



Selected Topics in Visual Recognition using Deep Learning

Homework 3 announcement

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Homework 2 reminder

- **Deadline: 11/26** (one week left)
- **submission:**
 - Remove your duplicate submission file and rename your filename of final submission without postfix
 - Also backup your final submission file incase someone mis-delete your file

Name ↑



mAP_0.45_0610001.json



Code readability: suggestion

- It's hard to reproduce the results without proper config management
- Use configuration tools such as [yacs config](#) or config.py to track your model config
- Develop your model code by jupyter notebook is good. But train your model with python scripts is recommended

☐  config

☐  logs

☐  inspect_data_result.ipynb

☐  dataloader.py

☐  parse_annotations_to_data.py

☐  README.MD







☐  train.py





Homework 3: Instance segmentation

- **Deadline: 12/11, Thr at 23:59**
 1. Upload your **submission file** in this [Google drive](#)
 2. Upload your **report.pdf** to E3

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			



HW3 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!





Process the PASCAL VOC dataset

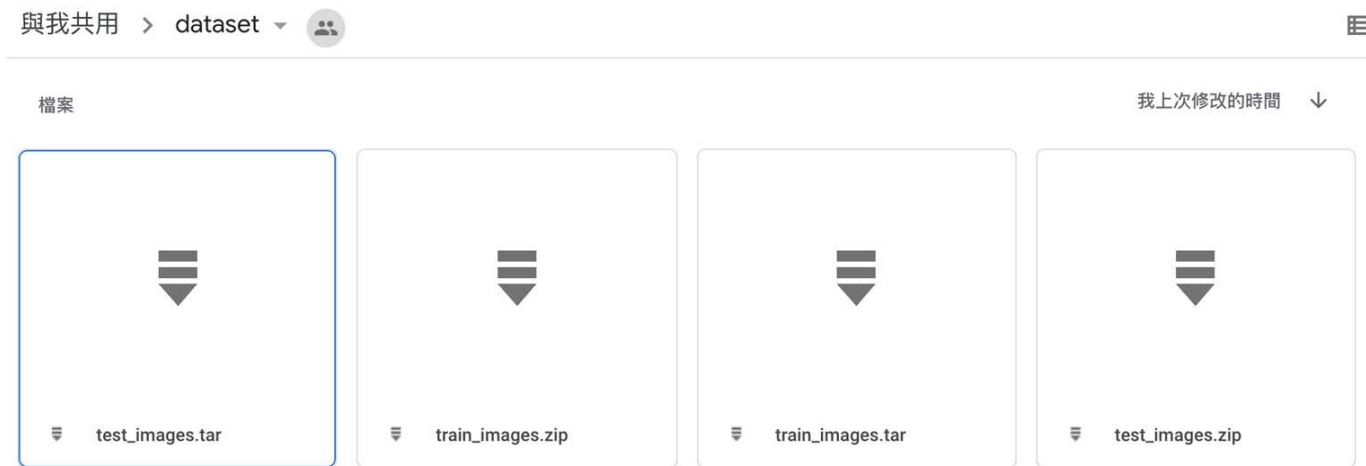
- PASCAL VOC dataset is widely used for evaluating the detection/segmentation models

	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



Get the dataset

- Download the dataset from this [Google Drive](#)
- The annotations are saved in **json file**. You can use [pycocotools](#) to read the 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. We will evaluate it and return the performance on your filename
- Filename should be **STUDENT_ID.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 ↑

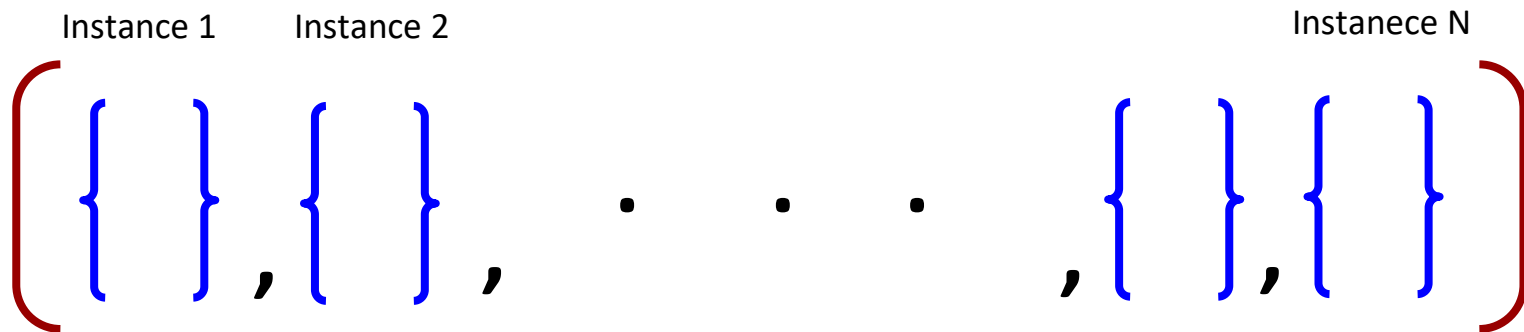


mAP_0.45_0610001.json



Submission.json file format ([COCO style format](#))

- List of dictionaries, $\text{len}(\text{list}) = \text{number of detected instances in all test images}$. Find demo code [here](#) in *Prepare submission file*
- Each dictionary contains four 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

- Widely used 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

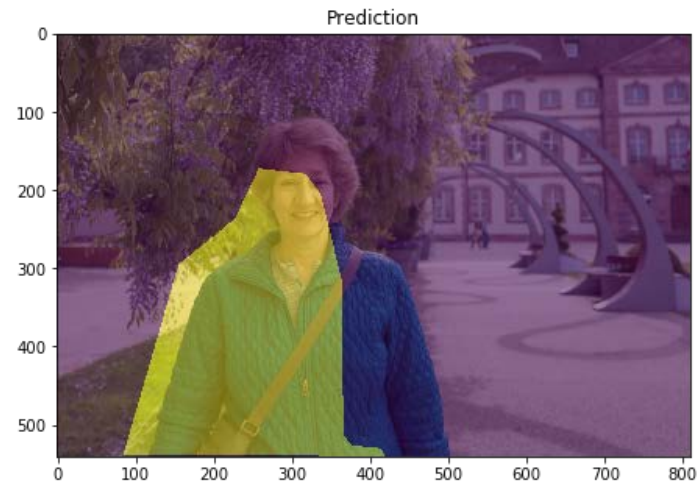
TABLE 11: Detection results on PASCAL VOC 2007 test.





Evaluation metrics: mean Average Precision

- Use mask to measure the Intersection-of-Union between predictions and ground truth
- We use average precision at $\text{IoU}=0.5$ to evaluate your results





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)
```

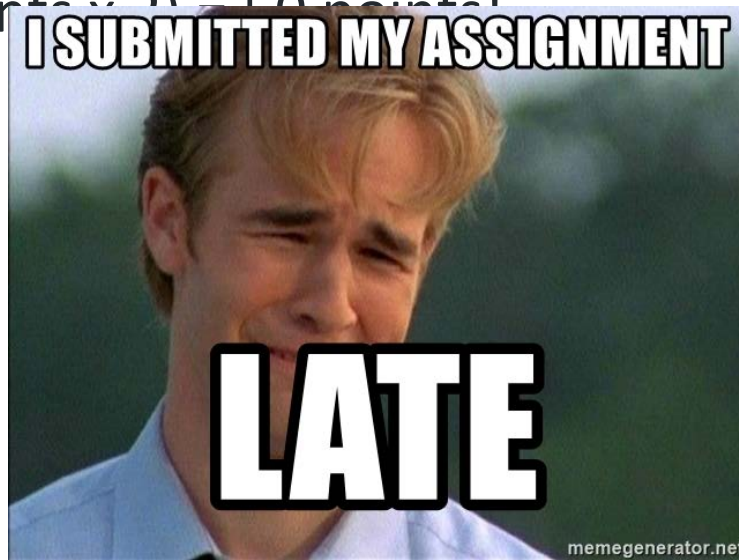
Not recommended

```
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- $(20 \text{ points} \times 2) = 50 \text{ points}$



Keywords

- Beat the baseline
 - Mask R-CNN
 - Data-augmentation
 - Regularization (Dropout, L2-norm,...)
- Rank Top 3!
 - Read some paper (Instance segmentation) from CVPR2020, ECCV2020 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 (3pm-4pm, Thur., EC118)
 - 佳諭: mylifeai1116@gmail.com (3pm-4pm, Thur., EC118)
 - 玉霖: oscar861201@gmail.com (3pm-4pm, Thur., EC118)



Have fun!

