Effective Training Strategies for Deep Learning-based Precipitation Nowcasting and Estimation (Software User Guide)

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1 General Information

• Version: 1.0

2 Introduction

In this software, we provided the following implementations in PyTorch:

- Model and Dataset implementation (model.py and dataset.py): U-Net based model (1x1 resolution) for precipitation nowcasting and estimation, and the dataset for each target task.
- Codes for experiments: Codes for pre-training and fine-tuning the model, and evaluating the performance of the trained model.

Detailed information about the implementations is explained in the following paper:

• Jihoon Ko*, Kyuhan Lee*, Hyunjin Hwang*, Seok-Woo Son, and Kijung Shin, "Effective training strategies for deep learning-based precipitation nowcasting and estimation", Under Review

3 Requirements

In order to execute the codes, you need to install following packages.

- NumPy
- PyTorch
- tqdm
- pyProj

4 Input File Format

To run the provided code, you need to prepare radar images and ground-truth rainfall data and then designate each dataset's path.

4.1 Radar Images

The only supported type is *.bin.gz containing an array of 16bit integers. The dBZ of the i-th pixel multiplied by 100 should be stored in the 2i-1 to 2i-th bytes of the file. For running experiments, you need to specify center, radius, and original_radar_size of the datasets, which indicates center coordinates of the cropped radar image, half of the size of the radar image, and the original image size before cropping. See dataset.py for more details. Our implementation assumes the input image has a size of 1468x1468. You should change the architecture if you want to use the model with different input sizes of images.

4.2 Ground-truth Rainfall Data

The only supported type is *.pkl containing a serialized list of ground-truth rainfall information. Each element in the list should be a timestamp pair with format YYYYM-MDDhhmm and the corresponding rainfall information. The rainfall information is a list of tuples consist of coordinates (y-axis and x-axis) of pixels consistent with the coordinates of radar images and the corresponding rainfall information. Here is an example of the format of the list.

[['202006010000', [(
$$Y_1$$
, X_1 , V_1), (Y_2 , X_2 , V_2), \cdots], ['202006010100', [(Y_3 , Y_3 , Y_3), (Y_4 , Y_4 , Y_4), \cdots]

After preparing the data, the last step of processing the dataset is to serialize the data. pickle.dump(open("sampled.pkl", "wb"), list) serializes the list and stores the serialized data in sampled.pkl.

5 How to Execute

- 5.1 Pretraining (pretrain.py)
 - -h, --help: Show help message
 - --lr LR: Set learning rate as LR
 - --n-steps N_STEPS: Set number of training steps as N_STEPS
 - --batch-size BATCH_SIZE: Set batch size as BATCH_SIZE
 - --data-path DATA_PATH: Set path of the radar input images as DATA_PATH
 - --gpus GPUS: Set which GPUs to use. GPUS should be comma-separated string or single GPU id. If GPUS is not specified, only CPU is used during execution.

- 5.2 Finetuning (finetuning-[CE|focal|ours].py for Nowcasting / finetuning.py for Estimation)
 - -h, --help: Show help message
 - --lr, --n-steps, --batch-size, --data-path, --gpus: See Section 5.1
 - --sampled-path SAMPLED_PATH: Set path of the sampled rainfall data as SAMPLED_PATH
 - \bullet --pretrained-weights-path WEIGHTS_PATH : Set path of pretrained weights as WEIGHTS_PATH
- 5.3 Evaluation (evaluation.py)
 - -h, --help: Show help message
 - --batch-size, --data-path, --gpus: See Section 5.1
 - --sampled-path: See Section 5.2
 - --finetuned-weights-path WEIGHTS_PATH : Set path of finetuned weights as WEIGHTS_PATH
- 5.4 Inference (inference.py)
 - -h, --help: Show help message
 - latitude: Latitude of target point (as a float value)
 - longitude: Longitude of target point (as a float value)
 - time: Timestamp of the last input radar image (Format: YYYYMMDDHH)
 - **gap**: Gap between given images and target time (hour), the value should be 1, 2, 3, 4, 5, or 6
 - --data-path, --finetuned-weights-path, --gpus : See Section 5.1 and 5.3

6 Examples

We prepared sample radar images and sample data for running the codes in example_data folder.

1. Pretraining example:

python pretrain.py --data-path ../example_data --gpus 0,1,2,3

2. Finetuning example:

python finetuning-ours.py --data-path ../example_data --sampled-path ../example_data/sampled.pkl --pretrained-weights-path checkpoints/pretraining_10.pkt --gpus 0.1,2,3

3. Evaluation example:

python evaluation.py --data-path ../example_data --sampled-path ../example_data/sampled.pkl --finetuned-weights-path checkpoints/finetuning_ours_10.pkt --gpus 0.1,2,3

4. Inference example:

python inference.py 35.0 126.0 2019060101 1 --data-path ../example_data --finetuned-weights-path checkpoints/finetuned.pkt --gpus 0,1,2,3