# Video Streaming and tracking

HW1 - Classification

Deadline: 2022/10/10

### Objective

- Train a neural network to do classification.
  - Implement a VGG19-like Network.
  - Environment: Anaconda
  - Framework: pytorch
- Goal: Top-1 Accuracy >= 55%
- Fewer parameters can get higher score!

#### Not allow !!!

- 1. Use pretrained weight
- 2. Call torchvision build model.

### **Environment Version Constraint**

You can use the following four package version (Suggestion).

Python: 3.8

Pytorch: 1.9.1

Pandas: 1.4.4

Matplotlib: 3.5.3

Scikit-Image: 0.19.3

If you have other package, you need to descript them in your report.

You can provide your environment (Optional).

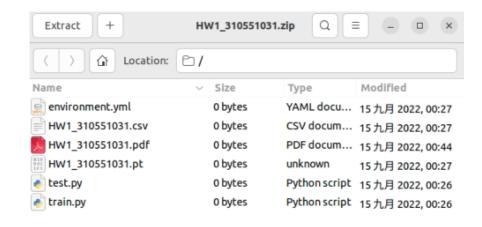
### **Dataset**

Dataset Link: <a href="https://reurl.cc/YXn934">https://reurl.cc/YXn934</a>

- The dataset consists 10 classes of 224 × 224 RGB images.
- We will give you 850 images for training, 285 for validation, 287 for testing.
- For training and validation data, we will give you labels in train.csv and val.csv.
- For testing data, we will not give you labels.

### **Submission**

- Your submission should contain:
  - Training Code and Testing Code
  - Model Weight(HW1\_StudentID.pt)
  - Testing Result(HW1\_StudentID.csv)
  - Report (HW1\_StudentID.pdf)



Compress them into one zip file name HW1\_StudentID.zip.

(Do **not** contain dataset in your submission)

# Submission - Testing Result

- Please predict the label for testing data(in order) into a txt file named HW1\_ StudentID.csv (ex: HW1\_310551031.csv)
- The format description is as follows.
- Put it in the HW1\_StudentID.csv.
   (if the path or format is wrong, you will get -10 points)
- TA will test the csv file.

8.jpg, 1 14.jpg, 3 25.jpg, 5 32.jpg, 43

# Submission - Report

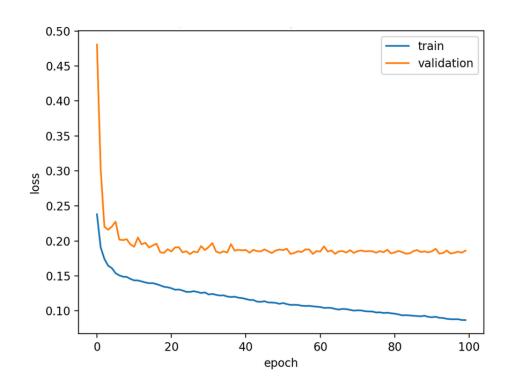
- How to reproduce your result. (Including package, environment, reproduced method) (4 points)
   ex: python test.py
- 2. Number of Model parameters (4 points)

```
ex: number_of_params = sum(p.numel() for p in model.parameters() if p.requires_grad)
```

- 3. Explain model structure (as detail as possible) (4 points)
  - ex: The layer number, convolution layer, channel number, loss function... revised from VGGnet19.
- 4. Results (training and validation loss curve, training and validation accuracy curve) (4 points)
- 5. Problems encountered and discussion (4 points)

# Submission - Report

### Loss Curve



# Grading

- Top-1 Accuracy >= 55% (65 points)
  - 50% <= Accuracy < 55% (55 points)</li>
  - 35% <= Accuracy < 50% (50 points)</li>
  - 35% > Accuracy (0 points)
- Number of parameters (15 points)
  - Compare with your classmates(fewer parameters gets higher score)
  - Top 1/5 get 15 points, second 1/5 get 12 points, etc.
- Report (20 points)

# Penalty

Format penalty - 10 points (Maximum)

If you have any incorrect file format or name, then you will get -5 points.

- Submit the result in the wrong name -5 points
- Submit the result in the wrong format -5 points

Late penalty - 20% per day

- 1 day => 80%, 2 day => 60%...

# Reference



### ANACONDA DISTRIBUTION

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ANACONDA

The world's most popular opensource Python distribution platform



1. Create Environment

conda env create -n [environment name] python=[python version]

2. Activate Environment

conda activate [environment name]

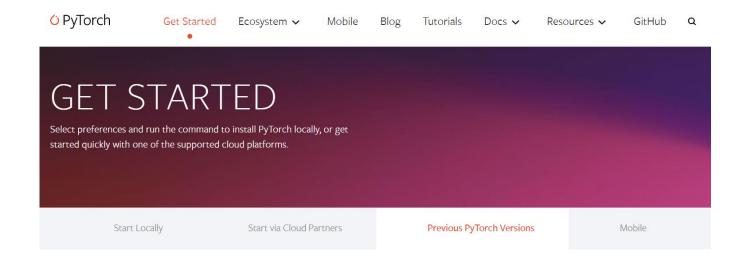
3. Export Environment (You should activate your current environment then you can export your environment)

conda env export > environment.yml

## **Pytorch**

Tutorial1: <a href="https://www.youtube.com/watch?v=85uJ9hSaXig">https://www.youtube.com/watch?v=85uJ9hSaXig</a>

Tutorial2: <a href="https://www.youtube.com/watch?v=VbqNn20FoHM">https://www.youtube.com/watch?v=VbqNn20FoHM</a>



#### INSTALLING PREVIOUS VERSIONS OF PYTORCH

We'd prefer you install the latest version, but old binaries and installation instructions are provided below for your convenience.

COMMANDS FOR VERSIONS >= 1.0.0

## Pytorch – Data Process

#### 1. Import package

```
from torch.utils import data
```

#### 2. Construct dataset

```
class SportLoader(data.Dataset):
   def init (self, mode, transform=None):
       self mode = mode
       self.sport = # read the file from csv file
                                                                            Initialize image and label data
       self.img name = # image name list from the csv file
       self.label = # label list from the csv file
       self.transform = transform # image transform
   def len (self):
                                                                             Numbers of image and label
       return len(self.img name)
   def getitem (self, index):
       image path = self.mode+"/"+self.img name[index]
       self.img = io.imread(image path)
       self.target = self.label[index]
                                                                             Return single image and label
       if self.transform:
           self.img = self.transform(self.img)
       return self.img, self.target
```

## Pytorch – Data Loader

1. Import package

```
from torch.utils.data import DataLoader
```

2. Select your mode in datasets(Page14)

```
train_dataset=SportLoader("train")
valid_dataset=SportLoader("val")
test_dataset=SportLoader("test")
```

3. Setting batch\_size in the dataloader

```
train_loader = DataLoader(train_dataset, batch_size=64, shuffle=True)
valid_loader = DataLoader(valid_dataset, batch_size=64, shuffle=True)
test_loader = DataLoader(test_dataset, batch_size=64, shuffle=True)
```

## Pytorch – Build Model

#### 1. Import package

```
import torch.nn as nn
import torch.nn.functional as F
```

#### 2. Build model

```
class Network(nn.Module):
   def init (self):
       super(Network, self).__init__()
       self.conv1 = nn.Conv2d(in channels=3, out channels=12, kernel size=5, stride=1, padding=1)
       self.conv2 = nn.Conv2d(in channels=12, out channels=12, kernel size=5, stride=1, padding=1)
       self.pool = nn.MaxPool2d(2,2)
       self.conv4 = nn.Conv2d(in channels=12, out channels=24, kernel size=5, stride=1, padding=1)
       self.conv5 = nn.Conv2d(in_channels=24, out_channels=24, kernel_size=5, stride=1, padding=1)
       self.fc1 = nn.Linear(24*10*10, 10)
   def forward(self, input):
       output = F.relu(self.conv1(input))
       output = F.relu(self.conv2(output))
       output = self.pool(output)
       output = F.relu(self.conv4(output))
       output = F.relu(self.conv5(output))
       output = output.view(-1, 24*10*10)
       output = self.fc1(output)
        return output
```

# Pytorch – Conv2D

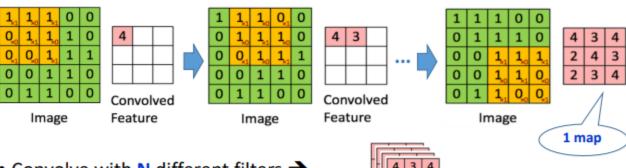
### **Convolution** – Example

• Convolve with a 3x3 filter

r	1	0	1	
	0	1	0	
	1	0	1	

#### Convolution

- Element-wise multiplication and sum of a filter and the signal (image)
- Convolve (slide) over all spatial locations
- The # of output maps is equal to # of filters



- Convolve with N different filters →
- What filters to convolve? To be learned

### Pytorch – Conv2D

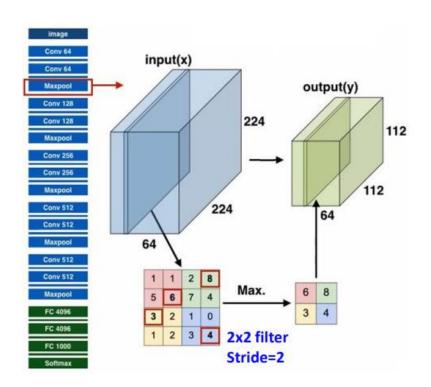
#### CONV2D

```
CLASS torch.nn.Conv2d(in_channels, out_channels, kernel_size, stride=1, padding=0, dilation=1, groups=1, bias=True, padding_mode='zeros', device=None, dtype=None) [SOURCE]
```

#### Parameters

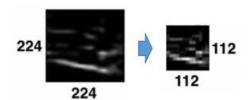
- in\_channels (int) Number of channels in the input image
- out\_channels (int) Number of channels produced by the convolution
- kernel\_size (int or tuple) Size of the convolving kernel
- stride (int or tuple, optional) Stride of the convolution. Default: 1
- padding (int, tuple or str, optional) Padding added to all four sides of the input. Default: 0
- padding\_mode (string, optional) 'zeros', 'reflect', 'replicate' or 'circular'. Default: 'zeros'
- dilation (int or tuple, optional) Spacing between kernel elements. Default: 1
- groups (int, optional) Number of blocked connections from input channels to output channels. Default: 1
- bias (bool, optional) If True, adds a learnable bias to the output. Default: True

## Pytorch – MaxPool2D



## Max Pooling

- 1. Reduce dimension
- 2. produce an abstracted form of the representation
- avoid over-fitting



### Pytorch – MaxPool2D

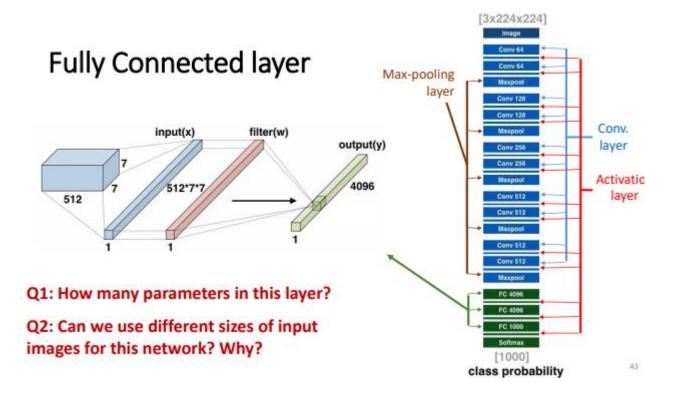
#### MAXPOOL2D

```
CLASS torch.nn.MaxPool2d(kernel_size, stride=None, padding=0, dilation=1, return_indices=False, ceil_mode=False) [SOURCE]
```

#### Parameters

- kernel\_size the size of the window to take a max over
- stride the stride of the window. Default value is kernel size
- . padding implicit zero padding to be added on both sides
- dilation a parameter that controls the stride of elements in the window
- return\_indices if True, will return the max indices along with the outputs. Useful for torch.nn.MaxUnpool2d later
- ceil\_mode when True, will use ceil instead of floor to compute the output shape

# Pytorch – Linear



# Pytorch – Linear

#### LINEAR

CLASS torch.nn.Linear(in\_features, out\_features, bias=True, device=None, dtype=None) [SOURCE]

#### Parameters

- in\_features size of each input sample
- out\_features size of each output sample
- . bias If set to False, the layer will not learn an additive bias. Default: True

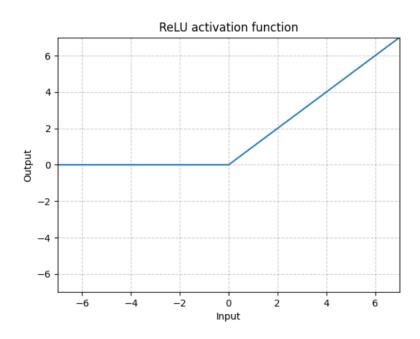
# Pytorch – relu

### **RELU**

CLASS torch.nn.ReLU(inplace=False) [SOURCE]

#### Parameters

inplace - can optionally do the operation in-place. Default: False



# Pytorch – CrossEntropyLoss

### Entropy, Cross entropy, KL divergence

### Entropy

$$H_p = E_{x \sim p}[\log_2 1/p(x)]$$

$$= \sum_{i=1}^{n} p_i \log \frac{1}{p_i}$$

	X <sub>1</sub>	X <sub>2</sub>	X <sub>3</sub>	X <sub>4</sub>	
Probability p	0.5	0.25	0.125	0.125	
$I = \log_2(1/p)$	1	2	3	3	
Entropy H <sub>P</sub>	0.5x1 + 0.25x2 + 0.125x3 + 0.125x3				
	= 1.75				

- Expected Entropies across all choices
- Skewed Probability Distribution: Low entropy
- Balanced Probability Distribution: High entropy

## Pytorch – CrossEntropyLoss

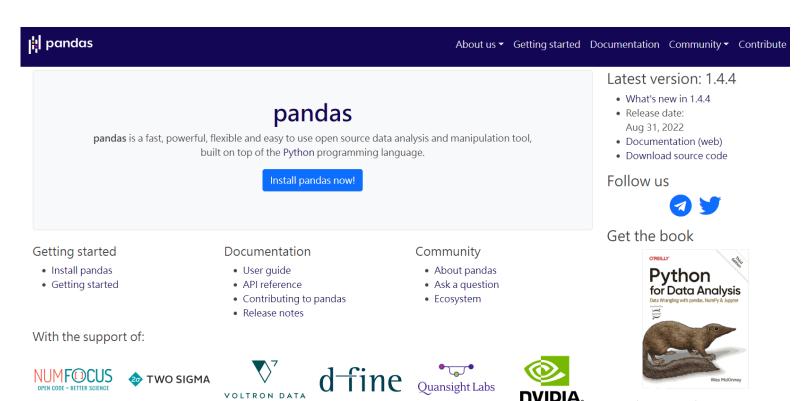
### CROSSENTROPYLOSS

```
CLASS torch.nn.CrossEntropyLoss(weight=None, size_average=None, ignore_index=- 100, reduce=None, reduction='mean', label_smoothing=0.0) [SOURCE]
```

```
# Example of target with class indices
loss = nn.CrossEntropyLoss()
input = torch.randn(3, 5, requires_grad=True)
target = torch.empty(3, dtype=torch.long).random_(5)
output = loss(input, target)
output.backward()
```

### **Pandas**

### Tutorial: <a href="https://pandas.pydata.org/docs/user\_guide/10min.html">https://pandas.pydata.org/docs/user\_guide/10min.html</a>



Previous versions

### Matplotlib

### Tutorial: <a href="https://matplotlib.org/stable/tutorials/index.html">https://matplotlib.org/stable/tutorials/index.html</a>



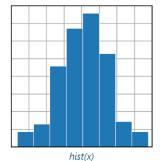
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Matplotlib is a comprehensive library for creating static, animated, and interactive visualizations in Python. Matplotlib makes easy things easy and hard things possible.

- · Create publication quality plots.
- Make interactive figures that can zoom, pan, update.
- · Customize visual style and layout.
- . Export to many file formats.
- Embed in JupyterLab and Graphical User Interfaces.
- Use a rich array of third-party packages built on Matplotlib.

Try Matplotlib (on Binder)















Cheat Sheets

Examples

### Reference

- Pytorch tutorial1: <a href="https://www.youtube.com/watch?v=85uJ9hSaXig">https://www.youtube.com/watch?v=85uJ9hSaXig</a>
- Pytorch tutorial2: <a href="https://www.youtube.com/watch?v=VbqNn20FoHM">https://www.youtube.com/watch?v=VbqNn20FoHM</a>
- Official tutorial: <a href="https://pytorch.org/tutorials/beginner/blitz/cifar10">https://pytorch.org/tutorials/beginner/blitz/cifar10</a> tutorial.html