SKIN ABNORMAL DETECTION

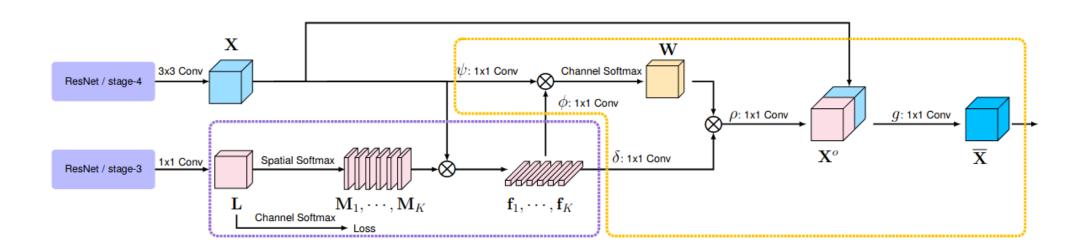
With mmsegmentation

Method used: OCRNET

Config file: ocnfigs/ocrnet/ocrnet_hr18s_4xb2-40k_cityscapes-512x1024.py

Checkpoin file:

 $https://download.openmmlab.com/mmsegmentation/v0.5/ocrnet/ocrnet_hr18s_4xb2-40k_cityscapes-512x1024/ocrnet_hr18s_4xb2-40k_cityscapes-512x1024_20230227_145026-6c052a14.pth$



Config file modified:

- Change data path, datatype
- Define number of batches and gpu
- Define default hook for logging and saving checkpoint
- Define dataset (type, root, prefix, pipeline)
- Define param_scheduler for auto modify learning rate
- Define the output num for two decoders
- Load check point pretrained
- Define epoch and iteration

```
    #Define num of class

• cfg.model.decode head[0].num classes = 2
• cfg.model.decode head[1].num classes = 2

    #Define batch norm

• cfg.norm cfg = dict(type='BN', requires grad=True)
• cfg.model.backbone.norm cfg = cfg.norm cfg
• cfg.model.decode head[0].norm cfg = cfg.norm cfg
• cfg.model.decode head[1].norm cfg = cfg.norm cfg
• # Modify dataset type and path
• cfg.dataset type = 'ISICDATASET '
• cfq.data root = '/content/dataset'
• cfq.train dataloader.dataset.type = 'ISICDATASET'
• cfg.train dataloader.dataset.data root = '/content/dataset'
• cfg.train dataloader.dataset.data prefix = dict(img path='images/ISIC2018 Task1-2 Training Input',
  seg map path='groundTruth/ISIC2018 Task1 Training GroundTruth')
• cfq.train dataloader.dataset.pipeline = cfq.train pipeline
  cfq.val dataloader.dataset.type = 'ISICDATASET '
• cfg.val dataloader.dataset.data root = '/content/dataset'
• cfg.val dataloader.dataset.data prefix = dict(img path='images/ISIC2018 Task1-2 Validation Input',
  seg map path='groundTruth/ISIC2018 Task1 Validation GroundTruth')
• cfg.val dataloader.dataset.pipeline = cfg.test pipeline
```

```
• cfg.test dataloader.num workers = 2
• cfg.test dataloader.batch size = 1
  cfg.test dataloader.dataset.data root = '/content/dataset'
• cfg.test_dataloader.dataset.data_prefix = dict(img_path='images/ISIC2018_Task1-2_Test_Input', seg_map_path='groundTruth/ISIC20T8_Task1_Test_GroundTruth')
• cfg.test dataloader.dataset.pipeline = cfg.test pipeline
• cfg.test evaluator = dict(
• type='IoUMetric',
iou metrics=['mIoU'],

    format only=False,

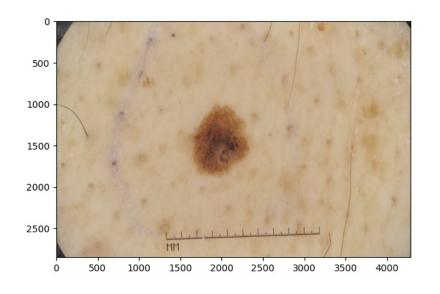
• output dir='work dirs/format results'

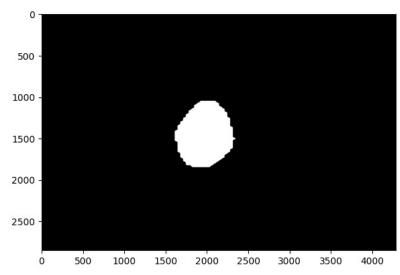
    # the number of samples and workers per GPU

  cfg.train dataloader.batch size = 4
• cfg.train dataloader.num workers = 1
  cfg.work dir = './work dirs/final'
 cfg.train cfg = dict(
• type='EpochBasedTrainLoop', max epochs=5, val begin=1, val interval=1)
```

```
• cfg.param scheduler = [
• dict(type='LinearLR', by epoch=False, start factor=0.1, begin=0, end=200),
• dict(
• type='PolyLR',
• eta min=0.0001,
• power=0.9,
• begin=0,
• end=160,
• by epoch=False)
• cfg.default hooks = dict(
• timer=dict(type='IterTimerHook'),
• logger=dict(type='LoggerHook', interval=50, log metric by epoch=False),
• param scheduler=dict(type='ParamSchedulerHook'),
 checkpoint=dict(type='CheckpointHook', interval = 1000, by epoch=False),
 sampler seed=dict(type='DistSamplerSeedHook'))
  cfg.log processor = dict(by epoch=True)
 cfg['randomness'] = dict(seed=32)
• cfg.dump('/content/mmsegmentation/configs/ocrnet/ocrnet khanh.py')
  #Load pretrain model
 cfg.load from = "https://download.openmmlab.com/mmsegmentation/v0.5/ocrnet/ocrnet hr18s 4xb2-40k cityscapes-
  512x1024/ocrnet hr18s 4xb2-40k cityscapes-512x1024 20230227 145026-6c052a14.pth"
```

Dataset





- Add custom palatte for each label
- Modify file images suffix and seg_map_suffix

Train command

```
from mmengine.runner import Runner
runner = Runner.from_cfg(cfg)
runner.train()
```

Evaluation command

```
cfg =
   Config.fromfile('/content/mmsegmentation/configs/ocrnet/ocrnet_kha
   nh.py')

checkpoint_path =
   '/content/mmsegmentation/work_dirs/tutorial/iter_1000.pth'

cfg.model.pretrained = checkpoint_path

runner_1 = Runner.from_cfg(cfg)

runner.test()
```

Review:

The model does extremely good at predict the abnormal region of for all the test images. At the early stage of trainning, model reach high accuracy, and then quickly to converge.

