

Pull Requests

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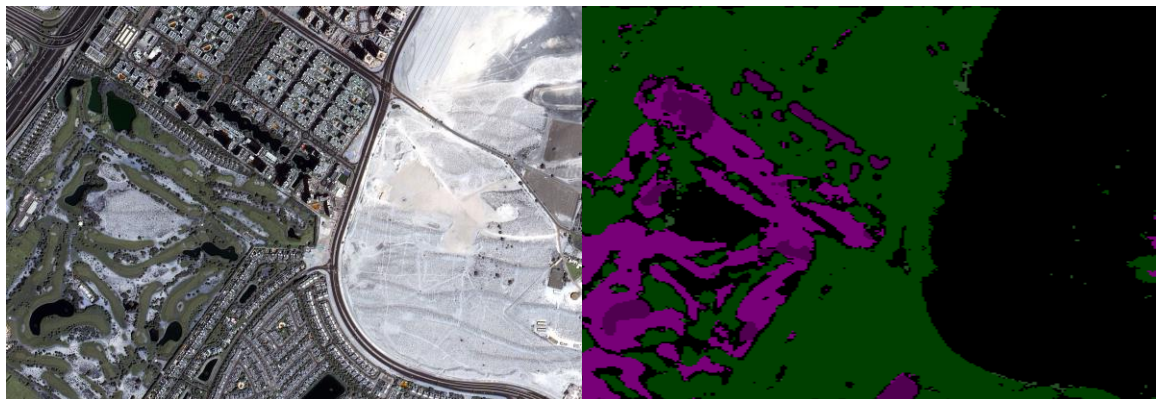
2022/07/06

Created PRs List

- 2022-U-Net: Add support for MBRSC dataset. Train model on this dataset.
- 2022-Flooding: Add support for tensorboard
- 2022-BackgroundMatting: Use Hydra for configuration management.

2022-U-Net

- This repo is about semantic segmentation using U-Net
- Problem: Only test on KITTI dataset.
- **My PRs:**
 - **Add support for MBRSC dataset.**
 - **Data collection related code**
 - Dataset structure is different
 - **Mask image translation related code**
 - Mask image format is different
 - **Train & Test model on MBRSC dataset.**
 - Train for 50 epoch, gain test accuracy of 0.7654



Input RGB image

Output Segmentation Mask

A screenshot of the Kaggle website showing a dataset page. The page title is "Semantic segmentation of aerial imagery" and it is described as "Satellite images of Dubai, the UAE segmented into 6 classes". The page includes a search bar, navigation links, and a list of dataset details. The "About Dataset" section includes a context paragraph, a content list, and a usability score of 7.50. The "Content" section lists six classes: Building, Land (unpaved area), Road, Vegetation, Water, and Unlabeled, each with a unique color code.

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Semantic segmentation of aerial imagery
Satellite images of Dubai, the UAE segmented into 6 classes

[Data](#) [Code \(8\)](#) [Discussion \(1\)](#) [Metadata](#)

About Dataset Usability 7.50

Context
Humans in the Loop is publishing an open access dataset annotated for a joint project with the Mohammed Bin Rashid Space Center in Dubai, the UAE.

Content
The dataset consists of aerial imagery of Dubai obtained by MBRSC satellites and annotated with pixel-wise semantic segmentation in 6 classes. The total volume of the dataset is 72 images grouped into 6 larger tiles. The classes are:

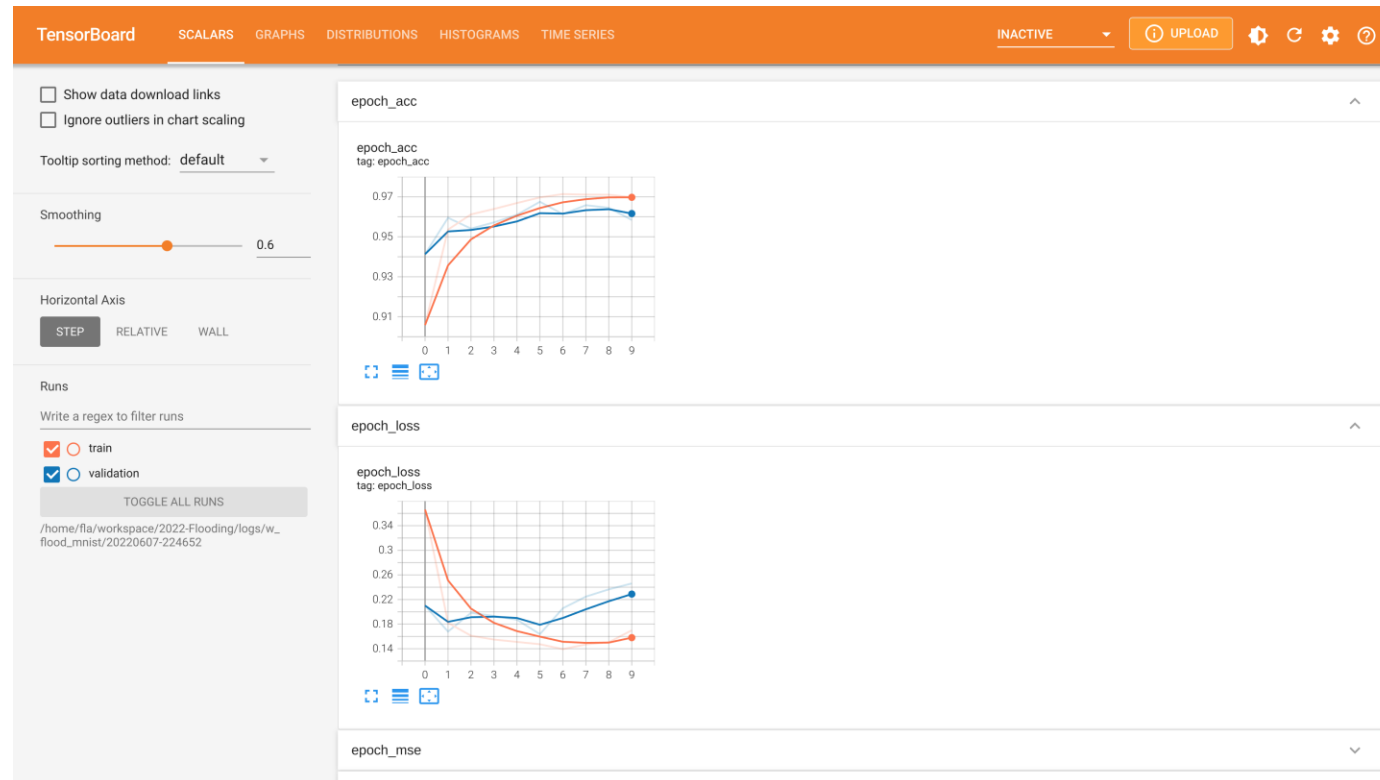
1. Building: #3C1098
2. Land (unpaved area): #8429F6
3. Road: #6EC1E4
4. Vegetation: #FEDD3A
5. Water: #E2A929
6. Unlabeled: #9B9B9B

View Active Events

MBRSC Dataset from Kaggle

2022-Flooding

- This repo is about using 'flooding' to improve performance of ANN.
- Problem: Training without logger.
- **My PRs:**
 - **Add support for Tensorboard.**



Tensorboard

2022-BackgroundMatting

- This repo is about background matting, reimplemented using Pytorch.
- Problem: Multi-Model & Multi-experiment involved, complex parameter configuration.
- **My PRs:**
 - **Using Hydra to handle configuration management.**
 - **Modify Action codes & README.md correspondingly**

2022-BackgroundMatting

Original: Using Command Line for configuration of each experiment

Training Base Network

The Base Network includes ASPP module from DeepLabV3. I used pretrained DeepLabV3 weight([best_deeplabv3_resnet50_voc_os16.pth](#)).

```
usage: train_base.py [-h] [--train_rgb_path TRAIN_RGB_PATH] [--train_alp_path TRAIN_ALP_PATH] [--train_bck_path TRAIN_BCK_PATH] [--valid_rgb_path VALID_RGB_PATH] [--valid_alp_path VALID_ALP_PATH] [--valid_bck_path VALID_BCK_PATH] [--checkpoint_path CHECKPOINT_PATH] [--logging_path LOGGING_PATH] [--batch_size BATCH_SIZE] [--num_workers NUM_WORKERS] [--pretrained_model PRETRAINED_MODEL] [--epochs EPOCHS]

optional arguments:
  -h, --help            show this help message and exit
  --train_rgb_path TRAIN_RGB_PATH
                        foreground data directory path for training
  --train_alp_path TRAIN_ALP_PATH
                        alpha matte data directory path for training
  --train_bck_path TRAIN_BCK_PATH
                        background data directory path for training
  --valid_rgb_path VALID_RGB_PATH
                        foreground data directory path for validation
  --valid_alp_path VALID_ALP_PATH
                        alpha matte data directory path for validation
  --valid_bck_path VALID_BCK_PATH
                        background data directory path for validation
  --checkpoint_path CHECKPOINT_PATH
                        checkpoint saving dir path
  --logging_path LOGGING_PATH
                        path to save logs
  --batch_size BATCH_SIZE
                        batch size
  --num_workers NUM_WORKERS
                        num workers
  --pretrained_model PRETRAINED_MODEL
                        pretrained model path
  --epochs EPOCHS      epochs to train
```

Training Whole Network (Refinement Network)

After training the Base Network, train the Base Network and Refinement Network jointly.

```
usage: train_refine.py [-h] [--train_rgb_path TRAIN_RGB_PATH] [--train_alp_path TRAIN_ALP_PATH] [--train_bck_path TRAIN_BCK_PATH] [--valid_rgb_path VALID_RGB_PATH] [--valid_alp_path VALID_ALP_PATH] [--valid_bck_path VALID_BCK_PATH] [--checkpoint_path CHECKPOINT_PATH] [--logging_path LOGGING_PATH] [--batch_size BATCH_SIZE] [--num_workers NUM_WORKERS] [--pretrained_model PRETRAINED_MODEL] [--epochs EPOCHS]

optional arguments:
  -h, --help            show this help message and exit
  --train_rgb_path TRAIN_RGB_PATH
                        foreground data directory path for training
  --train_alp_path TRAIN_ALP_PATH
                        alpha matte data directory path for training
  --train_bck_path TRAIN_BCK_PATH
                        background data directory path for training
  --valid_rgb_path VALID_RGB_PATH
                        foreground data directory path for validation
  --valid_alp_path VALID_ALP_PATH
                        alpha matte data directory path for validation
  --valid_bck_path VALID_BCK_PATH
                        background data directory path for validation
  --checkpoint_path CHECKPOINT_PATH
                        checkpoint saving dir path
  --logging_path LOGGING_PATH
                        path to save logs
  --batch_size BATCH_SIZE
                        batch size
  --num_workers NUM_WORKERS
                        num workers
  --pretrained_model PRETRAINED_MODEL
                        pretrained model path
  --epochs EPOCHS      epochs to train
```

Test Image Background Matting

You can download my trained weight form [here](#).

Using trained weight, you can test image background matting.

Make sure that related image and background data are same order in each directory.

```
usage: test_image.py [-h] [--pretrained_model PRETRAINED_MODEL] [--output_path OUTPUT_PATH] src_path bck_path {com,alp,fg,ref}

positional arguments:
  src_path            source directory path
  bck_path            background directory path
  {com,alp,fg,ref}   choose output types from [composite layer, alpha matte, foreground residual,]

optional arguments:
  -h, --help            show this help message and exit
  --pretrained_model PRETRAINED_MODEL
                        pretrained model path
  --output_path OUTPUT_PATH
                        output directory path
```

2022-BackgroundMatting

Now: Using configuration files.

```
! test_image.yaml | train_base.yaml | train_refine.yaml | default.yaml X
app > configs > data > ! default.yaml
1 # @package _global_
2
3 # Dataset path setting
4 original_work_dir: ${hydra:runtime.cwd}
5
6 data_root: ${original_work_dir}/data
7
8 rgb_data_dir: ${data_root}/VideoMatte240K_JPEG_SD
9 bck_data_dir: ${data_root}/Backgrounds
10
11 train_rgb_path: ${rgb_data_dir}/train/fgr
12 train_alp_path: ${rgb_data_dir}/train/pha
13 valid_rgb_path: ${rgb_data_dir}/test/fgr
14 valid_alp_path: ${rgb_data_dir}/test/pha
15
16 train_bck_path: ${bck_data_dir}/train
17 valid_bck_path: ${bck_data_dir}/test
```

Dataset path setting

```
! test_image.yaml | train_base.yaml X | train_refine.yaml | default.yaml
app > configs > ! train_base.yaml
1 # Experiment setting
2
3 checkpoint_path: ${original_work_dir}/result/checkpoint/basenet
4 |
5 logging_path: ${original_work_dir}/logs/basenet
6
7 batch_size: 2
8 num_workers: 2
9 epochs: 10
10 pretrained_model: ${original_work_dir}/data/pretrained/best_deeplabv3_resnet50_voc_os16.pth
11
12 defaults:
13 | - data: default.yaml
```

Experiment: Base-Model Training

```
! test_image.yaml | train_base.yaml | train_refine.yaml X | default.yaml
app > configs > ! train_refine.yaml
1 # Experiment setting
2
3 checkpoint_path: ${original_work_dir}/result/checkpoint/basenet
4
5 logging_path: ${original_work_dir}/logs/basenet
6
7 batch_size: 2
8 num_workers: 2
9 epochs: 10
10 pretrained_model: ${original_work_dir}/data/pretrained/best_deeplabv3_resnet50_voc_os16.pth
11
12 defaults:
13 | - data: default.yaml
```

Experiment: Refine-Model Training

```
! test_image.yaml X | train_base.yaml | train_refine.yaml | default.yaml
app > configs > ! test_image.yaml
1 # Experiment setting
2 original_work_dir: ${hydra:runtime.cwd}
3 pretrained_model: ${original_work_dir}/app/checkpoint_epoch0_iter54999.pth
4
5 src_path: ${original_work_dir}/tiny_dataset/test/img
6 bck_path: ${original_work_dir}/tiny_dataset/test/bgr
7 output_path: ${original_work_dir}/outputs
8 output_type: com alp fgr err ref # 'com', 'alp', 'fgr', 'err', 'ref'
```

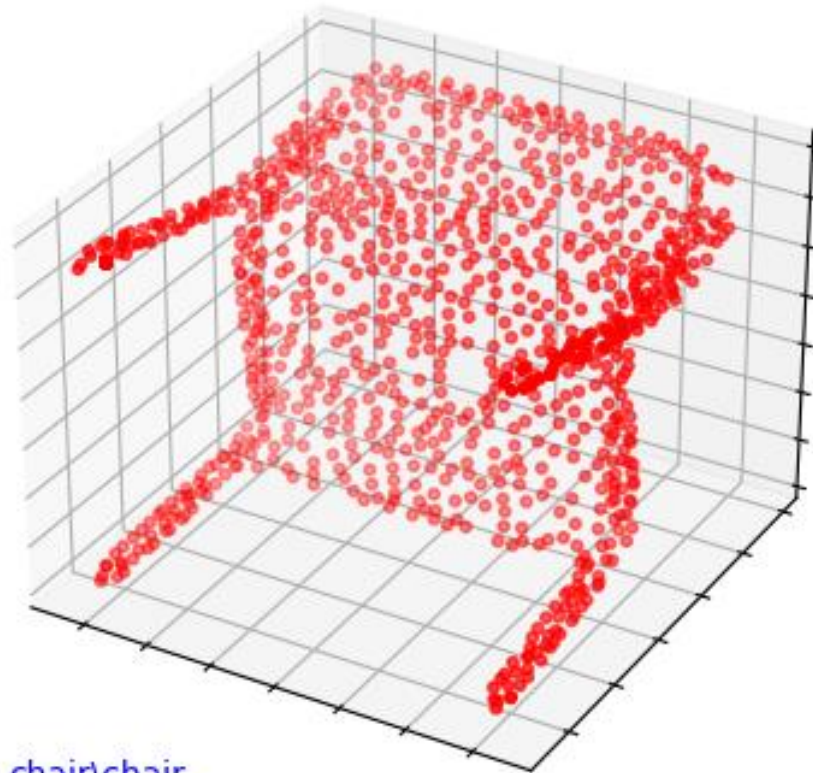
Experiment: Model Testing

Reviewed PR

- 2022-PCT-Lightning: Alternative Visualization from @Yoharol

2022-PCT-Lightning

- This repo is about Point Cloud Classification
- Problem: Visualization result is not so clear. Using C++ to render, need compiling.
- **Recieved PR:**
 - **Using Matplotlib to visualize the result.**



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