

電腦視覺與深度學習

(Computer Vision and Deep Learning)

Homework 1

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Office Hour: 14:00~16:00, Mon.

10:00~12:00, Fri.

At CSIE 9F Robotics Lab.

Notice (1/2)

- Copying homework is strictly prohibited!! **Penalty: Both individuals will receive a score of 0!!**
- Due date => **09:00:00, 2023/11/09 (Thu.)**
 - Do not submit late**, or the following points will be deducted:
 - Submit within seven days after the deadline, and your score will be reduced by half.
 - If you submit after this period, you will receive a score of 0.
- You must **attend the demonstration**, otherwise your score will be 0. The demonstration schedule **will be announced on NCKU Moodle**.
- You must **create GUI**, otherwise your point will be **deducted**.
- Upload to => **140.116.154.28 -> Upload/Homework/Hw1**
 - **User ID: cvdl2023 Password: RL2023cvdl**
- Format
 - Filename: **Hw1_StudentID_Name_Version.rar**
 - **Ex: Hw1_F71234567_林小明_V1.rar**
 - If you want to update your file, you should update your version to be V2,
 - **Ex: Hw1_F71234567_林小明_V2.rar**
 - Content: **Project folder** *(Excluding the pictures)
 - *Note: Remove your “Debug” folder to reduce file size.

Notice (2/2)

- Python (recommended):
 - Python 3.8 (<https://www.python.org/downloads/>)
 - **Opencv-contrib-python (3.4.2.17)**
 - Matplotlib 3.7.3
 - UI framework: pyqt5 (5.15.10)
 - Pytorch 2.1.0
 - Torchvision 0.16.0
 - Torchsummary 1.5.1
 - Tensorboard 2.14.0
 - Pillow 10.1.0

Assignment scoring (Total: 100%)

1. (20%) Camera Calibration (出題：Shan)

- 1.1 (4%) Corner detection
- 1.2 (4%) Find the intrinsic matrix
- 1.3 (4%) Find the extrinsic matrix
- 1.4 (4%) Find the distortion matrix
- 1.5 (4%) Show the undistorted result

2. (20%) Augmented Reality (出題：Eric)

- 2.1 (10%) Show words on board
- 2.2 (10%) Show words vertically

3. (20%) Stereo Disparity Map (出題：Eric)

- 3.1 (10%) Stereo Disparity Map
- 3.2 (10%) Checking the Disparity Value

4. (20%) SIFT (出題：Chen)

- 4.1 (10%) Keypoints
- 4.2 (10%) Matched Keypoints

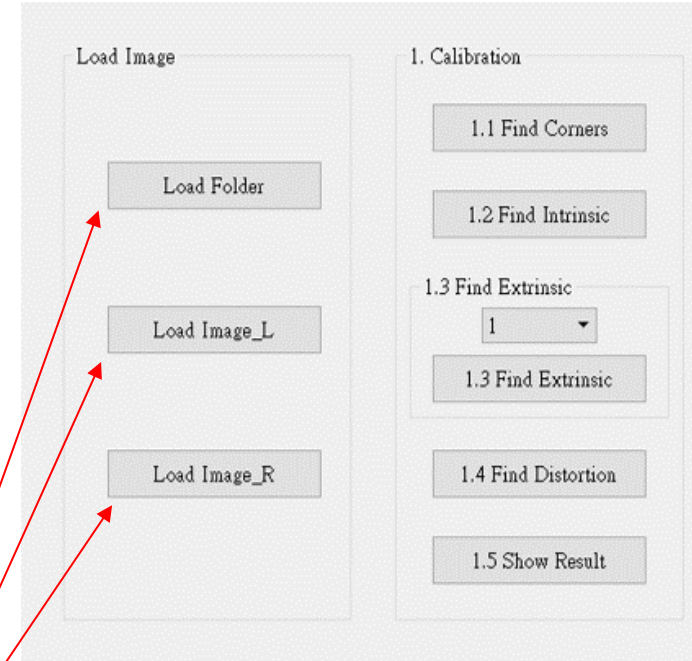
5. (20%) Training a CIFAR10 Classifier Using VGG19 with BN (出題：Hsiang)

- 5.1 Load CIFAR10 and show 9 Augmented Images with Labels. (4%)
- 5.2 Load Model and Show Model Structure. (4%)
- 5.3 Show Training/Validating Accuracy and Loss. (6%)
- 5.4 Use the Model with Highest Validation Accuracy to Run Inference, Show the Predicted Distribution and Class Label. (6%)

* Don't fix your image path
(There is another dataset for demonstration)

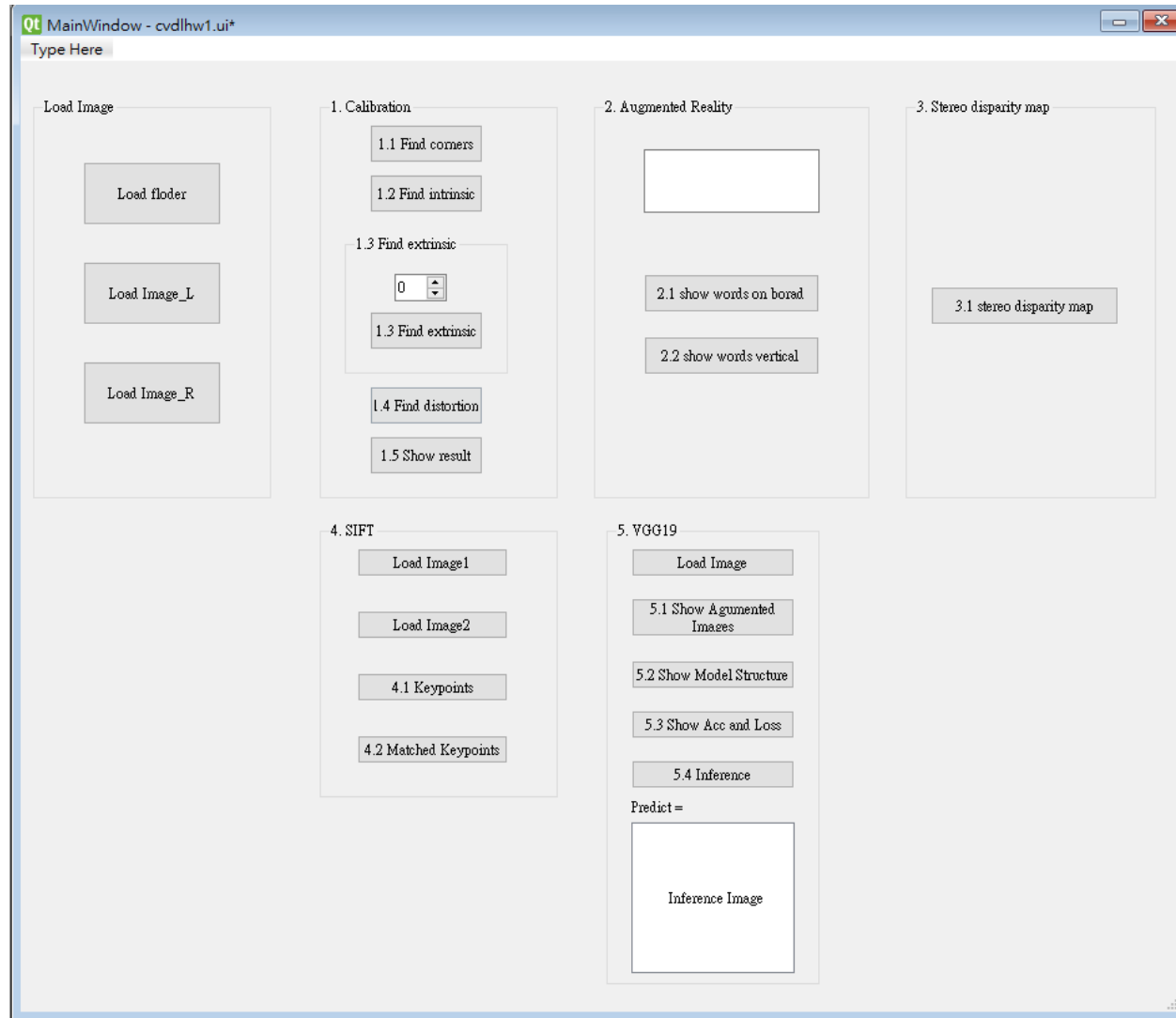
Load image please use the following function to read the path.

[QFileDialog.getOpenFileName](#)



Assignment scoring (Total: 100%)

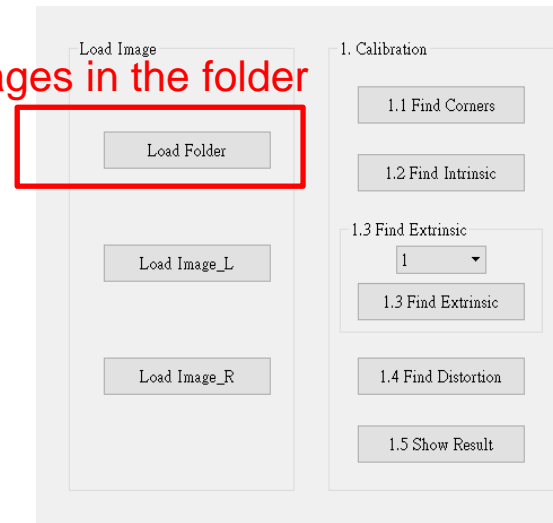
- Use one UI to present 5 questions.



1. (20%) Camera Calibration

- 1.1 (4%) Corner detection
- 1.2 (4%) Find the intrinsic matrix
- 1.3 (4%) Find the extrinsic matrix
- 1.4 (4%) Find the distortion matrix
- 1.5 (4%) Show the undistorted result

Load all images in the folder



1.1 Corner Detection (4%)

❑ Given: 15 images, 1.bmp ~ 15.bmp

❑ Q1:

1) Find and draw the corners on the chessboard for each image.

`cv2.findChessboardCorners(grayimg, (width, high), None)`

`cv2.cornerSubPix(image, corners, winSize, zeroZone, criteria)`

winSize = (5, 5) **zeroZone** = (-1, -1), this parameter means to ignore.

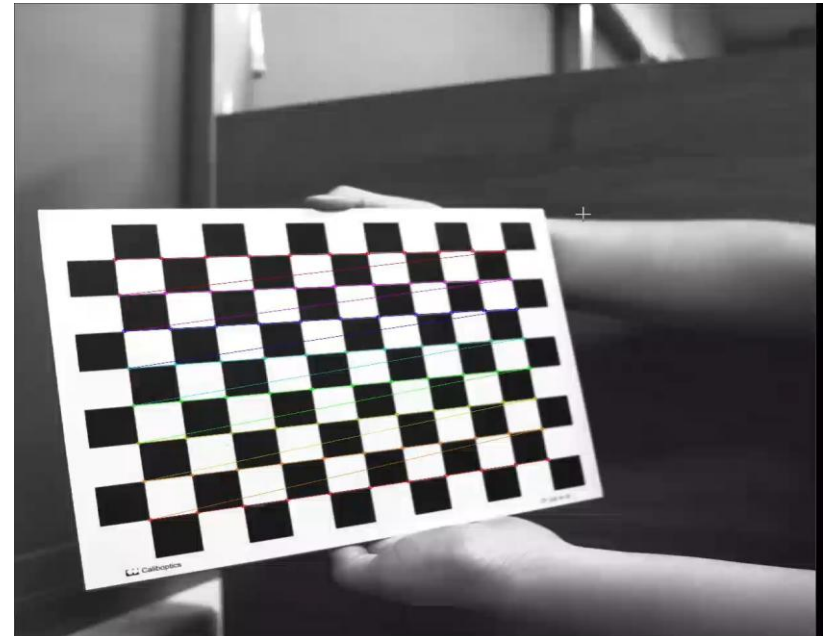
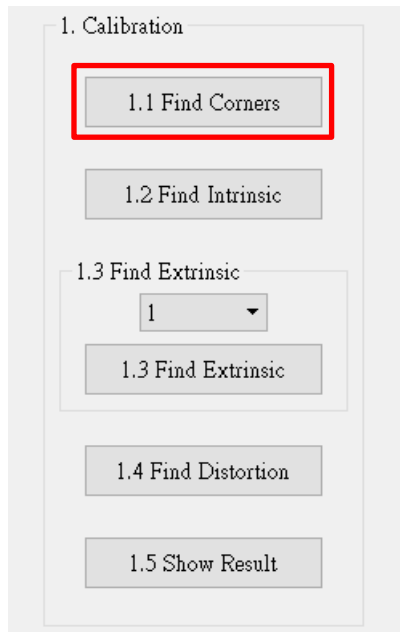
criteria = (cv2.TERM_CRITERIA_EPS + cv2.TERM_CRITERIA_MAX_ITER, 30, 0.001)

2) Click button “1.1 Find Corners” to show each picture.

❑ Hint:

OpenCV Textbook Chapter 11 (p. 398 ~ p. 399)

❑ Ex:



1.2 Find the Intrinsic Matrix (4%)

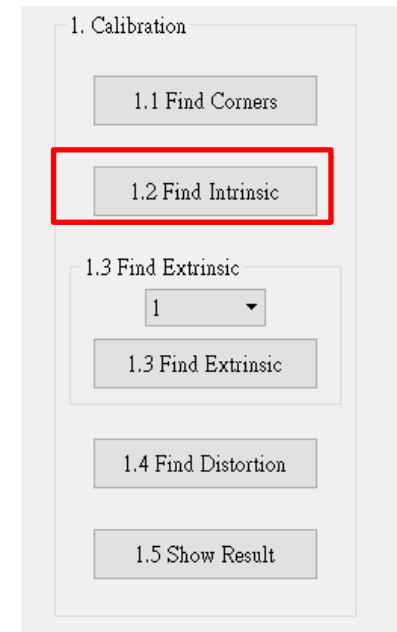
❑ Given: 15 images, 1.bmp ~ 15.bmp

❑ Q2:

1) Find the intrinsic matrix:
$$\begin{bmatrix} \alpha & \gamma & u_0 \\ 0 & \beta & v_0 \\ 0 & 0 & 1 \end{bmatrix}$$

`cv2.calibrateCamera (objectPoints, imagePoints, (width, high) ,None, None)`

2) Click button “1.2 Find Intrinsic” and then show the result on the console window.



❑ Output format:

```
Intrinsic:
[[2.22370244e+03 0.00000000e+00 1.03021663e+03]
 [0.00000000e+00 2.22296836e+03 1.03752624e+03]
 [0.00000000e+00 0.00000000e+00 1.00000000e+00]]
```

(Just an example)

❑ Hint: OpenCV Textbook Chapter 11 (p.398 ~ p.400)

1.3 Find the Extrinsic Matrix (4%)

- ❑ Given: Intrinsic parameters, distortion coefficients, and the list of 15 images
- ❑ Q3:

1) Find the extrinsic matrix of the chessboard for each of the 15 images, respectively:

$$\begin{bmatrix} R_{11} & R_{12} & R_{13} & T_1 \\ R_{21} & R_{22} & R_{23} & T_2 \\ R_{31} & R_{32} & R_{33} & T_3 \end{bmatrix}$$

2) Click button “1.3 Find Extrinsic” and then show the result on the console window.

- ❑ Output format:

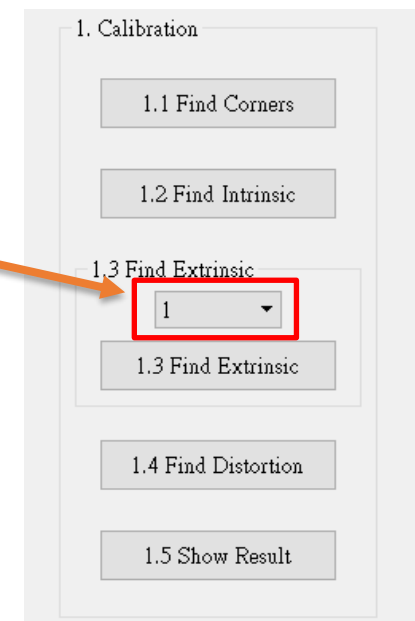
```
Extrinsic:
[[ -0.8767247  -0.23001438  0.4224301  4.39838495]
 [  0.19727469 -0.97293475 -0.12033563  0.68022105]
 [  0.43867585 -0.02216645  0.89837194 16.22126   ]]
```

(Just an example)

- ❑ Hint: OpenCV Textbook Chapter 11, (p.370 ~ p.402)

(1) List of numbers: 1~15

(2) Select 1, then 1.bmp will be applied, and so on



1.4 Find the Distortion Matrix (4%)

(出題 : shan)

☐ Given: 15 images

☐ Q4:

1) Find the distortion matrix: $[k_1, k_2, p_1, p_2, k_3]$

2) Click button “1.4 Find Distortion” to show the result on the console window.

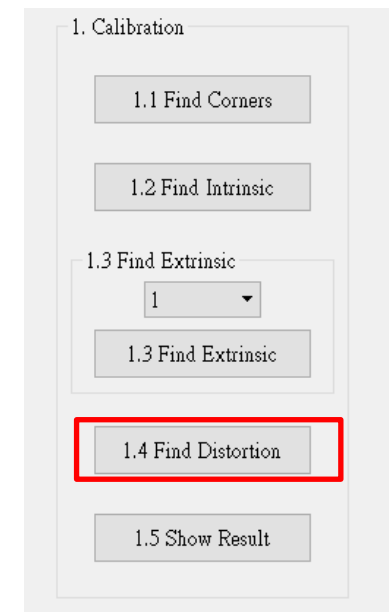
☐ Output format:

```
Distortion:
[[-0.11868112  0.02776881 -0.00092036  0.00047227  0.11793646]]
```

 (Just an example)

☐ Hint:

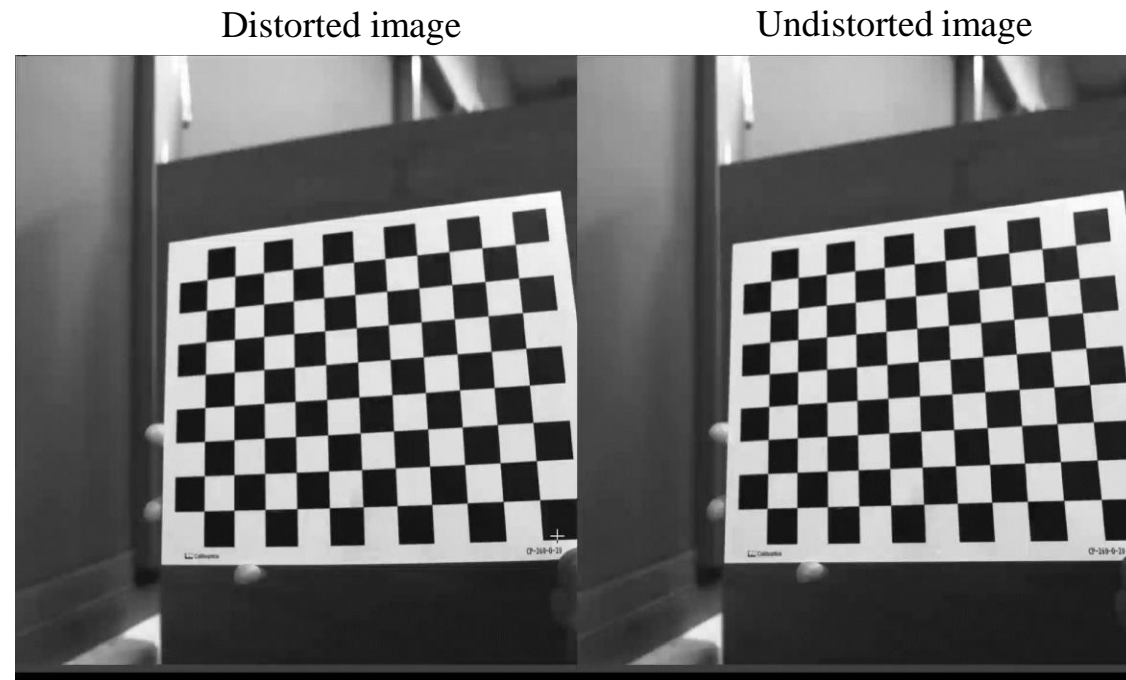
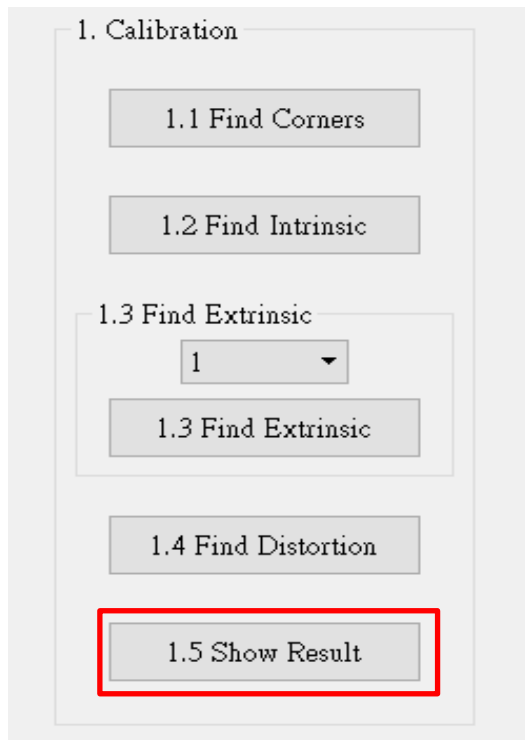
- Distortion coefficients can be obtained simultaneously with intrinsic.
- OpenCV Textbook Chapter 11 (p.398 ~ p.400)



1.5 Show the Undistorted Result (4%)

(出題 : shan)

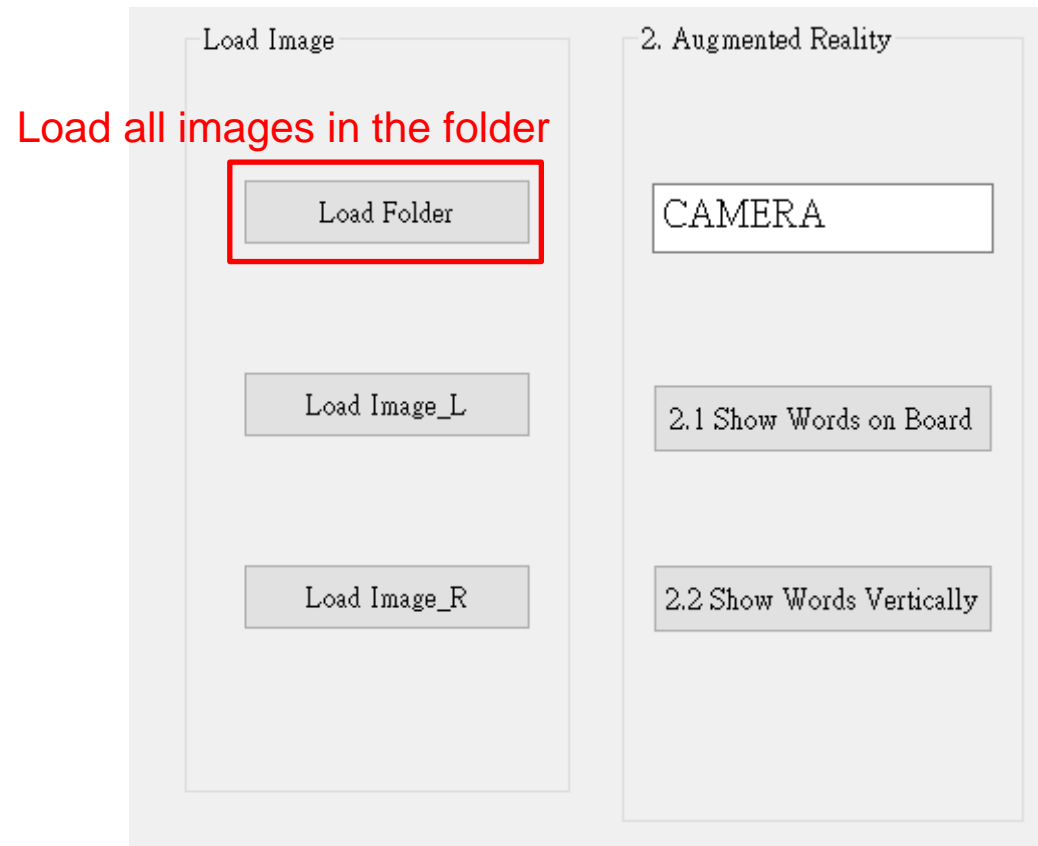
- ☐ Given: 15 images
- ☐ Q5:
 - 1) Undistort the chessboard images.
 - 2) Click button “1.5 Show Result” to show distorted and undistorted images
- ☐ Hint:
 - `cv2.undistort(...)` or `cv2.initUndistortRectifyMap(...)`
 - OpenCV Textbook Chapter 11 (p.398 ~ p.400)



2. (20%) Augmented Reality

2.1 (10%) Show words on board

2.2 (10%) Show words vertically



2. (20%) Augmented Reality

(出題：Eric)

➤ Given: 5 images (1~5.bmp), alphabet_lib_onboard.txt and alphabet_lib_vertical.txt.

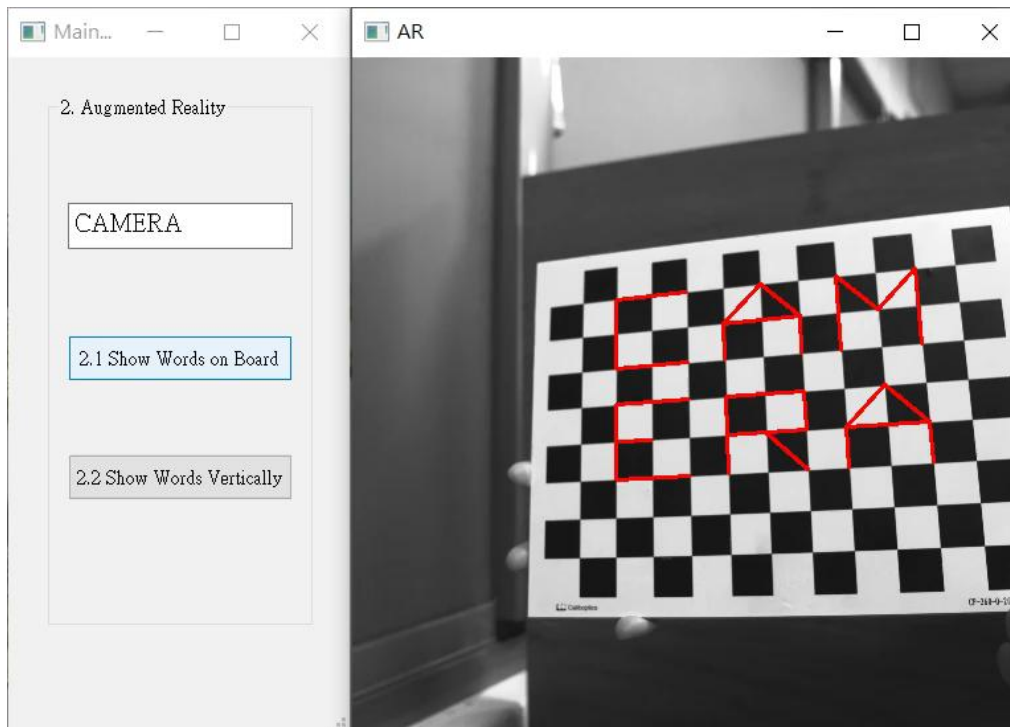
Q1: Show a Word (e.g. CAMERA) on chessboard

- 1) Calibrate 5 images to get intrinsic, distortion and extrinsic parameters.
- 2) Input a word **less than 6 char in English** in textEdit box.
- 3) Derive the shape of the word by **using the provided library (alphabet_lib_onboard.txt)**.
- 4) Click button “2.1 Show Words on Board” to show result, each picture for 1 second (total 5 images).

Q2: Show a Word (e.g. CAMERA) **vertically** on chessboard

- 1) Calibrate 5 images to get intrinsic, distortion and extrinsic parameters.
- 2) Input a word **less than 6 char in English** in textEdit box (Accept the same word entered in as Q1).
- 3) Derive the shape of the word by **using the provided library (alphabet_lib_vertical.txt)**.
- 4) Click button “2.2 Show Words Vertically” to show result, each picture for 1 second (total 5 images).

➤ Result Video:



➤ Hint: OpenCV Textbook Chapter 11 (P.387~395)
cv2.calibrateCamera()
OpenCV Textbook Chapter 11 (P.405~412)
cv2.projectPoints()

2. (20%) Augmented Reality

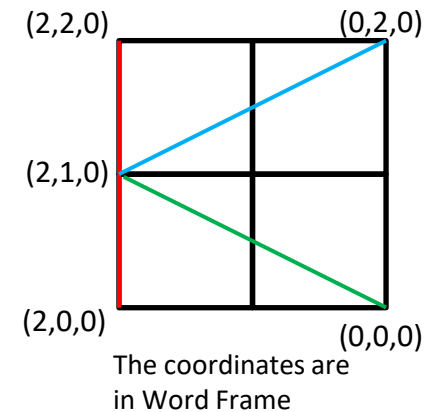
➤ Guides and Requirements:

- How to use the library: (alphabet_lib_onboard.txt, alphabet_lib_vertical.txt)
 - Use OpenCV function to read and derive the array or matrix of the char
Here take 'K' in 'alphabet_lib_onboard.txt' for example

e.g. (Python):

```
fs = cv2.FileStorage('alphabet_lib_onboard.txt', cv2.FILE_STORAGE_READ)
ch = fs.getNode('K').mat() ➔ get the lines of 'K'
```

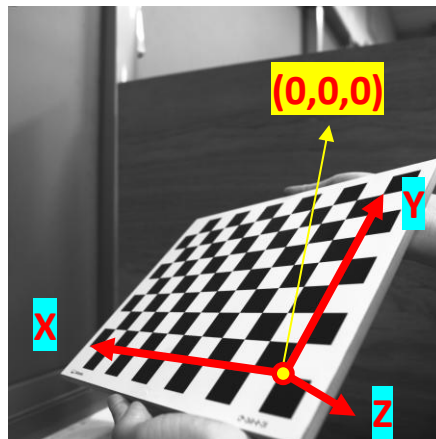
```
ch = [[[2, 2, 0], [2, 0, 0]],
      [[0, 2, 0], [2, 1, 0]],
      [[2, 1, 0], [0, 0, 0]]]
```



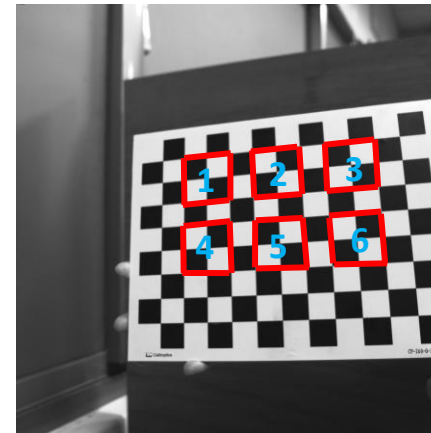
- 'K' consist of 3 lines, so the 'ch array' consists 3 pairs of 3D coordinates in Word Frame representing two ends of the line shown in the upper right image.

2) Chessboard Coordinates

- The chessboard x, y, z axis and (0,0,0) coordinate are shown in the bottom left image
- Each Char should be place in the order and position shown in the bottom right image



Chessboard Frame

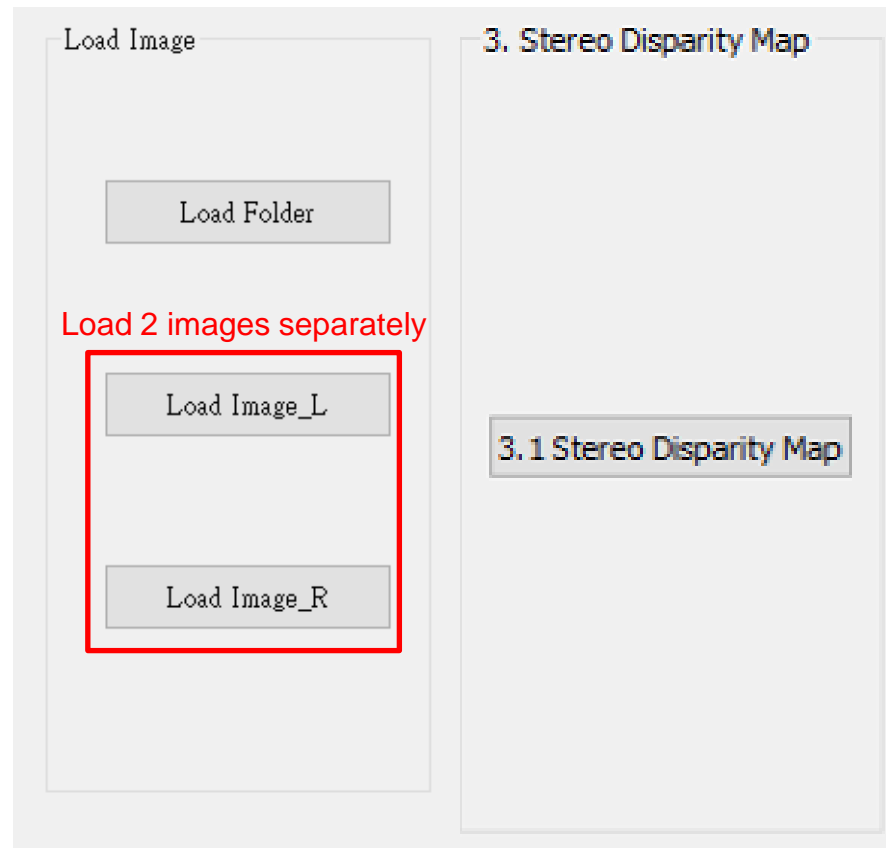


Position and Order

3. (20%) Stereo Disparity Map

3.1 (10%) Stereo Disparity Map

3.2 (10%) Checking the Disparity Value



3.1 (10%) Stereo Disparity Map

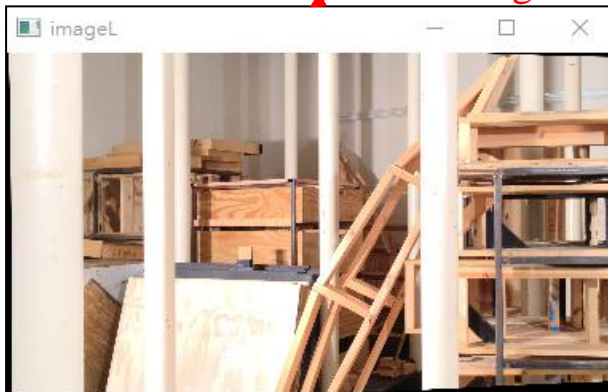
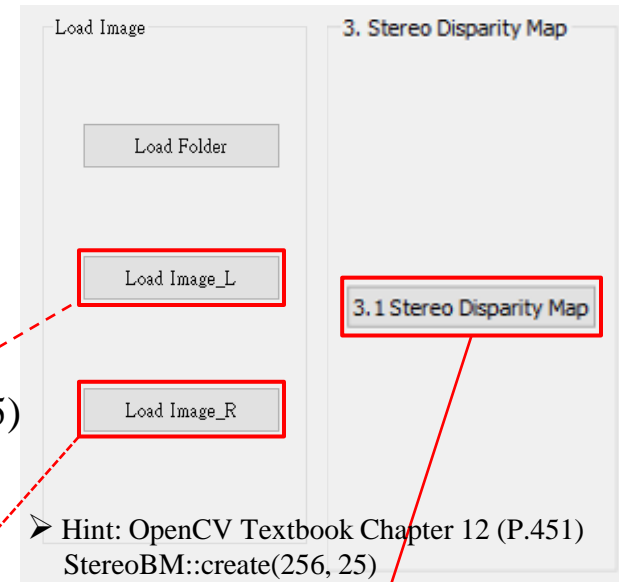
➤ Given: a pair of images, imL.png and imR.png (have been rectified).

Q: Find **the disparity map/image** based on Left and Right stereo images

- 1) Load imL.png (click “Load Image_L”).
- 2) Load imR.png (click “Load Image_R”).
- 3) Click button “3.1 Stereo Disparity Map” to show result.

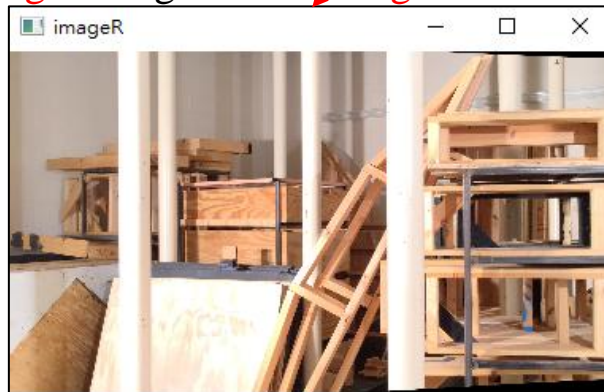
➤ Hint:

- Use OpenCV StereoBM class to build StereoBM objects.
- `stereo = cv2.StereoBM_create(numDisparities=256, blockSize=25)`
- The above parameters can be freely changed according to the following rules.
- **numDisparities**: The disparity search range, Must be **positive** and **divisible by 16**. In addition, map **disparity range** to **gray value range** 0~255 for the purpose of **visualization**.
- **blockSize**: The linear size of blocks compared by the algorithm, Must be **odd** and within the range **[5, 50]**. Larger block size implies smoother but less accurate disparity map. Smaller block size gives finer disparity details, yet increase the likelihood of algorithmic misalignment.
- If the **left image** is the **reference image** (the one used to cal. depth info for each pixel of that Img), then **the search direction** at **right image** will go **from the right to left** direction.



imL.png

Left Image (Reference Image)



imR.png

Right Image



Result

3.2 (10%) Checking the Disparity Value

(出題 : Eric)

➤ Given: a pair of images, imL.png and imR.png and disparity map from 3.1

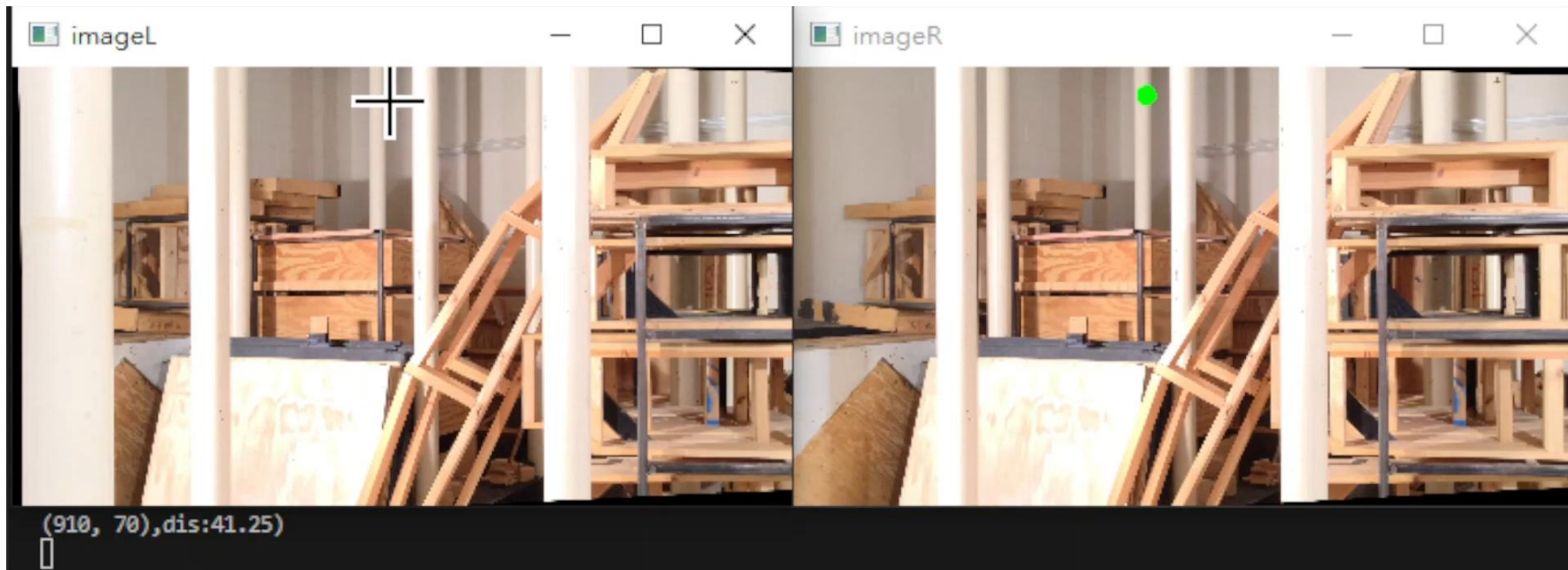
Q: Click at left image and **draw the corresponding dot** at right image.

1) Click at left image and draw the dot on the right image at accurate position.

2) User should be allowed to **repeat the action 1**).

➤ Hint: Click at gray position at disparity map result from Q3.1, ignore the position with 0 disparity (e.g. Failure case).

➤ Result Video:

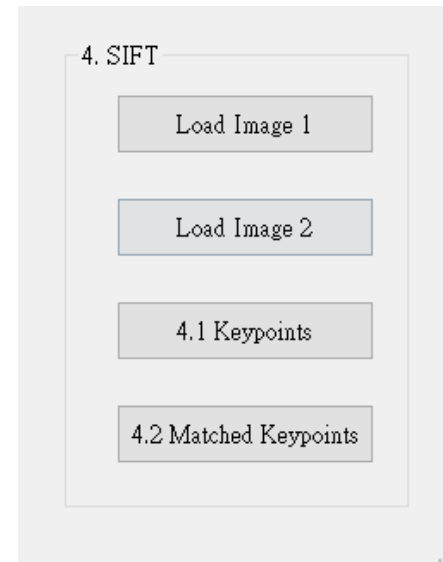


4. (20%) SIFT

4.1 (10%) Keypoints

4.2 (10%) Matched Keypoints

(出題：Chen)

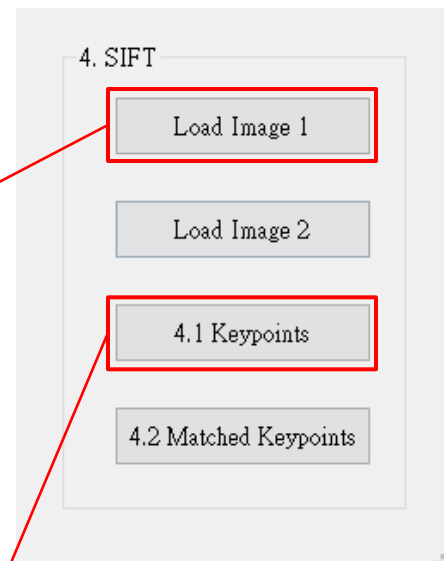


4.1 (10%) Keypoints

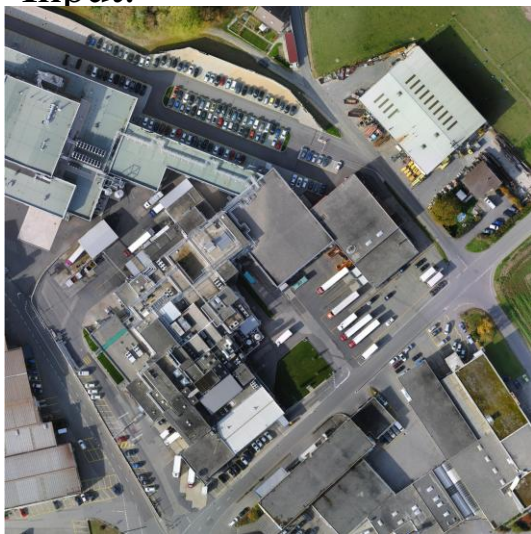
➤ Click button “4.1 Keypoints” to show:

- 1) Load Left.jpg (click “Load Image 1”)
- 2) Based on SIFT algorithm, find keypoints on Left.jpg.
 - Hint: Use OpenCV SIFT detector to detect keypoints and descriptors.
 - `sift = cv.SIFT_create()` # Create a SIFT detector
 - `sift.detectAndCompute(image, None)`
- 3) Then **draw the keypoints** of Left.jpg as I_1 .
 - Hint: `cv.drawKeypoints(gray, kp, None, color=(0,255,0))`

Please show image I_1

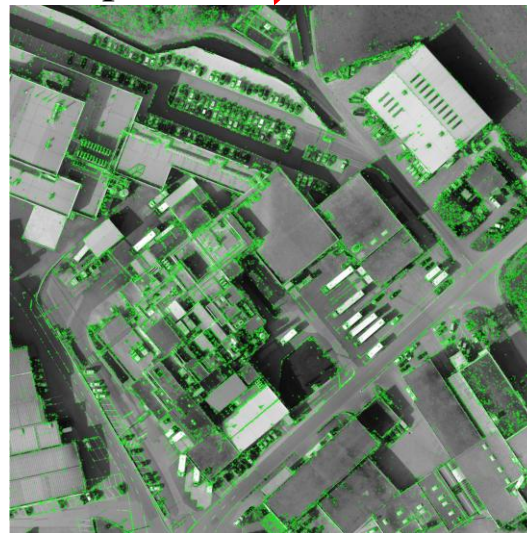


Input:



Left.jpg

Output:



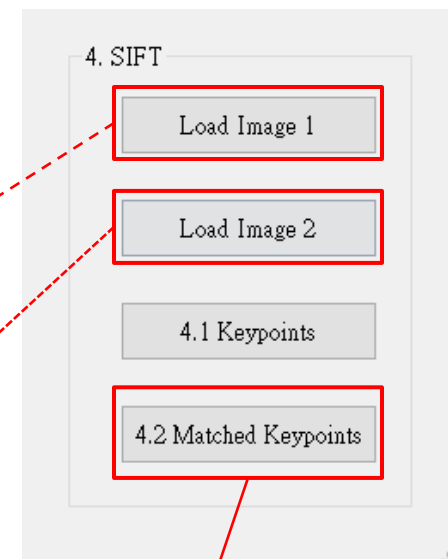
I_1

4.2 (10%) Matched Keypoints

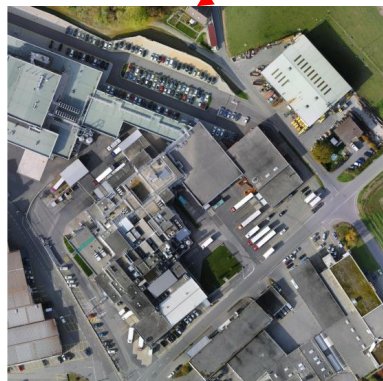
- Click button “4.2 Matched Keypoints”,
- 1) Load Image 1 (Left.jpg) and Image 2 (Right.jpg)
 - 2) Based on SIFT algorithm, find the keypoints and descriptors at Image 1 and Image 2 (same as question 1)
 - 3) Find match keypoint of two images
 - Hint: Use `BFMatcher.knnMatch(descriptors 1, descriptors 2,k=2)` to locate the matched keypoints.
 - 4) Extract Good Matches from 3) result:
 - Hint:

```
for m,n in matches:
    if m.distance < 0.75*n.distance:
        good.append([m])
```
 - 5) Draw the matched feature points between two image
 - Hint: Use “`cv2.drawMatchesKnn()`” to draw the matches,

```
cv.drawMatchesKnn(img1,kp1,img2,kp2,mat,good,flags=cv.DrawMatchesFlags_NOT_DRAW_SINGLE_POINTS)
```



Input:

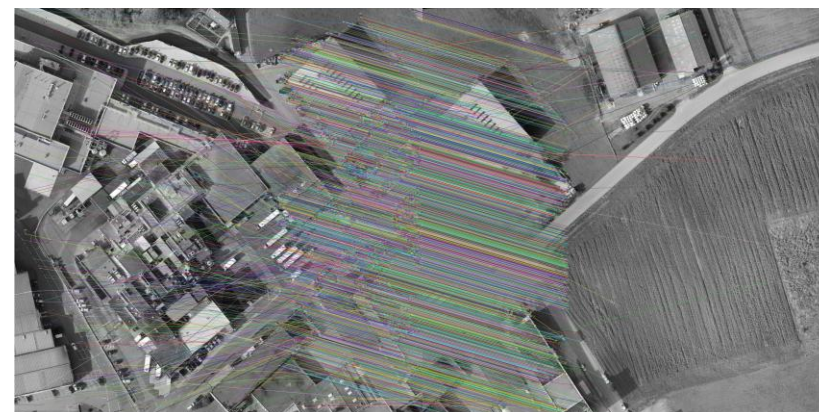


Left.jpg



Right.jpg

Output:



I_2

Please show image I_2

5. Training a CIFAR10 Classifier Using VGG19 with BN (20%)

5.1 Load CIFAR10 and show 9 Augmented Images with Labels. (4%) (出題：Hsiang)

5.2 Load Model and Show Model Structure. (4%)

5.3 Show Training/Validating Accuracy and Loss. (6%)

5.4 Use the Model with Highest Validation Accuracy to Run Inference, Show the Predicted Distribution and Class Label. (6%)

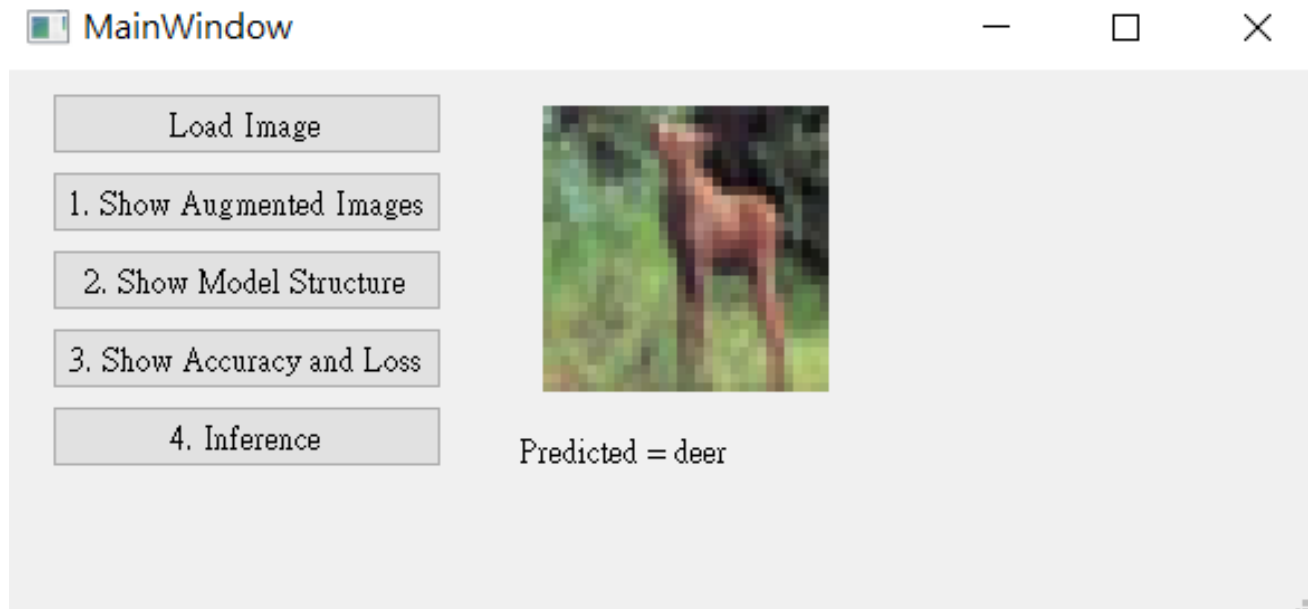


Figure: GUI example

5. Training a CIFAR10 Classifier Using VGG19 with BN (20%)

(出題：Hsiang)

1. Objective

- 1) Learn how to train a VGG19 with BN (Batch Normalization) model to **classify 10 different classes images** of CIFAR10.

2. VGG19 with BN

- 1) VGG19: A convolutional neural network that is 19 layers deep.
- 2) BN (Batch Normalization): used to make training of artificial neural networks faster and more stable.

3. CIFAR10

- 1) A collection of 60,000 **32x32 color images** in **10 different classes** that is commonly used to train machine learning and computer vision algorithms.
- 2) 10 classes:
airplane, automobile, bird, cat, deer, dog, frog, horse, ship, truck
- 3) Datasets
 - (1) Training dataset: **50000** images in total.
 - (2) Validation dataset: **10000** images in total.
 - (3) Testing dataset: **10 images in total**. (Generating from validation dataset.)

airplane
automobile
bird
cat
deer
dog
frog
horse
ship
truck

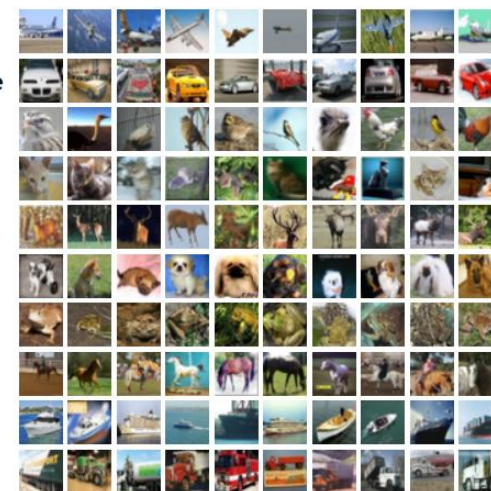


Figure1: CIFAR10

R. Reference

- 1) [VGG19](#)
- 2) [Batch Normalization](#)

5. Training a CIFAR10 Classifier Using VGG19 with BN (20%)

(出題：Hsiang)

4. Requirements

- 1) Train VGG model with batch normalization (BN) using **PyTorch**.
- 2) In the submitted file, you need to include
 - A. Weight file for VGG19 with BN in **.pth** format. (File size is approximately 540MB)
 - B. Figure of training/validating loss and accuracy in **.jpg** or **.png** format.
 - C. Code for your GUI program
 - D. Code for model training.
- 3) **Please do not include image data in the submitted file.**

5. Homework Images

- 1) There are 2 different folders in 'Q5_image'.
- 2) In the subfolder 'Q5_image/Q5_1,' there are 9 different images used in Q5-1. When demoing, use the same images.
- 3) In the subfolder 'Q5_image/Q5_4,' there are 9 different images used in Q5-4. These images are used for testing your program. When demoing, we will use different images for the demonstration.

5.1 Show 9 Augmented Images with Labels (3%)

(出題：Hsiang)

Q5.1

1) At home:

- (1) Use [PIL.Image.open\(\)](#) to load 9 images in `/Q5_image/Q5_1/` folder.
- (2) Apply **at least 3 different** type of data augmentation ([tutorial](#)).
 - A. `transforms.RandomHorizontalFlip()`
 - B. `transforms.RandomVerticalFlip()`
 - C. `transforms.RandomRotation(30)`

Notice: This is an example; you can use different data augmentation techniques

2) When the demo:

- (1) Click the button “1. Show Augmentation Images”
- (2) Load 9 images in `/Q5_image/Q5_1/` folder
- (3) Apply data augmentation on 9 images.
- (4) Show 9 **augmented images with label** in a new window

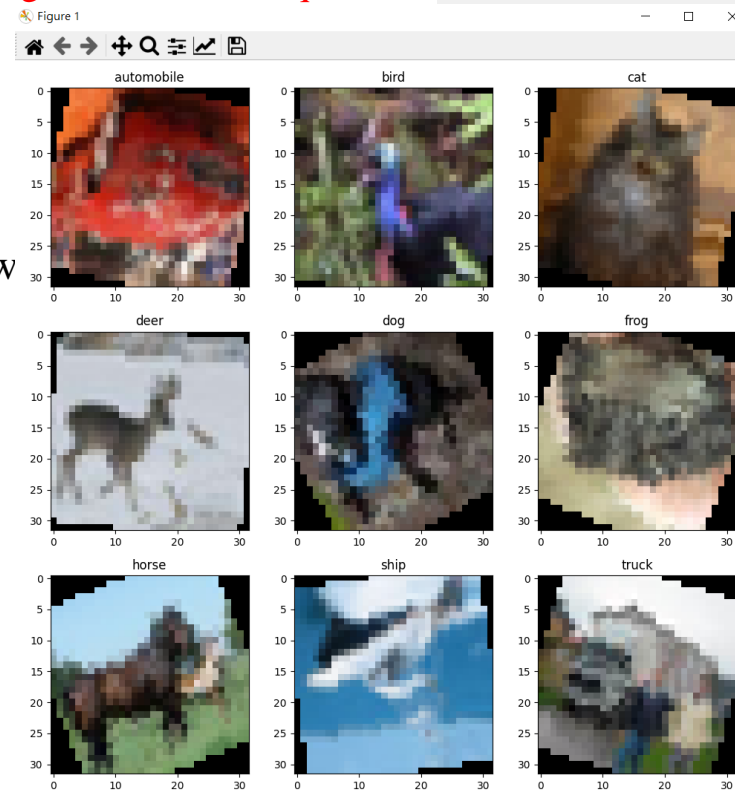
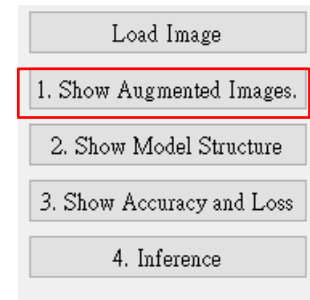


Figure1: 9 Augmented images

Notice: this is an example, the images might differ

5.2 Show the Structure of VGG19 with BN (3%)

Q5.2

1) At home:

- (1) Use `torchvision.models.vgg19_bn(num_classes=10)` to build a VGG19 with batch normalization (BN) model.
- (2) Use `torchsummary.summary` to show the structure in the terminal.

2) When the demo:

- (1) Click the button "2. Show Model Structure"
- (2) Run the function to show the structure in the terminal.

The -1 indicates that the actual size of batch size can vary.

Load Image

1. Show Augmented Images.

2. Show Model Structure

3. Show Accuracy and Loss

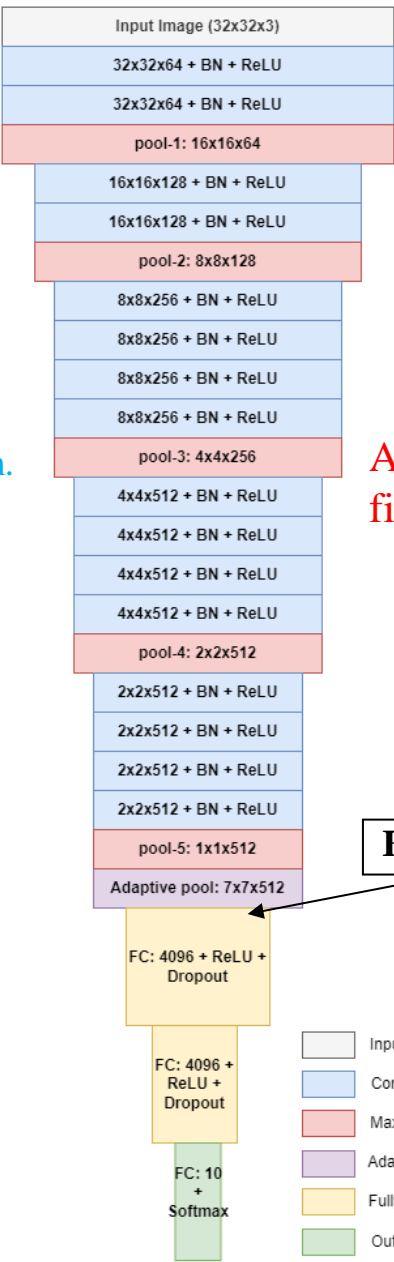
4. Inference

Layer (type)	Feature map shape (Batch, Channels, Height, Width)	Num. of param.
BatchNorm2d-38	[-1, 512, 4, 4]	1,024
ReLU-39	[-1, 512, 4, 4]	0
MaxPool2d-40	[-1, 512, 2, 2]	0
Conv2d-41	[-1, 512, 2, 2]	2,359,808
BatchNorm2d-42	[-1, 512, 2, 2]	1,024
ReLU-43	[-1, 512, 2, 2]	0
Conv2d-44	[-1, 512, 2, 2]	2,359,808
BatchNorm2d-45	[-1, 512, 2, 2]	1,024
ReLU-46	[-1, 512, 2, 2]	0
Conv2d-47	[-1, 512, 2, 2]	2,359,808
BatchNorm2d-48	[-1, 512, 2, 2]	1,024
ReLU-49	[-1, 512, 2, 2]	0
Conv2d-50	[-1, 512, 2, 2]	2,359,808
BatchNorm2d-51	[-1, 512, 2, 2]	1,024
ReLU-52	[-1, 512, 2, 2]	0
MaxPool2d-53	[-1, 512, 1, 1]	0
AdaptiveAvgPool2d-54	[-1, 512, 7, 7]	0
Linear-55	[-1, 4096]	102,764,544
ReLU-56	[-1, 4096]	0
Dropout-57	[-1, 4096]	0
Linear-58	[-1, 4096]	16,781,312
ReLU-59	[-1, 4096]	0
Dropout-60	[-1, 4096]	0
Linear-61	[-1, 10]	40,970

Total params: 139,622,218
Trainable params: 139,622,218
Non-trainable params: 0

Input size (MB): 0.01
Forward/backward pass size (MB): 7.55
Params size (MB): 532.62
Estimated Total Size (MB): 540.18

Figure: the Structure of VGG19 with BN



All convolution filter size is 3x3

Flatten Here

Figure: VGG19 with BN model structure

5.3 Show Training/Validating Accuracy and Loss (6%)

(出題：Hsiang)

Q5.3

1) At home:

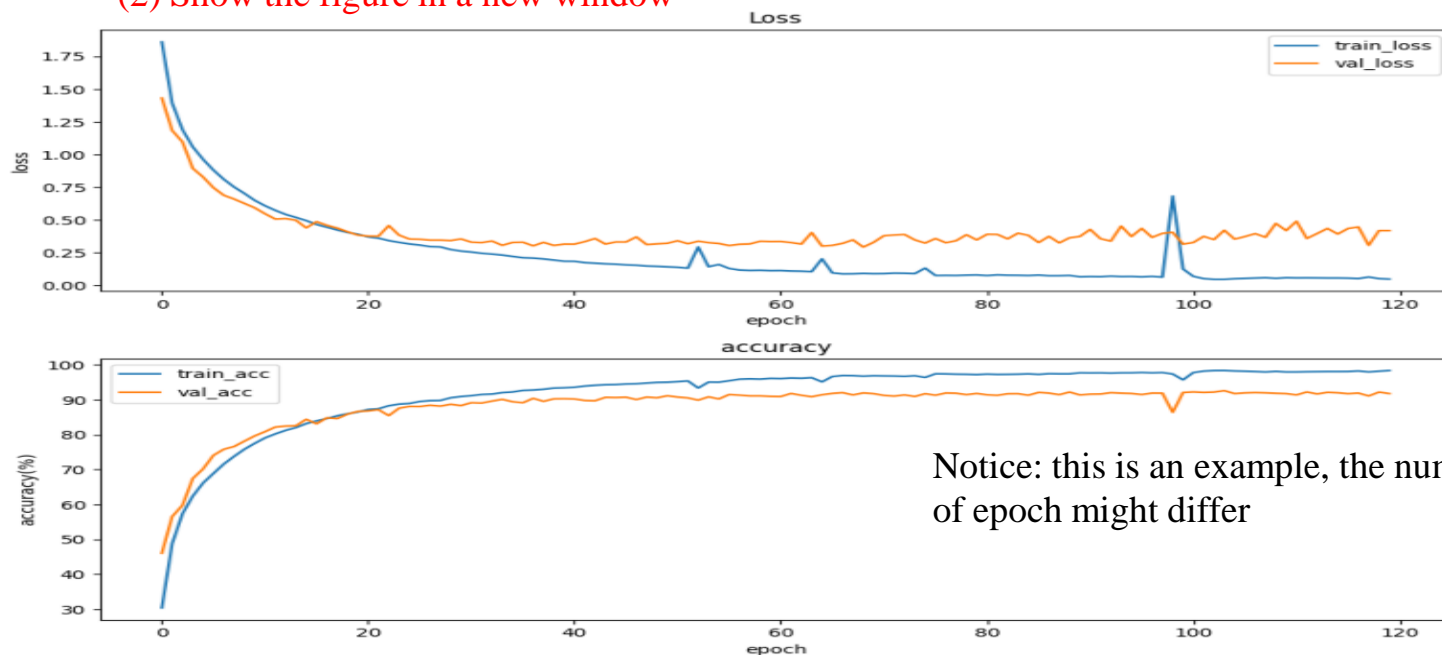
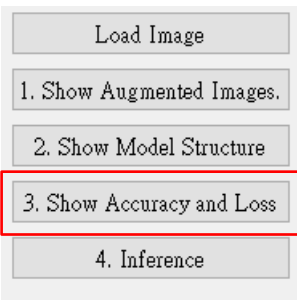
- (1) Use `torchvision.datasets.CIFAR10` to load the training and validation datasets. ([tutorial](#))
- (2) Training and validating VGG19 with BN at least 40 epochs at home ([tutorial](#)) and record the training/validating accuracy and loss in each epoch ([tutorial](#)).
- (3) Notice: If your validation accuracy is low, you can try
 - A. Adjust the **learning rate** of the optimizer.
 - B. Change the **data augmentation** techniques used.
- (4) Save weight file with highest validation accuracy .
- (5) Use [matplotlib.pyplot.plot\(\)](#) to create a line chart for the **training and validating loss and accuracy** values.
- (6) Save the figure

2) When the demo:

- (1) Click the button “3. Show Accuracy and Loss”
- (2) Show the **saved figure** of Training/Validating loss and accuracy in a new window

(2) Show the figure in a new window

(1) Click the button.



5.4 Use the Model with Highest Validation Accuracy to Run Inference

Show the Predicted Distribution and Class Label. (6%) (出題：Hsiang)

Q5.4

1) At home:

- (1) Load the model which trained at home
- (2) Click the button “Load Image” to display a new file selection dialog
- (3) Select 1 image arbitrarily.
- (4) Show the loaded image on the GUI. (In order to make it visually clear on the UI, use [QtGui.QPixmap.scaled](#) to scale the image to 128x128 when displaying it.)
- (5) Click the button “4. Inference” to run inference on the image. ([tutorial](#))
- (6) Show the predicted class label on the GUI.
- (7) Show the probability distribution of model predictions using a histogram in a new window.

2) When the demo: repeat the process

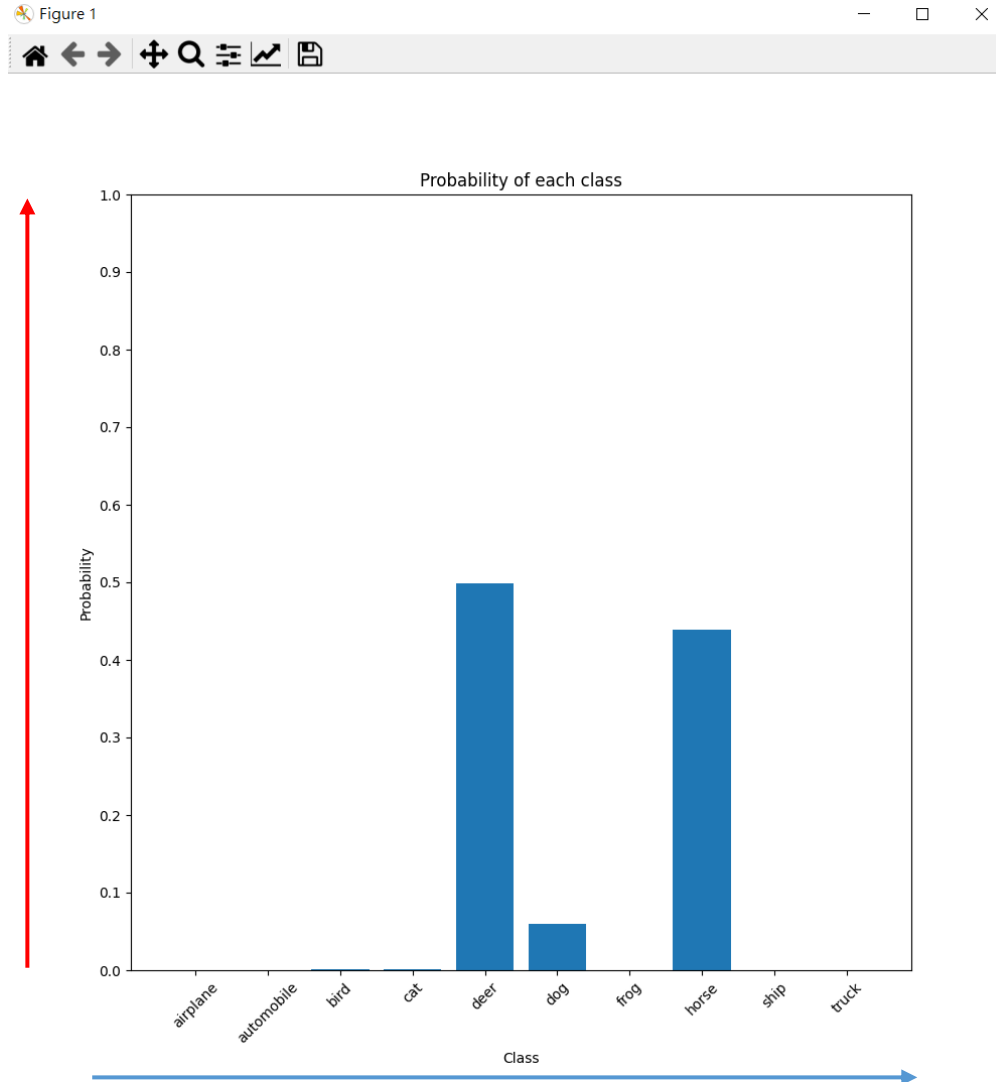
The screenshot shows the GUI with several components and annotations:

- File Selection Dialog:** A file explorer window titled "Open file" showing a directory with various image files. The file "cat.png" is selected, indicated by a red box and the annotation "(3) Select 1 image arbitrarily".
- MainWindow:** The main application window with a title bar "MainWindow". It contains a "Load Image" button (highlighted with a red box and annotation "(2) Click the button."), a list of buttons: "1. Show Augmented Images.", "2. Show Model Structure", "3. Show Accuracy and Loss", and "4. Inference" (highlighted with a red box and annotation "(5) Run inference on the loaded image").
- Loaded Image:** A small image of a deer is displayed on the GUI, highlighted with a red box and annotation "(4) Show the loaded image (When displaying loaded image, resizing the loaded image to 128x128)".
- Predicted Class Label:** Below the image, the text "Predicted = deer" is shown, with "deer" highlighted by a red box and annotation "(6) Show the predicted class label".
- Probability Distribution Histogram:** A new window titled "Figure 1" displays a bar chart titled "Probability of each class". The x-axis is labeled "Class" and lists: airplane, automobile, bird, cat, deer, dog, frog, horse, ship, truck. The y-axis is labeled "Probability" and ranges from 0.0 to 1.0. The bars show the following approximate probabilities: airplane (0.0), automobile (0.0), bird (0.0), cat (0.0), deer (0.5), dog (0.05), frog (0.0), horse (0.45), ship (0.0), truck (0.0).

(7) Show the probability distribution of model prediction in new window.

5.4 Use the Model with Highest Validation Accuracy to Run Inference Show the Predicted Distribution and Class Label. (6%) (出題：Hsiang)

- The probability distribution of model prediction using a histogram.



Y-axis: probability

X-axis: class name

5. Training a CIFAR10 Classifier Using VGG19 – Example Video

(出題：Hsiang)

- This is an example illustrating the objectives from 5.1 ~ 5.4.

