

CV Practice Class

2. OpenCV Tutorial, Edge Detection, Image Pyramid, Template Matching

2017-07-04

FIRA 인공지능 에이전트 과정
SNUVL Lab

실습수업 구성

- 실습수업시간
 - 과제정답확인 및 질문답변
 - OpenCV API 소개 및 코드 작성
 - 과제 출제
- 과제
 - API로 구현된 함수를 numpy랑 python으로 직접 구현하기

실습코드 받는법

<https://github.com/bckim92/fira-2017-cv>

```
$ git clone https://github.com/bckim92/fira-2017-cv.git  
$ cd fira-2017-cv/practice2_tutorial  
$ jupyter notebook
```

This repository

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Repository for FIRA Computer Vision (2017 Spring) Practice

Add topics

2 commits 1 branch 0 releases 1 contributor

Branch: master New pull request

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bckim92 Update practice2_tutorial

practice2_tutorial	Update practice2_tutorial
.gitignore	Update practice2_tutorial
README.md	first commit
requirements.txt	Update practice2_tutorial

Clone with HTTPS Use SSH

Use Git or checkout with SVN using the web URL.

<https://github.com/bckim92/fira-2017-cv>

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13 hours ago

Help people interested in this repository understand your project by adding a README.

Add a README

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1. OpenCV-Python
2. Edge Detection
3. Image Pyramid
4. Template Matching

OpenCV-Python

Introduction

Image manipulation

Draw objects

OpenCV-Python

- OpenCV
 - **Computer vision library** started from 1995(Intel)
 - Now supports a multitude of algorithms related to CV and ML (a little of)
- OpenCV-Python
 - OpenCV is basically written in C++
 - OpenCV-Python is a **Python wrapper** of OpenCV
- Prior knowledge of **Python** and **Numpy** is needed
 - A Quick guide to Python - [A Byte of Python](#)
 - [Numpy Quickstart Tutorial](#) / [Justin Johnson's Numpy Tutorial](#)



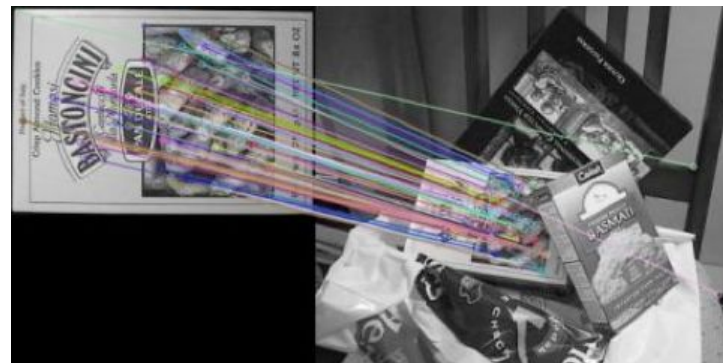
OpenCV-Python

- Examples of algorithms with OpenCV

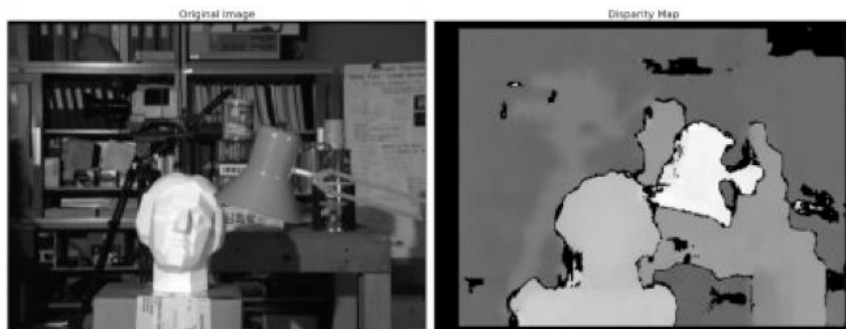
- Face Detection



- Feature extraction/matching



- Depth map for stereo images



- Image inpainting



Using OpenCV-Python

- Import OpenCV-Python package “cv2”

```
import numpy as np
import cv2 # OpenCV-Python
%matplotlib inline
import matplotlib.pyplot as plt
```

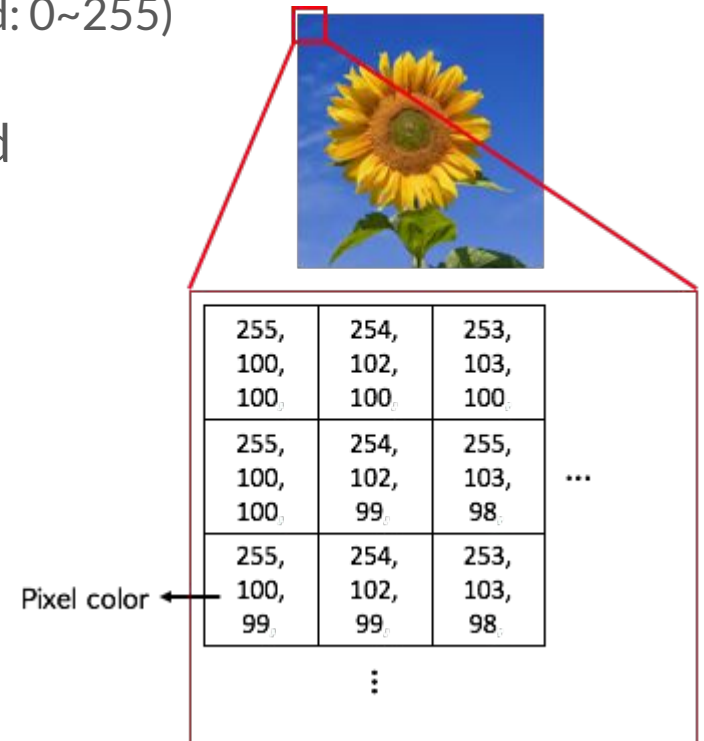
- **Numpy** arrays are data structure used in cv2
 - Converted from/to CvMat(OpenCV in C++) by OpenCV-Python
 - Also used in many python packages

Open/Display an Image

- Open an image

```
img = cv2.imread('image.jpg', cv2.IMREAD_COLOR)
```

- The output is a Numpy array
 - 3D (H x W x C for color) / 2D (H x W for grayscale)
 - Top-left to bottom-right
 - Data type (dtype): np.uint8 (1-byte unsigned: 0~255)
- Flag specifies the way image should be read
 - cv2.IMREAD_COLOR
 - cv2.IMREAD_GRAYSCALE
 - cv2.IMREAD_UNCHANGES

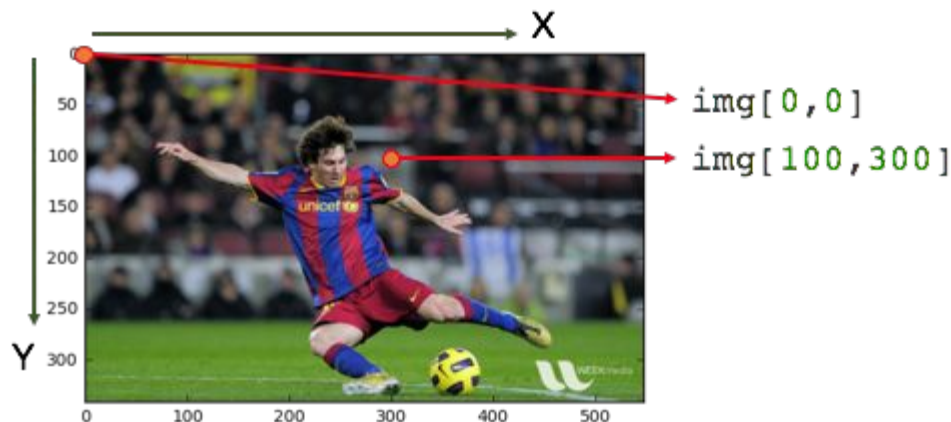


Open/Display an Image

- Display an image using Matplotlib

```
# display an image using matplotlib  
plt.imshow(img) # => The output in wrong color!!  
plt.imshow(cv2.cvtColor(img, cv2.COLOR_BGR2RGB))
```

- `plt.imshow(img)` displays an (RGB, RGBA, grayscale) image
- OpenCV represents RGB images as Numpy arrays in **REVERSE** order (**BGR** not RGB)
- `cv2.cvtColor(img, conversion)` provides conversion among many colortypes



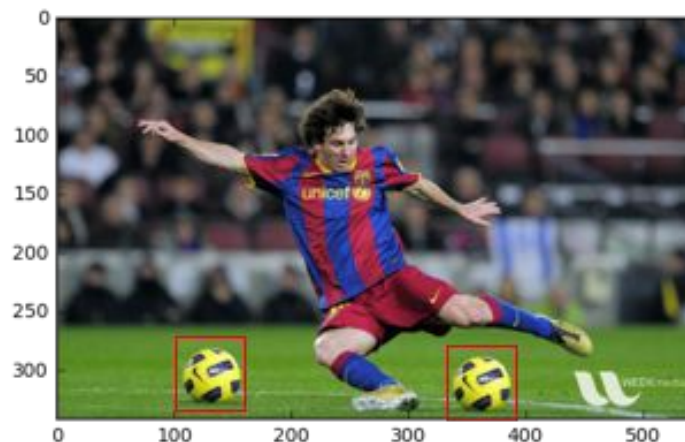
Modify Pixels & ROI

- Pixel and ROI(Region of Interest) can be accessed by Numpy indexing
 - [row, column] ordering - same as matrix indexing

```
# Access a pixel value (BGR order)  
img[50, 235]
```

```
=> array([27, 25, 24], dtype=uint8)
```

```
# ROI is obtained using Numpy indexing  
ball = img[280:340, 330:390]  
img[273:333, 100:160] = ball
```



Draw Objects

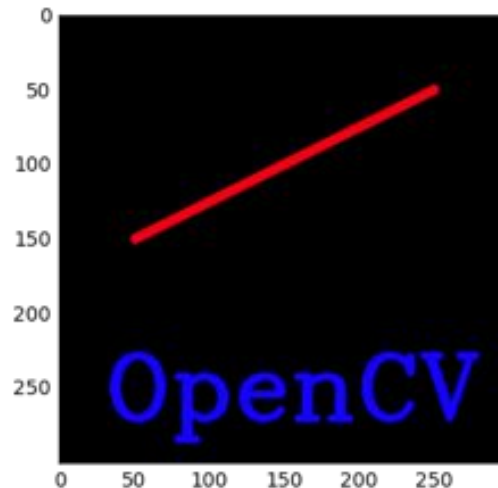
- Draw object (line, rectangle, circle, ellipse, polygon)
 - `cv2.line()`, `cv2.rectangle()`, `cv2.circle()`, `cv2.ellipse()`, `cv2.polylines()`
- Put some text
 - `cv2.putText()`
- Arguments
 - `cv2.function(image, {properties of object})`
 - **RGB** order in color (not BGR)
 - **X, Y** order in position (not row, column)

Draw Objects

- Example

```
# cv2.line(image, startPoint, endPoint, rgb, thickness)
cv2.line(img, (50,150), (250,50), (255,0,0), 5)

# cv2.putText(image, text, bottomLeft, fontType, fontScale,
rgb, thickness, lineType)
font = cv2.FONT_HERSHEY_COMPLEX
cv2.putText(img, 'OpenCV', (30,270), font, 2, (0,0,255), 3,
cv2.LINE_AA)
```



Let's Check the Code

1_getting_started.ipynb

Edge Detection

Image gradient

Sobel operator

Canny edge detection

Edge Detection

- Finding discontinuity of the intensity in images
 - A salient feature of image
 - Much more compact representation than raw pixels
- HOW?
 - Image gradient: DoG, LoG, Sobel operator
 - Dealing with real, noisy images: Canny edge detector

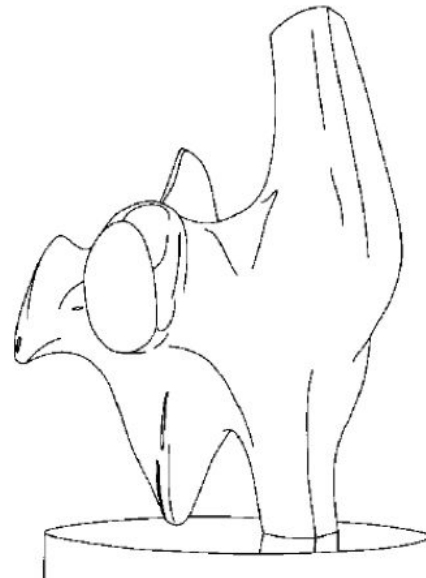
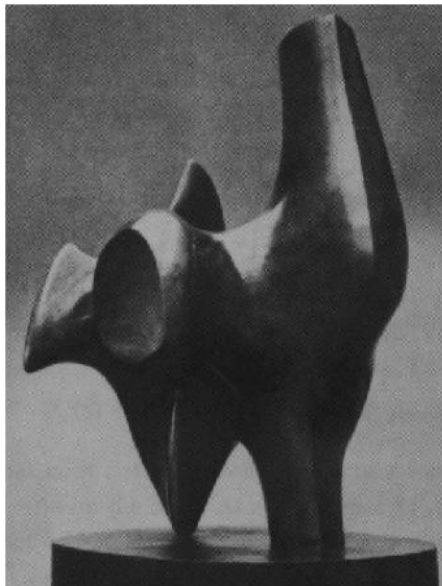


Image Gradient

- The image gradient is approximated by filtering

$$\frac{\partial I}{\partial x} \approx \frac{1}{2\varepsilon} \left((I_{i+1,j+1} - I_{i,j+1}) + (I_{i+1,j} - I_{i,j}) \right)$$
$$\frac{\partial I}{\partial y} \approx \frac{1}{2\varepsilon} \left((I_{i+1,j+1} - I_{i+1,j}) + (I_{i,j+1} - I_{i,j}) \right)$$

$$\frac{\partial I}{\partial x} \approx \frac{1}{2\varepsilon} \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix}$$

$$\frac{\partial I}{\partial y} \approx \frac{1}{2\varepsilon} \begin{bmatrix} 1 & 1 \\ -1 & -1 \end{bmatrix}$$

- Sobel operator
 - A better approximation filter of size 3, 5, 7, ...
 - OpenCV provides Sobel filters of high-dimensional derivatives.

-1	-2	-1
0	0	0
1	2	1

Sobel 3x3 filter

-1	-4	-6	-4	-1
-2	-8	-12	-8	-2
0	0	0	0	0
2	8	12	8	2
1	4	6	4	1

Sobel 5x5 filter

Image Gradient

- Sobel operator in OpenCV

```
dst = cv2.Sobel(src, ddepth, dx, dy, ksize=3, scale=1.0)
```

- src: input image(grayscale, 2-dim array [HxW])
- ddepth: output image depth
- dx: order of the derivative x
- dy: order of the derivative y
- ksize: size of the extended Sobel kernel; it must be 1, 3, 5 or 7
- scale: (optional) scale factor for the computed derivative values

$$dst \approx \frac{\partial^{dx+dy}}{\partial^{dx}x \partial^{dy}y} src$$

Input image



Sobel x operation



Sobel y operation



Sobel intensity



Image Gradient

- Sobel operator in OpenCV (example)

```
img_color = cv2.imread('images/stitch.jpg', cv2.IMREAD_COLOR)
img_gray = cv2.cvtColor(img_color, cv2.COLOR_BGR2GRAY)

sobelx = cv2.Sobel(img_gray, cv2.CV_64F, 1, 0)
```

Grayscale input
image

Input image



64-bit float output

Sobel x operation



cv2.CV_64F
(Gray when 0)

1st order derivative
in x-direction

Sobel x operation



cv2.CV_8U
(Black when 0)

Canny Edge Detector

- A popular edge detection algorithm
- The detection consists of the following procedures (multi-stage algorithm)
 - Noise reduction (5x5 Gaussian filter)
 - Finding image gradient from Sobel-x/y operator
 - Non-maximum suppression (edge-thinning)
 - Hysteresis thresholding

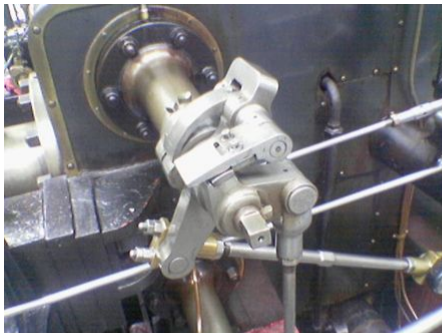


Canny Edge Detector

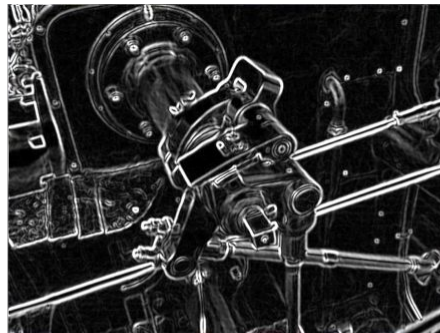
- Sobel operator in OpenCV

```
edges = cv2.Canny(image, threshold1, threshold2, apertureSize=3,  
L2gradient=False)
```

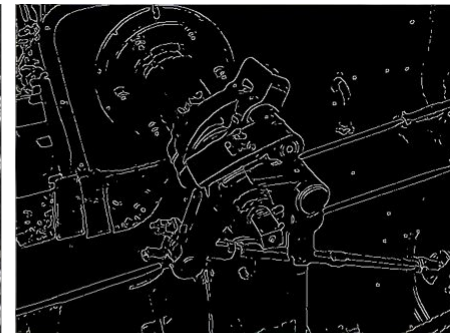
- image: 8-bit grayscale input image
- threshold1/threshold2: thresholds for the hysteresis procedure
- apertureSize: aperture size for the Sobel() operator
- L2gradient: A flag. True to use L_2 -norm of gradients. False for L_1 -norm



Input image



Sobel operator



Canny edge detector

Let's Check the Code

2_edge_detection.ipynb

Image Pyramid

Multi-resolution image pyramids

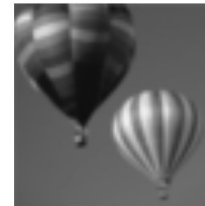
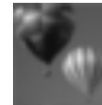
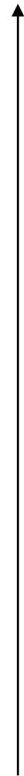
Image blending

Template matching

Multi-resolution Image Pyramids

- Inspired by human visual encoding
- Computationally efficient

Low resolution



High resolution

Multi-resolution Image Pyramids

- (Gaussian) Pyramid operator in OpenCV

```
lower_reso = cv2.pyrDown(higher_reso)
```

- `higher_reso`: $[M \times N]$ input image
- `lower_reso`: $[M/2 \times N/2]$ output image



Multi-resolution Image Pyramids

- (Gaussian) Pyramid operator in OpenCV

```
higher_reso2 = cv2.pyrDown(lower_reso)
```

- `lower_reso`: $[M \times N]$ input image
- `higher_reso2`: $[2M \times 2N]$ output image



Laplacian Image Pyramids

```
high_res_img2 = cv2.pyrUp(low_res_img)
laplacian = cv2.subtract(high_res_img, high_res_img2)
```

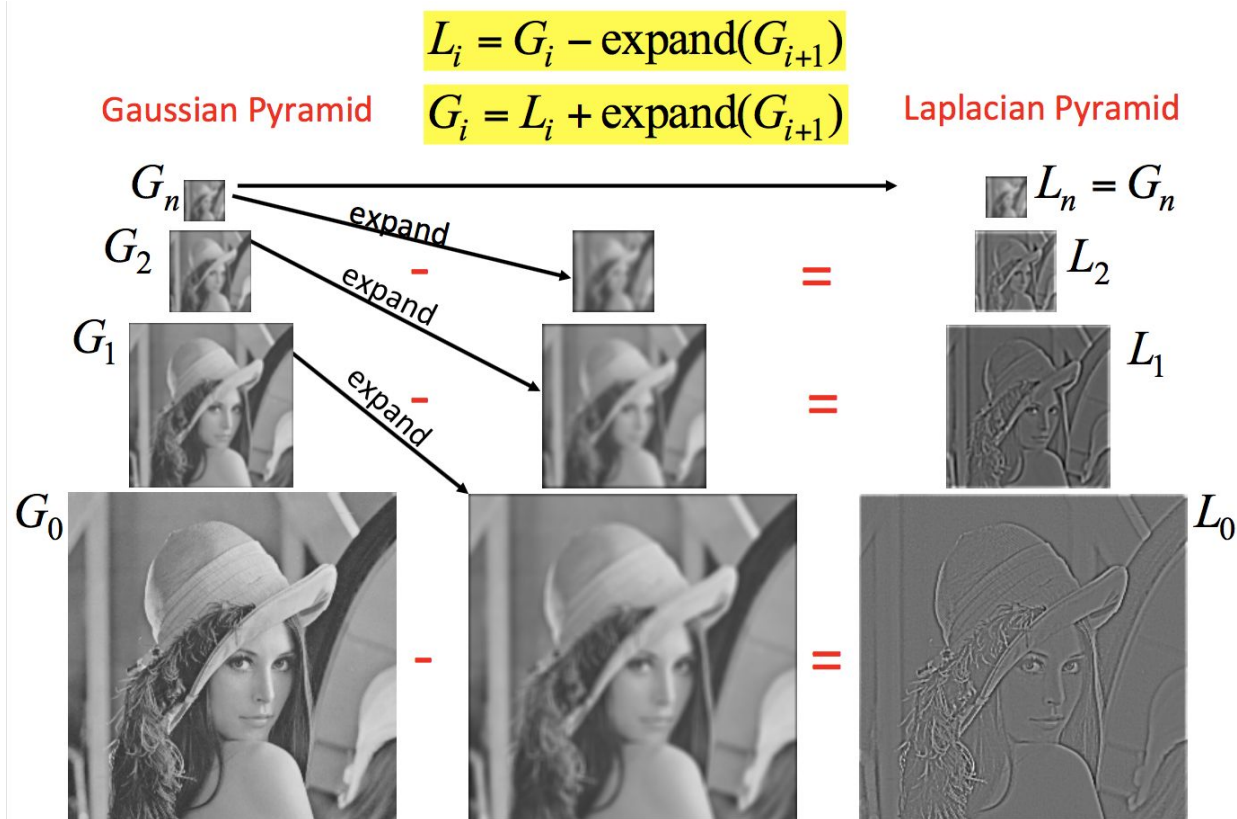


Image Blending using Pyramids

- One application of pyramids is image blending
- Seamless blending without leaving much data in the images

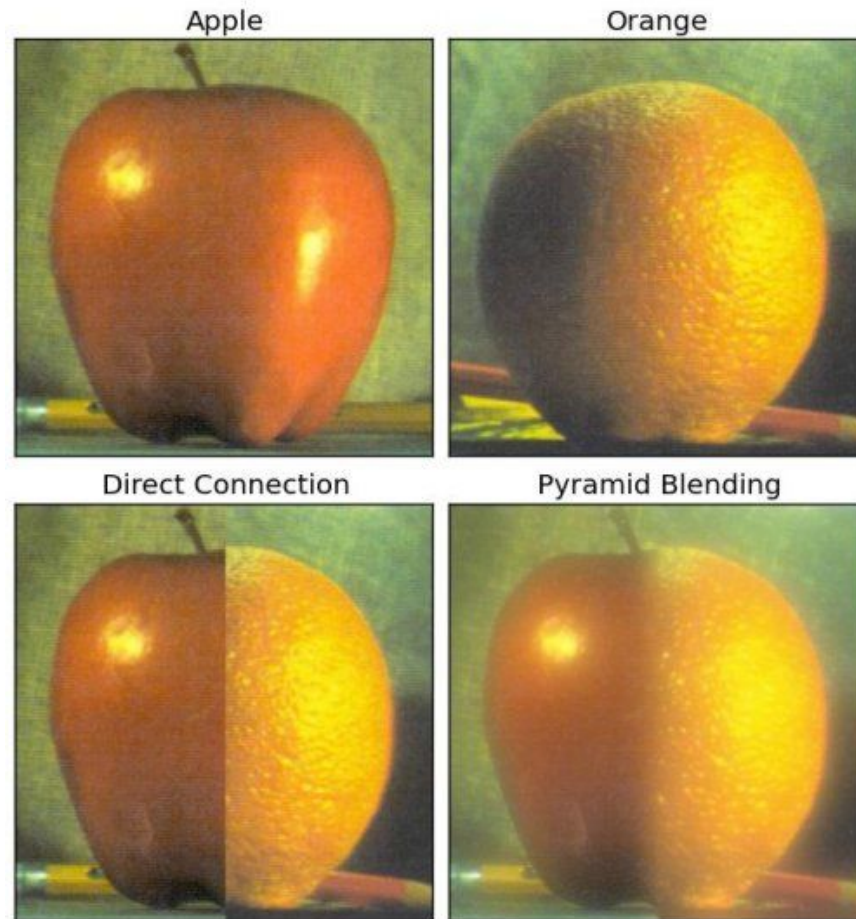


Image Blending using Pyramids

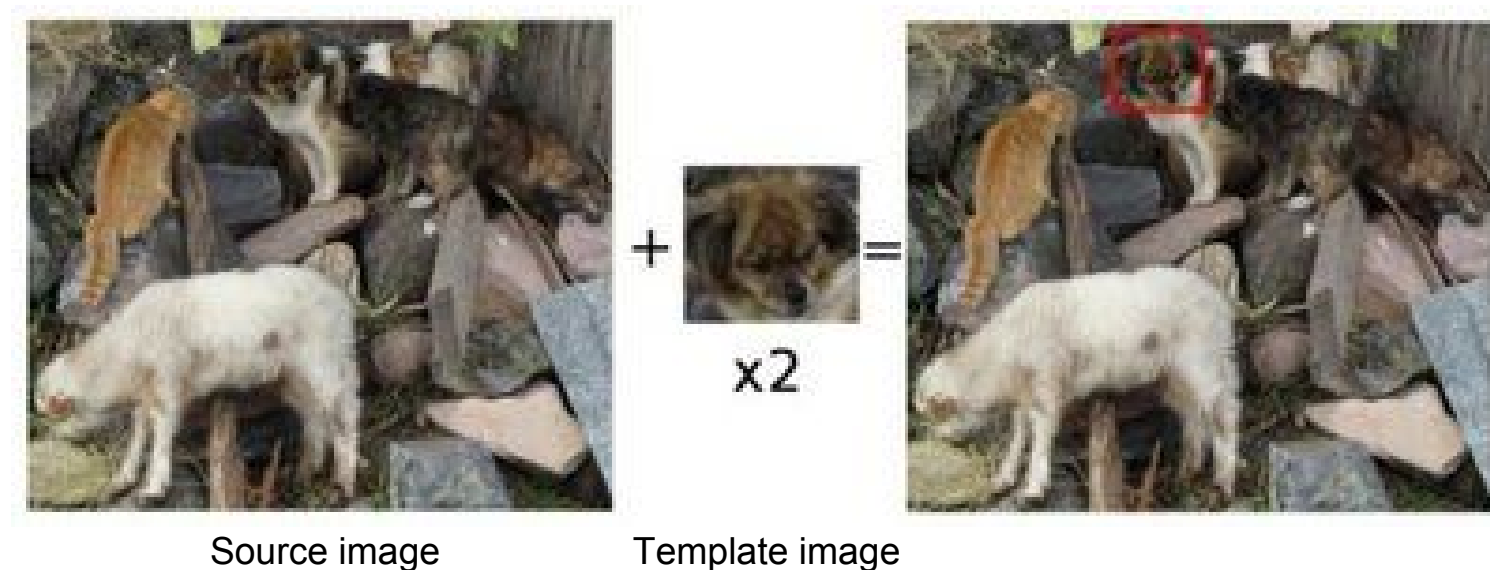
1. Load the two images of apple and orange
2. Find the Gaussian Pyramids for apple and orange
3. From Gaussian Pyramids, find their Laplacian Pyramids
4. Now join the left half of apple and right half of orange in each levels of Laplacian Pyramids
5. Finally from this joint image pyramids, reconstruct the original image

Let's Check the Code

3_image_blending.ipynb

Template Matching

- Technique for finding areas of an image that match (are similar) to a template image
- Two primary components
 - **Source image (I)**: image to match the template image
 - **Template image (T)**: patch image to compare
- **Goal**: detect the highest matching area



Template Matching

- Template matching operator in OpenCV

```
matching_result = cv2.matchTemplate(img, template, match_method)
```

- `img`: image to match the template image
- `template`: patch image to compare
- `match_method`: comparison method
 - `cv2.TM_CCOEFF`
 - `cv2.TM_CCOEFF_NORMED`
 - `cv2.TM_CCORR`
 - `cv2.TM_CCORR_NORMED`
 - `cv2.TM_SQDIFF`
 - `cv2.TM_SQDIFF_NORMED`
- `matching_result`: grayscale image; each pixel denotes how much does the template match with image

Template Matching

- Template matching operator in OpenCV

```
min_val, max_val, min_loc, max_loc =  
cv2.minMaxLoc(matching_result)
```

- `matching_result`: input single-channel array
- `min_val`: minimum value
- `max_val`: maximum value
- `min_loc`: minimum location
- `max_loc`: maximum location

Matching Result



Detected Point



Fast Template Matching

Template



Search Region

Original Image



Let's Check the Code

4_template_matching.ipynb

Homework

See `5_homework.ipynb`

Reference

- OpenCV-Python Tutorials

http://docs.opencv.org/3.0-beta/doc/py_tutorials/py_tutorials.html

Q&A

Any Question?