Video Streaming and Tracking

Homework 2 - Object Detection

Deadline: 2024/10/21 23:55

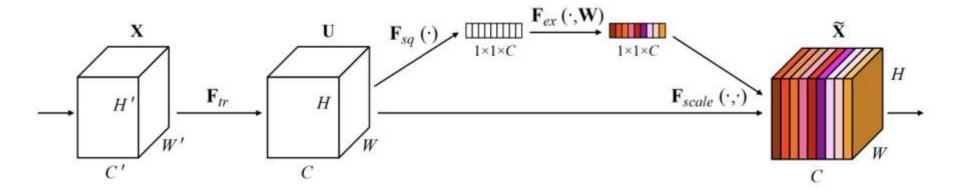
Outline

- Introduction
- Dataset
- Evaluation Metrics
- Hand in Rules
- Grading Policy

Introduction

- Train a neural network to do detection on our own dataset
- The pretrained model weights are available
- Model : object detection algorithms
 - YOLOX-s (we use the <u>official code</u> to set the baseline)
- Add SE module or Inception module to your network
- Framework : PyTorch

Squeeze-and-Excitation Networks



Ftr(): convolution operation

F_{sq}(): avg_pool2d

Fex(): Linear → ReLU → Linear → Sigmoid

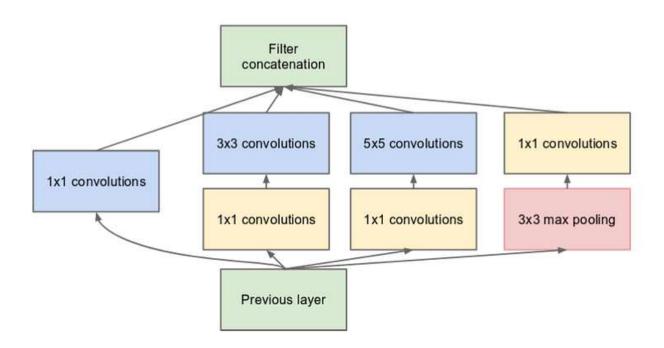
Sample code

Conv2d → SELayer → Conv2d

```
class SELayer(nn.Module):
def init (self, channel, reduction=16):
     super(SELayer, self). init ()
     self.avg pool = nn.AdaptiveAvgPool2d(1)
     self.fc = nn.Sequential(
        nn.Linear(channel, channel // reduction),
        nn.ReLU(inplace=True),
        nn.Linear(channel // reduction, channel),
        nn.Sigmoid()
def forward(self, x):
    b, c, , = x.size()
    y = self.avg pool(x).view(b, c)
    y = self.fc(y).view(b, c, 1, 1)
    return x * v
```

from torch import nn

Inception Module



Sample code

MaxPool2d → Inception

```
\rightarrow Inception \rightarrow MaxPool2d
```

```
class Inception(nn.Module):
def init (self,in c,c1,c2,c3,c4):
    super(Inception, self).__init__()
     self.p1 = nn.Sequential(
         nn.Conv2d(in c,c1,kernel size=1),
         nn.ReLU(inplace=True)
     self.p2 = nn.Sequential(
         nn.Conv2d(in c,c2[0],kernel size=1),
         nn.ReLU(inplace=True),
         nn.Conv2d(c2[0], c2[1], kernel_size=3,padding=1),
         nn.ReLU(inplace=True)
     self.p3 = nn.Sequential(
         nn.Conv2d(in c, c3[0], kernel size=1),
         nn.ReLU(inplace=True),
         nn.Conv2d(c3[0], c3[1], kernel size=5,padding=2),
         nn.ReLU(inplace=True)
     self.p4 = nn.Sequential(
         nn.MaxPool2d(kernel size=3, stride=1, padding=1),
         nn.Conv2d(in c,c4,kernel size=1),
         nn.ReLU(inplace=True)
def forward(self, x):
    p1 = self.p1(x)
    p2 = self.p2(x)
    p3 = self.p3(x)
    p4 = self.p4(x)
    return torch.cat((p1,p2,p3,p4),dim=1)
```

Dataset

- GTA video dataset
- You only need to detect car (Only one class)
- 1596 training images, labels
- 227 validation images, labels
- 456 testing images



Labels

- ./HW2_ObjectDetection_2024/{train, val}_labels/
- Each row is [class x_center y_center width height] (0~1 range) format (use 0 to represent car)
- Converting the dataset to the YOLOX-compatible format(e.g. COCO)



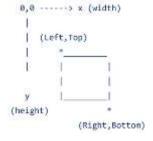
Evaluation Metrics: mAP (mean Average Precision)

- Most common metric for object detection
- In this HW, mAP defined in the PASCAL VOC 2012 competition is used
- We will use the following github repo to calculate your score

https://github.com/rafaelpadilla/Object-Detection-Metrics

- It also contains some explanations about how to calculate it
- We will set IoU threshold greater than 0.85 to calculate the testing score
 E.g. python pascalvoc.py -t 0.85 -gtformat xyrb -detformat xyrb -np

Hand in Rules (1/3)



- You should hand in your result by detecting the testing data through your model
- Format [class confidence left top right bottom] in pixel wise (1920x1080)

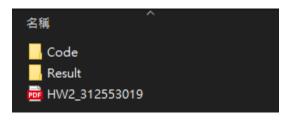


Hand in Rules (2/3)

- Store each detection result in [image_name].txt
- You should have a model: with the SE module or Inception module
- Use this model to detect on the testing set
- Submit results in the folder
- You should hand in
 - Result folder contain testing 456 results
 - Folder should be named: Result

Hand in Rules (3/3)

- Your submission should contain
 - Result folder contain testing 456 results
 - Report (in pdf)
 - Code (include your environment and checkpoints). Do not contain dataset and pretrained model weights.
 - Please submit the code that can generate the prediction results.
- Compress them into one zip file name HW2_[studentID].zip



Grading Policy (1/2)

- Model implementation 65 points
 - Implement on your own or clone from Github then run on our dataset and pass the baseline (mAP = 0.8) by using the code we provide (set IoU threshold to 0.85) to evaluate on the validation set 50 points
 - Add the **SE module** or **Inception module** to your model 15 points
- Model performance 15 points
 - The points will determined by the rank with your classmates
 - You can use SE module or Inception module + other module to improve your performance
 - You should use YOLOX-s as the baseline model
 - Ranking the average mAP on testing set you will get 15 / 10 / 5 / 0 points base on your rank in the class

Grading Policy (2/2)

- Report 20 points
 - Experiment Setup (Data pre-process, Hyperparameters,...) 5 points
 - Brief explain your code 5 points
 - which layer you add modules
 - training / inference command line
 - Screenshot your validation results on your models 10 points
- If you used code from GitHub, provide reference

Penalty

- Format penalty 5 points
 - Submit the result in the wrong name, format, etc.
 - Submit the report not in pdf format
- No validation results are shown in the report 10 points
- Deadline: 2024/10/21 23:59
- Late penalty 20% per day
 - 1 day => 80%, 2 day => 60%...
- You can use any code from Github, but don't copy from your classmate!

References

- https://arxiv.org/pdf/1709.01507.pdf
- https://github.com/Megvii-BaseDetection/YOLOX
- https://github.com/rafaelpadilla/Object-Detection-Metrics