## IGVC Full Dataset

March 20, 2020

```
[0]: # Mount Drive
    from google.colab import drive
    drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).
```

```
[0]: # Restart runtime after running once

# Library for pretrained segmentation models fo PyTorch
!pip install segmentation-models-pytorch==0.1.0

# Catalyst library
!pip install -U catalyst
```

Collecting segmentation-models-pytorch==0.1.0

Downloading https://files.pythonhosted.org/packages/70/88/763a25dfe076a9 f30f33466b1bd0f2d31b915b88d4cb4481fe4043cf26b4/segmentation\_models\_pytorch-0.1.0 -py3-none-any.whl (42kB)

```
| | 51kB 1.9MB/s
```

Collecting pretrainedmodels==0.7.4

Downloading https://files.pythonhosted.org/packages/84/0e/be6a0e58447ac1 6c938799d49bfb5fb7a80ac35e137547fc6cee2c08c4cf/pretrainedmodels-0.7.4.tar.gz (58kB)

```
| 61kB 3.8MB/s
```

Collecting efficientnet-pytorch>=0.5.1

Downloading https://files.pythonhosted.org/packages/b8/cb/0309a6e3d404862ae4bc 017f89645cf150ac94c14c88ef81d215c8e52925/efficientnet\_pytorch-0.6.3.tar.gz Requirement already satisfied: torchvision>=0.3.0 in

/usr/local/lib/python3.6/dist-packages (from segmentation-models-pytorch==0.1.0) (0.5.0)

Requirement already satisfied: torch in /usr/local/lib/python3.6/dist-packages (from pretrainedmodels==0.7.4->segmentation-models-pytorch==0.1.0) (1.4.0) Collecting munch

Downloading https://files.pythonhosted.org/packages/cc/ab/85d8da5c9a45e072301b eb37ad7f833cd344e04c817d97e0cc75681d248f/munch-2.5.0-py2.py3-none-any.whl Requirement already satisfied: tqdm in /usr/local/lib/python3.6/dist-packages (from pretrainedmodels==0.7.4->segmentation-models-pytorch==0.1.0) (4.38.0)

```
Requirement already satisfied: numpy in /usr/local/lib/python3.6/dist-packages
(from torchvision>=0.3.0->segmentation-models-pytorch==0.1.0) (1.18.2)
Requirement already satisfied: six in /usr/local/lib/python3.6/dist-packages
(from torchvision>=0.3.0->segmentation-models-pytorch==0.1.0) (1.12.0)
Requirement already satisfied: pillow>=4.1.1 in /usr/local/lib/python3.6/dist-
packages (from torchvision>=0.3.0->segmentation-models-pytorch==0.1.0) (7.0.0)
Building wheels for collected packages: pretrainedmodels, efficientnet-pytorch
   Building wheel for pretrainedmodels (setup.py) ... done
   Created wheel for pretrainedmodels: filename=pretrainedmodels-0.7.4-cp36-none-
any.whl size=60962
\verb|sha| 256 = 8796 = \verb|a1a7273a| 0 = 5773a \\ 1176 f c \\ 77412 d 6 = 94d \\ 706 501116 \\ a 0 b 6 b 3 b 6 a 652 b d 2656 \\ a 65
   Stored in directory: /root/.cache/pip/wheels/69/df/63/62583c096289713f22db605a
a2334de5b591d59861a02c2ecd
   Building wheel for efficientnet-pytorch (setup.py) ... done
   Created wheel for efficientnet-pytorch:
filename=efficientnet_pytorch-0.6.3-cp36-none-any.whl size=12422
Stored in directory: /root/.cache/pip/wheels/42/1e/a9/2a578ba9ad04e776e80bf0f7
0d8a7f4c29ec0718b92d8f6ccd
Successfully built pretrainedmodels efficientnet-pytorch
Installing collected packages: munch, pretrainedmodels, efficientnet-pytorch,
segmentation-models-pytorch
Successfully installed efficientnet-pytorch-0.6.3 munch-2.5.0
pretrainedmodels-0.7.4 segmentation-models-pytorch-0.1.0
Collecting catalyst
   Downloading https://files.pythonhosted.org/packages/c9/02/8328fb3b01b7c1
28a4f8abfe2df9508bcae64866a8c41c44ea9f7981f44f/catalyst-20.3.2-py2.py3-none-
any.whl (388kB)
                                        | 389kB 3.4MB/s
Collecting crc32c>=1.7
   Downloading https://files.pythonhosted.org/packages/ab/82/f60248c01a8a23ae07bd
4c43d78d69b20ffe324311db3b0785e391aa09d2/crc32c-2.0-cp36-cp36m-manylinux1_x86_64
.whl
Requirement already satisfied, skipping upgrade: tensorboard>=1.14.0 in
/tensorflow-1.15.0/python3.6 (from catalyst) (1.15.0)
Requirement already satisfied, skipping upgrade: torchvision>=0.2.1 in
/usr/local/lib/python3.6/dist-packages (from catalyst) (0.5.0)
Requirement already satisfied, skipping upgrade: PyYAML in
/usr/local/lib/python3.6/dist-packages (from catalyst) (3.13)
Collecting Pillow<7
   Downloading https://files.pythonhosted.org/packages/8a/fd/bbbc569f98f478
13c50a116b539d97b3b17a86ac7a309f83b2022d26caf2/Pillow-6.2.2-cp36-cp36m-manylinux
1_x86_64.whl (2.1MB)
                                        | 2.1MB 38.6MB/s
Requirement already satisfied, skipping upgrade: scikit-image>=0.14.2 in
/usr/local/lib/python3.6/dist-packages (from catalyst) (0.16.2)
Requirement already satisfied, skipping upgrade: opency-python in
/usr/local/lib/python3.6/dist-packages (from catalyst) (4.1.2.30)
```

## Collecting deprecation

```
Downloading https://files.pythonhosted.org/packages/b9/2a/d5084a8781398cea745c
01237b95d9762c382697c63760a95cc6a814ad3a/deprecation-2.0.7-py2.py3-none-any.whl
Requirement already satisfied, skipping upgrade: packaging in
/usr/local/lib/python3.6/dist-packages (from catalyst) (20.3)
Requirement already satisfied, skipping upgrade: tqdm>=4.33.0 in
/usr/local/lib/python3.6/dist-packages (from catalyst) (4.38.0)
Requirement already satisfied, skipping upgrade: torch>=1.0.0 in
/usr/local/lib/python3.6/dist-packages (from catalyst) (1.4.0)
Requirement already satisfied, skipping upgrade: plotly>=4.1.0 in
/usr/local/lib/python3.6/dist-packages (from catalyst) (4.4.1)
Requirement already satisfied, skipping upgrade: numpy>=1.16.4 in
/usr/local/lib/python3.6/dist-packages (from catalyst) (1.18.2)
Requirement already satisfied, skipping upgrade: pandas>=0.22 in
/usr/local/lib/python3.6/dist-packages (from catalyst) (0.25.3)
Requirement already satisfied, skipping upgrade: imageio in
/usr/local/lib/python3.6/dist-packages (from catalyst) (2.4.1)
Requirement already satisfied, skipping upgrade: seaborn in
/usr/local/lib/python3.6/dist-packages (from catalyst) (0.10.0)
Collecting tensorboardX
  Downloading https://files.pythonhosted.org/packages/35/f1/5843425495765c
8c2dd0784a851a93ef204d314fc87bcc2bbb9f662a3ad1/tensorboardX-2.0-py2.py3-none-
any.whl (195kB)
                       | 204kB 36.0MB/s
Collecting GitPython>=2.1.11
  Downloading https://files.pythonhosted.org/packages/d3/2f/6a366d56c9b135
5b0880be9ea66b166cb3536392638d8d91413ec66305ad/GitPython-3.1.0-py3-none-any.whl
(450kB)
                       | 460kB 41.2MB/s
Requirement already satisfied, skipping upgrade: scikit-learn>=0.20 in
/usr/local/lib/python3.6/dist-packages (from catalyst) (0.22.2.post1)
Requirement already satisfied, skipping upgrade: matplotlib in
/usr/local/lib/python3.6/dist-packages (from catalyst) (3.2.0)
Requirement already satisfied, skipping upgrade: ipython in
/usr/local/lib/python3.6/dist-packages (from catalyst) (5.5.0)
Requirement already satisfied, skipping upgrade: markdown>=2.6.8 in
/usr/local/lib/python3.6/dist-packages (from tensorboard>=1.14.0->catalyst)
Requirement already satisfied, skipping upgrade: absl-py>=0.4 in
/usr/local/lib/python3.6/dist-packages (from tensorboard>=1.14.0->catalyst)
(0.9.0)
Requirement already satisfied, skipping upgrade: wheel>=0.26; python_version >=
"3" in /usr/local/lib/python3.6/dist-packages (from
tensorboard>=1.14.0->catalyst) (0.34.2)
Requirement already satisfied, skipping upgrade: setuptools>=41.0.0 in
/usr/local/lib/python3.6/dist-packages (from tensorboard>=1.14.0->catalyst)
(46.0.0)
Requirement already satisfied, skipping upgrade: protobuf>=3.6.0 in
```

```
/usr/local/lib/python3.6/dist-packages (from tensorboard>=1.14.0->catalyst)
(3.10.0)
Requirement already satisfied, skipping upgrade: six>=1.10.0 in
/usr/local/lib/python3.6/dist-packages (from tensorboard>=1.14.0->catalyst)
(1.12.0)
Requirement already satisfied, skipping upgrade: grpcio>=1.6.3 in
/usr/local/lib/python3.6/dist-packages (from tensorboard>=1.14.0->catalyst)
(1.24.3)
Requirement already satisfied, skipping upgrade: werkzeug>=0.11.15 in
/usr/local/lib/python3.6/dist-packages (from tensorboard>=1.14.0->catalyst)
(1.0.0)
Requirement already satisfied, skipping upgrade: scipy>=0.19.0 in
/usr/local/lib/python3.6/dist-packages (from scikit-image>=0.14.2->catalyst)
(1.4.1)
Requirement already satisfied, skipping upgrade: networkx>=2.0 in
/usr/local/lib/python3.6/dist-packages (from scikit-image>=0.14.2->catalyst)
(2.4)
Requirement already satisfied, skipping upgrade: PyWavelets>=0.4.0 in
/usr/local/lib/python3.6/dist-packages (from scikit-image>=0.14.2->catalyst)
(1.1.1)
Requirement already satisfied, skipping upgrade: pyparsing>=2.0.2 in
/usr/local/lib/python3.6/dist-packages (from packaging->catalyst) (2.4.6)
Requirement already satisfied, skipping upgrade: retrying>=1.3.3 in
/usr/local/lib/python3.6/dist-packages (from plotly>=4.1.0->catalyst) (1.3.3)
Requirement already satisfied, skipping upgrade: pytz>=2017.2 in
/usr/local/lib/python3.6/dist-packages (from pandas>=0.22->catalyst) (2018.9)
Requirement already satisfied, skipping upgrade: python-dateutil>=2.6.1 in
/usr/local/lib/python3.6/dist-packages (from pandas>=0.22->catalyst) (2.8.1)
Collecting gitdb<5,>=4.0.1
  Downloading https://files.pythonhosted.org/packages/1e/f5/8f84b3bf9d94bd
f2454a302f2fa375832b53660ea532586b8a55ff16ae9a/gitdb-4.0.2-py3-none-any.whl
(63kB)
                       | 71kB 9.8MB/s
     1
Requirement already satisfied, skipping upgrade: joblib>=0.11 in
/usr/local/lib/python3.6/dist-packages (from scikit-learn>=0.20->catalyst)
(0.14.1)
Requirement already satisfied, skipping upgrade: cycler>=0.10 in
/usr/local/lib/python3.6/dist-packages (from matplotlib->catalyst) (0.10.0)
Requirement already satisfied, skipping upgrade: kiwisolver>=1.0.1 in
/usr/local/lib/python3.6/dist-packages (from matplotlib->catalyst) (1.1.0)
Requirement already satisfied, skipping upgrade: pickleshare in
/usr/local/lib/python3.6/dist-packages (from ipython->catalyst) (0.7.5)
Requirement already satisfied, skipping upgrade: pygments in
/usr/local/lib/python3.6/dist-packages (from ipython->catalyst) (2.1.3)
Requirement already satisfied, skipping upgrade: pexpect; sys_platform !=
"win32" in /usr/local/lib/python3.6/dist-packages (from ipython->catalyst)
(4.8.0)
Requirement already satisfied, skipping upgrade: prompt-toolkit<2.0.0,>=1.0.4 in
```

```
/usr/local/lib/python3.6/dist-packages (from ipython->catalyst) (1.0.18)
    Requirement already satisfied, skipping upgrade: simplegeneric>0.8 in
    /usr/local/lib/python3.6/dist-packages (from ipython->catalyst) (0.8.1)
    Requirement already satisfied, skipping upgrade: traitlets>=4.2 in
    /usr/local/lib/python3.6/dist-packages (from ipython->catalyst) (4.3.3)
    Requirement already satisfied, skipping upgrade: decorator in
    /usr/local/lib/python3.6/dist-packages (from ipython->catalyst) (4.4.2)
    Collecting smmap<4,>=3.0.1
      Downloading https://files.pythonhosted.org/packages/35/d2/27777ab463cd44842c78
    305fa8097dfba0d94768abbb7e1c4d88f1fa1a0b/smmap-3.0.1-py2.py3-none-any.whl
    Requirement already satisfied, skipping upgrade: ptyprocess>=0.5 in
    /usr/local/lib/python3.6/dist-packages (from pexpect; sys_platform !=
    "win32"->ipython->catalyst) (0.6.0)
    Requirement already satisfied, skipping upgrade: wcwidth in
    /usr/local/lib/python3.6/dist-packages (from prompt-
    toolkit<2.0.0,>=1.0.4->ipython->catalyst) (0.1.8)
    Requirement already satisfied, skipping upgrade: ipython-genutils in
    /usr/local/lib/python3.6/dist-packages (from traitlets>=4.2->ipython->catalyst)
    (0.2.0)
    ERROR: albumentations 0.1.12 has requirement imgaug<0.2.7,>=0.2.5, but
    you'll have imgaug 0.2.9 which is incompatible.
    Installing collected packages: crc32c, Pillow, deprecation, tensorboardX, smmap,
    gitdb, GitPython, catalyst
      Found existing installation: Pillow 7.0.0
        Uninstalling Pillow-7.0.0:
          Successfully uninstalled Pillow-7.0.0
    Successfully installed GitPython-3.1.0 Pillow-6.2.2 catalyst-20.3.2 crc32c-2.0
    deprecation-2.0.7 gitdb-4.0.2 smmap-3.0.1 tensorboardX-2.0
[0]: # Dependencies
     # Handles data
     import json
     import numpy as np
     import matplotlib.pyplot as plt
     import cv2
     import glob
     from operator import itemgetter
     # Torch utilities
     from typing import List
     from pathlib import Path
     from torch.utils.data import Dataset
     import torch
     # Data Loader utilities
     import collections
```

```
from sklearn.model_selection import train_test_split

from torch.utils.data import DataLoader

# Model building and training
import segmentation_models_pytorch as smp
from torch import nn

from catalyst.contrib.nn import DiceLoss, IoULoss
from torch import optim
from catalyst import utils

from catalyst.contrib.nn import RAdam, Lookahead
from catalyst.dl import SupervisedRunner

from catalyst.dl.callbacks import DiceCallback, IouCallback, \
CriterionCallback, AccuracyCallback, MulticlassDiceMetricCallback
```

```
[0]: # Sets a seed for better reproducibility
SEED = 42
utils.set_global_seed(SEED)
utils.prepare_cudnn(deterministic=True)
```

<IPython.core.display.HTML object>

```
[0]: # Defines and establishes a dataset class
     class SegmentationDataset(Dataset):
         def __init__(
             self.
             image_arr_path,
             mask_arr_path,
         ) -> None:
             self.images = np.load(image_arr_path)
             self.masks = np.load(mask_arr_path)
         def __len__(self) -> int:
             return len(self.images)
         def __getitem__(self, idx: int) -> dict:
             image = self.images[idx]
             image = np.swapaxes(image, 2, 0)
             image = np.swapaxes(image, 2, 1)
             image = torch.from_numpy(image).float()
             result = {"image": image}
             if self.masks is not None:
```

```
mask = self.masks[idx]
mask = np.swapaxes(mask, 2, 0)
mask = np.swapaxes(mask, 2, 1)
mask = torch.from_numpy(mask).float()
result["mask"] = mask
return result
```

[0]: # Loading once to enable visualizion prior to model training

dset = SegmentationDataset(image\_arr\_path="/content/drive/My Drive/RoboJackets/

→Split\_Data/train\_images.npy", mask\_arr\_path="/content/drive/My Drive/

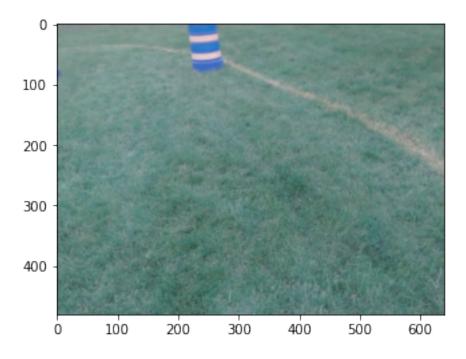
→RoboJackets/Split\_Data/train\_masks.npy")

```
[0]: # Show sizes of the image and mask
out = dset[0]
out["image"].shape, out["mask"].shape, len(dset)
```

[0]: (torch.Size([3, 480, 640]), torch.Size([1, 480, 640]), 592)

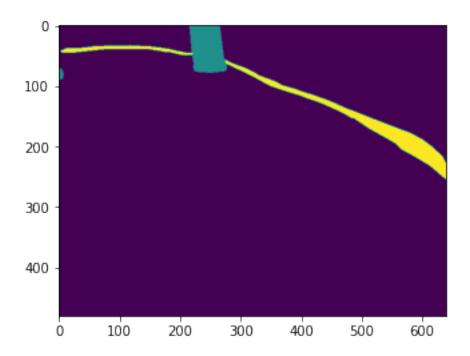
```
[0]: # Show an image
show_image = np.asarray(dset[40]['image'])
show_image = np.swapaxes(show_image, 2, 0)
show_image = np.swapaxes(show_image, 1, 0)
show_image = show_image.astype(np.uint8)
np.shape(show_image)
plt.imshow(show_image)
```

[0]: <matplotlib.image.AxesImage at 0x7ff0343bdfd0>



```
[0]: # Show associated mask
show_mask = np.squeeze(dset[40]['mask'])
plt.imshow(show_mask)
```

## [0]: <matplotlib.image.AxesImage at 0x7ff0736afef0>



```
[0]: # Set up transfer learning system

ENCODER = 'efficientnet-b3'
ENCODER_WEIGHTS = 'imagenet'
DEVICE = 'cuda'

# ACTIVATION = 'softmax'
ACTIVATION = None

model = smp.Unet(
    encoder_name=ENCODER,
    encoder_weights=ENCODER_WEIGHTS,
    classes=3,
    activation=ACTIVATION,
)
```

```
[0]: # Define loaders
     def get_loaders(
         images: List[Path],
         masks: List[Path],
         image_arr_path: str,
         mask_arr_path: str,
         random_state: int,
         valid_size: float = 0.1,
         batch_size: int = 12,
         num_workers: int = 4,
         # train_transforms_fn = None,
         # valid_transforms_fn = None,
         ) -> dict:
         indices = np.arange(len(images))
         train_indices, valid_indices = train_test_split(
           indices, test_size=valid_size, random_state=random_state, shuffle=True
         )
         np_images = np.array(images)
         np_masks = np.array(masks)
         #print(np_images.shape, np_masks.shape)
         #print(train_indices)
         train_dataset = SegmentationDataset(image_arr_path, mask_arr_path)
         train dataset.images = np images[train indices]
         train dataset.masks = np masks[train indices]
         #print(len(train_dataset))
         #print(train_dataset.images.shape)
         #print(train_dataset.masks.shape)
         valid_dataset = SegmentationDataset(image_arr_path, mask_arr_path)
         valid_dataset.images = np_images[valid_indices]
         valid_dataset.masks = np_masks[valid_indices]
         #print(len(valid_dataset))
         #print(valid_dataset.images.shape)
         #print(valid_dataset.masks.shape)
         train_loader = DataLoader(
           train_dataset,
           batch_size=batch_size,
```

```
shuffle=False,
num_workers=num_workers,
drop_last=False,
)

valid_loader = DataLoader(
  valid_dataset,
  batch_size=batch_size,
  shuffle=False,
  num_workers=num_workers,
  drop_last=False,
)

loaders = collections.OrderedDict()
loaders["train"] = train_loader
loaders["valid"] = valid_loader

return loaders
```

```
(592, 480, 640, 3) (592, 480, 640, 1)

[412 85 558 199 435 91 212 43 47 245 400 480 500 347 337 202 142 64 45 455 149 70 173 384 219 227 176 120 208 572 80 555 263 420 378 11 40 36 69 254 130 410 554 121 548 84 411 195 49 427 12 503 339 128 297 82 61 445 491 273 467 260 537 164 391 48 562 65 41 541 213 580 233 413 416 90 0 285 458 497 311 426 526 472 255 259 116 134 204 579 186 281 225 383 459 25 545 95 470 32 529 156 44 333 162 181 372 544 494 550 531 387 161 557 179 473 271 356 13 180 570 417 286 177 381 140 106 457 553 418 581 250 589 137 210 117 315 546 486 346 460 144 159 287 340 290 489 216 275 19 135 355 207 229 215 591 523 436 83 112 167 328 133 113 319 214 56 86 364 584 336 561 122 375 53 289 477 353 332 101 222 320 34 183 343 79 404 123 18 453 362 87 228 155 67 291 165 351
```

```
153 349 42 23 124 429 270 143 71 590 476 17 567 57 508 171 348 540
     514 444 421 578 395 365 475 110 262 312 466 78 484 350 288 274 29 236
     576 114 295 170 166 205 193 585 490 283 401 511 380 419 536 582 298
     518 240 573 21 322 430 423 24 256 432 20 15 509 111 97 438 376
     515 139
              31 342 542 146 402 530 408 428 163 501 234 556 249
     104 329
              66 434 409
                           2 118 507 55 502 310
                                                   6 253 373 483 564 524 251
     293 451 330
                   4 493 454 257
                                  62 358 533 172 547
                                                    94 538 393 587 235 568
     248 33 175 325
                     39 187 126 77 354 38 247 276 334
                                                         75 148
     437 261 127 469 450 226 299 318 341 68 496 488 528 506 324 89
     267 447 74 583 399 513 335 46 382 302 359 22 244 105 258 345 178 499
      99 505 252 327 482 115 174 189 16 296 326 282 269 439 54 492 27 521
     396 405 232 566 465 192 462 217 243 371 456 520 51 313 389 203 100 471
     125 102 363 218 52 138 534 168 415 323 268 440 169
                                                         81 481 504 532 368
     495 145 294 188 385 209 158 360 37
                                         30 279 338 160 221 303 109 571 93
     388 58 238 301 10 321 292 551 516 88 224
                                                   9 306 563 560 407 510 151
     397 277 200 485
                      8 422 468 414 304 190 565 425 331 552 374 317 424 361
       3 237 527 449 398 63 92 98 370 369 196 103 223 73 344 569 379 136
     431 367 535 28 461 308 246 517 59 377 386 392 107 543 76 559 197 549
     448 182 152 230 266 154 185 575 390 72]
    (532, 480, 640, 3)
    (532, 480, 640, 1)
    (60, 480, 640, 3)
    (60, 480, 640, 1)
[0]: # Helpful code taken from ---
          Copyright 2019 Division of Medical Image Computing, German Cancer Research ∪
      → Center (DKFZ), Heidelberg, Germany
         Licensed under the Apache License, Version 2.0 (the "License");
     #
          you may not use this file except in compliance with the License.
          You may obtain a copy of the License at
     #
     #
     #
             http://www.apache.org/licenses/LICENSE-2.0
     #
          Unless required by applicable law or agreed to in writing, software
          distributed under the License is distributed on an "AS IS" BASIS,
          WITHOUT WARRANTIES OR CONDITIONS OF ANY KIND, either express or implied.
          See the License for the specific language governing permissions and
          limitations under the License.
    import torch
    from torch import nn
    import numpy as np
    def sum tensor(inp, axes, keepdim=False):
        axes = np.unique(axes).astype(int)
         if keepdim:
```

```
for ax in axes:
            inp = inp.sum(int(ax), keepdim=True)
    else:
        for ax in sorted(axes, reverse=True):
            inp = inp.sum(int(ax))
    return inp
def softmax_helper(x):
    rpt = [1 for _ in range(len(x.size()))]
    rpt[1] = x.size(1)
    x_max = x.max(1, keepdim=True)[0].repeat(*rpt)
    e_x = torch.exp(x - x_max)
    return e_x / e_x.sum(1, keepdim=True).repeat(*rpt)
class CrossentropyND(nn.CrossEntropyLoss):
    Network has to have NO NONLINEARITY!
    def forward(self, inp, target):
        target = target.long()
        num_classes = inp.size()[1]
        i0 = 1
        i1 = 2
        while i1 < len(inp.shape): # this is ugly but torch only allows to_\sqcup
\rightarrow transpose two axes at once
            inp = inp.transpose(i0, i1)
            iO += 1
            i1 += 1
        inp = inp.contiguous()
        inp = inp.view(-1, num_classes)
        target = target.view(-1,)
        return super(CrossentropyND, self).forward(inp, target)
def get_tp_fp_fn(net_output, gt, axes=None, mask=None, square=False):
    net\_output must be (b, c, x, y(, z)))
    gt must be a label map (shape (b, 1, x, y(, z)) OR shape (b, x, y(, z))) or\Box
\rightarrow one hot encoding (b, c, x, y(, z))
    if mask is provided it must have shape (b, 1, x, y(, z))
    :param net_output:
    :param gt:
    :param axes:
```

```
:param mask: mask must be 1 for valid pixels and 0 for invalid pixels
   :param square: if True then fp, tp and fn will be squared before summation
   :return:
   HHHH
   if axes is None:
       axes = tuple(range(2, len(net_output.size())))
   shp_x = net_output.shape
   shp_y = gt.shape
   with torch.no_grad():
       if len(shp_x) != len(shp_y):
           gt = gt.view((shp_y[0], 1, *shp_y[1:]))
       if all([i == j for i, j in zip(net_output.shape, gt.shape)]):
           # if this is the case then qt is probably already a one hot encoding
           y_onehot = gt
       else:
           gt = gt.long()
           y_onehot = torch.zeros(shp_x)
           if net_output.device.type == "cuda":
               y_onehot = y_onehot.cuda(net_output.device.index)
           y_onehot.scatter_(1, gt, 1)
  tp = net_output * y_onehot
   fp = net_output * (1 - y_onehot)
   fn = (1 - net_output) * y_onehot
   if mask is not None:
       tp = torch.stack(tuple(x_i * mask[:, 0] for x_i in torch.unbind(tp,_
\rightarrowdim=1)), dim=1)
       fp = torch.stack(tuple(x_i * mask[:, 0] for x_i in torch.unbind(fp,__
\rightarrowdim=1)), dim=1)
       fn = torch.stack(tuple(x_i * mask[:, 0] for x_i in torch.unbind(fn, __
\rightarrowdim=1)), dim=1)
   if square:
       tp = tp ** 2
       fp = fp ** 2
       fn = fn ** 2
   tp = sum_tensor(tp, axes, keepdim=False)
   fp = sum_tensor(fp, axes, keepdim=False)
   fn = sum_tensor(fn, axes, keepdim=False)
   return tp, fp, fn
```

```
class SoftDiceLoss(nn.Module):
    def __init__(self, apply_nonlin=None, batch_dice=False, do_bg=True,
                 smooth=1., square=False):
        11 11 11
        super(SoftDiceLoss, self).__init__()
        self.square = square
        self.do_bg = do_bg
        self.batch_dice = batch_dice
        self.apply_nonlin = apply_nonlin
        self.smooth = smooth
    def forward(self, x, y, loss_mask=None):
        shp_x = x.shape
        if self.batch_dice:
            axes = [0] + list(range(2, len(shp_x)))
        else:
            axes = list(range(2, len(shp_x)))
        if self.apply_nonlin is not None:
            x = self.apply_nonlin(x)
        tp, fp, fn = get_tp_fp_fn(x, y, axes, loss_mask, self.square)
        dc = (2 * tp + self.smooth) / (2 * tp + fp + fn + self.smooth)
        if not self.do_bg:
            if self.batch_dice:
                dc = dc[1:]
            else:
                dc = dc[:, 1:]
        dc = dc.mean()
        return -dc
class DC_and_CE_loss(nn.Module):
    def __init__(self, soft_dice_kwargs, ce_kwargs, aggregate="sum"):
        super(DC_and_CE_loss, self).__init__()
        self.aggregate = aggregate
        self.ce = CrossentropyND(**ce_kwargs)
        self.dc = SoftDiceLoss(apply_nonlin=softmax_helper, **soft_dice_kwargs)
    def forward(self, net_output, target):
```

```
[0]: # Define loss criterion
     criterion = {
         "CE": CrossentropyND(),
     }
     from torch.optim import AdamW
     # Set up optimization
     learning_rate = 0.001 #0.001
     encoder_learning_rate = 0.0005
     encoder_weight_decay = 0.00003 #0.00003
     optimizer_weight_decay = 0.0003 #0.0003
     optim_factor = 0.25 #0.25
     optim_patience = 2 #2
     optimizer = AdamW(model.parameters(), lr=0.001, betas=(0.9, 0.999), eps=1e-08,
     →weight_decay=0.01, amsgrad=False)
     scheduler = optim.lr_scheduler.ReduceLROnPlateau(optimizer,__
     →factor=optim_factor, patience=optim_patience)
     num_epochs = 10
     device = utils.get_device()
     runner = SupervisedRunner(device=device, input_key="image",_
      →input_target_key="mask")
```

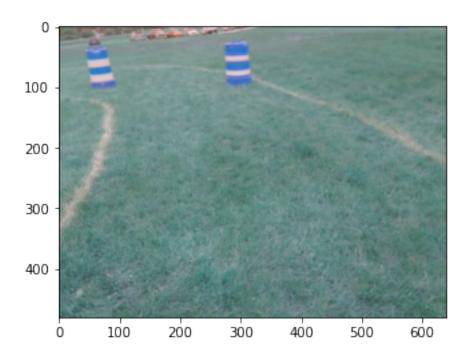
```
[0]: runner.train(
         model=model,
         criterion=criterion,
         optimizer=optimizer,
         scheduler=scheduler,
         loaders=loaders,
         callbacks=callbacks,
         logdir='content/full_model2', #this logdir must be changed with every new_
      \hookrightarrow run
         num_epochs=num_epochs,
         main_metric="loss",
         minimize_metric=True,
         fp16=None,
         monitoring_params=None,
         verbose=True,
     )
    1/10 * Epoch (train): 100% 178/178 [01:41<00:00, 1.75it/s, _timer/_fps=0.821,
    _timer/batch_time=3.654, _timer/data_time=0.312, _timer/model_time=3.341,
    loss=0.053
    1/10 * Epoch (valid): 100% 20/20 [00:03<00:00, 6.29it/s, _timer/_fps=20.898,
    _timer/batch_time=0.144, _timer/data_time=0.012, _timer/model_time=0.131,
    loss=0.112]
    [2020-03-20 03:33:53,038]
    1/10 * Epoch 1 (train): _timer/_fps=6.3847 | _timer/batch_time=0.5372 |
    _timer/data_time=0.3472 | _timer/model_time=0.1900 | dice_0=0.9635 |
    dice_1=0.0130 | dice_2=0.5910 | dice_mean=0.5225 | loss=0.2064
    1/10 * Epoch 1 (valid): _timer/_fps=19.6645 | _timer/batch_time=0.1569 |
    _timer/data_time=0.0235 | _timer/model_time=0.1333 | dice_0=0.9769 |
    dice_1=0.5987 | dice_2=0.6905 | dice_mean=0.7554 | loss=0.1282
    2/10 * Epoch (train): 100% 178/178 [01:22<00:00, 2.15it/s, timer/_fps=7.819,
    _timer/batch_time=0.384, _timer/data_time=0.286, _timer/model_time=0.098,
    loss=0.044
    2/10 * Epoch (valid): 100% 20/20 [00:03<00:00, 5.98it/s, _timer/_fps=20.212,
    _timer/batch_time=0.148, _timer/data_time=0.013, _timer/model_time=0.136,
    loss=0.073
    [2020-03-20 03:35:21,168]
    2/10 * Epoch 2 (train): _timer/_fps=6.4760 | _timer/batch_time=0.4641 |
    _timer/data_time=0.3083 | _timer/model_time=0.1557 | dice_0=0.9895 |
    dice_1=0.5373 | dice_2=0.8154 | dice_mean=0.7808 | loss=0.0591
    2/10 * Epoch 2 (valid): _timer/_fps=18.6243 | _timer/batch_time=0.1648 |
    _timer/data_time=0.0237 | _timer/model_time=0.1412 | dice_0=0.9869 |
    dice_1=0.6184 | dice_2=0.7951 | dice_mean=0.8001 | loss=0.0757
    3/10 * Epoch (train): 100% 178/178 [01:22<00:00, 2.15it/s, _timer/_fps=7.283,
```

```
_timer/batch_time=0.412, _timer/data_time=0.305, _timer/model_time=0.107,
loss=0.037]
3/10 * Epoch (valid): 100% 20/20 [00:03<00:00, 6.26it/s, timer/fps=21.101,
_timer/batch_time=0.142, _timer/data_time=0.011, _timer/model_time=0.131,
loss=0.0391
[2020-03-20 03:36:49,210]
3/10 * Epoch 3 (train): timer/ fps=6.4597 | timer/batch time=0.4649 |
_timer/data_time=0.3103 | _timer/model_time=0.1546 | dice_0=0.9910 |
dice_1=0.6875 | dice_2=0.8353 | dice_mean=0.8379 | loss=0.0467
3/10 * Epoch 3 (valid): _timer/_fps=19.6169 | _timer/batch_time=0.1574 |
_timer/data_time=0.0193 | _timer/model_time=0.1380 | dice_0=0.9890 |
dice_1=0.8331 | dice_2=0.8027 | dice_mean=0.8749 | loss=0.0630
4/10 * Epoch (train): 100% 178/178 [01:22<00:00, 2.15it/s, _timer/_fps=7.647,
_timer/batch_time=0.392, _timer/data_time=0.298, _timer/model_time=0.094,
loss=0.034
4/10 * Epoch (valid): 100% 20/20 [00:03<00:00, 6.08it/s, _timer/_fps=21.047,
_timer/batch_time=0.143, _timer/data_time=0.013, _timer/model_time=0.130,
loss=0.031]
[2020-03-20 03:38:17,556]
4/10 * Epoch 4 (train): timer/ fps=6.4678 | timer/batch time=0.4644 |
_timer/data_time=0.3086 | _timer/model_time=0.1558 | dice_0=0.9927 |
dice 1=0.7995 | dice 2=0.8587 | dice mean=0.8837 | loss=0.0378
4/10 * Epoch 4 (valid): _timer/_fps=19.1945 | _timer/batch_time=0.1616 |
_timer/data_time=0.0231 | _timer/model_time=0.1385 | dice_0=0.9900 |
dice_1=0.9001 | dice_2=0.8102 | dice_mean=0.9001 | loss=0.0585
5/10 * Epoch (train): 100% 178/178 [01:22<00:00, 2.16it/s, _timer/_fps=7.183,
_timer/batch_time=0.418, _timer/data_time=0.318, _timer/model_time=0.100,
loss=0.029]
5/10 * Epoch (valid): 100% 20/20 [00:03<00:00, 6.25it/s, _timer/_fps=20.178,
_timer/batch_time=0.149, _timer/data_time=0.012, _timer/model_time=0.137,
loss=0.024
[2020-03-20 03:39:45,143]
5/10 * Epoch 5 (train): _timer/_fps=6.4919 | _timer/batch_time=0.4626 |
_timer/data_time=0.3076 | _timer/model_time=0.1550 | dice_0=0.9939 |
dice 1=0.8967 | dice 2=0.8694 | dice mean=0.9200 | loss=0.0325
5/10 * Epoch 5 (valid): _timer/_fps=19.5679 | _timer/batch_time=0.1575 |
timer/data time=0.0209 | timer/model time=0.1366 | dice 0=0.9879 |
dice_1=0.8508 | dice_2=0.7516 | dice_mean=0.8635 | loss=0.0770
6/10 * Epoch (train): 100% 178/178 [01:22<00:00, 2.14it/s, _timer/_fps=7.483,
_timer/batch_time=0.401, _timer/data_time=0.308, _timer/model_time=0.093,
loss=0.023]
6/10 * Epoch (valid): 100% 20/20 [00:03<00:00, 6.27it/s, timer/fps=20.726,
_timer/batch_time=0.145, _timer/data_time=0.011, _timer/model_time=0.134,
loss=0.023]
[2020-03-20 03:41:12,348]
6/10 * Epoch 6 (train): _timer/_fps=6.4583 | _timer/batch_time=0.4650 |
_timer/data_time=0.3094 | _timer/model_time=0.1556 | dice_0=0.9944 |
dice_1=0.9358 | dice_2=0.8749 | dice_mean=0.9350 | loss=0.0297
```

```
6/10 * Epoch 6 (valid): _timer/_fps=19.6176 | _timer/batch_time=0.1571 |
_timer/data_time=0.0196 | _timer/model_time=0.1375 | dice_0=0.9893 |
dice_1=0.9125 | dice_2=0.7859 | dice_mean=0.8959 | loss=0.0633
7/10 * Epoch (train): 100% 178/178 [01:22<00:00, 2.15it/s, _timer/_fps=7.636,
timer/batch time=0.393, timer/data time=0.289, timer/model time=0.104,
loss=0.028]
7/10 * Epoch (valid): 100% 20/20 [00:03<00:00, 6.30it/s, timer/fps=20.504,
_timer/batch_time=0.146, _timer/data_time=0.012, _timer/model_time=0.135,
loss=0.034]
[2020-03-20 03:42:39,445]
7/10 * Epoch 7 (train): _timer/_fps=6.4820 | _timer/batch_time=0.4635 |
_timer/data_time=0.3078 | _timer/model_time=0.1558 | dice_0=0.9946 |
dice_1=0.9381 | dice_2=0.8802 | dice_mean=0.9376 | loss=0.0281
7/10 * Epoch 7 (valid): timer/ fps=19.7017 | timer/batch time=0.1563 |
_timer/data_time=0.0187 | _timer/model_time=0.1376 | dice_0=0.9833 |
dice_1=0.5378 | dice_2=0.7334 | dice_mean=0.7515 | loss=0.1073
8/10 * Epoch (train): 100% 178/178 [01:22<00:00, 2.15it/s, _timer/_fps=7.823,
_timer/batch_time=0.383, _timer/data_time=0.292, _timer/model_time=0.091,
loss=0.022]
8/10 * Epoch (valid): 100% 20/20 [00:03<00:00, 6.25it/s, timer/fps=21.096,
_timer/batch_time=0.142, _timer/data_time=0.012, _timer/model_time=0.130,
loss=0.022]
[2020-03-20 03:44:06,497]
8/10 * Epoch 8 (train): _timer/_fps=6.4862 | _timer/batch_time=0.4632 |
_timer/data_time=0.3083 | _timer/model_time=0.1549 | dice_0=0.9942 |
dice_1=0.9063 | dice_2=0.8724 | dice_mean=0.9243 | loss=0.0305
8/10 * Epoch 8 (valid): _timer/_fps=19.5776 | _timer/batch_time=0.1573 |
_timer/data_time=0.0196 | _timer/model_time=0.1377 | dice_0=0.9876 |
dice_1=0.8488 | dice_2=0.7385 | dice_mean=0.8583 | loss=0.0897
9/10 * Epoch (train): 100% 178/178 [01:22<00:00, 2.15it/s, _timer/_fps=7.189,
_timer/batch_time=0.417, _timer/data_time=0.313, _timer/model_time=0.104,
loss=0.019]
9/10 * Epoch (valid): 100% 20/20 [00:03<00:00, 6.15it/s, timer/fps=20.706,
_timer/batch_time=0.145, _timer/data_time=0.012, _timer/model_time=0.133,
loss=0.022]
[2020-03-20 03:45:33,709]
9/10 * Epoch 9 (train): timer/ fps=6.4782 | timer/batch time=0.4637 |
_timer/data_time=0.3089 | _timer/model_time=0.1547 | dice_0=0.9951 |
dice_1=0.9631 | dice_2=0.8868 | dice_mean=0.9483 | loss=0.0245
9/10 * Epoch 9 (valid): _timer/_fps=19.2671 | _timer/batch_time=0.1602 |
_timer/data_time=0.0216 | _timer/model_time=0.1386 | dice_0=0.9901 |
dice_1=0.9512 | dice_2=0.8037 | dice_mean=0.9150 | loss=0.0617
10/10 * Epoch (train): 100% 178/178 [01:22<00:00, 2.15it/s, _timer/_fps=7.131,
_timer/batch_time=0.421, _timer/data_time=0.317, _timer/model_time=0.103,
loss=0.018]
10/10 * Epoch (valid): 100% 20/20 [00:03<00:00, 6.21it/s, _timer/_fps=20.265,
_timer/batch_time=0.148, _timer/data_time=0.013, _timer/model_time=0.135,
loss=0.019]
```

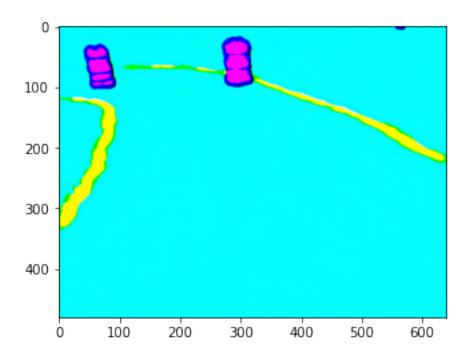
```
[2020-03-20 03:47:01,070]
    10/10 * Epoch 10 (train): _timer/_fps=6.4627 | _timer/batch_time=0.4648 |
    _timer/data_time=0.3094 | _timer/model_time=0.1554 | dice_0=0.9955 |
    dice_1=0.9705 | dice_2=0.8971 | dice_mean=0.9544 | loss=0.0218
    10/10 * Epoch 10 (valid): timer/ fps=19.4141 | timer/batch time=0.1585 |
    _timer/data_time=0.0214 | _timer/model_time=0.1371 | dice_0=0.9904 |
    dice 1=0.9289 | dice 2=0.8159 | dice mean=0.9117 | loss=0.0619
    Top best models:
    content/full_model2/checkpoints/train.4.pth
                                                    0.0585
[0]: # Test model on test dataset
     test_data = SegmentationDataset("/content/drive/My Drive/RoboJackets/Split_Data/
      →test images.npy", "/content/drive/My Drive/RoboJackets/Split Data/test masks.
      [0]: infer_loader = DataLoader(
         test data,
         batch_size=12,
         shuffle=False,
         num_workers=4
[0]: # get predictions on test data
     predictions = runner.predict_loader(
         model=model,
         loader=infer_loader,
         resume=f"content/full_model2/checkpoints/best.pth",
         verbose=False,
     )
     print(type(predictions))
     print(predictions.shape)
    => loading checkpoint content/full model2/checkpoints/best.pth
    loaded checkpoint content/full_model2/checkpoints/best.pth (global epoch 4,
    epoch 4, stage train)
    <class 'numpy.ndarray'>
    (149, 3, 480, 640)
[0]: show_image = np.asarray(test_data[30]['image'])
     show_image = np.swapaxes(show_image, 2, 0)
     show_image = np.swapaxes(show_image, 1, 0)
     show_image = show_image.astype(np.uint8)
     np.shape(show image)
     plt.imshow(show_image)
```

[0]: <matplotlib.image.AxesImage at 0x7fefe3a8a4e0>



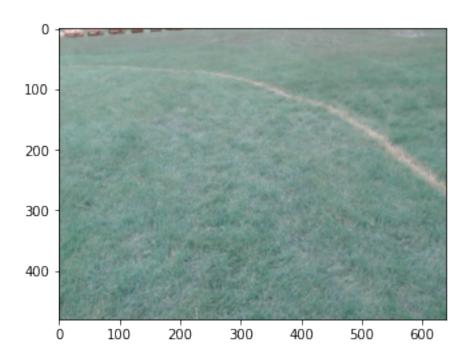
```
[0]: show_image = np.asarray(predictions[30])
show_image = np.swapaxes(show_image, 2, 0)
show_image = np.swapaxes(show_image, 1, 0)
show_image = show_image.astype(np.uint8)
np.shape(show_image)
plt.imshow(show_image)
```

[0]: <matplotlib.image.AxesImage at 0x7fefe39e4b38>



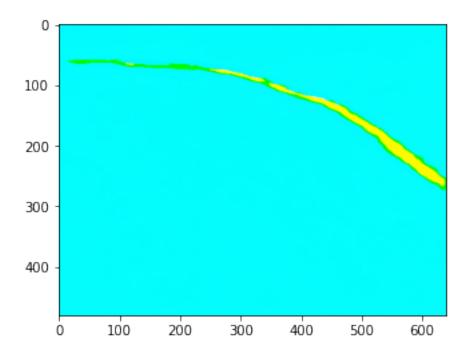
```
[0]: show_image = np.asarray(test_data[19]['image'])
show_image = np.swapaxes(show_image, 2, 0)
show_image = np.swapaxes(show_image, 1, 0)
show_image = show_image.astype(np.uint8)
np.shape(show_image)
plt.imshow(show_image)
```

[0]: <matplotlib.image.AxesImage at 0x7fefe38de978>



```
[0]: show_image = np.asarray(predictions[19])
show_image = np.swapaxes(show_image, 2, 0)
show_image = np.swapaxes(show_image, 1, 0)
show_image = show_image.astype(np.uint8)
np.shape(show_image)
plt.imshow(show_image)
```

[0]: <matplotlib.image.AxesImage at 0x7fefe3820748>



```
[0]: %load_ext tensorboard %tensorboard --logdir {'content/full_model2'}
```

<IPython.core.display.HTML object>

```
[0]: LOG_DIR = './content/full_model2'

get_ipython().system_raw(
        'tensorboard --logdir {} --host 0.0.0.0 --port 6007 &'
        .format(LOG_DIR)
)

# Install
! npm install -g localtunnel

# Tunnel port 6006 (TensorBoard assumed running)
get_ipython().system_raw('lt --port 6007 >> url.txt 2>&1 &')

# Get url
! cat url.txt
```

```
/tools/node/bin/lt ->
/tools/node/lib/node_modules/localtunnel/bin/lt.js
+ localtunnel@2.0.0
updated 1 package in 2.06s
```

```
Update available 5.7.1 \rightarrow 6.14.3
```

Run  $npm \ i \ -g \ npm \ to \ update$ 

## [0]: !npm i -g npm

/tools/node/bin/npm -> /tools/node/lib/node\_modules/npm/bin/npm-cli.js /tools/node/bin/npx -> /tools/node/lib/node\_modules/npm/bin/npx-cli.js + npm@6.14.3 added 325 packages from 161 contributors, removed 423 packages and updated 59 packages in 14.469s