

# SKIN ABNORMAL DETECTION

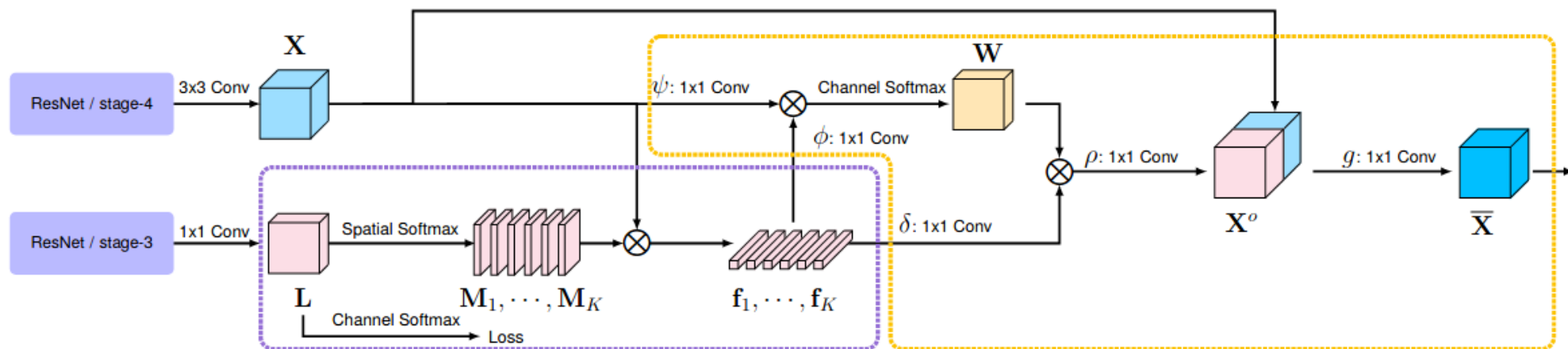
With mmsegmentation

# Method used: OCRNET

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

Checkpoint file:

`https://download.openmmlab.com/mmdetection/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_'`
- `cfg.data_root = '/content/dataset'`
- `cfg.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')`
- `cfg.train_dataloader.dataset.pipeline = cfg.train_pipeline`
- `cfg.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/ISIC2018_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)
.
```

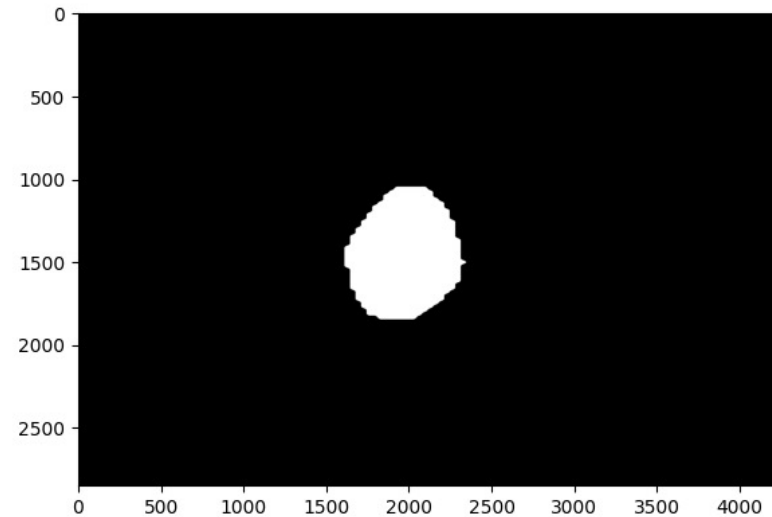
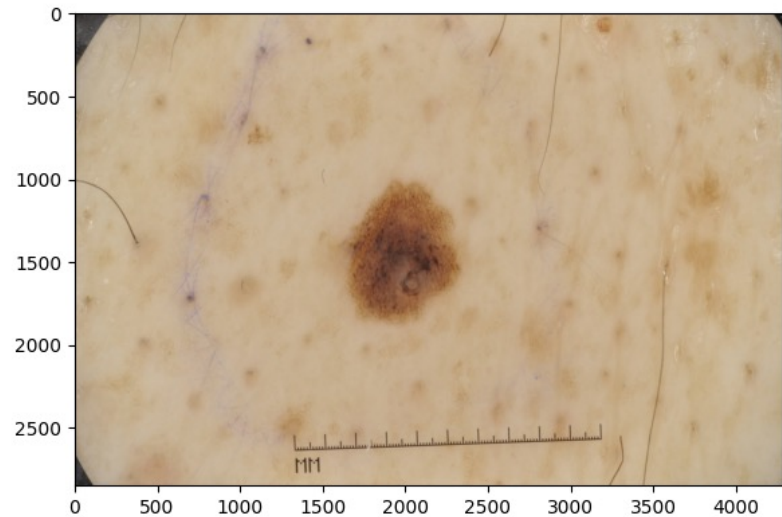
```
cfg.param_scheduler = [
dict(type='LinearLR', by_epoch=False, start_factor=0.1, begin=0, end=200),
dict(
type='PolyLR',
eta_min=0.0001,
power=0.9,
begin=0,
end=160,
by_epoch=False)
]
cfg.default_hooks = dict(
timer=dict(type='IterTimerHook'),
logger=dict(type='LoggerHook', interval=50, log_metric_by_epoch=False),
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/mmdetection/configs/ocrnet/ocrnet_khanh.py')

#Load pretrain model

cfg.load_from = "https://download.openmmlab.com/mmdetection/v0.5/ocrnet/ocrnet_hr18s_4xb2-40k_cityscapes-512x1024/ocrnet_hr18s_4xb2-40k_cityscapes-512x1024_20230227_145026-6c052a14.pth"
```

# Dataset



- Add custom palette 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/mmdetection/configs/ocrnet/ocrnet_khanh.py')  
checkpoint_path =  
'/content/mmdetection/work_dirs/tutorial/iter_1000.pth'  
cfg.model.pretrained = checkpoint_path  
runner_1 = Runner.from_cfg(cfg)  
runner.test()
```

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# Review:

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

