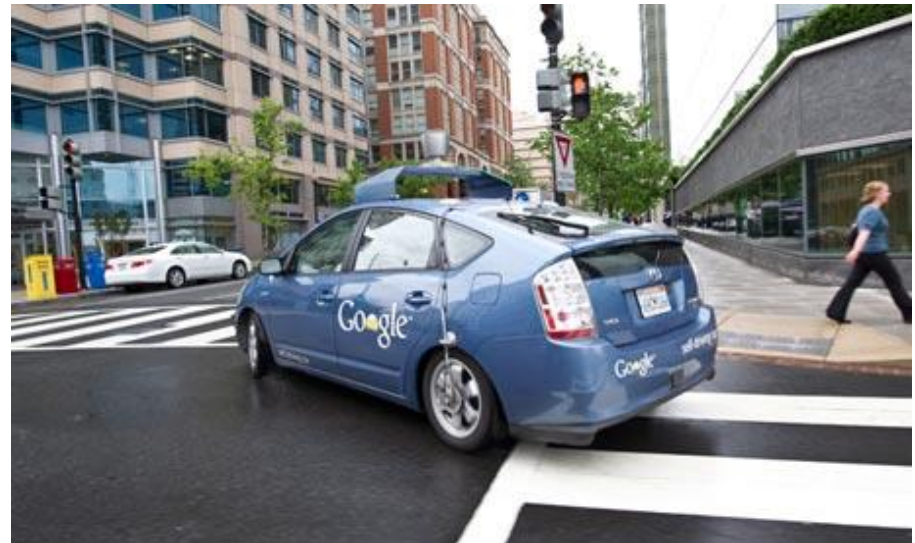
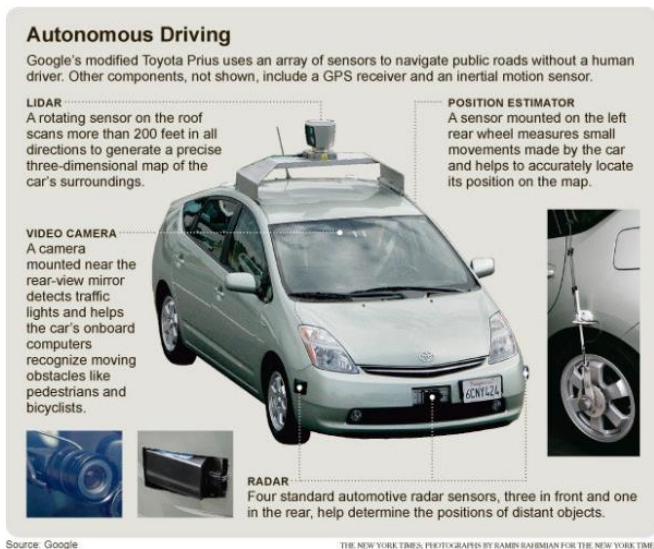
자동차가 어디에 있는가?

건물이 얼마나 멀리 있는가?

:

컴퓨터비전 응용사례

• 구글 자동운전 자동차



※ 비디오 카메라 뿐 아니라 레이저를 이용한 LIDAR, 전파를 이용하는 RADAR에도 본 자료 내의 기술이 적용됨

이 강연에서...

- Data Type
 - Image (O)
 - Video (X)

NOT

- GPU(CUDA), 3D, 얼굴인식, 머신러닝 관련기술, 주파수 영역분석

- Language
 - Python (2.7.X)



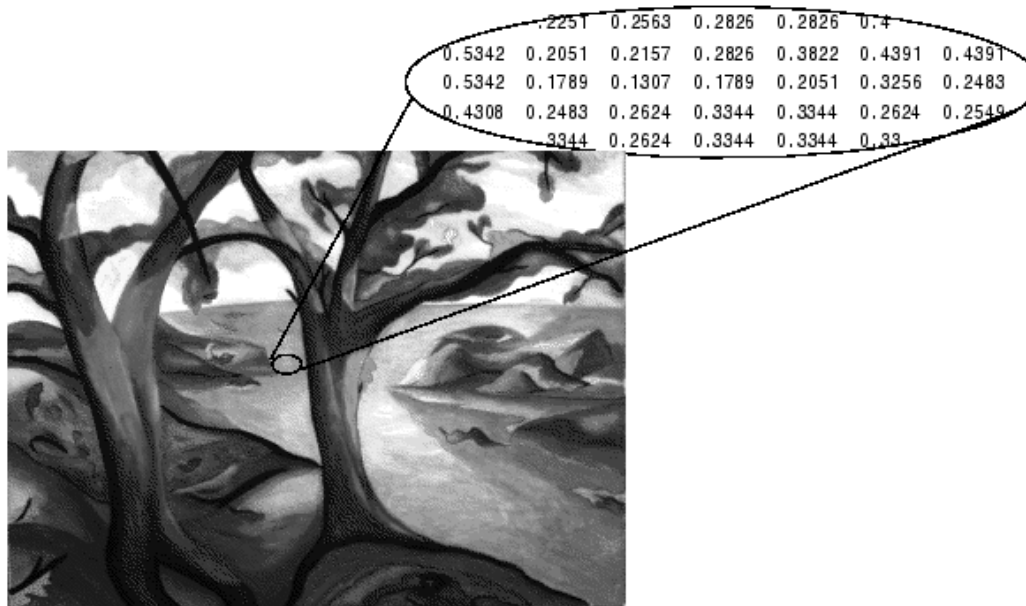
- Library (패키지)
 - scikit-image



※ 컴퓨터비전 기술의 시도와 활용을
극히 쉽게 만들어줌

영상 인지의 특성

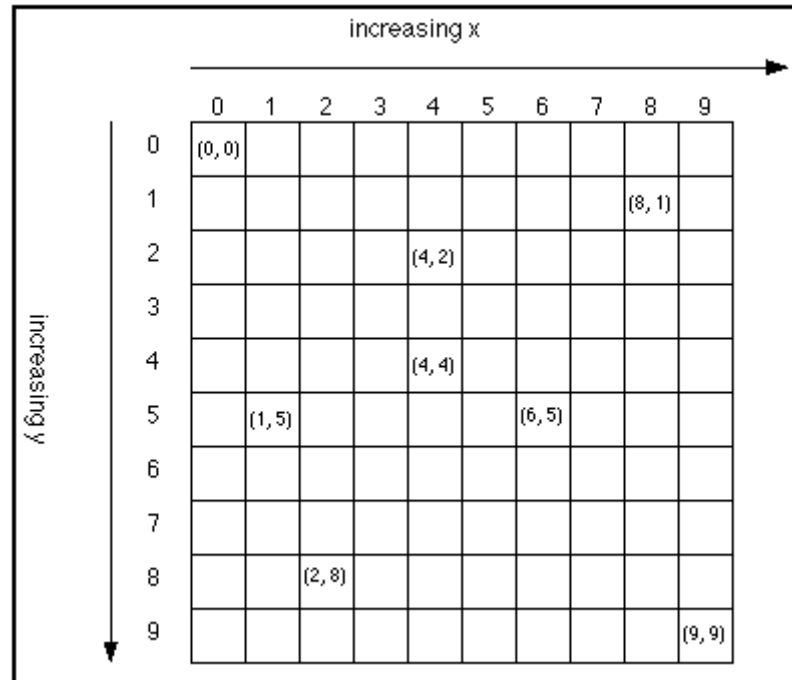
- Digitizing



※ 가상의 격자. 이 격자로 인해 원하지 않게 문제점이 발생하기도 함

영상 인지의 특성

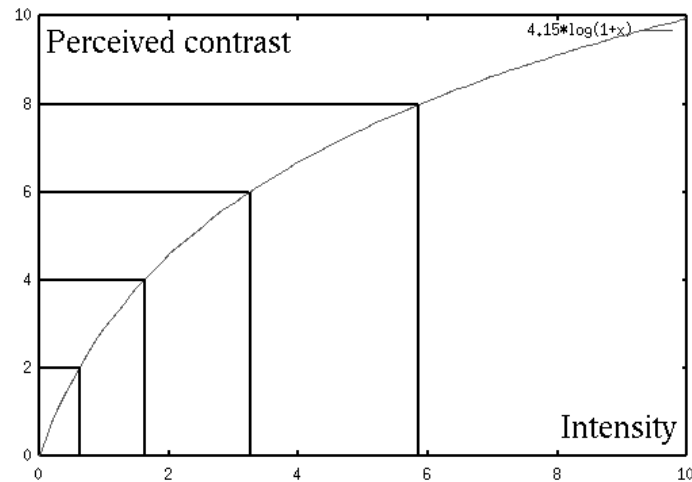
- 영상좌표계



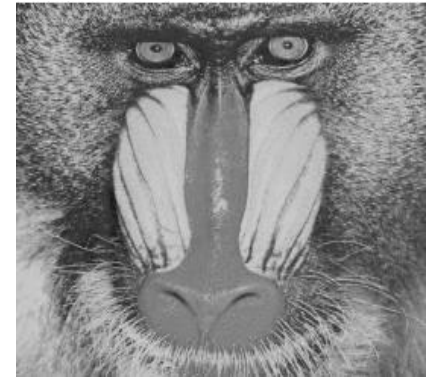
1. 좌측상단이 원점
2. 좌표계 관례 확인 필요

영상 인지의 특성

- 256 단계 픽셀



Weber-Fechner law



“한 자극의 다음 자극은 훨씬 더 커야 한다” (지수적 증가)
모든 종류의 감각자극이 이 법칙을 따름

영상 파일 포맷

- 주요 영상 파일 포맷 설명

| | 데이터의 손실 | 데이터 용량 | 인지적 우수성 | 특성 |
|------|---------|-------------|---------|---------------------------------------|
| BMP | 無 | 큼 (용량고정) | N/A | 색 재현성 우수, Windows OS용 |
| TIFF | | | | 투명도 고려 가능, <u>가장 선호됨</u> |
| PNG | | | | 위 둘보다 헤더 구조가 단순 |
| GIF | 有 | 작음 | 낮음 | 경계선이 부드러움, 동영상처럼 만들 수 있음(:모션 GIF) |
| JPG | | 매우 작음 | 매우 우수 | 데이터 용량이 가장 작아 <u>네트워 크에서의 전송에 유리함</u> |

차 례

- Basic Processing
- Region Processing
- Edge/Contour
- Point/Line/Corner
- Matching
- Preprocessing
- Color

※ 내용이 어려워지는 순서

Basic Processing

"이것만 하면 다 할 수 있다."
(단, 전문가라면...)

Load/Save and Display

- PIL (Python Image Library)

- Import Module

```
>> from PIL import Image  
>> from pylab import *
```

- Load

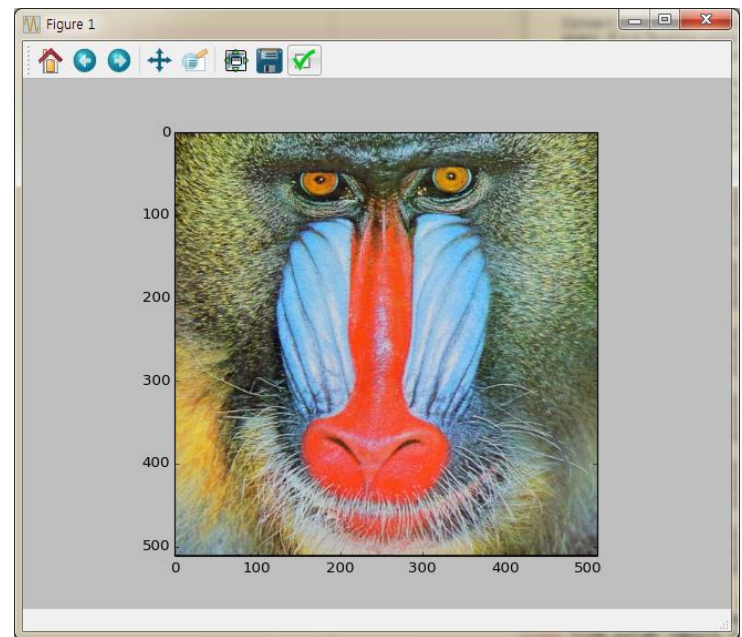
```
>> img = array(Image.open('C:\Wmandrill.bmp'))
```

- Display

```
>> imshow(img)
```

- Save

```
>> Img = Image.fromarray(img)  
>> Img.save('C:\Wmandrill_new.bmp')
```



코드 안내

- 그래픽 출력용 모듈인 matplotlib와 그 서브모듈인 pylab은 임포트를 생략함
- numpy의 경우 생략
- 각 장에서, 이미 앞 부분에서 설명된 처리를 위한 모듈의 경우 임포트가 생략되었을 수 있음
- 코드 내용의 자료 기입 이후의 수정으로 일부 비작동 사례 있을 수 있음 (발견 시, neuralix@gmail.com 로 inform 요망)

Load/Save and Display

- skimage

```
>> from skimage import io
```

```
>> camera = io.imread( filename )
```

※ URL의 입력도 가능

```
>> imshow( img )
```

```
>> io.imsave( filename, logo )
```

- opencv

```
>> import numpy as np  
>> import cv2
```

```
>> img = cv2.imread( filename, 0 )
```

```
>> cv2.imshow('image', img)
```

```
>> cv2.imwrite( filename, img)
```

Pixel Operation

- Get pixel value

```
>> print img[ i ][ j ][ k ]
```

- Put Pixel Value

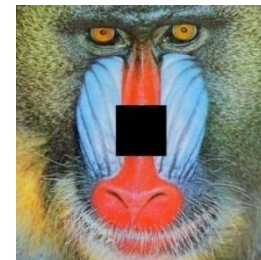
```
>> img[ i ][ j ][ k ] = 0
```

- Area Operation

```
for i in range(img.shape[0]):  
    for j in range(img.shape[1]):  
        for k in range(img.shape[2]):  
            if i>=200 and i<300:  
                if j>=200 and j<300:  
                    img[i][j][k] = 0 * img[i][j][k]
```

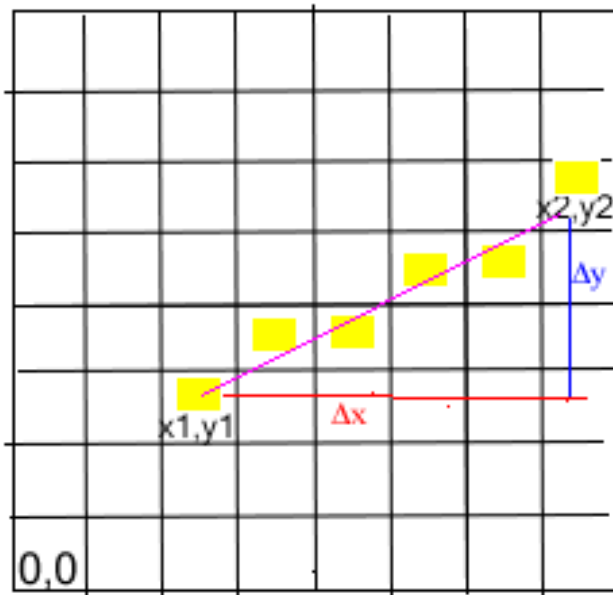
or Simply,

```
>> img[200:300][200:300] =  $\mathbb{W}$   
    0 * img[200:300][200:300]
```



Line Drawing

- DDA (Digital Differential Analyzer)



※ 가장 단순한 선긋기 방법

```
1 ▾ DDA(float x1, float y1 float x2, float y2) {  
2  
3     float x_start, y_start;  
4     float xincr, yincr;  
5     int steps;  
6  
7 ▾     /* steps는 x, y중 더 큰 차이인 쪽으로 선택한다. */  
8     steps = Round(x2) - Round(x1);  
9     xincr = (x2 - x1) / steps;  
10    yincr = (y2 - y1) / steps;  
11    x_start = x1  
12    y_start = y1;  
13  
14    MarkPixel(Round(x_start),Round(y_start))  
15  
16 ▾    for (i=0; i<steps1 i++) {  
17        x = x + xincr;  
18        y = y + yincr;  
19        MarkPixel(Round(x),Round(y));  
20    }  
21 }
```

```
>> from skimage.draw import line, set_color
```

```
>> rr, cc = line(x1, y1, x2, y2)
```

```
>> set_color(img, (rr, cc), 1)
```

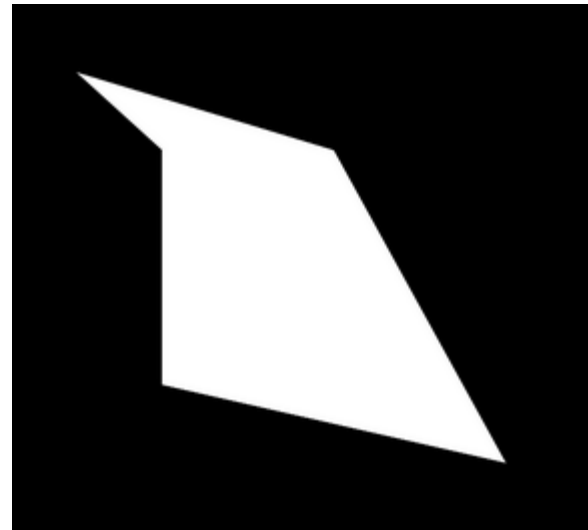
```
>> imshow(img)
```

Polygon Drawing

- 다각형 그리기

```
>> from skimage.draw import polygon
```

```
>> img = np.zeros((width, height), dtype=np.uint8)  
>> x = np.array([x1, x2, x3, x4, ...])  
>> y = np.array([y1, y2, y3, y4, ...])  
>> rr, cc = polygon(y, x)  
>> img[rr, cc] = 1
```



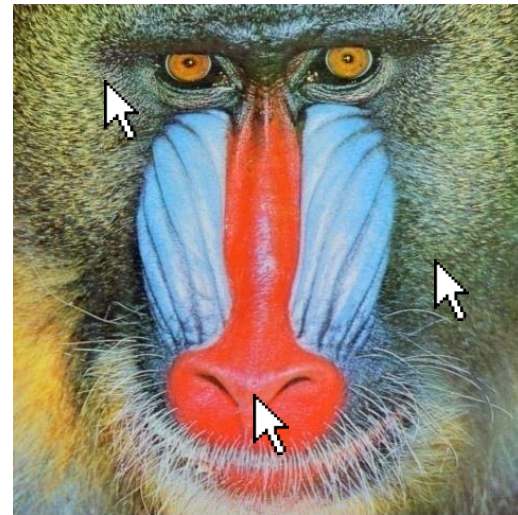
Picking Points ("ginput")

- 클릭한 지점의 좌표를 배열로 리턴

```
>> from PIL import Image
>> from pylab import *

>> img = array(Image.open('C:\mandrill.bmp'))
>> imshow(img)
>> print 'Please click 3 points'
>> x = ginput(3)
>> print 'you clicked:',x
>> show()
```

```
>> x
you clicked: [(16.737903225806448, 18.705645161
290249), (406.85887096774195, 23.2419354838709
18), (227.22177419354841, 242.79838709677415)]
```



※ 이름의 유래는 MATLAB의 함수 이름 (Graphical INPUT)에서. 워낙 많이 쓰이고 편리하여 보통명사처럼 됨.

Region Processing

※ 영역의 추출과 처리

Thresholding (Binarizing)

- Otsu's (Automatic) Thresholding

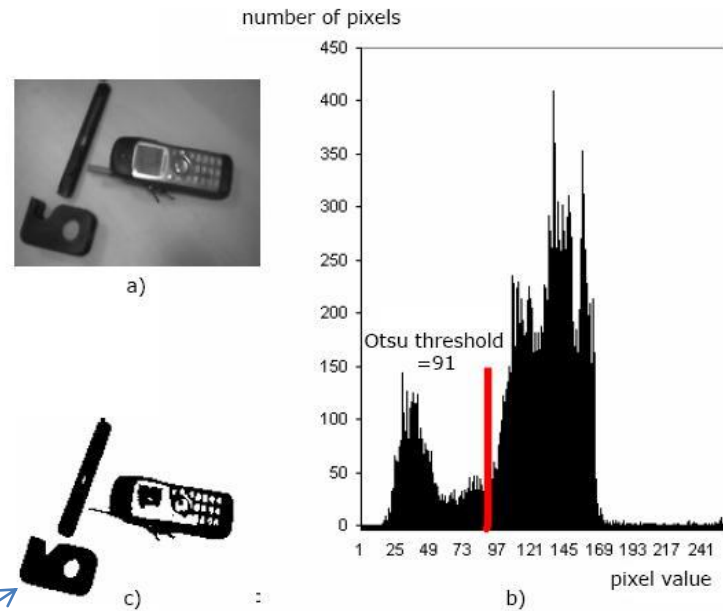
```
>> from skimage import data
>> from skimage.filter import threshold_otsu, \
    threshold_adaptive

>> from skimage import color

>> img = io.imread('C:\Mandrill.bmp')
>> img_g = color.rgb2grey(img)

>> val_thresh = threshold_otsu(img_g)
>> img_bw = img_g > val_thresh

>> io.imshow( img_bw )
```



"Blob"

Labeling

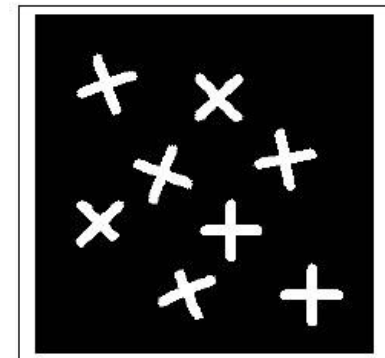
- Connected Components Labeling

```
>> from skimage.morphology import label, closing, square
>> from skimage.color import label2rgb
...
>> val_thresh = threshold_otsu(img)

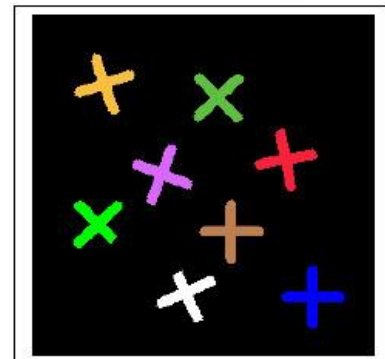
>> img_label = label(val_thresh)
>> image_label_rgb = label2rgb(label_image)
>> imshow(image_label_rgb)
```

※ Check neighborhood convention first!
(8-neighborhood is complete, 4-neighbor is not.)

Original Image



Labeled Components



Morphological Operation

- 1) 블랍을 <뚱뚱하게/날씬하게>, 2) 서로 다른 블랍을 <붙게/떨어뜨리게>, 3) <구멍을 메우게/돌출부를 평평하게>

```
>> from skimage.morphology import erosion, dilation, opening, closing, white_tophat
>> from skimage.morphology import black_tophat, skeletonize, convex_hull_image
>> from skimage.morphology import disk
...
>> selem = disk(6)

>> img_eroded = erosion(img, selem)
>> imshow(img_eroded, cmap=plt.cm.gray)

>> img_dilated = dilation(img, selem)
>> imshow(img_dilated, cmap=plt.cm.gray)
```

※ Structuring element (:selem)의 모양과 크기에 의해 조정됨

※ LIDAR/RADAR 데이터의 처리에서 중요

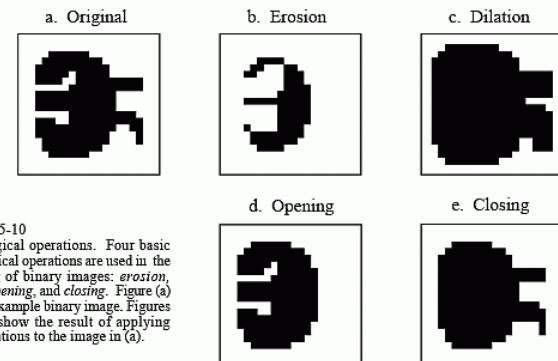
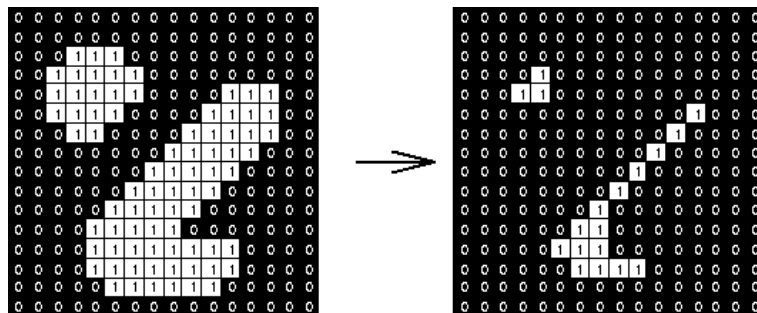
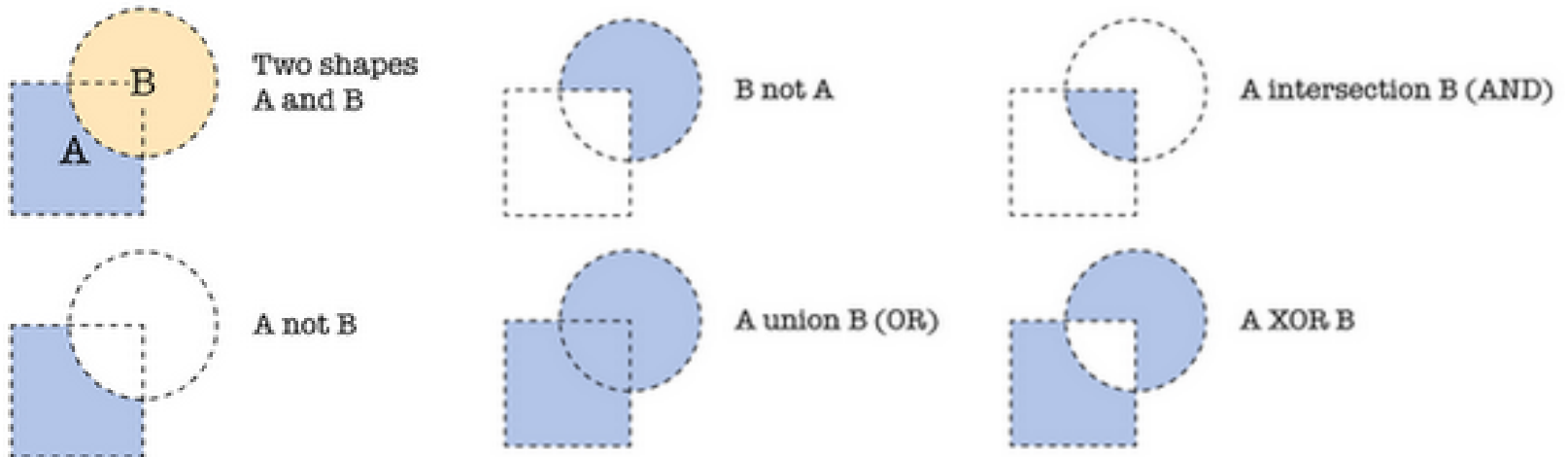


FIGURE 25-10
Morphological operations. Four basic morphological operations are used in the processing of binary images: *erosion*, *dilation*, *opening*, and *closing*. Figure (a) shows an example binary image. Figures (b) to (e) show the result of applying these operations to the image in (a).

Boolean Operation

- Image Boolean Operations

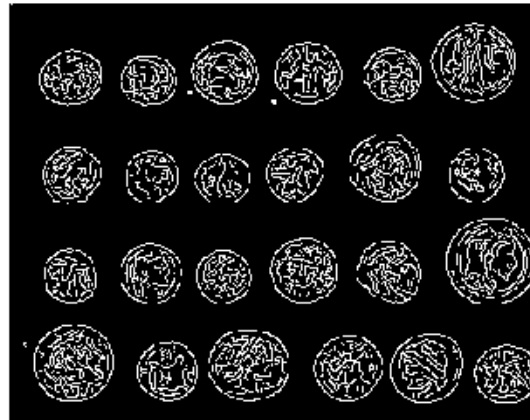


Flood Fill

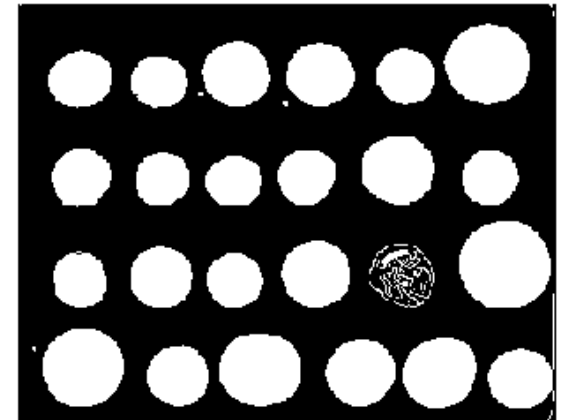
- Filling the holes



Canny detector



Filling the holes



```
>> from scipy import ndimage  
>> from skimage.filter import canny  
  
>> img_edges = canny(img/255.)  
>> img_filled = ndimage.binary_fill_holes(img_edges)
```

Segmentation

- Mean Shift

```
>> import cv2
>> import pymeanshift as pms

>> img_orig = cv2.imread("example.png")

>> (img_segmented, img_label, number_regions) = pms.segment(img_orig, spatial_radius=6,
                                                             range_radius=4.5, min_density=50)
```

※ 구획화(Segmentation)는 경계를 인지, 이용하는 방향과 내부 유사성을 인지, 이용하는 두 계열이 있음. GraphCuts, GrabCuts는 전자, Mean Shift(:Moving Averaging)은 후자, Chan's Active Contour without Edges는 둘을 모두 추구함



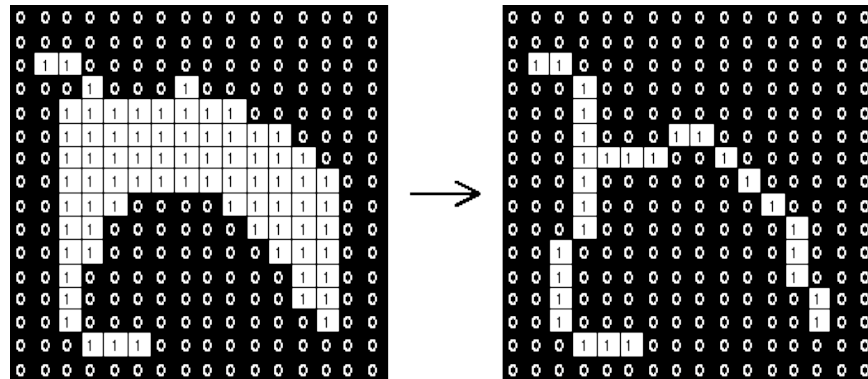
Edge / Contour

※ 임의의 선, 곡선들

Skeletonization

- Skeletonization

```
>> from skimage.morphology import skeletonize  
>> img_skeleton = skeletonize(image)  
>> imshow(img_skeleton, cmap=plt.cm.gray)
```



Contour Finding

- Contour Finding

```
>> import numpy as np
>> import matplotlib.pyplot as plt
>> from skimage import measure

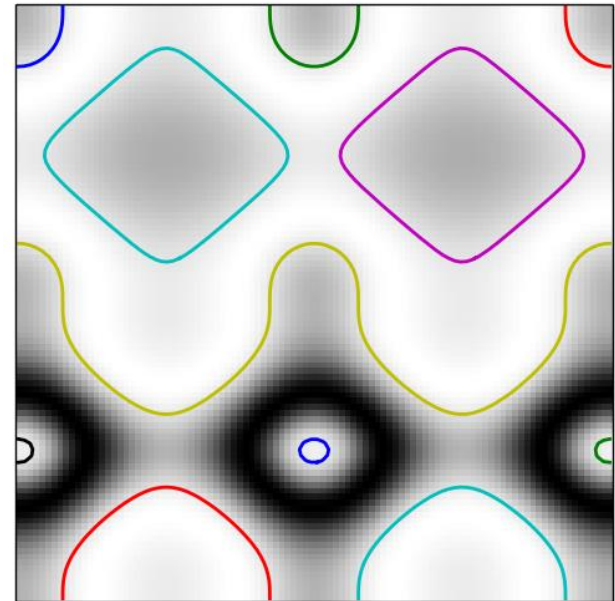
>> x, y = np.ogrid[-np.pi:np.pi:100j, -np.pi:np.pi:100j]
>> r = np.sin(np.exp((np.sin(x)**3 + np.cos(y)**2)))

>> contours = measure.find_contours(r, 0.8)

>> fig, ax = plt.subplots()
>> ax.imshow(r, interpolation='nearest', cmap=plt.cm.gray)

>> for n, contour in enumerate(contours):
>>     ax.plot(contour[:, 1], contour[:, 0], linewidth=2)

>> plt.show()
```



※ 콘투어의 각 점들을 배열로 추출해줌

Edge Detection (1 of 2)

- Canny Edge Detection

```
>> import numpy as np
>> import matplotlib.pyplot as plt
>> from scipy import ndimage

>> from skimage import filter

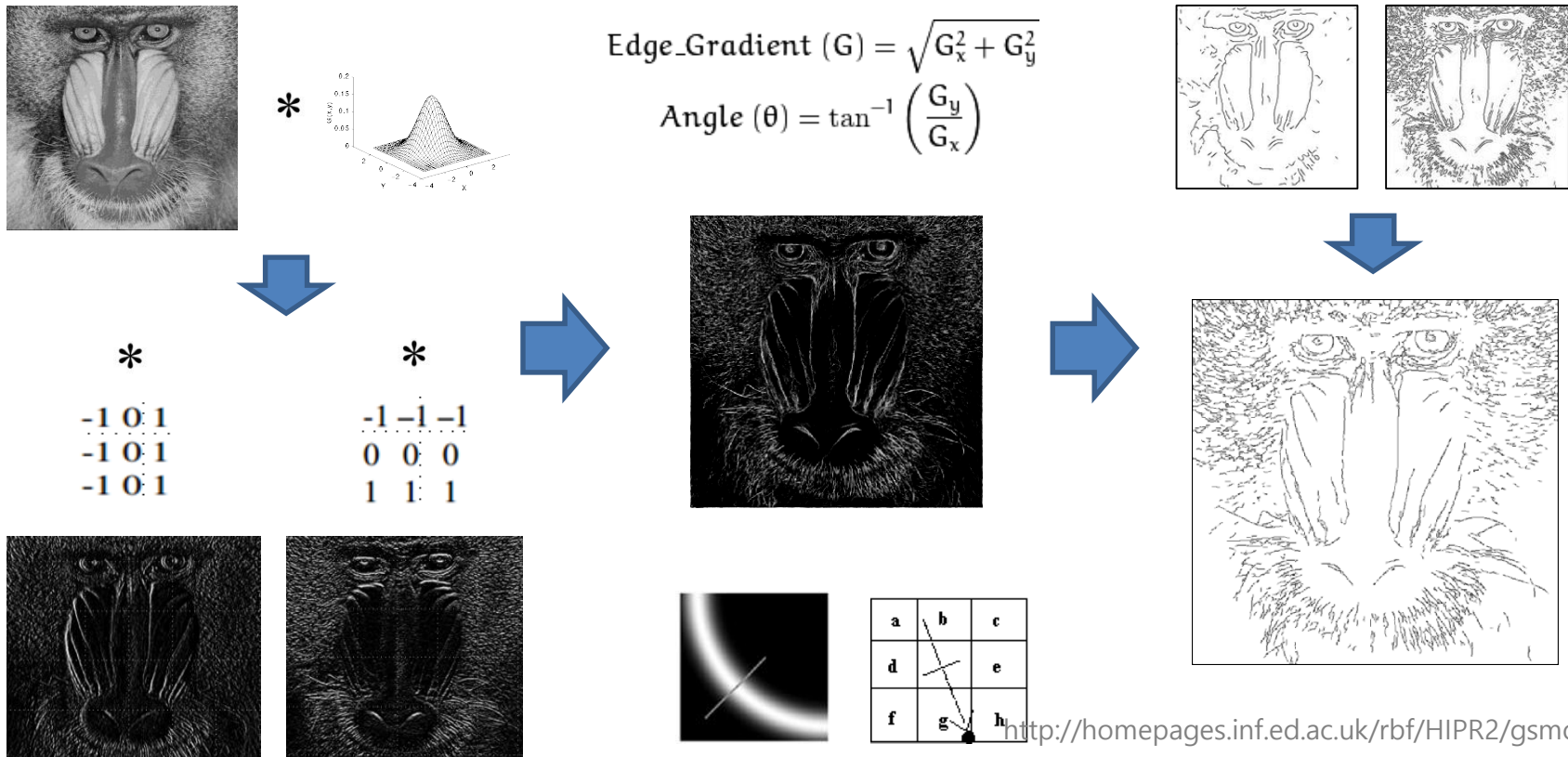
>> img_edges = filter.canny(img, sigma=3)
>> imshow(img_edges, cmap=plt.cm.gray)
```

※ 콘투어가 전체에서 동일한 값인 지점을 주는 것이라면 에지는 국소적인 조사로 최적경계를 찾으므로 대개의 경우에 더 효과적임 (1차미분과 유사함).
※ 위의 경우는 Thresholding 후 에지 검출이 가능



Edge Detection (2of2)

- Canny Edge Detection ※ 에지 검출 기법의 대명사와 같음 ("Canny")



<http://homepages.inf.ed.ac.uk/rbf/HIPR2/gsmooth.htm>

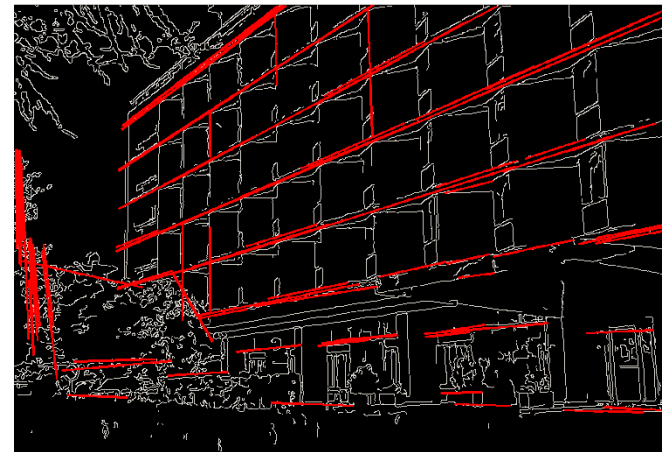
http://news.mynavi.jp/photo/series/computer_vision/040/images/014l.jpg

http://homepages.inf.ed.ac.uk/rbf/CVonline/LOCAL_COPIES/MARBLE/low/edges/canny.htm

Point / Line / Corner

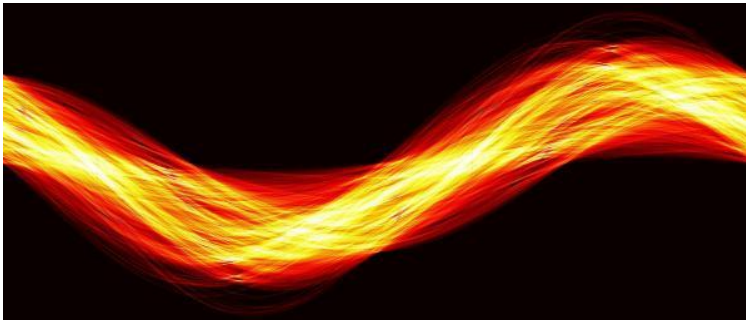
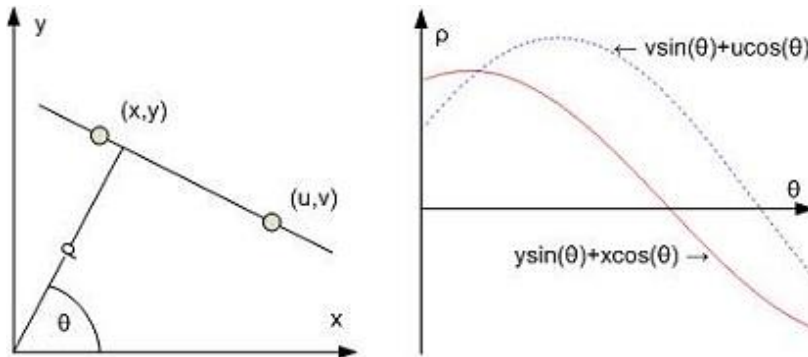
Hough Transform (1 of 2)

- Hough Transform



Hough Transform (2of2)

- Hough Transform



```
>> from skimage.transform import (hough_line, hough_line_peaks, probabilistic_hough_line)
```

```
>> edges = canny(image, 2, 1, 25)
```

```
>> lines = probabilistic_hough_line(edges, threshold=10, line_length=5, line_gap=3)
```

```
>> for line in lines:
```

```
    p0, p1 = line
```

```
    plt.plot((p0[0], p1[0]), (p0[1], p1[1]))
```

```
>> plt.show()
```

※ rho, theta 평면의 한 지점은 x, y 평면의 하나의 직선에 대응함

※ 두 점을 잇는 모든 직선을 rho, theta 평면에서 voting하여 피크지점을 추출하여 직선을 추정함

Corner Detection (1 of 2)

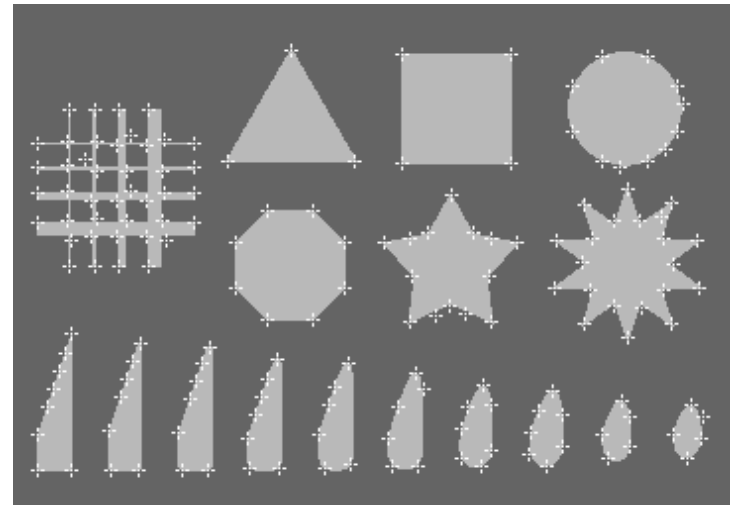
- Harris Corner Detection

```
>> from skimage.feature import corner_harris, corner_subpix, corner_peaks

>> coords = corner_peaks(corner_harris(image), min_distance=5)
>> coords_subpix = corner_subpix(image, coords, window_size=13)

>> fig, ax = plt.subplots()
>> ax.imshow(image, interpolation='nearest', cmap=plt.cm.gray)
>> ax.plot(coords[:, 1], coords[:, 0], 'b', markersize=3)
>> ax.plot(coords_subpix[:, 1], coords_subpix[:, 0], 'r', markersize=15)
>> ax.axis((0, 350, 350, 0))

>> plt.show()
```

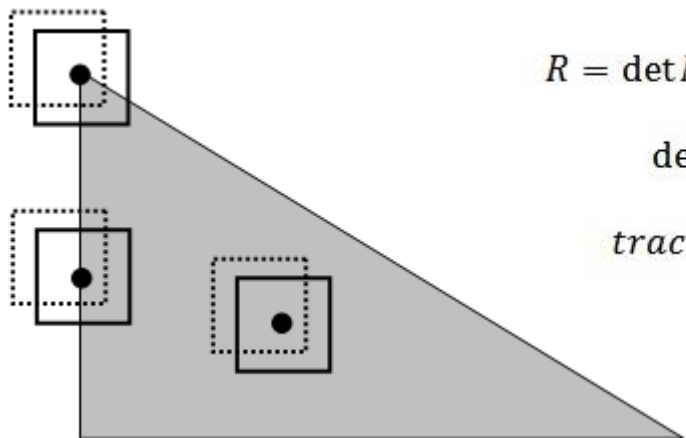


$$\mathbf{M} = \begin{bmatrix} \left(\frac{\partial f}{\partial x}\right)^2 & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial x \partial y} & \left(\frac{\partial f}{\partial y}\right)^2 \end{bmatrix}$$

$$R = \det \mathbf{M} - k \operatorname{Trace}^2 \mathbf{M}$$

Corner Detection (2of2)

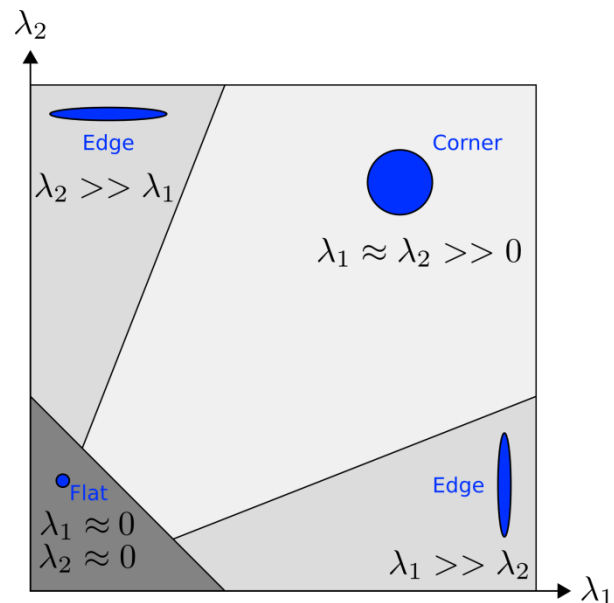
- Harris Corner Detection



$$R = \det M - k(\text{trace } M)^2$$

$$\det M = \lambda_1 \lambda_2$$

$$\text{trace } M = \lambda_1 + \lambda_2$$



※ 사각형 영역을 조사해보아 주성분분석을 하였을 때, 최대값인 아이겐밸류 λ_1, λ_2 가 모두 0보다 꽤 크고, 서로간에 값이 비슷하면 Corner라 판단해도 좋다.

Matching / Registration

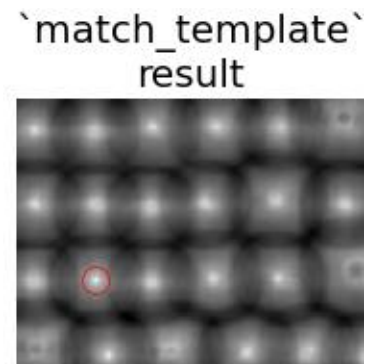
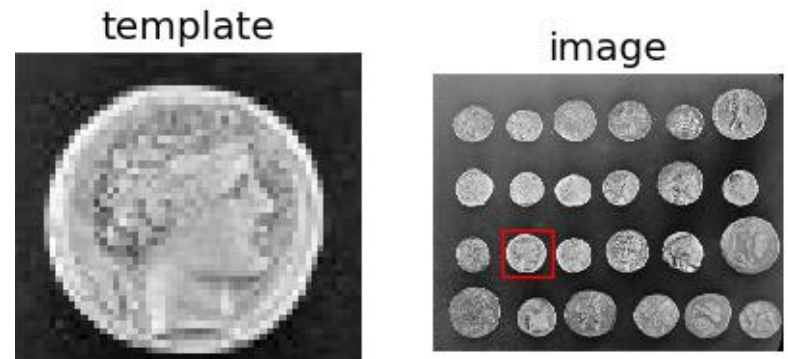
Image Matching

- Template Matching

```
>>import numpy as np
>>import matplotlib.pyplot as plt

>>from skimage import data
>>from skimage.feature import match_template
```

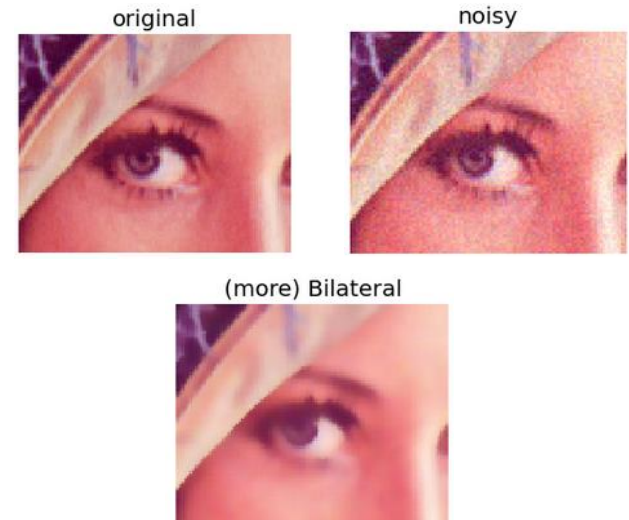
```
>>result = match_template( img_whole, img_template )
```



Noise Suppression

- Denoising

```
>> from skimage import data, img_as_float
>> from skimage.restoration import denoise_tv_ch
    ambolle, denoise_bilateral
...
>> imshow(denoise_bilateral(img, sigma_range=0.1,
    sigma_spatial=15))
```



- Median

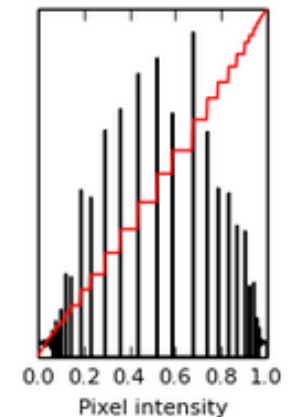
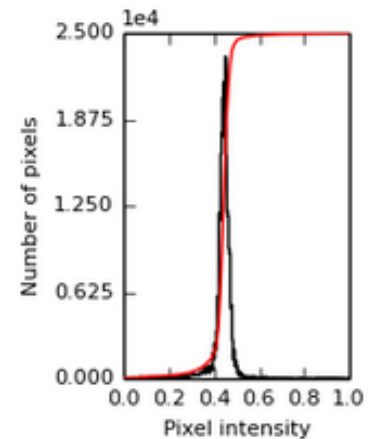
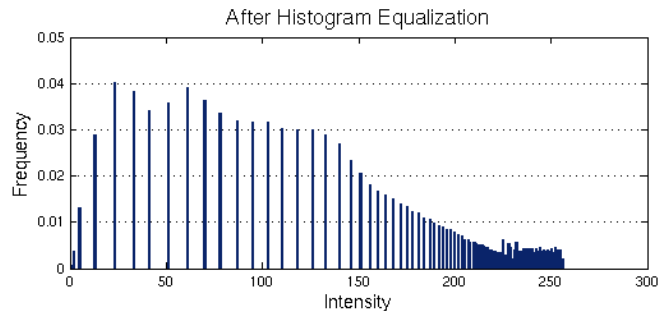
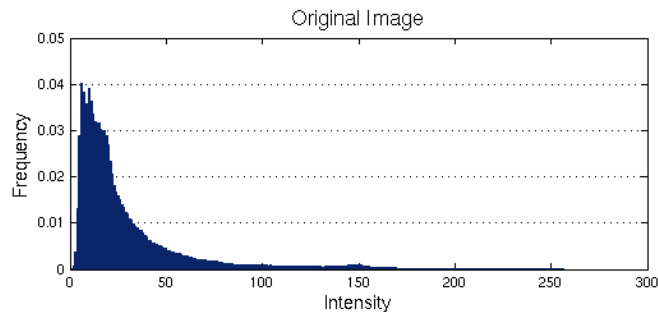
```
>> from skimage.morphology import disk
>> from skimage.filter.rank import median
...
>> img_med = median(img, disk(5))
```



Contrast

- Histogram Equalizing

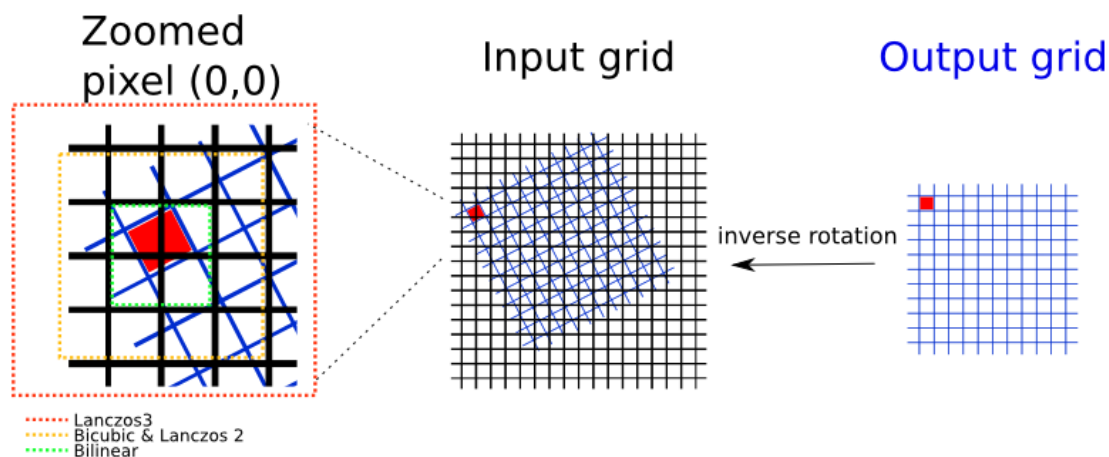
```
>> from skimage import exposure  
...  
>> img_eq = exposure.equalize_hist(img)
```



※ 히스토그램 변환규칙을 만들고 그 히스토그램을 참고해 모든 픽셀의 값을 변경함. 여러 기법 중 히스토그램 이퀄라이징은 CDF가 강제로 직선이도록 변환규칙을 만들고 이를 따르도록 하여 매 영상마다 최적에 가깝게 콘트라스트를 개선함.

Rotation

- Rotation with Interpolation



※ 모든 rotation과 resizing은 격자 불일치를 초래해 interpolation에 의한 보정을 필요로 함. Skimage는 이의 정밀한 보정에서 아직 빈약하여 opencv를 권장. 이때 `image.shape, flags` 변수의 적절한 설정이 필요함.

```
>> from skimage.transform import rotate
...
>> img = rotate(img, 90, resize=True).shape
```

```
>> import numpy as np
...
>> coord_center = tuple(np.array(image.shape)/2)
>> rot_mat = cv2.getRotationMatrix2D(coord_center, angle, 1.0)
>> result = cv2.warpAffine(image, rot_mat, image.shape, flags=cv2.INTER_LINEAR)
```

Color (1 of 3)

- RGB → Grey



Original
(Color)



Luminosity
(Grey Value)

$$0.21 R + 0.72 G + 0.07 B$$



Average

$$(R + G + B) / 3$$



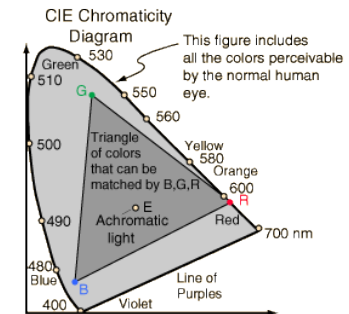
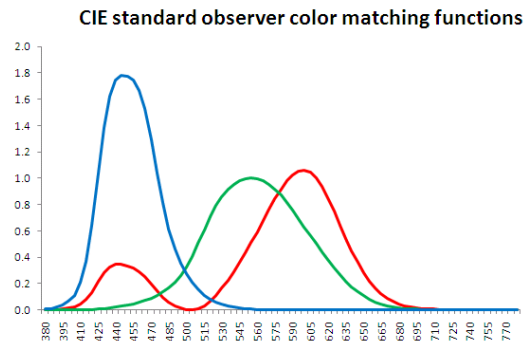
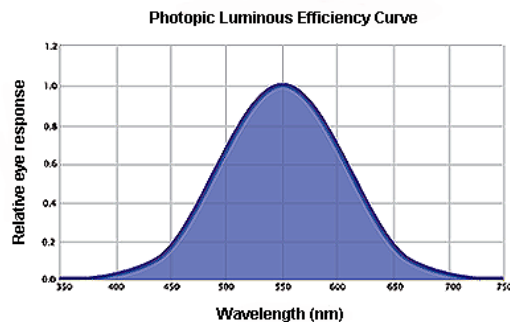
Lightness

$$(\max(R, G, B) + \min(R, G, B)) / 2$$

```
>> from skimage.color import rgb2gray
>> from skimage import data
...
>> img_gray = rgb2gray(img)    // Luminosity
```

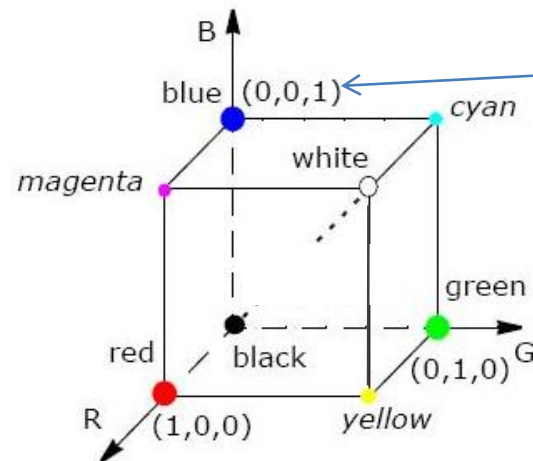
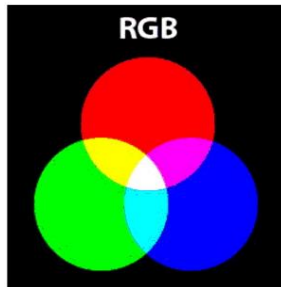
Color (2of3)

- Radiometry → Photometry → Colorimetry

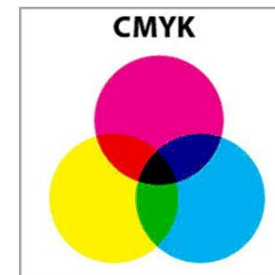


Color (3of3)

- Color Space



※ 슬라이드에서는 1이 255에 해당



※ RGB와 CMY는 반대
(Cyan: R=0, G=255, B=255)

Kalman Filter

(linear quadratic estimation)

Kalman Filtering

• Kalman Filtering

※ 시계열 데이터의 예측과
평활화에 적용

※ “뉴턴 물리학과 칼만 필터로 인간을 달에 보냈다.” (NASA)
“Stanley는 모든게 칼만 필터였다.” (세바스찬 스런)

Kalman filter example demo in Python

A Python implementation of the example given in pages 11-15 of "An
Introduction to the Kalman Filter" by Greg Welch and Gary Bishop,
University of North Carolina at Chapel Hill, Department of Computer
Science, TR 95-041,
<http://www.cs.unc.edu/~welch/kalman/kalmanIntro.html>

by Andrew D. Straw

```
import numpy
import pylab
```

```
# initial parameters
n_iter = 50
sz = (n_iter,) # size of array
x = -0.37727 # truth value (typo in example at top of p. 13 calls this z)
z = numpy.random.normal(x,0.1,size=sz) # observations (normal about x, sigma=0.1)
```

```
Q = 1e-5 # process variance
```

```
# allocate space for arrays
xhat=numpy.zeros(sz) # a posteri estimate of x
P=numpy.zeros(sz) # a posteri error estimate
xhatminus=numpy.zeros(sz) # a priori estimate of x
Pminus=numpy.zeros(sz) # a priori error estimate
K=numpy.zeros(sz) # gain or blending factor
```

```
R = 0.1**2 # estimate of measurement variance, change to see effect
```

※ z : 노이즈가 가미된 입력
 $xhat$: 추정된 출력

```
# initial guesses
xhat[0] = 0.0
P[0] = 1.0
```

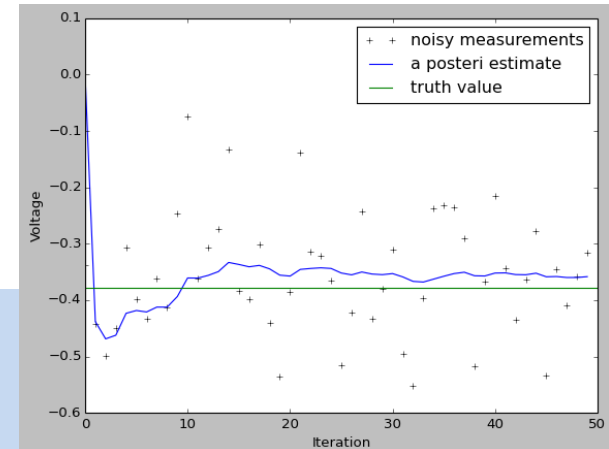
```
for k in range(1,n_iter):
```

```
    # time update
    xhatminus[k] = xhat[k-1]
    Pminus[k] = P[k-1]+Q
```

```
    # measurement update
    K[k] = Pminus[k]/( Pminus[k]+R )
    xhat[k] = xhatminus[k]+K[k]*(z[k]-xhatminus[k])
    P[k] = (1-K[k])*Pminus[k]
```

```
pylab.figure()
pylab.plot(z,'k+',label='noisy measurements')
pylab.plot(xhat,'b-',label='a posteri estimate')
pylab.axhline(x,color='g',label='truth value')
pylab.legend()
pylab.xlabel('Iteration')
pylab.ylabel('Voltage')
```

```
pylab.figure()
valid_iter = range(1,n_iter) # Pminus not valid at step 0
pylab.plot(valid_iter,Pminus[valid_iter],label='a priori error estimate')
pylab.xlabel('Iteration')
pylab.ylabel('$(Voltage)^2$')
pylab.setp(pylab.gca(),'ylim',[0,.01])
pylab.show()
```



부가자료

- 파이썬용 Package, Tools
 - Anaconda Distribution <https://store.continuum.io/cshop/anaconda>
※ opencv, scikit-image, spyder를 모두 담음
 - Scikit-Image Package
 - <http://scikit-image.org>
 - http://scikit-image.org/docs/dev/auto_examples
 - Pycharm IDE <http://www.jetbrains.com/pycharm>
 - 기타 Spyder IDE, iPython QtConsole, Sublime Text 2 등
- 책 (국내 및 번역서)
 - http://book.naver.com/bookdb/book_detail.nhn?bid=6191572
 - http://book.naver.com/bookdb/book_detail.nhn?bid=6802764
 - http://book.naver.com/bookdb/book_detail.nhn?bid=6332184
 - <http://kangcom.com/sub/view.asp?sku=200311040002> (컬러)
- 책 (원서)
 - <http://www.amazon.com/Machine-Vision-Ramesh-Jain/dp/0070320187>
 - http://www.amazon.com/Computer-Vision-Algorithms-Applications-Science/dp/1848829345/ref=sr_1_1?s=books&ie=UTF8&qid=1401379429&sr=1-1&keywords=szeliski+computer+vision

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