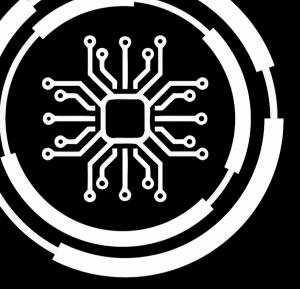


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1. Train Summary

1.1 Overview

- Train name: ISIC2019II
- Creator of the train: Karl Kindermann
- Location: https://git.rwth-aachen.de/padme-development/padme-train-depot/-/archive/main/padme-train-depot-main.zip?path=ISIC2019II

1.2 Image Building File

```
Listing 1.1: Image building file

FROM pytorch/pytorch:2.0.1-cuda11.7-cudnn8-runtime

RUN apt-get update && apt-get install -y git

WORKDIR /usr/src/app

LABEL "envs"="[{\"name\":\"FHIR_SERVER\",\"type\":\"string
\",\"required\":true},{\"name\":\"FHIR_PORT\",\"type\":\"
number\",\"required\":true},{\"name\":\"BATCH_SIZE\",\"type
\":\"number\",\"required\":true},{\"name\":\"LR\",\"type
\":\"string\",\"required\":true},{\"name\":\"EIGTH_DECAY
\",\"type\":\"number\",\"required\":true}]"

COPY requirements.txt .

RUN pip install --no-cache-dir -r requirements.txt

COPY . .
```

1.3 Requirements & Dependencies

Listing 1.2: List of all Requirements

```
requests
Pillow
pandas
numpy
git+https://github.com/beda-software/fhir-py.git
torchvision
datetime
pytz
tqdm
pytest-shutil
```

```
Legend: env query retrieve_previous_results execute_analysis save log
```

```
1 from fhirpy import SyncFHIRClient
2 import requests
3 from io import BytesIO
4 import numpy as np
5 import pandas as pd
6 from PIL import Image
7 from torch.utils.data import Dataset
8 import torchvision.transforms as transforms
10 class ISICDataset(Dataset):
      classes = {'NV': 0, 'MEL': 1, 'BKL': 2, 'DF': 3, 'SCC': 4, 'BCC': 5, '
     VASC': 6, 'AK': 7}
12
      """ISIC dataset."""
      def __init__(self, fhir_server, fhir_port, split='train', input_size
14
      =256):
15
          Args:
16
17
              fhir_server (string): Address of FHIR Server.
              fhir_port (string): Port of FHIR Server.
19
20
          self.fhir_server = fhir_server
          self.fhir_port = fhir_port
21
          self.split = split
22
          self.input_size = input_size
          # Create an instance
24
          client = SyncFHIRClient('http://{}:{}/fhir'.format(fhir_server,
      fhir_port))
          # Search for patients
26
          patients = client.resources('Patient') # Return lazy search set
27
          patients_data = []
29
          for patient in patients:
30
              patient_birthDate = None
31
32
                  patient_birthDate = patient.birthDate
              except:
33
                  pass
34
              # patinet_id, gender, birthDate
              patients_data.append([patient.id, patient.gender,
36
      patient_birthDate])
          patients_df = pd.DataFrame(patients_data, columns=["patient_id", "
      gender", "birthDate"])
          # Search for media
```

```
media_list = client.resources('Media').include('Patient', 'subject'
39
40
          media_data = []
           for media in media_list:
               media_bodySite = None
               media_reasonCode = None
               media_note = None
44
45
               try:
                   media_bodySite = media.bodySite.text
46
               except:
47
                   pass
48
49
               try:
50
                   media_reasonCode = media.reasonCode[0].text
51
               except:
52
                   pass
               try:
53
                   media_note = media.note[0].text
54
55
               except:
56
                   pass
               media_data.append([media.subject.id, media.id, media_bodySite,
57
      media_reasonCode, media_note, media.content.url])
          media_df = pd.DataFrame(media_data, columns=["patient_id", "
58
      media_id", "bodySite", "reasonCode", "note", "image_url"])
59
           self.data_df = pd.merge(patients_df, media_df, on='patient_id', how
      ='outer')
60
          self.data_df = self.data_df[self.data_df['note'].notna()].
      reset_index()
          self.trans = transforms.Compose([transforms.RandomHorizontalFlip(),
61
                                              transforms.RandomVerticalFlip(),
62
                                              transforms.ColorJitter(brightness
63
      =32. / 255., saturation=0.5),
                                              transforms.Resize(self.input_size)
64
                                              transforms.ToTensor()])
65
66
      def __len__(self):
          if self.split == "train":
68
69
               return int(len(self.data_df) * 0.8)
           elif self.split == "val":
70
               return int(len(self.data_df) * 0.2)
72
      def __getitem__(self, idx):
73
           val_start_id = int(len(self.data_df) * 0.8)
74
           if self.split == "train":
75
               idx = idx
76
           else:
77
               idx = idx + val_start_id
78
           img_url = self.data_df.loc[idx, 'image_url']
79
          note = self.data_df.loc[idx, 'note']
80
          y = self.classes[note]
81
          img_res = requests.get(img_url)
82
           if img_res.status_code == 200:
83
               x = Image.open(BytesIO(img_res.content))
84
               x = self.center_crop(x)
85
               x = self.trans(x)
86
               # Transform y
87
               y = np.int64(y)
               return {"image": x, "label": y}
89
90
               raise RuntimeError("Image url {} is not reachable!".format(
91
      img_url))
```

```
92
93
      def center_crop(self, pil_img):
           img_width, img_height = pil_img.size
           if img_width > img_height:
               crop_size = img_height
           else:
97
               crop_size = img_width
           return pil_img.crop(((img_width - crop_size) // 2,
99
                                 (img_height - crop_size) // 2,
100
                                 (img_width + crop_size) // 2,
101
                                 (img_height + crop_size) // 2))
102
                                Listing 1.3: isicdataset.py
1 import os
2 import os.path as osp
3 import numpy as np
5 import glob
6 import tqdm
7 import shutil
8 import pytz
9 import datetime
11 import torch
12 import torch.nn as nn
13 from torch.autograd import Variable
15 from isicnet import ISICNet
{\scriptsize 16} from isicdataset import ISICDataset
18 ## Define evaluation function
19 def _fast_hist(label_true, label_pred, n_class):
      hist = np.bincount(
20
           n_class * label_true.astype(int) +
21
           label_pred.astype(int), minlength=n_class ** 2).reshape(n_class,
22
      n_class)
      return hist
25 def label_accuracy_score(label_trues, label_preds, n_class=8):
26
      hist = np.zeros((n_class, n_class))
      hist += _fast_hist(label_trues, label_preds, n_class)
      acc = np.diag(hist).sum() / hist.sum()
28
      with np.errstate(divide='ignore', invalid='ignore'):
29
          precision = np.diag(hist) / hist.sum(axis=1)
30
      mean_precision = np.nanmean(precision)
31
      with np.errstate(divide='ignore', invalid='ignore'):
           recall = np.diag(hist) / hist.sum(axis=0)
33
      mean_recall = np.nanmean(recall)
34
35
      with np.errstate(divide='ignore', invalid='ignore'):
           iou = np.diag(hist) / (hist.sum(axis=1) + hist.sum(axis=0) - np.
36
      diag(hist))
37
      mean_iou = np.nanmean(iou)
      with np.errstate(divide='ignore', invalid='ignore'):
38
           f1 = (2 * np.diag(hist)) / (hist.sum(axis=1) + hist.sum(axis=0) + 2
39
      * np.diag(hist))
      mean_f1 = np.nanmean(f1)
40
      return acc, mean_precision, mean_recall, mean_iou, mean_f1
41
43 ## Define directory of output
44 here = osp.dirname(osp.abspath(__file__))
```

```
45 out_dir = osp.join(here, 'output')
46 if not os.path.exists(out_dir):
      os.makedirs(out_dir)
48 runs = sorted(glob.glob(os.path.join(out_dir, 'run_*')))
49 run_id = int(runs[-1].split(',_')[-1]) + 1 if runs else 0
50 experiment_dir = os.path.join(out_dir, 'run_{}'.format(str(run_id)))
51 if not os.path.exists(experiment_dir):
      os.makedirs(experiment_dir)
54 ## Define (input) variables from Docker Container environment variables
55 fhir_server = str(os.environ['FHIR_HOST_NAME'])
56 fhir_port = str(os.environ['FHIR_PORT'])
57 # num_station = int(os.environ['NUM_STATION'])
58 # sid = int(os.environ['SID'])
59 #Hyperparameters
60 batch_size = int(os.environ.get('BATCH_SIZE', '1'))
61 num_epoch = int(os.environ.get('NUM_EPOCH', '1'))
62 lr = float(os.environ.get('LR', '0.01'))
63 weight_decay = float(os.environ.get('WEIGHT_DECAY', '0.005'))
64 model_name = "resnet18"
66 ## Define (output) file formats
67 if not osp.exists(osp.join(experiment_dir, 'val_log.csv')):
      with open(osp.join(experiment_dir, 'val_log.csv'), 'w') as f:
      header = ['epoch', 'Loss', 'Acc', 'Precision', 'Recall', 'Iou', '
F1Score', 'train/Loss', 'elapsed_time']
          header = map(str, header)
          f.write(','.join(header) + '\n')
71
          print("Initial Log file")
72
74 cuda = torch.cuda.is_available()
75 torch.manual_seed(1337)
76 if cuda:
      torch.cuda.manual_seed(1337)
79 ## Initial Model
80 print("Initial Model")
81 model = ISICNet(backbone=model_name)
82 print("Initial Model {}".format(model_name))
83 if cuda:
      print("Cuda:", cuda)
84
      model = model.cuda()
85
86 else:
      print("Running on CPU")
87
88 ## Initial Datasets of train and val on station 1, 2, 3 and test
89 kwargs = {'num_workers': 4, 'pin_memory': True} if cuda else {}
90 print ("Initial Training Dataset")
91 train_dataloader = torch.utils.data.DataLoader(ISICDataset(fhir_server,
      fhir_port, split='train'), batch_size=batch_size, shuffle=True, **
      kwargs)
92 print("Initial Val Dataset")
93 val_dataloader = torch.utils.data.DataLoader(ISICDataset(fhir_server,
      fhir_port, split='val'), batch_size=batch_size, shuffle=False, **kwargs
94 ## Initial criterion (Cross Entropy Loss)
95 print("Initial Loss function")
96 criterion = nn.CrossEntropyLoss()
97 ## Initial Optimizers for station
98 print("Initial Optimizer")
99 optim = torch.optim.Adam(model.parameters(), lr=lr, weight_decay=
      weight_decay)
```

```
100
101 ## Load model from previous train
102 if run_id > 0:
       prev_experiment_dir = osp.join(out_dir, 'run_{}'.format(str(run_id - 1))
      ))
       if osp.exists(osp.join(prev_experiment_dir, 'best_model.pth.tar')):
104
           prev_best_model = torch.load(osp.join(prev_experiment_dir, ')
105
      best_model.pth.tar'))
           model.load_state_dict(prev_best_model['model_state_dict'])
106
           optim.load_state_dict(prev_best_model['optim_state_dict'])
107
           shutil.copy(osp.join(prev_experiment_dir, 'best_model.pth.tar'),
108
                        osp.join(experiment_dir, 'best_model.pth.tar'))
109
           print("Model loaded from previous train.")
           print("No previous best model found!")
112
113 else:
       torch.save({
114
115
           'epoch': 0,
           'optim_state_dict': optim.state_dict(),
116
           'model_state_dict': model.state_dict(),
           'best_acc': 0.0,
118
       }, osp.join(experiment_dir, 'best_model.pth.tar'))
119
120
121 timestamp_start = datetime.datetime.now(pytz.timezone('Asia/Tokyo'))
122 best_acc = 0.0
123 ## Run the training processing on the station
124 for epoch in range(num_epoch):
       model.train()
       train_loss = 0.0
126
       for batch_idx, sample in tqdm.tqdm(enumerate(train_dataloader), total=
127
      len(train_dataloader), desc='Station Train epoch=%d' % epoch, ncols=80,
       leave=False):
           assert model.training
128
           img, lbl = sample['image'], sample['label']
129
130
           if cuda:
               img, lbl = img.cuda(), lbl.cuda()
           img, lbl = Variable(img), Variable(lbl)
133
           optim.zero_grad()
134
           pred = model(img)
           loss = criterion(pred, lbl)
           train_loss = train_loss + loss.data.item()
136
           loss.backward()
           optim.step()
138
139
       train_loss = train_loss / len(train_dataloader)
140
       print("Train epoch {} finished with average train loss of {}.".format(
      epoch, train_loss))
142
143
       model.eval()
       val_loss = 0.0
144
       label_trues, label_preds = [], []
145
       for batch_idx, sample in tqdm.tqdm(enumerate(val_dataloader), total=len
146
      (val_dataloader), desc='Station Val epoch=%d' % epoch, ncols=80, leave=
      False):
           img, lbl = sample['image'], sample['label']
147
148
           if cuda:
               img, lbl = img.cuda(), lbl.cuda()
           img, lbl = Variable(img), Variable(lbl)
150
           with torch.no_grad():
151
               pred = model(img)
           loss = criterion(pred, lbl)
153
```

```
val_loss = val_loss + loss.data.item()
154
           lbl = lbl.data.cpu().numpy()
           pred = pred.data.max(1)[1].cpu().numpy()
156
           label_trues = np.concatenate((label_trues, lbl), axis=0)
           label_preds = np.concatenate((label_preds, pred), axis=0)
      val_loss = val_loss / len(val_dataloader)
159
      acc, mean_precision, mean_recall, mean_iou, mean_f1 =
160
      label_accuracy_score(label_trues, label_preds)
      with open(osp.join(experiment_dir, 'val_log.csv'), 'a') as f:
161
           elapsed_time = (datetime.datetime.now(pytz.timezone('Asia/Tokyo'))
162
      - timestamp_start).total_seconds()
163
           log = [epoch, val_loss, acc, mean_precision, mean_recall, mean_iou,
       mean_f1, train_loss, elapsed_time]
           log = map(str, log)
164
           f.write(','.join(log) + '\n')
165
166
      is_best = acc > best_acc
167
      if is_best:
168
          best_acc = acc
169
      torch.save({
170
           'epoch': epoch,
           'optim_state_dict': optim.state_dict(),
           'model_state_dict': model.state_dict(),
173
174
           'best_acc': best_acc,
175
      }, osp.join(experiment_dir, 'checkpoint.pth.tar'))
176
      if is_best:
           shutil.copy(osp.join(experiment_dir, 'checkpoint.pth.tar'), osp.
      join(experiment_dir, 'best_model.pth.tar'))
      print("Station Val epoch {} finished with loss of {}, acc of {},
178
      precision of {}, recall of {}, iou of {}, f1-score of {}.".format(epoch
       val_loss, acc, mean_precision, mean_recall, mean_iou, mean_f1))
179 print("Finished training process")
                                  Listing 1.4: main.py
1 import torch
2 import torch.nn as nn
3 from torchvision.models import resnet
5 class ISICNet(nn.Module):
      def __init__(self, n_feature=3, n_class=8, backbone='resnet18'):
           super(ISICNet, self).__init__()
           self.n_feature = n_feature
          self.n_class = n_class
           self.backbnone = backbone
10
           if self.backbnone == 'resnet18':
               base = resnet.resnet18(pretrained=True)
13
               self.resnet_expansion = 1
           self.in_block = nn.Sequential(
14
               nn.Conv2d(self.n_feature, 64, kernel_size=(7, 7), stride=(2, 2)
15
      , padding=(3, 3), bias=False),
16
               base.bn1,
               base.relu,
               base.maxpool)
18
           self.encoder1 = base.layer1
19
           self.encoder2 = base.layer2
20
           self.encoder3 = base.layer3
           self.encoder4 = base.layer4
           self.avgpool = base.avgpool
23
           self.flatten = nn.Flatten()
```

self.fc = nn.Linear(512*self.resnet_expansion, self.n_class , bias=

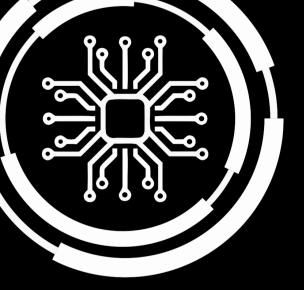
```
True)
26
      def forward(self, x):
27
          h = self.in_block(x)
          h = self.encoder1(h)
          h = self.encoder2(h)
          h = self.encoder3(h)
31
          h = self.encoder4(h)
32
          y = self.fc(self.flatten(self.avgpool(h)))
33
          return y
34
35
36 if __name__ == "__main__":
     model = ISICNet().cuda()
37
      x = torch.rand(2, 3, 512, 512).cuda()
38
      y = model(x)
40
      print(y)
      print(y.data.max(1)[1])
41
      \# t = torch.tensor([1.0, 0.0, 1.0])
42
      # y = torch.tensor([0.02, 0.05, 0.99])
43
      # bce = nn.BCELoss()
      \# 1 = \text{torch.log(torch.tensor(0.02))} + \text{torch.log(torch.tensor(1-0.05))} +
45
      torch.log(torch.tensor(0.99))
46
      # print(bce(y, t))
      # print(1)
                                 Listing 1.5: isicnet.py
1 {"resources":[{"id":"5bce186a-5005-4ca8-878e-9d8d4e0747c0","datasets":[{"id
      ":"0ed80d8f-6f34-45c7-979f-714681f59046a"},{"id":"a80bac0f-7f44-4fdd-
      a3c5-45f9aad175b6"}]}], "route":[{"id":"5bce186a-5005-4ca8-878e-9
      d8d4e0747c0","ownEnvs":[]}]}
```

Listing 1.6: route.json

1.5 Connection Parameters

Overall, 2 connection parameters have been found in the train:

Name	Type	Required?	
FHIR_SERVER	string	True	
FHIR_PORT	number	True	



2. Static Train Analysis

Summary 2.1

• Number of Lines: 223754

• Vulnerabilities per Line: 0.0

• Analysis Score: 1.0

Total

Summary: Low: 0 Medium: 0 High: 0

2.1 SAST

Summary: Low: 0 Medium: 0 High: 0

Detailed Vulnerabilities:

Vulnerability 2.1

Test-specific ID: 62d9a0c9581c461f41278053d31b06a1ee7b663f285f97c18d2b90af109a170c

Severity: Info **Identifiers:**

• semgrep_id: bandit.B101

• cwe: 754

• bandit_test_id: B101

Message:Improper Check for Unusual or Exceptional Conditions

File:ISIC2019II/main.py Start line: 128

2.2 Secret Detection

Summary: Low: 0 Medium: 0 High: 0

Detailed Secret Detection:

2.3 Dependency Analysis

Summary: Low: 0 Medium: 0 High: 0

Detailed Dependency Analysis:

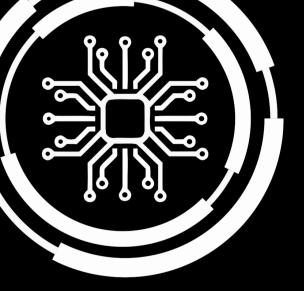
2.4 Blacklist Detection

The following patterns have been configured:



The following matches have been found:

None



3. Train Image Analysis

3.1 Vulnerabilities

Summary: Low: 22 Medium: 8 High: 0 Critical: 0

Detailed Vulnerabilities:

Critical:

High:

Medium:

Vulnerability 3.1

Test-specific ID: SNYK-UBUNTU2004-GNUTLS28-6069042 Severity: medium

Identifiers: Message:Information Exposure

Vulnerability 3.2

Test-specific ID: SNYK-UBUNTU2004-GNUTLS28-6172192 Severity: medium

Identifiers: Message:Information Exposure

Vulnerability 3.3

Test-specific ID: SNYK-UBUNTU2004-NCURSES-5423141 Severity: medium

Identifiers: Message:Out-of-bounds Write

Vulnerability 3.4

Test-specific ID: SNYK-UBUNTU2004-OPENSSH-6142083 Severity: medium

Identifiers: Message:CVE-2023-51767

Vulnerability 3.5

Test-specific ID: SNYK-UBUNTU2004-OPENSSL-5661533 **Severity:** medium **Identifiers: Message**:Allocation of Resources Without Limits or Throttling

Vulnerability 3.6

Test-specific ID: SNYK-UBUNTU2004-PAM-6170207 Severity: medium

Identifiers: Message:CVE-2024-22365

Vulnerability 3.7

Test-specific ID: SNYK-UBUNTU2004-TAR-6096088 Severity: medium

Identifiers: Message:CVE-2023-39804

Vulnerability 3.8

Test-specific ID: SNYK-UBUNTU2004-XZUTILS-5854646 Severity: medium

Identifiers: Message:CVE-2020-22916

Low:

Vulnerability 3.9

Test-specific ID: SNYK-UBUNTU2004-COREUTILS-583876 Severity: low

Identifiers: Message:Improper Input Validation

Vulnerability 3.10

Test-specific ID: SNYK-UBUNTU2004-GIT-580979 Severity: low

Identifiers: Message:Improper Input Validation

3.1 Vulnerabilities 17

Vulnerability 3.11

Test-specific ID: SNYK-UBUNTU2004-GLIBC-1297554 Severity: low

Identifiers: Message: Use After Free

Vulnerability 3.12

Test-specific ID: SNYK-UBUNTU2004-GLIBC-2415100 **Severity:** low **Identifiers: Message**:Allocation of Resources Without Limits or Throttling

Vulnerability 3.13

Test-specific ID: SNYK-UBUNTU2004-GLIBC-5901432 Severity: low

Identifiers: Message: Use After Free

Vulnerability 3.14

Test-specific ID: SNYK-UBUNTU2004-GLIBC-5901476 Severity: low

Identifiers: Message: Use After Free

Vulnerability 3.15

Test-specific ID: SNYK-UBUNTU2004-GNUPG2-3035407 Severity: low

Identifiers: Message:Out-of-bounds Write

Vulnerability 3.16

Test-specific ID: SNYK-UBUNTU2004-KRB5-579303 Severity: low

Identifiers: Message:Integer Overflow or Wraparound

Vulnerability 3.17

Test-specific ID: SNYK-UBUNTU2004-LIBPNG16-3124878 Severity: low

Identifiers: Message: NULL Pointer Dereference

Vulnerability 3.18

Test-specific ID: SNYK-UBUNTU2004-NCURSES-1656318 Severity: low

Identifiers: Message:Out-of-bounds Write

Vulnerability 3.19

Test-specific ID: SNYK-UBUNTU2004-NCURSES-2770341 Severity: low

Identifiers: Message:Out-of-bounds Read

Vulnerability 3.20

Test-specific ID: SNYK-UBUNTU2004-OPENSSH-1047872 Severity: low

Identifiers: Message:Information Exposure

Vulnerability 3.21

Test-specific ID: SNYK-UBUNTU2004-OPENSSL-5786273 Severity: low

Identifiers: Message: Inefficient Regular Expression Complexity

Vulnerability 3.22

Test-specific ID: SNYK-UBUNTU2004-OPENSSL-5811821 Severity: low

Identifiers: Message: Excessive Iteration

Vulnerability 3.23

Test-specific ID: SNYK-UBUNTU2004-PATCH-2325780 Severity: low

Identifiers: Message: Release of Invalid Pointer or Reference

Vulnerability 3.24

Test-specific ID: SNYK-UBUNTU2004-PATCH-582546 Severity: low

Identifiers: Message: Double Free

3.1 Vulnerabilities

Vulnerability 3.25

Test-specific ID: SNYK-UBUNTU2004-PCRE3-580031 Severity: low

Identifiers: Message: Uncontrolled Recursion

Vulnerability 3.26

Test-specific ID: SNYK-UBUNTU2004-PROCPS-5816664 Severity: low

Identifiers: Message:Out-of-bounds Write

Vulnerability 3.27

Test-specific ID: SNYK-UBUNTU2004-SHADOW-5425687 Severity: low

Identifiers: Message: Arbitrary Code Injection

Vulnerability 3.28

Test-specific ID: SNYK-UBUNTU2004-SHADOW-577863 Severity: low

Identifiers: Message: Time-of-check Time-of-use (TOCTOU)

Vulnerability 3.29

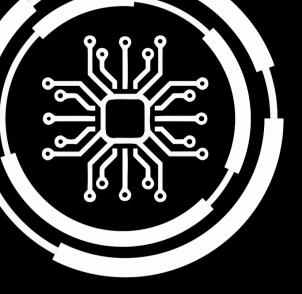
Test-specific ID: SNYK-UBUNTU2004-SYSTEMD-3339226 Severity: low

Identifiers: Message:CVE-2023-26604

Vulnerability 3.30

Test-specific ID: SNYK-UBUNTU2004-SYSTEMD-6137854 Severity: low

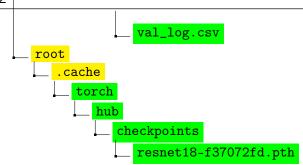
Identifiers: Message:CVE-2023-7008



4. Dynamic Train Analysis

- 4.1 Image size differences after simulation:
- 4.1.1 Simulated route:
 - Station 5bce186a-5005-4ca8-878e-9d8d4e0747c0: 225937765
- 4.1.2 Directory Overview of changed files:

```
Legend: Modified Added Deleted
  Train Root
       conda
         lib
             python3.10
                site-packages
                  numpy
                      __pycache__
                          _config__.cpython-310.pyc
     usr
       src
              __pycache__
                isicdataset.cpython-310.pyc
                 isicnet.cpython-310.pyc
              output
                run_0
                   best_model.pth.tar
                   checkpoint.pth.tar
```



4.1.3 ./simulationResults/_config_.cpython-310.pyc

No Preview Available! https://git.rwth-aachen.de/padme-development/padme-train-depot/-/jobs/4570446/artifacts/file/simulationResults/_config__.cpython-310.pyc

4.1.4 ./simulationResults/isicdataset.cpython-310.pyc

No Preview Available! https://git.rwth-aachen.de/padme-development/padme-train-depot/-/jobs/4570446/artifacts/file/simulationResults/isicdataset.cpython-310.pyc

4.1.5 ./simulationResults/isicnet.cpython-310.pyc

No Preview Available! https://git.rwth-aachen.de/padme-development/padme-train-depot/-/jobs/4570446/artifacts/file/simulationResults/isicnet.cpython-310.pyc

4.1.6 ./simulationResults/best_model.pth.tar

No Preview Available! https://git.rwth-aachen.de/padme-development/padme-train-depot/-/jobs/4570446/artifacts/file/simulationResults/best_model.pth.tar

4.1.7 ./simulationResults/checkpoint.pth.tar

No Preview Available! https://git.rwth-aachen.de/padme-development/padme-train-depot/-/jobs/4570446/artifacts/file/simulationResults/checkpoint.pth.tar

4.1.8 ./simulationResults/val_log.csv

```
Listing 4.1: val_log.csv

epoch, Loss, Acc, Precision, Recall, Iou, F1Score, train/Loss,
    elapsed_time

0,714102.0,0.0,0.0,0.0,0.0,0.0,7.464000657200813,1.320086
```

4.1.9 ./simulationResults/resnet18-f37072fd.pth

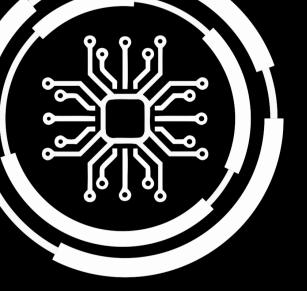
No Preview Available! https://git.rwth-aachen.de/padme-development/padme-train-depot/-/jobs/4570446/artifacts/file/simulationResults/resnet18-f37072fd.pth

4.2 Execution metrics:

- Maximum CPU Usage: 5.8371202223355425
- Maximum Memory Usage:562102272
- Maximum Number of PIDS:11

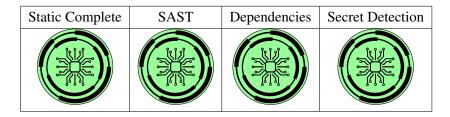
4.3 Network I/O:

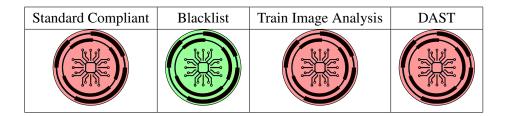
- RX (Reading):47501630
- TX (Transmitting):340878



5. Conclusion

The PADME Security Audit has come to the following conclusion:





Warnings

- Identified network traffic in dynamic analysis.
- Couldn't detect compliance with standards (usage of the python module padme_conductor [https://pypi.org/project/padme-conductor]).

Audit Result

