Introduction:

In this template, methods are provided to get you started on the task at hand (please see project description). Please implement your solution in the code cells marked with **TODO**. Most of the other code cells are hidden, feel free to explore and change these. These cells implement a basic pipeline for training your model but you may want to explore more complex procedures. **Make sure you run all cells before trying to implement your own solution!**

Imports and definitions:

```
1 #@title
 2 import numpy as np
 3 import requests
 4 import io
 5 from torch.utils.data import TensorDataset
 6 import matplotlib.pyplot as plt
 7 import torch
 8 from scipy.ndimage import rotate
 9 %matplotlib inline
10 from torch.utils.data import Dataset
11 import random
12 import torch
13 import torchvision.transforms as transforms
14 from sklearn.model selection import train test split
15 import torch.nn as nn
16 import torch.nn.functional as F
17 import torch.optim as optim
18 from torchsummary import summary
19 from tqdm import tqdm
20 import itertools
21
22
23
24 class BatchSampler():
25
26
    Implements an iterable which given a torch dataset and a batch size
2.7
    will produce batches of data of that given size. The batches are
28
    returned as tuples in the form (images, labels).
29
    Can produce balanced batches, where each batch will have an equal
30
    amount of samples from each class in the dataset. If your dataset is hear
31
    imbalanced, this might mean throwing away a lot of samples from
32
    ower-represented classes!
```

```
# Counting the ocurrence of the class labels:
40
         unique, counts = np.unique(self.dataset.targets, return counts=True)
41
42
         indexes = []
43
         # Sampling an equal amount from each class:
         for i in range(len(unique)):
44
           indexes.append(np.random.choice(np.where(self.dataset.targets == i)
45
46
         # Setting the indexes we will sample from later:
         self.indexes = np.concatenate(indexes)
47
48
      else:
49
         # Setting the indexes we will sample from later (all indexes):
50
         self.indexes = [i for i in range(len(dataset))]
51
52
53
     def len (self):
54
      return (len(self.indexes) // self.batch size) + 1
55
56
    def shuffle(self):
57
       # We do not need to shuffle if we use the balanced sampling method.
      # Shuffling is already done when making the balanced samples.
58
59
      if not self.balanced:
60
         random.shuffle(self.indexes)
61
62
    def iter (self):
63
      remaining = False
64
      self.shuffle()
65
      # Go over the datset in steps of 'self.batch size':
       for i in range(0, len(self.indexes), self.batch size):
66
67
           imgs, labels = [], []
68
           # If our current batch is larger than the remaining data, we quit:
           if i + self.batch size > len(self.indexes):
69
70
             remaining = True
71
             break
72
           # If not, we yield a complete batch:
73
           else:
74
             # Getting a list of samples from the dataset, given the indexes \( \)
75
             X batch = [self.dataset[self.indexes[k]][0] for k in range(i, i -
76
             Y batch = [self.dataset[self.indexes[k]][1] for k in range(i, i ·
77
             # Stacking all the samples and returning the target labels as a 1
78
             yield torch.stack(X batch).float(), torch.tensor(Y batch).long()
79
       # If there is still data left that was not a full batch:
80
      if remaining:
```

```
ے ر
 93
 94
     def init (self, x, y, transform=None, target transform=None):
 95
      self.targets = y
 96
      self.imgs = x
 97
       self.transform = transform
 98
       self.target transform = target transform
 99
100
     def len (self):
101
      return len(self.targets)
102
103
     def getitem (self, idx):
104
      image = torch.from numpy(self.imgs[idx] / 255).float()
105
      label = self.targets[idx]
106
       return image, label
107
108 def load numpy arr from url(url):
109
110
       Loads a numpy array from surfdrive.
111
112
       Input:
113
       url: Download link of dataset
114
115
       Outputs:
116
       dataset: numpy array with input features or labels
117
118
119
       response = requests.get(url)
120
       response.raise for status()
121
122
       return np.load(io.BytesIO(response.content))
123
124
125 class labels = {0: 'Atelectasis',
126
                   1: 'Effusion',
127
                   2: 'Infiltration',
128
                   3: 'No Finding',
```

```
1 #@title
2 # Downloading the images:
3 train_x = load_numpy_arr_from_url('https://surfdrive.surf.nl/files/index.pl
4 test_x = load_numpy_arr_from_url('https://surfdrive.surf.nl/files/index.ph;
```

changes on data set

Data Augmentation (not being used in base model)

```
1 # data augentation
 3 # defining transfrmation for data points
 5 preprocess aug = transforms.Compose([
 7
       #preprocess takes tensor x
       transforms.Lambda(lambda x: x.repeat(1, 3, 1, 1)),
 8
       transforms.Resize(299),
10
       transforms.CenterCrop(299),
       transforms.RandomResizedCrop(299),
11
12
       transforms.RandomHorizontalFlip(),
       transforms.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.224, 0.224]
13
14
       transforms.RandomResizedCrop((128,128))
15])
 1 #example of transformed data point
 2 \times eg = torch.cuda.FloatTensor(test \times [0].reshape(128,128))
```

```
1 # Function used for previous testing of model
 2
 3 def sizedSet(set x, set y):
       """Given a data set x this function makes all the labeled (by y) sub se
 5
       the same size as the smallest subset of x by removing the other points
 6
 7
 8
       # working out arrays to be used below
 9
10
      set arr = np.array([[test x[i], test y[i]] for i in range(len(test y))];
11
       set arr = set arr[set arr[:,1].argsort()]
12
      label sizes = np.array([np.count nonzero(test y==i) for i in class labe
13
       size = int(min(label sizes))
14
15
16
       # taking care of indeces for new set
17
18
      new arr=np.array([])
19
      index = 0
20
       for l in range(np.size(label sizes)):
21
           for i in range(size):
22
               new arr = np.append(new arr , [set arr[index+i,0],set arr[index
23
           index += label sizes[1]
```

```
6 xval= np.array([train_x[i] for i in xval_index])
7 yval= np.array([train_y[i] for i in xval_index])
8
9
10
11 train_index= [i for i in indexs if i not in xval_index]
12 train_xx= np.array([train_x[i] for i in train_index])
13 train_yy= np.array([train_y[i] for i in train_index])
14
15 val_dataset = ImageDataset(xval, yval) #test
16
17 # Comment out not to use validation split
18 train_x = train_xx
19 train_y = train_yy
20
```

Plotting the data distribution:

```
2 test dataset = ImageDataset(test_x, test_y)
```

Defining our model as a neural network:

TODO define your own model here, follow the structure as presented in the Pytorch tutorial (or see below as an example).

```
1 class Net(nn.Module):
2    def __init__(self, n_classes):
3        super(Net, self).__init__()
```

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```
46
47 # Make sure your model instance is assigned to a variable 'model':
48 model = Net(n classes = 6)
```

Moving model to CUDA, verifying model structure and

Data_Challenge_1_Model_(09_0_1_2) Final Version.ipynb - Colaboratory https://colab.research.google.com/drive/12pwTOsbB24Ij75eNiF5p_Rh7...

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