# Python OpenCV Lab 2021 Crash Course

NTU GIEE Media IC & System Lab 楊凱翔 2021/08/02

## Outline

- Prerequisite
  - Lab1: basic image processing
    - Image Filtering
    - Image PCA Analysis
  - Lab2: Homography

#### Outline

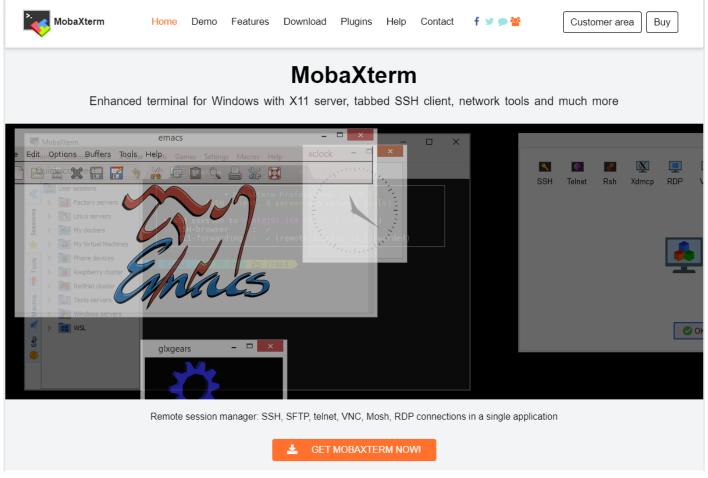
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#### Prerequisite

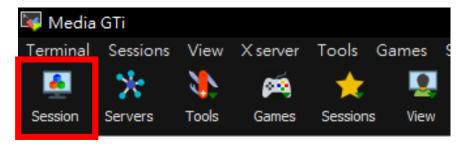
- SSH Client: MobaXterm
- Code editor: VS Code, Notepad++
- Language: Python3
- Library:
  - NumPy: array operation
  - OpenCV: computer vision task
  - Matplotlib: visualization in python

#### MobaXterm

Download from <a href="https://mobaxterm.mobatek.net/">https://mobaxterm.mobatek.net/</a>



#### MobaXterm

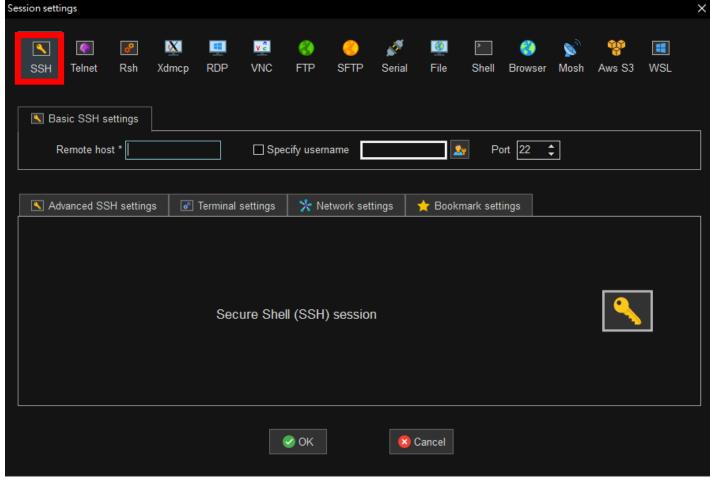


Refer to 421 wiki to get remote host and port.

#### MediaGTi:

remote host: 140.112.48.127

port: 10800

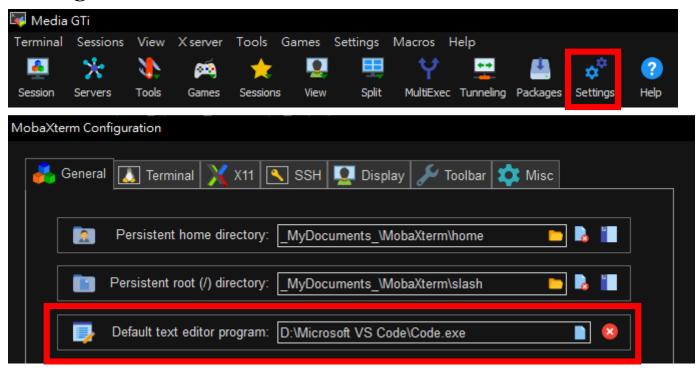


#### MobaXterm

Type \$ passwd to change password

```
# siangyang @ MediaGTi in ~ [15:23:09]
$ passwd
```

Change default text editor



#### Linux Commend

- Tutorial: <a href="https://blog.techbridge.cc/2017/12/23/linux-commnd-line-tutorial/">https://blog.techbridge.cc/2017/12/23/linux-commnd-line-tutorial/</a>, Google
- Basic
  - cd, ls, git, mkdir, mv, rm, ···
- Useful tool
  - tmux, nvidia-htop, nvidia-smi, jupyter notebook, anaconda, ···

## Python Grammar

- Tutorial: <a href="http://cs231n.github.io/python-numpy-tutorial/">http://cs231n.github.io/python-numpy-tutorial/</a>, Google
- Basic
  - print(), if else, for loop
- Useful
  - List, indexing of list
  - Ex: a = [1,2,4]; a[0] = 1; a[-1] = 4; a[:2] = [1,2]
- Optional
  - Function declaration (if some process is repeated)

#### NumPy

- Tutorial: <a href="http://cs231n.github.io/python-numpy-tutorial/">http://cs231n.github.io/python-numpy-tutorial/</a>, Google
- Import library:
  - import numpy as np
- Basic:
  - Array initialization, basic property (shape, data type), indexing
- Useful:
  - Build-in function for array operation: argmin, matmul

#### OpenCV

- Tutorial: <a href="https://docs.opencv.org/4.5.2/">https://docs.opencv.org/4.5.2/</a>, Google
- Basic
  - Image read, write, resize, color conversion, ···
- Useful
  - Padding, filtering, other CV tasks

# [Supplement] some tips for image read/write

- All image read method
  - OpenCV
  - Pillow
  - Matplotlib
  - Scikit-image
  - Imageio
  - Scipy.misc (unavailable after 1.1.0)
- Shape: (H, W, C)
- In OpenCV, the default color order is BGR
- Change to np.array and change dtype from uint8 to floatXX before some operations
- More examples in ImageRead.ipynb

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## Image Filtering

Weighted sum of the region of the input

$$g(x,y) = \frac{1}{W} \sum_{i,j \in [-r,r]} h(i,j) f(x-i,y-j)$$
  $W = \sum_{i,j \in [-r,r]} h(i,j)$ 

45	60	98	127	132	133	137	133						
46	65	98	123	126	128	131	133					200	
47	65	96	115	119	123	135	137		0.1	0.1	0.1		
47	63	91	107	113	122	138	134	*	0.1	0.2	0.1	=	
50	59	80	97	110	123	133	134		0.1	0.1	0.1		5
49	53	68	83	97	113	128	133					535	
50	50	58	70	84	102	116	126						
50	50	52	58	69	86	101	120						8

$$f(x,y)$$
  $h(x,y)$   $g(x,y)$ 

# Lab1: Image Filtering

Sobel filter: used in edge detection

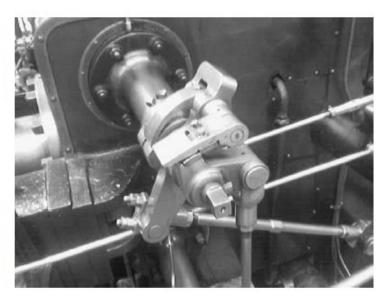
$$G_{x} = \begin{bmatrix} 1 & 0 & -1 \\ 2 & 0 & -2 \\ 1 & 0 & -1 \end{bmatrix} * I$$

$$G_{x} = \begin{bmatrix} 1 & 0 & -1 \\ 2 & 0 & -2 \\ 1 & 0 & -1 \end{bmatrix} * I \qquad G_{y} = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix} * I$$

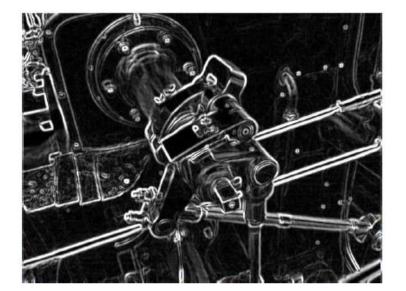
$$G = \sqrt{{G_x}^2 + {G_y}^2}$$

Input Gray-scale image

Output Gray-scale image

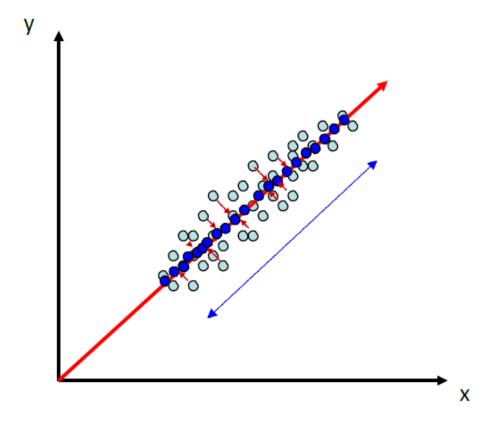






# Principal Component Analysis (PCA)

- Goal : determine the projection to maximize the variance of the projected data
- Linear dimension reduction



# Principal Component Analysis (PCA)

- Input:
  - A set of instances  $\{\vec{x}\}_{i=1}^N$ ,  $\vec{x}_i \in \mathbb{R}^d$
  - Zero mean:  $\vec{x}' = \vec{x} \vec{\mu}$ , where  $\vec{\mu} = \frac{1}{N} \sum_i \vec{x}_i$
- First component:
  - A unit vector  $\vec{w} \in \mathbb{R}^d$  that maximize the variance of the projected data  $\{\vec{w} \cdot \vec{x_i}'\}_{i=1}^N$
- Further components:
  - Derived from the data without the first component

$$\{\vec{x}_i'\}_{i=1}^N \to \{\vec{x}_i - (\vec{w} \cdot \vec{x}_i')\vec{w}\}_{i=1}^N$$

Mutually orthogonal

# Principal Component Analysis (PCA)

- Principal components (PCA Eigen-basis)  $\{\vec{w}_i\}_{i=1}^N$  (usually  $N < d \rightarrow K = N-1$ )
- Vector representation

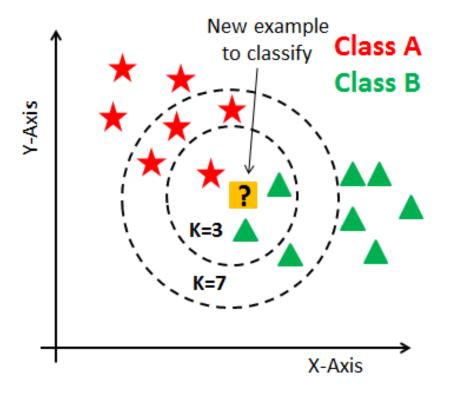
$$\vec{x}_i = \vec{\mu} + \sum_{i=1}^{N-1} (\vec{w}_i \cdot (\vec{x}_i - \vec{\mu})) \vec{w}_i$$

Vector approximation

$$\vec{x}_i \cong \vec{\mu} + \sum_{i=1}^k (\vec{w}_i \cdot (\vec{x}_i - \vec{\mu})) \vec{w}_i$$

#### KNN Classifier

• k nearest neighbors classifier



# Lab1: Image PCA Analysis

- Given face images  $\vec{x}_i$  with 40 classes, 10 images for each class (6 train, 4 test)
- Perform PCA on training set  $\rightarrow$  get the eigenfaces  $\vec{w}_i$
- Reconstructed an image with 3 or 100 eigenfaces and compute mean square error
   (MSE)
- Apply KNN classifier on test set

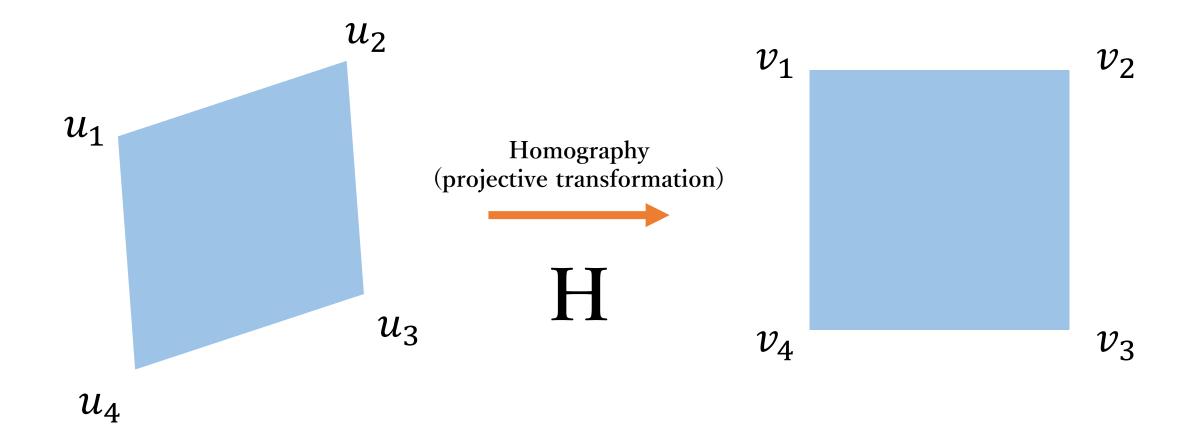
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# Lab2: Homography



## Recap of Homography

• Matrix form:

$$\begin{bmatrix} v_x \\ v_y \\ 1 \end{bmatrix} \sim \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} u_x \\ u_y \\ 1 \end{bmatrix}$$

• Equations:

$$v_x = \frac{h_{11}u_x + h_{12}u_y + h_{13}}{h_{31}u_x + h_{32}u_y + h_{33}}$$
$$v_y = \frac{h_{21}u_x + h_{22}u_y + h_{23}}{h_{31}u_x + h_{32}u_y + h_{33}}$$

## Recap of Homography

#### • Degree of freedom:

• 
$$9 - 1 = 8 \text{ DoF}$$

$$v_x = \frac{kh_{11}u_x + kh_{12}u_y + kh_{13}}{kh_{31}u_x + kh_{32}u_y + kh_{33}}$$

$$v_y = \frac{kh_{21}u_x + kh_{22}u_y + kh_{23}}{kh_{31}u_x + kh_{32}u_y + kh_{33}}$$



$$v_x = \frac{h_{11}u_x + h_{12}u_y + h_{13}}{h_{31}u_x + h_{32}u_y + h_{33}}$$

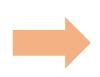
$$v_y = \frac{h_{21}u_x + h_{22}u_y + h_{23}}{h_{31}u_x + h_{32}u_y + h_{33}}$$

Constraint

$$h_{11}^2 + \dots + h_{33}^2 = 1$$

#### Solution

$$v_x = \frac{h_{11}u_x + h_{12}u_y + h_{13}}{h_{31}u_x + h_{32}u_y + h_{33}}$$
$$v_y = \frac{h_{21}u_x + h_{22}u_y + h_{23}}{h_{31}u_x + h_{32}u_y + h_{33}}$$



$$(h_{31}u_x + h_{32}u_y + h_{33})v_x = h_{11}u_x + h_{12}u_y + h_{13}$$
$$(h_{31}u_x + h_{32}u_y + h_{33})v_y = h_{21}u_x + h_{22}u_y + h_{23}$$



$$h_{11}u_x + h_{12}u_y + h_{13} - h_{31}u_xv_x - h_{32}u_yv_x - h_{33}v_x = 0$$
  
$$h_{21}u_x + h_{22}u_y + h_{23} - h_{31}u_xv_y - h_{32}u_yv_y - h_{33}v_y = 0$$

#### Solution

• Construct a linear system using N vertices:

$$\mathbf{A} \, \mathbf{h} = \mathbf{b}$$

- **b** is all zero
- Solve *h* 
  - Ah = 0
  - $A^TAh = 0$
  - SVD of  $A^T A = U \Sigma V^T$
  - Let h be the last column of U (unit eigenvector) associated with the smallest eigenvalue in  $\Sigma$

#### Lab2 Problem

#### Make the QR code frontal parallel



Calling function cv2.findHomography is FORBIDDEN!!!

## Backward Warping

- Prevent holes in output space
- Pixel value at sub pixel location like (30.21, 22.74)?
  - Bilinear interpolation
  - Nearest neighbor

#### Forward Warping

#### **Backward Warping**

