

影像處理、電腦視覺及深度學習概論 (Introduction to Image Processing, 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, 2023/11/07 (Tue.)**
 - 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: opencvdl2023 Password: RL2023opencvdl**
- 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%) Image Processing (出題：Chen)

- 1.1 (5%) Color Separation
- 1.2 (5%) Color Transformation
- 1.3 (5%) Color Extraction

2. (20%) Image Smoothing (出題：Shang)

- 2.1 (6%) Gaussian blur
- 2.2 (7%) Bilateral filter
- 2.3 (7%) Median filter

3. (20%) Edge Detection (出題：Zhong)

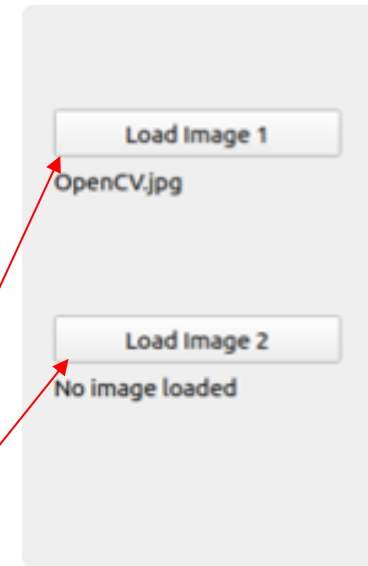
- 3.1 (5%) Sobel X
- 3.2 (5%) Sobel Y
- 3.3 (5%) Combination and Threshold
- 3.4 (5%) Gradient Angle

4. (20%) Transforms (出題：Jimmy)

- 4.1 (7%) Rotation
- 4.2 (7%) Scaling
- 4.3 (6%) Translate

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

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



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

Load image 請用下面Function 來讀取路徑
QFileDialog.getOpenFileName
獲取打開的檔路徑

Assignment scoring (Total: 100%)

- Use one UI to present 5 questions.

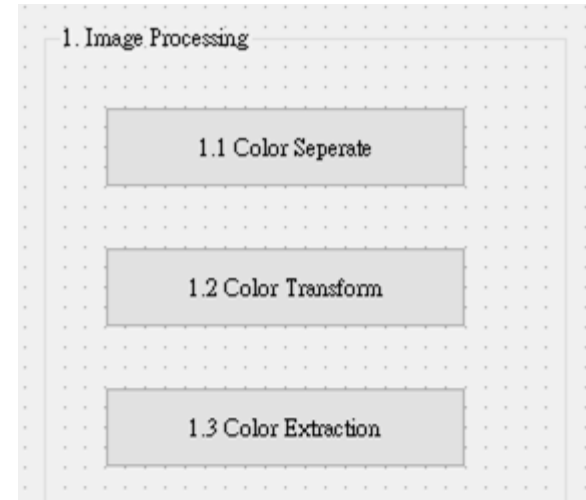


1. Image Processing (20%)

1.1 (6%) Color Separation

1.2 (6%) Color Transformation

1.3 (8%) Color Extraction

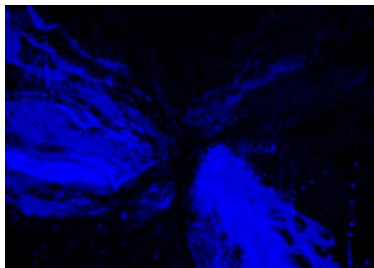


1.1 Color Separation (6%)

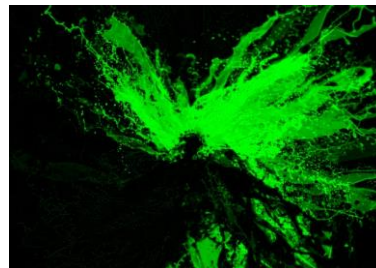
- Given: a color image, “rgb.jpg”
- Q: Extract 3 channels of the image **BGR** and show the result images.
 - 1) Use `cv2.split()` to get R G B gray scale images.
 - 2) Use `cv2.merge()` to turn each gray scale image back to bgr image.

Please show each R ,G ,B Image.

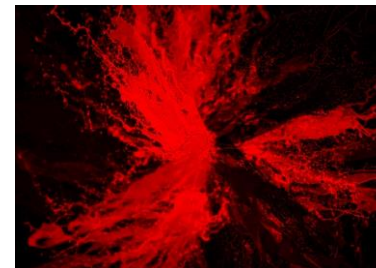
rgb.jpg



B channel



G channel



R channel



1.2 Color Transformation (6%)

➤ Given: 1 color image: “rgb.jpg”

➤ Q: Transform “rgb.jpg” into grayscale by

Q1): Calling OpenCV function `cv2.cvtColor(..., cv2.COLOR_BGR2GRAY)` on rgb.jpg to generate Image I_1

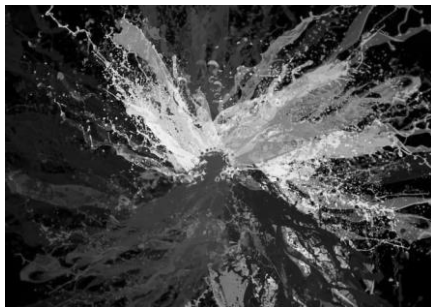
Q2): Merge **BGR** separated channel images from problem 1.1 to generate $I_2 = (R+G+B)/3$.

Please show above 2 images results.

rgb.jpg

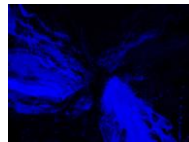


1) I_1 : OpenCV function

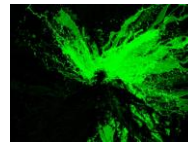


Perceptually weighted formula:
 $I_1 = 0.299 R + 0.587 G + 0.114 B$

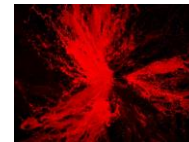
Result from problem 1.1



B channel



G channel



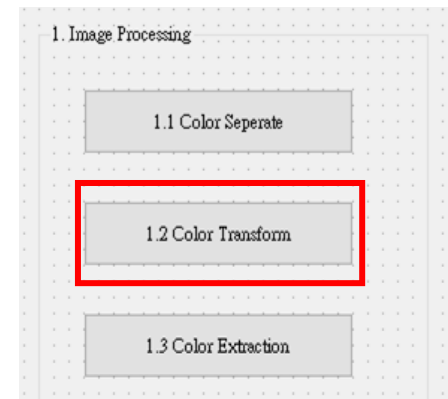
R channel



2) I_2 : Average



Average weighted formula:
 $I_2 = (R+G+B)/3$

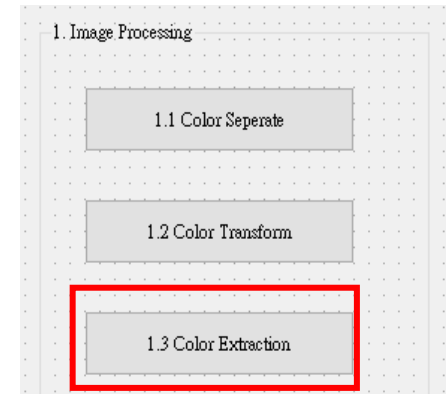
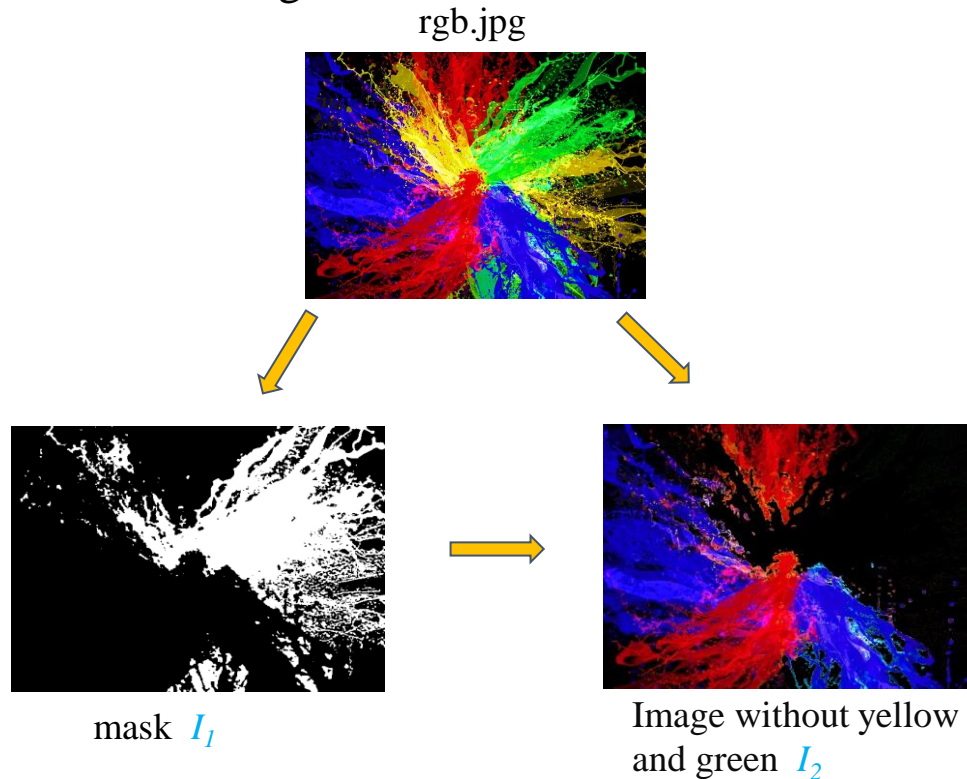


1.3 Color Extraction (8%)

➤ Given: 1 color image: “rgb.jpg”

- 1) Transform “rgb.jpg” from **BGR** format to **HSV** format: `cv2.cvtColor(bgr img, cv2.COLOR_BGR2HSV)`
- 2) Extract **Yellow-Green** I_1 mask by calling : `cv2.inRange(hsv img , lower bound , upper bound)`
- 3) Turn **Yellow-Green** mask into BGR format by calling : `cv2.cvtColor(..., cv2.COLOR_GRAY2BGR)`
- 4) Remove **Yellow** and **Green** color in the image to generate I_2 by calling : `cv2.bitwise_not(mask bgr ,bgr img ,mask)`

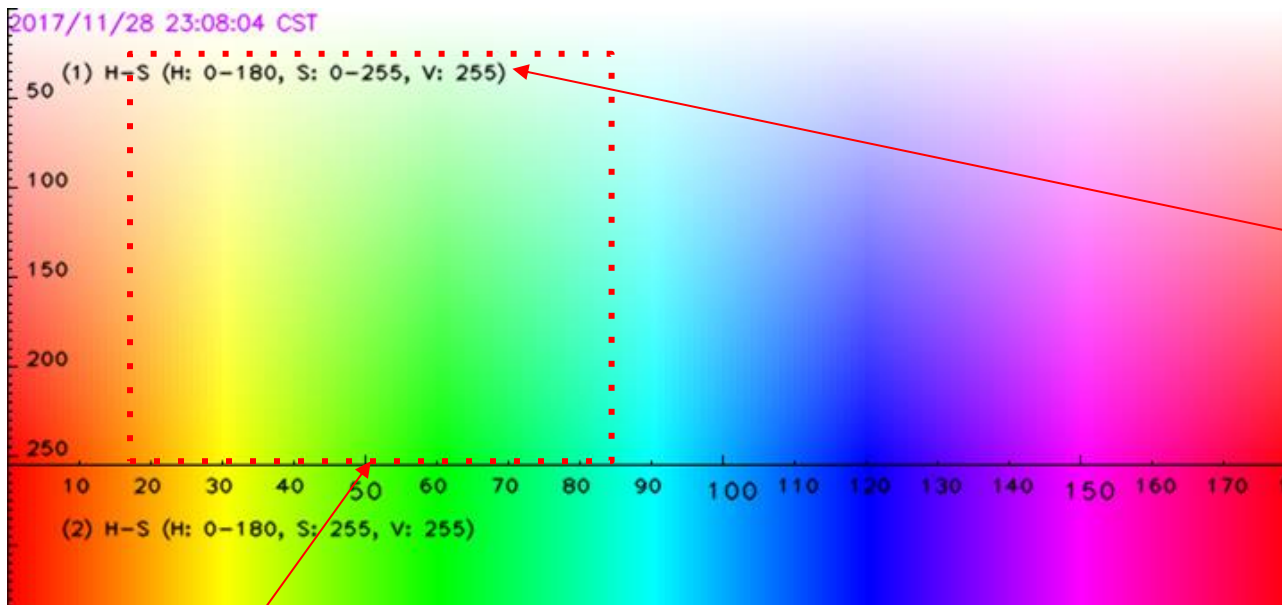
Please show above 2 image results.



1.3 Color Extraction (8%)

➤ Hint:

- `cv2.inRange(hsv_img , lower_bound(h,s,v=25) , upper_bound (h,s,v=255))`
 - Hue and Saturation please check below chart area , Value is between (25,255)



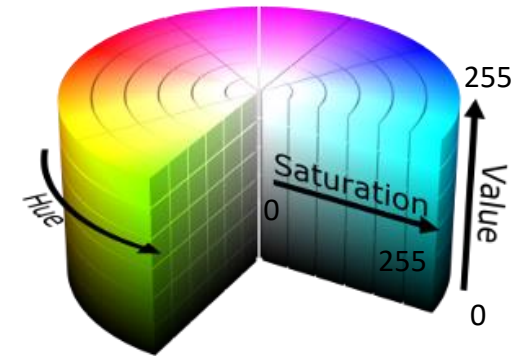
Yellow-Green mask range **Slight difference on range should be fine**

**HSV values ranges between
(h:0-180, s:0-255, v:0-255)**

H(Hue) : x axis

S(Saturation) : y axis

V(Value) : 255

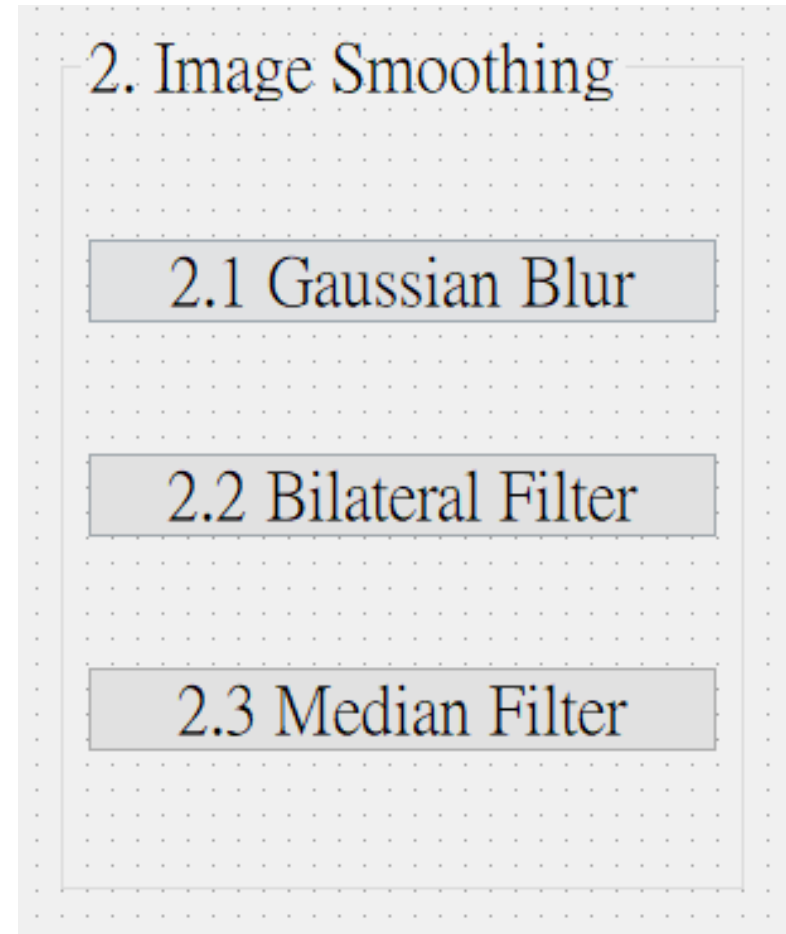


2. Image Smoothing (20%)

2.1 (6%) Gaussian blur

2.2 (7%) Bilateral filter

2.3 (7%) Median filter

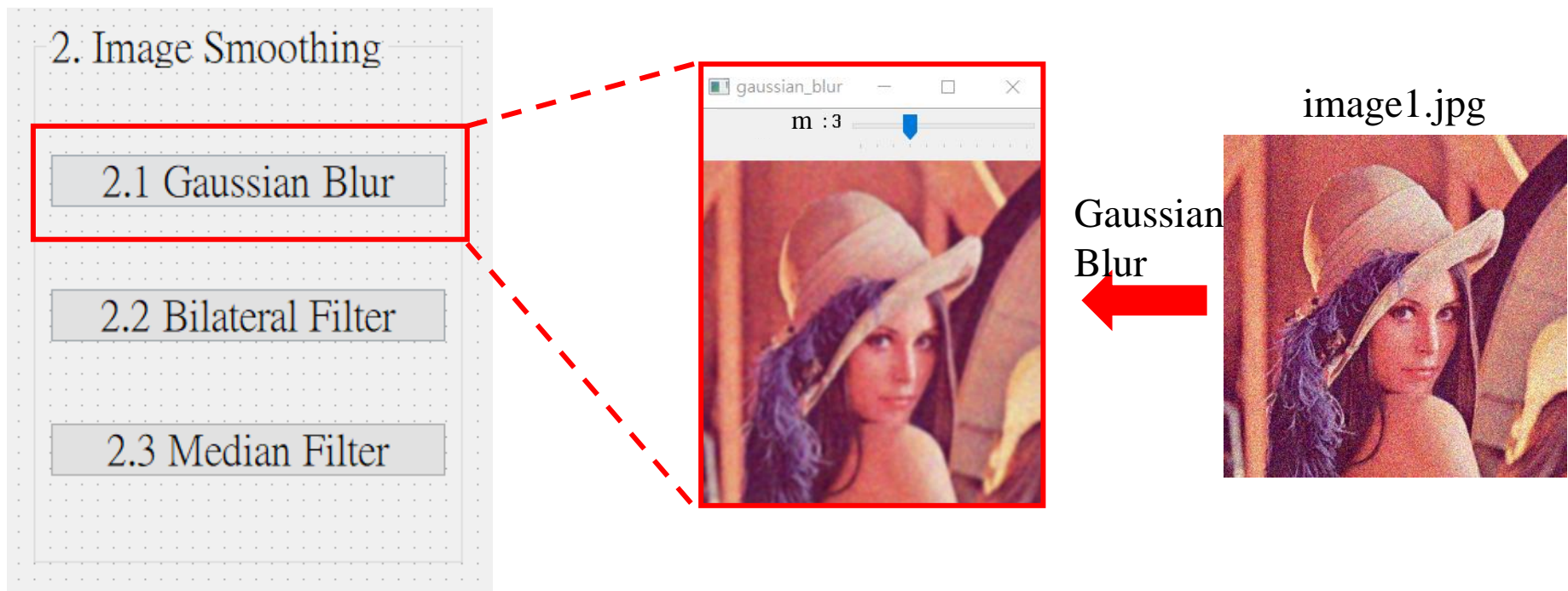


- Hint
 - 1) Textbook Chapter 3, p. 50 ~ 52, p.109~115

2.1 Gaussian Blur

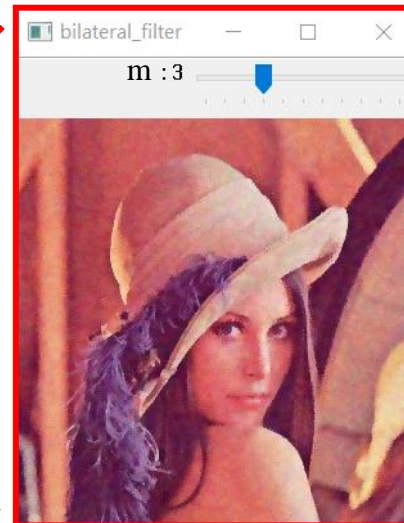
(出題 :
Shang)

1. Given: "image1.jpg"
2. Requirement:
 - 1) Using "Load image 1" button to load image.
 - 2) Click "2.1 Gaussian Blur" to show the popup window.
 - 3) Using `cv2.createTrackbar()` to create a trackbar on popup window.
 - 4) Using trackbar to change the **window radius** (m).
 - 5) The range of radius size is `[1, 5]`.
 - 6) Apply gaussian filter (`cv2.GaussianBlur()`) which kernel size is $(2m + 1) \times (2m + 1)$



2.2 Bilateral Filter

1. Given: “image1.jpg”
 2. Requirement:
 - 1) Using “Load image 1” button to load image.
 - 2) Click “2.2 Bilateral Filter” to show the popup window.
 - 3) Using `cv2.createTrackbar()` to create a trackbar on popup window.
 - 4) Using trackbar to change the **window radius** (m).
 - 5) The range of radius size is [1, 5].
 - 6) Apply bilateral filter (`cv2.bilateralFilter()`) which kernel size is $(2m + 1) \times (2m + 1)$ to
- Hint
 - 1) `simgaColor = 90, sigmaSpace = 90`



Bilateral
Blur



3. Edge Detection (20%)

3.1 (5%) Sobel X

3.2 (5%) Sobel Y

3.3 (5%) Combination and Threshold

3.4 (5%) Gradient Angle

3. Edge Detection

3.1 Sobel X

3.2 Sobel Y

3.3 Combination and Threshold

3.4 Gradient Angle

3.1 Sobel x (5%)

1. Given: A RGB image, “building.jpg”
2. Q: Generate Sobel x image for “building.jpg”
 - 1) Convert the RGB image into a grayscale image
 - 2) Smooth grayscale image with Gaussian smoothing filter.
 - 3) Use Sobel edge detection to detect **vertical edge** by your own 3x3 Sobel x operator. (**Can not use OpenCV Function cv2.Sobel and cv2.filter2D.**)
 - 4) Please show the result **with cv2.imshow** function.
- Hint: Textbook Chapter 6, p.144 ~ 149

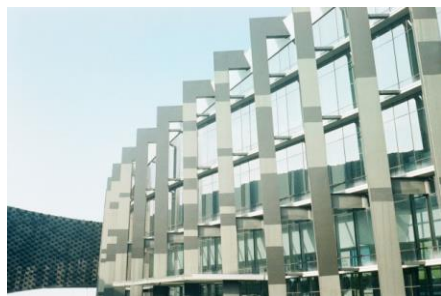
3. Edge Detection

3.1 Sobel X

3.2 Sobel Y

3.3 Combination and Threshold

3.4 Gradient Angle



building.jpg



Grayscale



Gaussian Smoothing

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



Sobel x Filter



Sobel x

3.2 Sobel y (5%)

1. Given: A RGB image, “building.jpg”
 2. Q: Generate Sobel y image for “building.jpg”
 - 1) Convert the RGB image into a grayscale image
 - 2) Smooth grayscale image with Gaussian smoothing filter.
 - 3) Use Sobel edge detection to detect **horizontal edge** by your own 3x3 Sobel x operator. (**Can not use OpenCV Function cv2.Sobel and cv2.filter2D.**)
 - 4) Please show the result **with cv2.imshow** function.
- Hint: Textbook Chapter 6, p.144 ~ 149

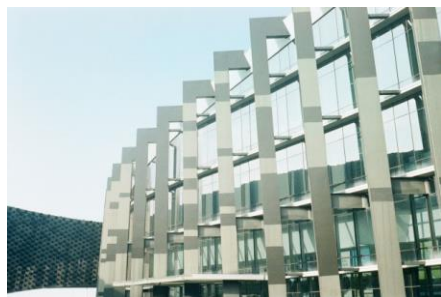
3. Edge Detection

3.1 Sobel X

3.2 Sobel Y

3.3 Combination and Threshold

3.4 Gradient Angle



building.jpg



Grayscale

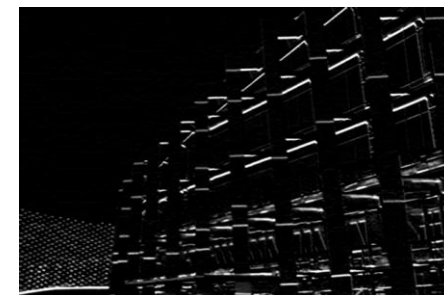


Gaussian Smoothing

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



Sobel y Filter



Sobel y

3.3 Combination and Threshold (5%)

1. Given: The result of 3.1) Sobel x and 3.2) Sobel y
 2. Q: Combine Sobel x and Sobel y, then set threshold for result
 - 1) New value of pixel = $\sqrt{Sobel_x^2 + Sobel_y^2}$
 - 2) Normalize combination result to 0~255
 - 3) Given threshold **128**. Set to 0 if pixel value is lower than threshold, otherwise, set to 255.
 - 4) Show both **combination** and **threshold result** with **cv2.imshow** function. Two results should be shown together.
- Hint: Textbook Chapter 6, p.148 ~ 149

3. Edge Detection

3.1 Sobel X

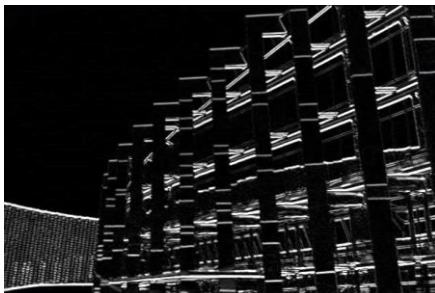
3.2 Sobel Y

3.3 Combination and Threshold

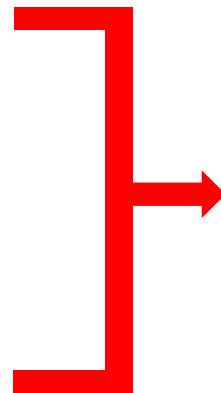
3.4 Gradient Angle



Sobel x



Sobel y



Combination of
Sobel x and Sobel y



Threshold result

3.4 Gradient Angle (5%)

1. Given: The result of 3.1) Sobel x and 3.2) Sobel y
2. Q: Calculate the gradient angle and show specific range of angle.
 - 1) Calculate the gradient angle by result of Sobel x and Sobel y
 - 2) Generate **two different masks** by given two different range of angle (1) $120^\circ \sim 180^\circ$ (2) $210^\circ \sim 330^\circ$. Set to 255 if pixel value is in range, otherwise set to 0.
 - 3) Generate results by calling `cv2.bitwise_and(combination, mask)` which **combination** is the result form 3.3
 - 4) Show both results **with `cv2.imshow`** function. Two results should be shown together.

3. Edge Detection

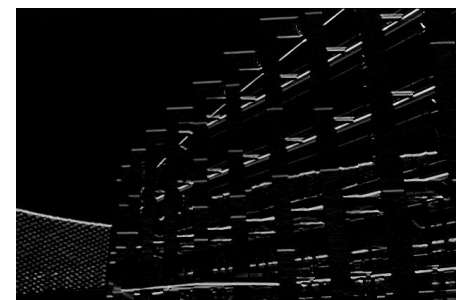
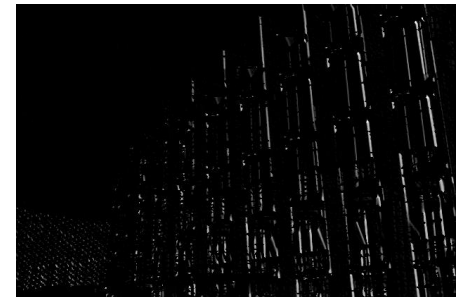
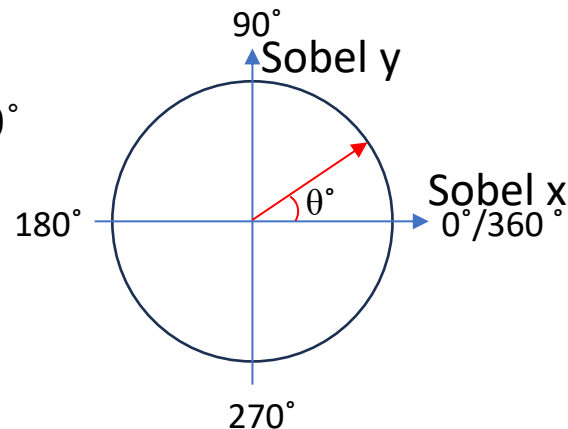
3.1 Sobel X

3.2 Sobel Y

3.3 Combination and Threshold

3.4 Gradient Angle

Hint: Gradient angle θ°



Output image

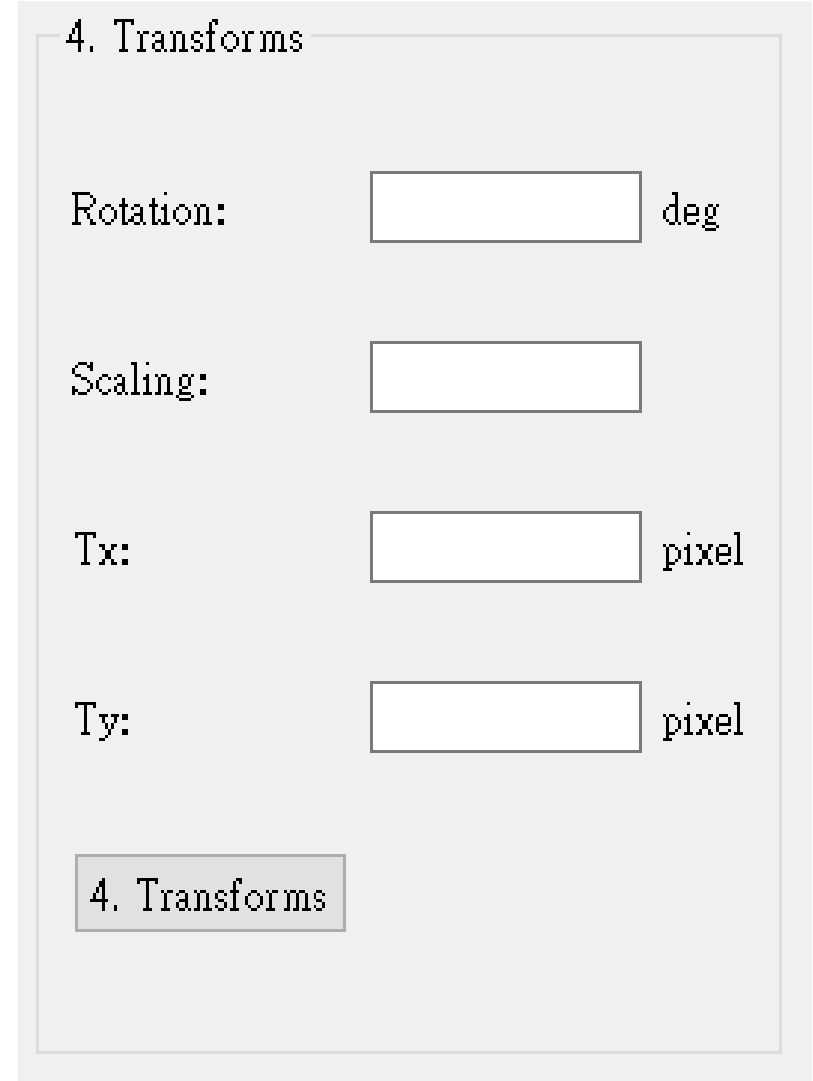
4. Transforms (20%)

4.1 (7%) Rotation

4.2 (7%) Scaling

4.3 (6%) Translate

UI Demo:



A UI demo for a '4. Transforms' panel. The panel has a title bar '4. Transforms' and contains four input fields with labels and units: 'Rotation:' with a unit of 'deg', 'Scaling:' with no unit, 'Tx:' with a unit of 'pixel', and 'Ty:' with a unit of 'pixel'. Each input field is a white rectangle with a thin black border. At the bottom of the panel is a button labeled '4. Transforms'.

4. Transforms

Rotation: deg

Scaling:

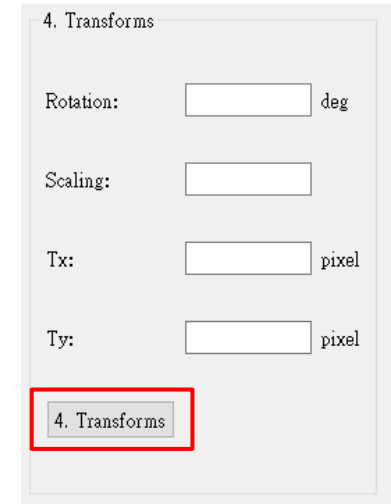
Tx: pixel

Ty: pixel

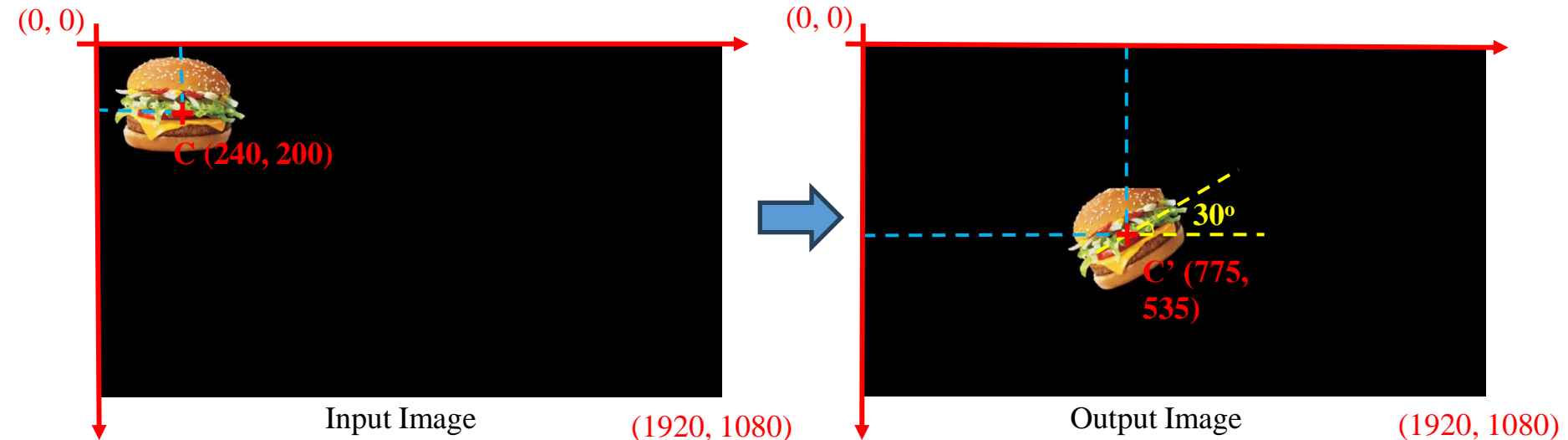
4. Transforms

4. Transforms (20%)

- Given: “burger.png”
 - Q: 1) Click button “4. Transforms”, burger.png should be showed.
2) Please **rotate**, **scale** and **translate** the burger (as image below) using **cv.warpAffine()** function with following parameters (set default values 0, should be manually adjusted in the GUI)
 - Angle = 30° (positive degree ☐ counter-clockwise)
 - Scale = 0.9,
 - Translation with:
 - $X_{new} = X_{old} + 535$ pixels = $240 + 535 = 775$
 - $Y_{new} = Y_{old} + 335$ pixels = $200 + 335 = 535$
 - Point C (240, 200) is center point of burger in original image
 - Point C' (775, 535) is center point of burger in result image
- Hint: Textbook Chapter 12, (p.407 ~ 412)
python: cv.warpAffine()
- Rotation, Scale: Object center not move
 - Translation: Object center move



Result Demo:



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%)

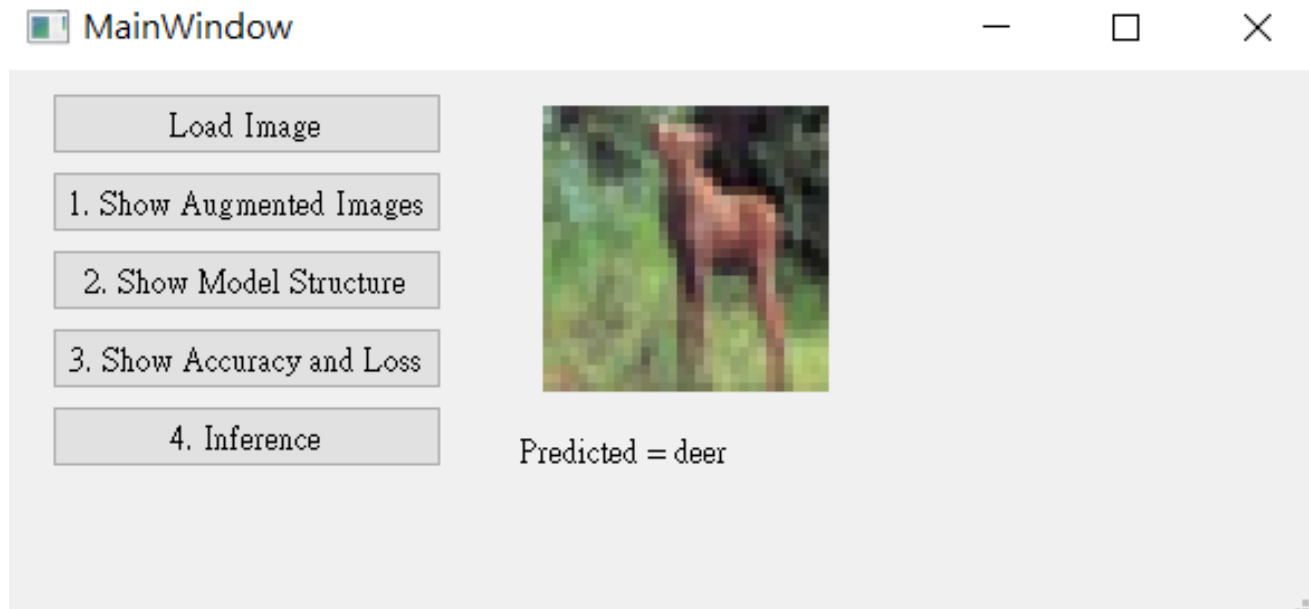


Figure: GUI example

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.)

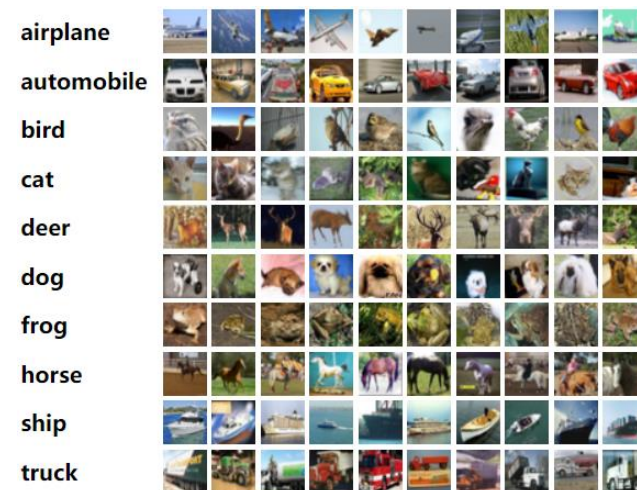


Figure1: CIFAR10

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

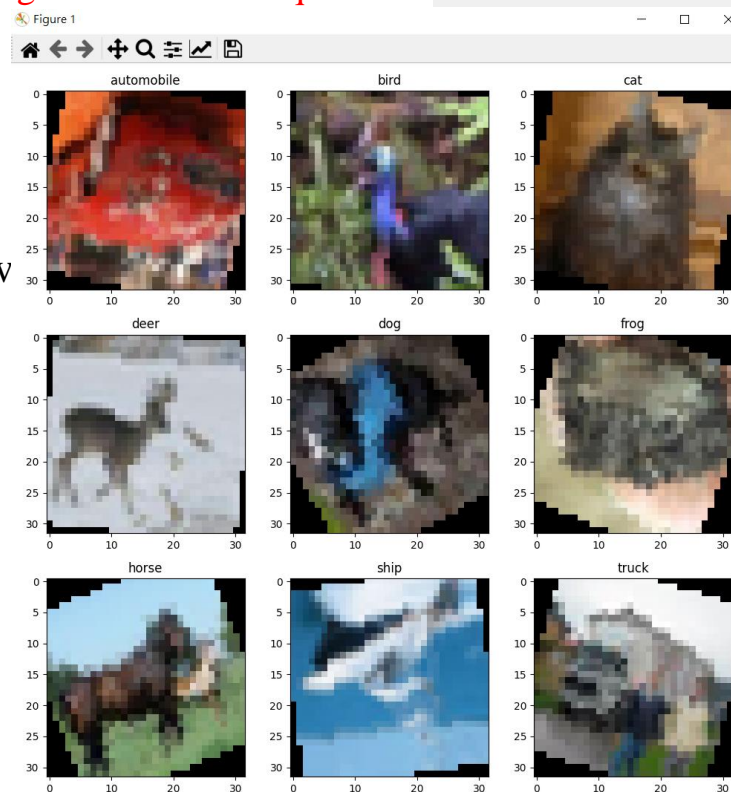
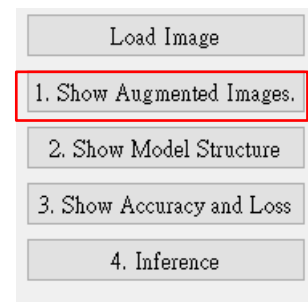


Figure1: 9 Augmented images

Notice: this is an example, the images might differ

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

Q5.2

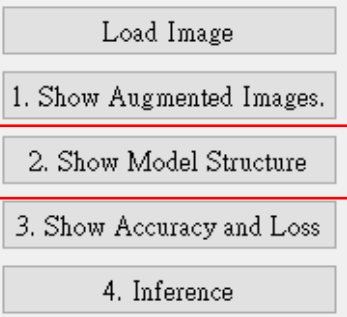
1) At home:

- (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.

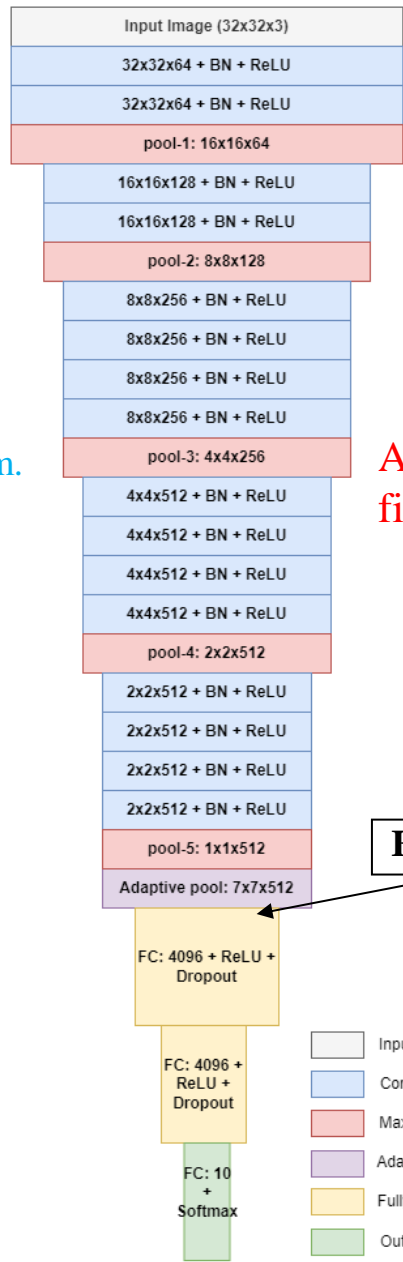


Layer (type)	Feature map shape (Batch, Channels, Height, Width)	Num. of param.
BatchNorm2d-38	[-1, 512, 4, 4]	1,024
ReLU-39	[-1, 512, 4, 4]	0
MaxPool2d-40	[-1, 512, 2, 2]	0
Conv2d-41	[-1, 512, 2, 2]	2,359,808
BatchNorm2d-42	[-1, 512, 2, 2]	1,024
ReLU-43	[-1, 512, 2, 2]	0
Conv2d-44	[-1, 512, 2, 2]	2,359,808
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 26

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.

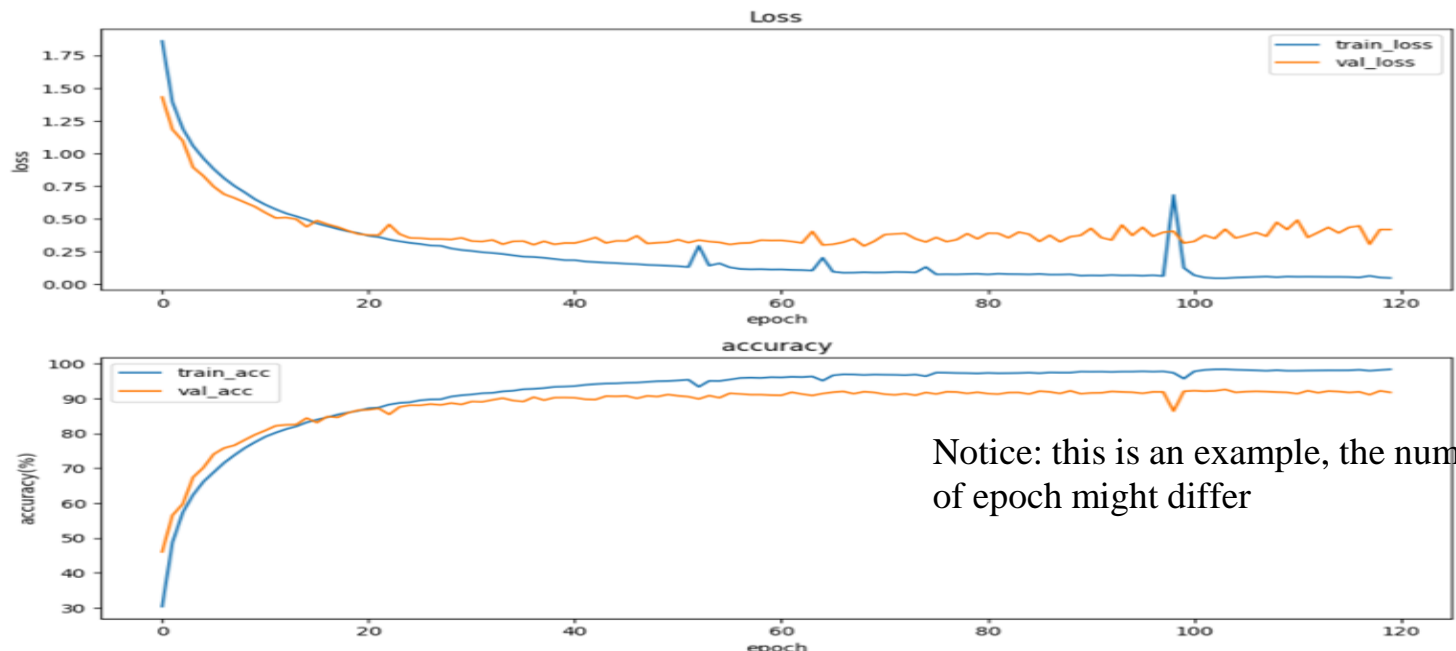
Load Image

1. Show Augmented Images.

2. Show Model Structure

3. Show Accuracy and Loss

4. Inference



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

(3) Select 1 image arbitrarily

(2) Click the button.

(4) Show the loaded image (When displaying loaded image, resizing the loaded image to 128x128)

Predicted = deer (6) Show the predicted class label

(5) Run inference on the loaded image

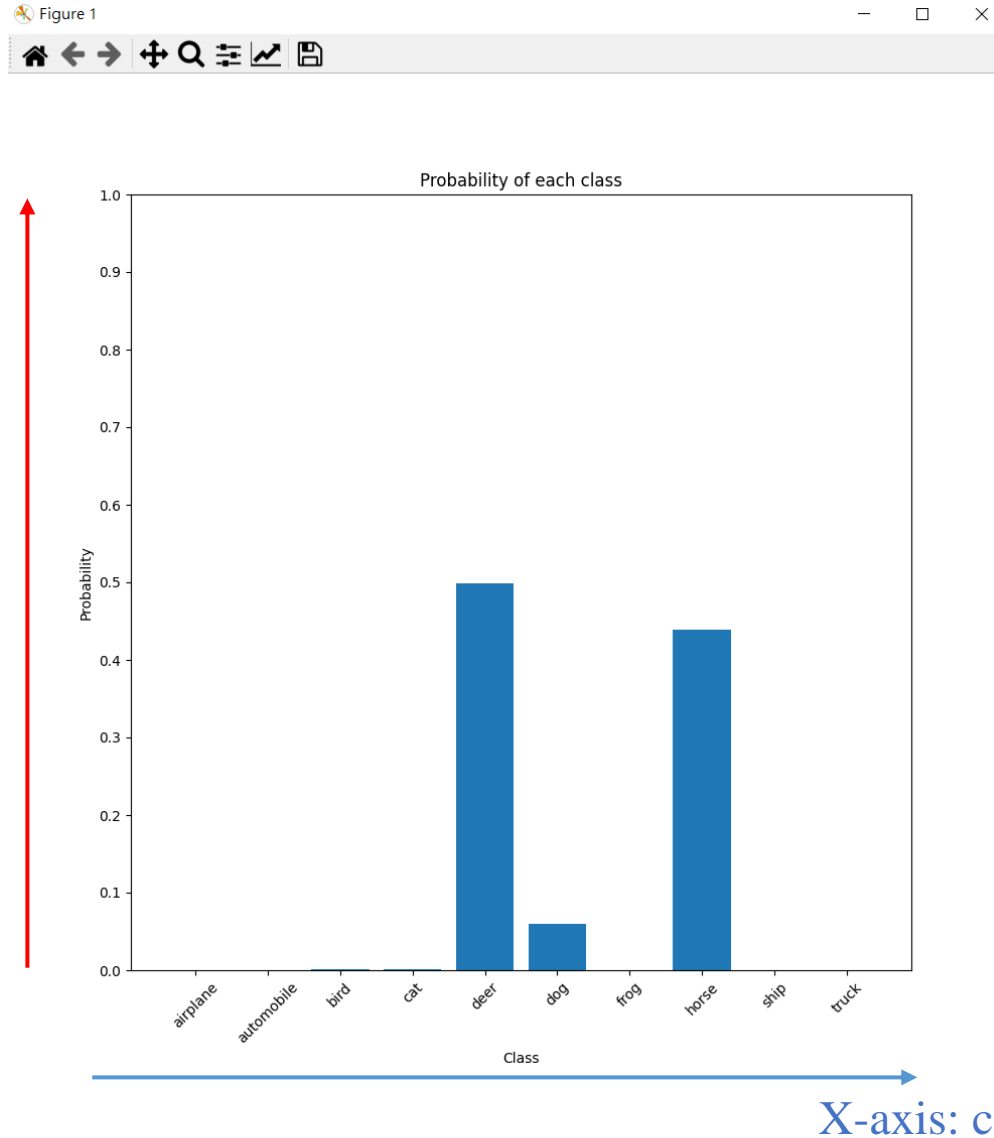
(7) Show the probability distribution of model prediction in new window.

Figure 1: Probability of each class

Class	Probability
airplane	0.00
automobile	0.00
bird	0.00
cat	0.00
deer	0.50
dog	0.05
horse	0.45
ship	0.00
truck	0.00

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.



5. Training a CIFAR10 Classifier Using VGG19 – Example Video

(出題：Hsiang)

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

