
Machine Learning HW3 - Image Classification

ML TAs

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Objective

1. Solve image classification with **convolutional neural networks**.
2. Improve the performance with **data augmentations**.
3. Understand how to **utilize unlabeled data** and how it benefits.

Task - Food Classification



Task - Food Classification

- The images are collected from the food-11 dataset classified into 11 classes.
- The dataset here is slightly modified:
- Training set: $280 * 11$ **labeled** images + 6786 **unlabeled** images
- Validation set: $60 * 11$ **labeled** images
- Testing set: 3347 images
- **DO NOT** utilize the original dataset or labels.
 - This is cheating.

↓
(3, 128, 128)

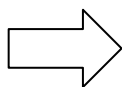
Task - Food Classification

$\{3, 128, 128\}$

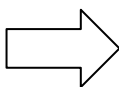


batch

$\{\text{batch}, 3, 128, 128\}$



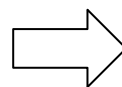
CNN
model



$\{\text{batch}, 11\}$
logits
(not normalized)

-0.25
+3.02
+0.56
-3.90
-0.01
+0.25
-0.47
+5.00
-1.32
+1.14
-0.28

softmax



$\{\text{batch}, 11\}$
probability
(normalized)

0.004
0.114
0.009
0.001
0.005
0.007
0.003
0.831
0.001
0.017
0.004

$\sum = 1$


Requirements

- This homework is in three levels:
 - Easy
 - Medium
 - Hard
- You can easily finish the easy level by running the example code.
- For the rest, we recommend you start with the same code.
 - We already prepared some TODO blocks for you.
- **DO NOT** pre-train your model on other datasets.
- If you use some well-known model architecture (e.g., ResNet), make sure **NOT** to load pre-trained weights as initialization.

Requirements - Easy

- Build a **convolutional neural network** using **labeled images** with provided codes.
- Public simple baseline: 44.862 (accuracy, %)

Requirements - Medium

- Improve the performance using **labeled images** with different model architectures or data augmentations. 
- Public medium baseline: 52.807 (accuracy, %)
- You can achieve the baseline by adding a few lines to the example code.

```
# It is important to do data augmentation in training.
# However, not every augmentation is useful.
# Please think about what kind of augmentation is helpful for food recognition.
train_tfm = transforms.Compose([
    # Resize the image into a fixed shape (height = width = 128)
    transforms.Resize((128, 128)),
    # You may add some transforms here.
    # ToTensor() should be the last one of the transforms.
    transforms.ToTensor(),
])
```


Requirements - Hard

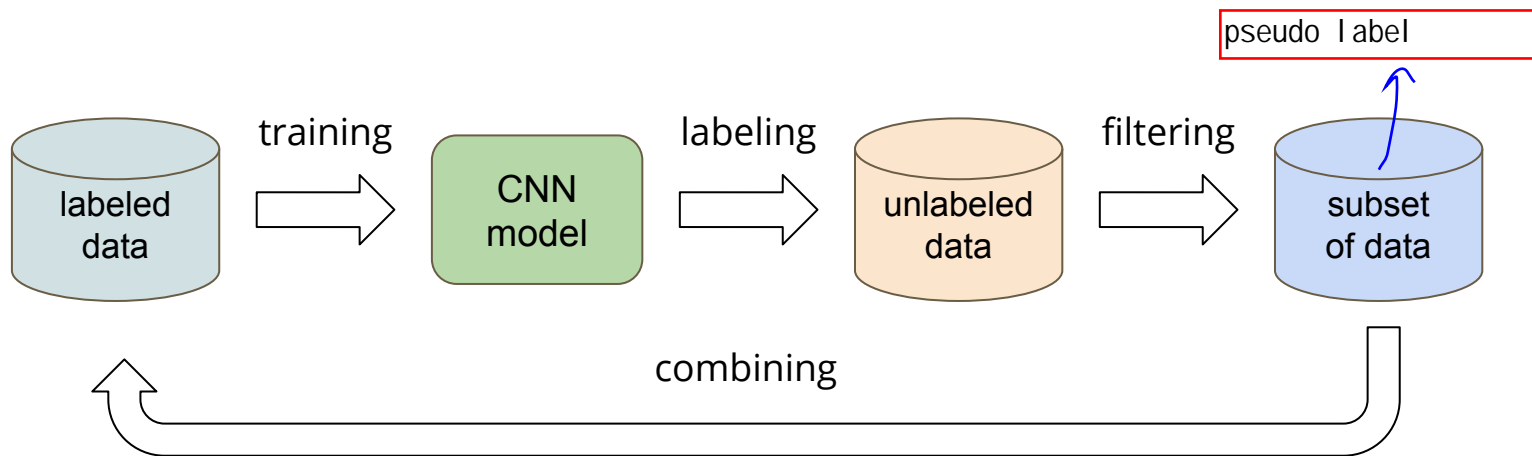
- Improve the performance with **additional unlabeled images.**
- Public strong baseline: 82.138 (accuracy, %)
- Do it on your own (by finishing TODO blocks in the example code).
- Using unlabeled testing data here is allowed.
- **Hint:** semi-supervised learning, self-supervised learning

```
def get_pseudo_labels(dataset, model, threshold=0.65):  
    # This functions generates pseudo-labels of a dataset using given model.  
    # It returns an instance of DatasetFolder containing images whose prediction confidences exceed a given threshold.  
    # You are NOT allowed to use any models trained on external data for pseudo-labeling.
```

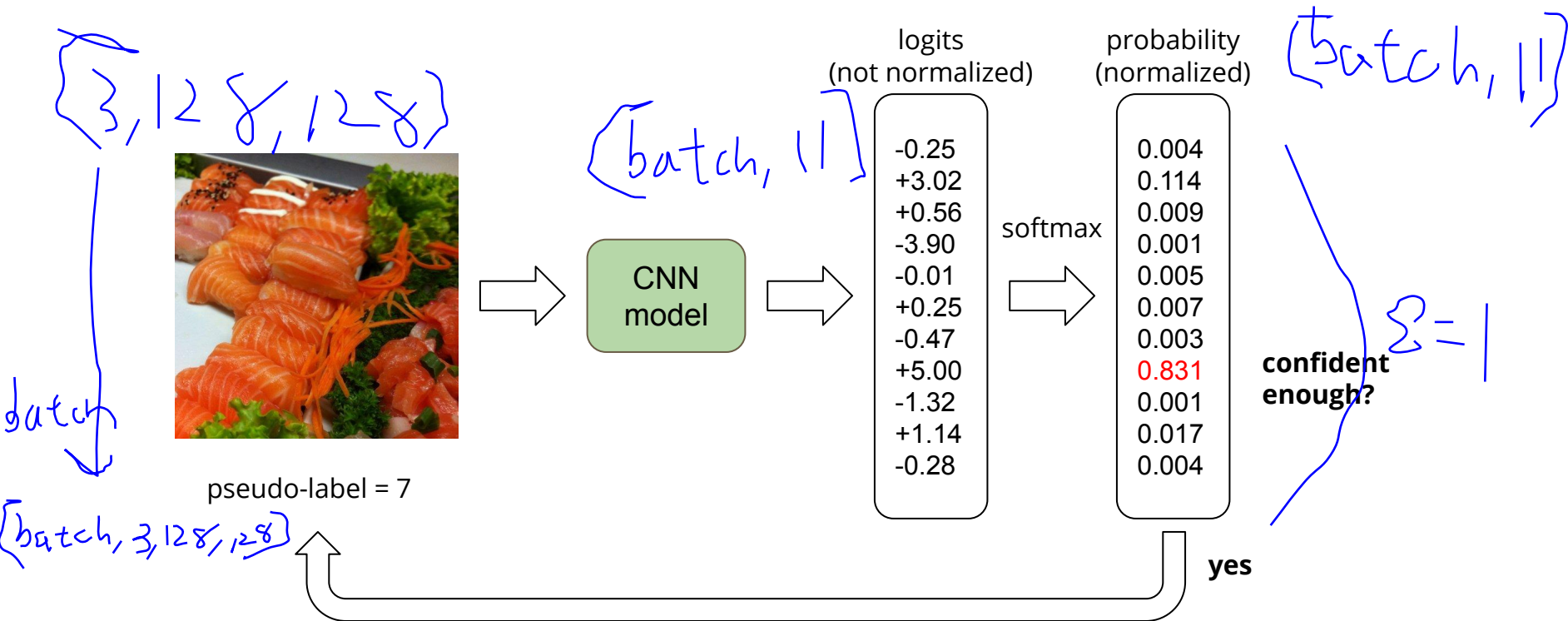
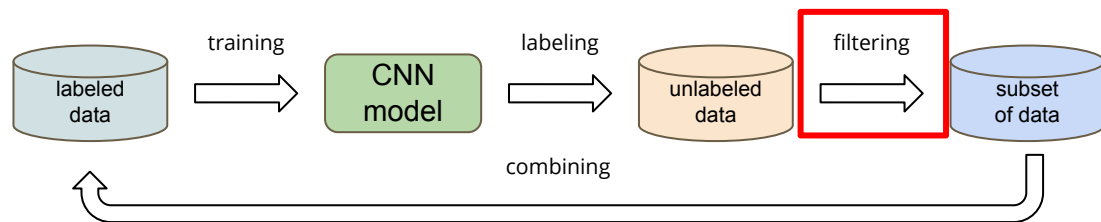
pseudo label

Semi-supervised Learning

- There are many ways to do semi-supervised learning.
- E.g., generate pseudo-labels for unlabeled data and train with them.



Pseudo-labels



Kaggle Submission Format

- The submitted predictions should be in **CSV** format.
- The first row is **"Id, Category"**
- The rest of rows are "{id}, {prediction}" (e.g., 0005, 8)
- There should be $(3347 + 1)$ rows in total.

| Id | Category |
|------|----------|
| 0001 | 0 |
| 0002 | 9 |
| 0003 | 4 |
| 0004 | 5 |

Grading Policy

- Public simple baseline: +1pt
- Public medium baseline: +1pt
- Public strong baseline: +1pt

- Private simple baseline: +1pt
- Private medium baseline: +1pt
- Private strong baseline: +1pt

- Submit your code: +4pt

Code Submission

- Submit your code via NTU COOL.

<student_id>_hw3.zip

- **DO**

- Specify the source of your code. (You may refer to [Academic Ethics Guidelines](#))
- Organize your code and make it easy to read (not necessary).

- **DO NOT**

- Submit empty or garbage files.
- Submit the dataset or model.
- Compress your codes into other formats like .rar or .7z and simply rename it to .zip.

- If we find you cheating or your code problematic, you will be punished.
 - Course final score * 0.9 for the first time, or fail the course otherwise.

Bonus

- If you successfully get 10 pts:
 - Your code will be made public to students.
 - You can submit a report in **PDF** format briefly describing what you have done (in English, less than 100 words) for **extra 0.5 pts**.
 - Reports will also be made public to students.
- [Report template](#)

Deadline

- Kaggle deadline: 2021/04/16 23:59:59
- Code submission: 2021/04/18 23:59:59
- Late submissions are **NOT** accepted.

Should You Have Any Questions...

- NTU COOL (recommended)
 - <https://cool.ntu.edu.tw/courses/4793>
- E-mail
 - ntu-ml-2021spring-ta@googlegroups.com
 - The title **must** start with **[hw3]**.
- TA hour
 - Fri. 14:00 - 18:00

Useful Resources

- Semi-supervised learning
 - [https://speech.ee.ntu.edu.tw/~tlkagk/courses/ML_2016/Lecture/semi%20\(v3\).pdf](https://speech.ee.ntu.edu.tw/~tlkagk/courses/ML_2016/Lecture/semi%20(v3).pdf)
 - https://www.youtube.com/watch?v=fX_guE7lNnY&ab_channel=Hung-yiLee
 - MixMatch: <https://arxiv.org/abs/1905.02249>
 - Noisy student: <https://arxiv.org/abs/1911.04252>
- PyTorch
 - <https://pytorch.org/>
- Torchvision
 - <http://pytorch.org/vision/stable/index.html>