



Selected Topics in Visual Recognition using Deep Learning

Homework 4 announcement

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Homework 4: Instance segmentation

- **Deadline: 12/19, Thr at 23:59**

1. Upload your **report.pdf** and **submission file** in this [Google drive](#)

| My Drive > CS_IOC5008 > HW4 ▾  | | | |  |  |
|---|-------|----------------|---|---|---|
| Name | Owner | Last modifi... | ↓ | | |
|  submission | me | 5:40 PM me | | | |
|  dataset | me | 5:40 PM me | | | |
|  reports | me | 5:39 PM me | | | |



HW4 Introduction: Tiny PASCAL VOC dataset

- Tiny VOC dataset contains only 1,349 training images, 100 test images with 20 common object classes
- **NO external data should be used and only ImageNet pre-trained model can be used**
- Deal with the overfitting problem!



Learn to process PASCAL VOC dataset

- PASCAL VOC dataset are often evaluated on current computer vision models

| | mean | aero plane | bicycle | bird | boat | bottle | bus | car |
|------------------------------|------|---------------|---------|------|------|--------|------|------|
| | ▼ | ▼ | ▼ | ▼ | ▼ | ▼ | ▼ | ▼ |
| RecoNet152_coco [?] | 89.0 | 97.3 | 80.4 | 96.5 | 83.8 | 89.5 | 97.6 | 95.4 |
| DeepLabv3+_JFT [?] | 89.0 | 97.5 | 77.9 | 96.2 | 80.4 | 90.8 | 98.3 | 95.5 |
| SRC-B-MachineLearningLab [?] | 88.5 | 97.2 | 78.6 | 97.1 | 80.6 | 89.7 | 97.4 | 93.7 |
| DeepLabv3+_AASPP [?] | 88.5 | 97.4 | 80.3 | 97.1 | 80.1 | 89.3 | 97.4 | 94.1 |
| SepaNet [?] | 88.3 | 97.2 | 80.2 | 96.2 | 80.0 | 89.2 | 97.3 | 94.7 |
| EMANet152 [?] | 88.2 | 96.8 | 79.4 | 96.0 | 83.6 | 88.1 | 97.1 | 95.0 |
| MSCI [?] | 88.0 | 96.8 | 76.8 | 97.0 | 80.6 | 89.3 | 97.4 | 93.8 |
| ExFuse [?] | 87.9 | 96.8 | 80.3 | 97.0 | 82.5 | 87.8 | 96.3 | 92.6 |
| DeepLabv3+ [?] | 87.8 | 97.0 | 77.1 | 97.1 | 79.3 | 89.3 | 97.4 | 93.2 |
| CFNet [?] | 87.2 | 96.7 | 79.7 | 94.3 | 78.4 | 83.0 | 97.7 | 91.6 |
| DeepLabv3-JFT [?] | 86.9 | 96.9 | 73.2 | 95.5 | 78.4 | 86.5 | 96.8 | 90.3 |



HW4 Get the dataset

- Download the dataset from this [Google Drive](#)
- The annotations are saved in **json file**. You can use [pycocotools](#) to read this file
- See [data_loader.ipynb](#) for more details



Upload your submission.json file [here](#)

- Free version Kaggle doesn't provide the metrics :(
- Upload your submission file into the Google Drive. I will evaluate and return the performance on your filename every midnight
- filename should be STUDENTID.json

My Drive > CS_IOC5008 > HW3 ▾



| Name ↑ | Owner | Last modified | File size |
|------------|-------|---------------|-----------|
| dataset | me | 1:00 AM me | — |
| submission | me | 1:52 AM me | — |

Name ↑



0610001.json



Name ↑

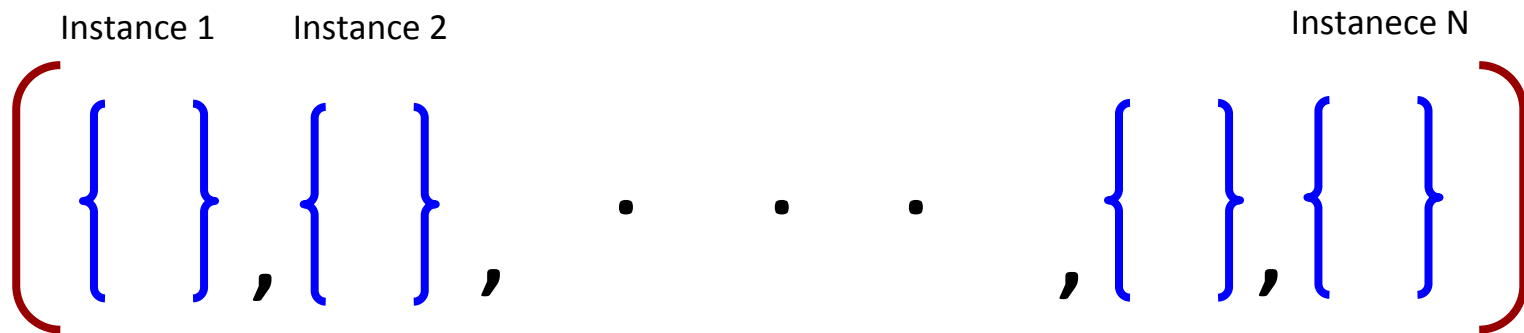


mAP_0.45_0610001.json



Submission.json file format

- List of dictionaries, $\text{len}(\text{list}) = \text{number of detected instance in all test images}$. Find pseudo code [here](#) in *Prepare submission file*
- Each dictionary contains three keys
 - “image_id”: id of test image, which is the key in “test.json”, **int**
 - “score”: probability for the class of this instance, **float**
 - “category_id”: category id of this instance, **int**
 - “segmentation”: Encode the mask in RLE by provide function, **str**



Evaluation metrics: mean Average Precision

- Most common metric for object detection/segmentation
- Measure the average precision on different threshold and also the IOU between GT and prediction

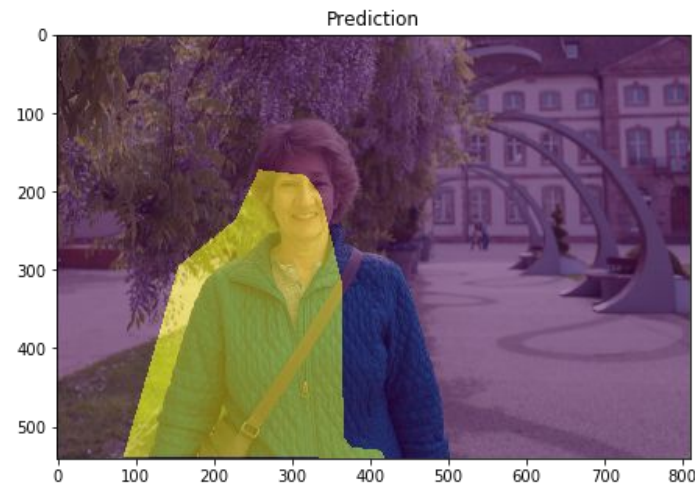
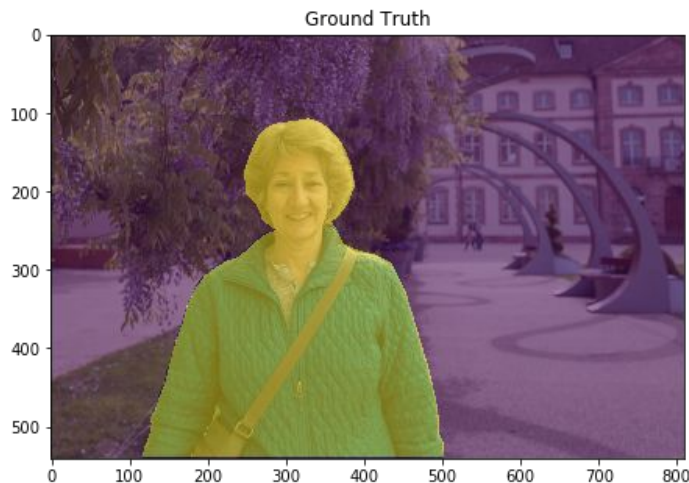
| | backbone | cascade | AP | AP ₅₀ | AP ₇₅ |
|--------------|------------|---------|------|------------------|------------------|
| Faster R-CNN | AlexNet | ✗ | 29.4 | 63.2 | 23.7 |
| | | ✓ | 38.9 | 66.5 | 40.5 |
| Faster R-CNN | VGG | ✗ | 42.9 | 76.4 | 44.1 |
| | | ✓ | 51.2 | 79.1 | 56.3 |
| R-FCN | RetNet-50 | ✗ | 44.8 | 77.5 | 46.8 |
| | | ✓ | 51.8 | 78.5 | 57.1 |
| R-FCN | ResNet-101 | ✗ | 49.4 | 79.8 | 53.2 |
| | | ✓ | 54.2 | 79.6 | 59.2 |

TABLE 11: Detection results on PASCAL VOC 2007 test.



Evaluation metrics: mean Average Precision

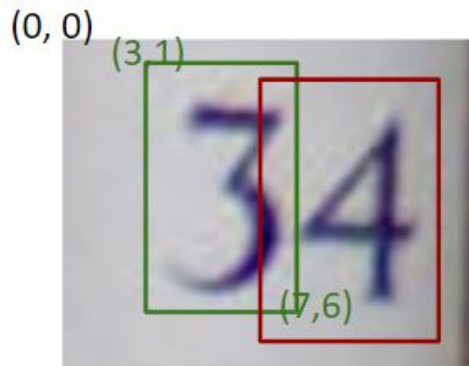
- Use mask to measure the Intersection-of-Union of predictions and ground truth
- We use average precision at IOU=0.5 to evaluate your results



Evaluation metrics: mean Average Precision

- Set IOU threshold=0.5
- We got 1 TP, 1 FP, precision=0.5, recall=0.5
- But if we change the probability threshold to 0.7, the red box will disappear, then we got 1 TP, precision=1, recall=0.5
- We also miss one GT, so when recall=1, precision=0.5
- E.g., your model output two boxes on image, the dictionary will be

```
dict = {"bbox": [(1, 3, 6, 7), (4, 5, 8, 12)],  
        "label": [3, 5],  
        "score": [0.87, 0.61]}
```



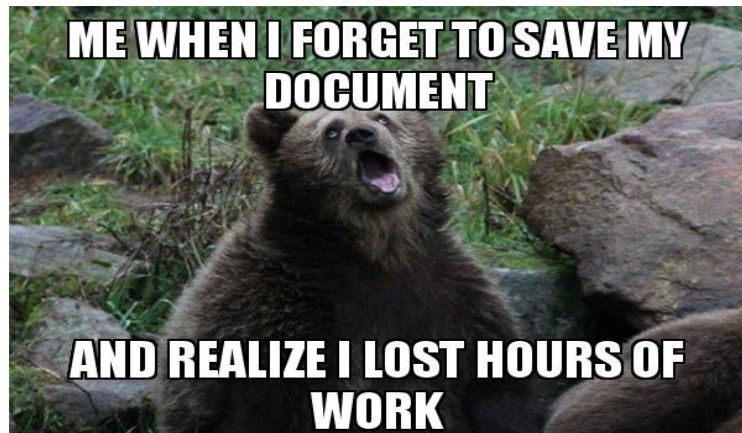
Grading policy: Model performance (70 points)

- Get at least 56% ($70\% \times 0.8$) by scoring over the baseline
- baseline (mAP@0.5): 0.247



Grading policy: Reports (20 points)

- Document your work (in PDF)
 - GitHub/ GitLab link of your code
 - **reference if you used code from GitHub**
 - Brief introduction
 - Methodology (Data pre-process, Model architecture, Hyperparameters,...)
 - Findings or Summary



Grading policy: Code readability (10 points)

- Write beautiful Python code with [PEP8 guidelines](#) for readability. Base requirement: use whitespace correctly!

Python

Recommended

```
def function(default_parameter=5):  
    # ...
```

Not recommended

```
def function(default_parameter = 5):  
    # ...
```

Python

Recommended

```
my_list = [1, 2, 3]
```

Not recommended

```
my_list = [ 1, 2, 3, ]
```

Python

```
x = 5
```

```
y = 6
```

Recommended

```
print(x, y)
```

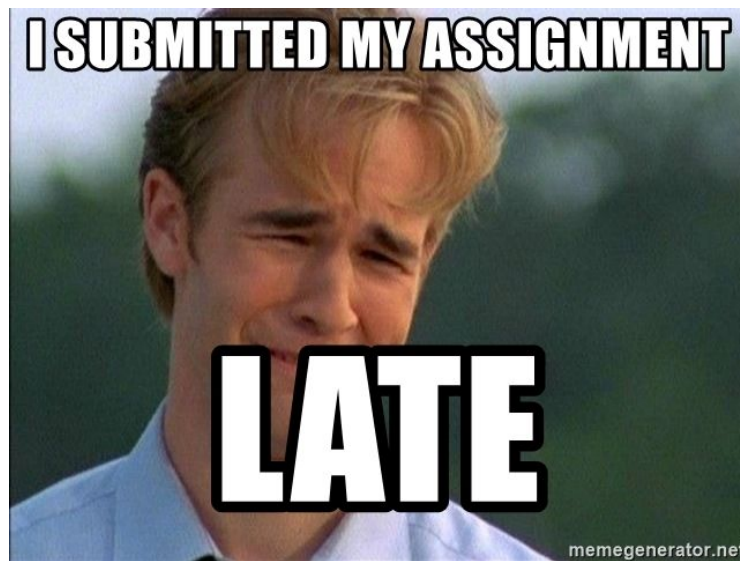
Not recommended

```
print(x , y)
```



Late Policy

- We will deduct a late penalty of 20 points per additional late day
- For example, If you get 90% of HW but delay for two days, your will get only 90 points- (20 points x 2) = 50 points!



Keywords

- Beat the baseline
 - Mask R-CNN
 - Data-augmentation
 - Regularization (Dropout, L2-norm,...)
- Rank Top 3!
 - Read some paper from CVPR2019, ICCV2019 and try to implement it!



FAQ

- Can I use any code/tools/Library from GitHub or other resources?
 - Yes! We encourage you to learn how to apply existing tools on your own task, such as [Keras-Mask R-CNN](#), [Pytorch-maskrcnn-benchmark](#), [TF-object-detection-API](#)
- **But DO NOT copy code from your classmate!**
- Why my testing results are so bad?
 - CNN model prone to overfitting with small dataset. Use some techniques such as regularization, data-augmentation to solve it!



Notice

- Check your email regularly, we will mail you if there are any updates or problems of the homework
- If you have any questions or comments for the homework, please mail me and cc Prof. Lin
 - Prof. Lin: lin@cs.nctu.edu.tw
 - Jimmy: d08922002@ntu.edu.tw



Have fun!

