```
In [1]:
import numpy as np
import cv2
# Open the video
cap = cv2.VideoCapture('C://Users/Nikolay/Downloads/video/Camera 3_20220526_003249(2).mp4')
# Initialize frame counter
cnt = 1
w_frame, h_frame = int(cap.get(cv2.CAP_PROP_FRAME_WIDTH)), int(cap.get(cv2.CAP_PROP_FRAME_HEIGHT))
fps, frames = cap.get(cv2.CAP_PROP_FPS), cap.get(cv2.CAP_PROP_FRAME_COUNT)
# define croping values
x,y,h,w = 115,210,235,235
# output
fourcc = cv2.VideoWriter_fourcc(*'XVID')
out = cv2. VideoWriter('result.avi', fourcc, fps, (w, h))
while(cap.isOpened()):
  ret, frame = cap.read()
  cnt += 1 # Counting frames
  # Avoid problems when video finish
  if ret==True:
    # Croping the frame
    crop_frame = frame[y:y+h, x:x+w]
    # Percentage
    xx = cnt *100/frames
    print(int(xx),'%')
    #Saving from the desired frames
    if cnt % 15 == 0:
       out.write(crop_frame)
    # see the video in real time
    cv2.imshow('frame',frame)
    cv2.imshow('croped',crop_frame)
    if cv2.waitKey(1) \& 0xFF == ord('q'):
       break
  else:
    break
cap.release()
out.release()
cv2.destroyAllWindows()
100 %
                                                                                                                                                            In [7]:
#size for reshape
down width = 116
down_height = 116
down_points = (down_width, down_height)
                                                                                                                                                            In [8]:
# reshape all images to 116*116
vidcap = cv2.VideoCapture('C://Users/Nikolay/jupyter notebooks/result.avi')
success,image = vidcap.read()
count = 0
while success:
  image = cv2.resize(image, down_points, interpolation= cv2.INTER_LINEAR)
  cv2.imwrite("frame%d.jpg" % count, image) # save frame as JPEG file
  success,image = vidcap.read()
  print('Read a new frame: ', success)
  count += 1
Read a new frame: True
Read a new frame: False
                                                                                                                                                           In [16]:
# checking the image
image = cv2.imread('C://Users/Nikolay/Desktop/test/1/frame9484.jpg')
cv2.imshow('Resized Down by defining height and width', image)
cv2.waitKey()
cv2.destroyAllWindows()
                                                                                                                                                            In [3]:
import torch
import random
#import numpy as np
import os
```

torch.backends.cudnn.deterministic = True from torchvision.datasets import ImageFolder

from torchvision.transforms import ToTensor from torchvision import datasets import torch import torch.nn as nn import torch.optim as optim from torch.utils.data import Dataset import torchvision.transforms as transforms from torch.utils.data import DataLoader import torchvision from skimage import io import pandas as pd # fix random seed random.seed(0) #np.random.seed(0) torch.manual_seed(0) torch.cuda.manual_seed(0) # using cuda device = "cuda" if torch.cuda.is_available() else "cpu" C:\Users\Nikolay\AppData\Local\Programs\Python\Python310\lib\site-packages\torch\random.py:42: UserWarning: Failed to initialize NumPy: module compiled against A PI version 0x10 but this version of numpy is 0xf (Triggered internally at ..\torch\csrc\utils\tensor_numpy.cpp:68.) return default_generator.manual_seed(seed) Making labels In [31]: X = os.listdir('C://Users/Nikolay/Desktop/test/no') y = os.listdir('C://Users/Nikolay/Desktop/test/target') In [32]: print(X[0], y[0]) frame0.jpg frame10214.jpg In [33]: len(X), len(y) Out[33]: (29105, 8578) In [54]: df_sit = pd.DataFrame({'Frame': y}) df_Notsit = pd.DataFrame({'Frame': X}) In [55]: df_sit['class'] = np.ones(len(y)) $df_Notsit['class'] = np.zeros(len(X))$ In [56]: df_Notsit.head() Out[56]: Frame class |**0**||frame0.jpg 0.0 1 frame1.jpg 0.0 2 frame10.jpg 0.0 3 frame100.jpg ||0.0 4 frame1000.jpg 0.0 In [59]: labels = pd.concat([df_sit, df_Notsit], axis = 0) labels.head() Out[59]: Frame class frame10214.jpg 1.0 1 frame10215.jpg 1.0 2 frame10216.jpg 1.0 3 frame10217.jpg 1.0 4 frame10218.jpg 1.0 In [136]: labels.to_excel('C://Users/Nikolay/jupyter_notebooks/labels.xlsx',header = True, index=['Frame', 'class'], columns=['Frame', 'class']) In [10]: pd.read_excel('C://Users/Nikolay/Desktop/test/images/labels.xlsx', index_col=0) Out[10]: Frame class 0 frame10214.jpg 1 frame10215.jpg | 1 1

2

frame10216.jpg 1

	Frame	class
3	frame10217.jpg	1
4	frame10218.jpg	1
29100	frame9995.jpg	0
29101	frame9996.jpg	0
29102	frame9997.jpg	0
29103	frame9998.jpg	0
29104	frame9999.jpg	0

37683 rows × 2 columns

```
making class for Dataset
                                                                                                                                                                In [9]:
class SitNotSitDataset(Dataset):
  def __init__(self, xlsx_file, root_dir, transform = None):
     self.annotations = pd.read_excel(xlsx_file, index_col=0)
     self.root dir = root dir
     self.transform = transform
  def __len__(self):
     return len(self.annotations)
  def __getitem__(self,index):
     img_path = os.path.join(self.root_dir, self.annotations.iloc[index, 0])
     image = io.imread(img_path)
     y_label = torch.tensor(int(self.annotations.iloc[index, 1]))
       image = self.transform(image)
     return(image, y_label)
                                                                                                                                                              In [10]:
dataset = SitNotSitDataset(xlsx file = 'C://Users/Nikolay/Desktop/test/images/labels.xlsx',
                root_dir = 'C://Users/Nikolay/Desktop/test/images',
                transform = transforms.ToTensor())
                                                                                                                                                              In [11]:
len(dataset)
                                                                                                                                                             Out[11]:
37683
train_test_split
                                                                                                                                                              In [12]:
train_size = int(0.7 * len(dataset))
test_size = len(dataset) - train_size
train_set, test_set = torch.utils.data.random_split(dataset, [train_size, test_size])
                                                                                                                                                               In [13]:
batch_size = 16
                                                                                                                                                               In [14]:
train_loader = DataLoader(dataset = train_set, batch_size = batch_size, shuffle = True)
test_loader = DataLoader(dataset = test_set, batch_size = batch_size, shuffle = True)
using googlenet
                                                                                                                                                               In [15]:
model = torchvision.models.googlenet(pretrained = True)
C:\Users\Nikolay\AppData\Local\Programs\Python\Python310\lib\site-packages\torchvision\models\_utils.py:208: UserWarning: The parameter 'pretrained' is deprecated
since 0.13 and will be removed in 0.15, please use 'weights' instead.
 warnings.warn(
```

GoogLeNet(

(conv3): BasicConv2d(

```
C:\Users\Nikolay\AppData\Local\Programs\Python\Python310\lib\site-packages\torchvision\models\utils.py:223: UserWarning: Arguments other than a weight enum or `
None` for 'weights' are deprecated since 0.13 and will be removed in 0.15. The current behavior is equivalent to passing `weights=GoogLeNet_Weights.IMAGENET1K_V
1`. You can also use `weights=GoogLeNet_Weights.DEFAULT` to get the most up-to-date weights.
 warnings.warn(msg)
                                                                                                                                                          In [18]:
model.to(device)
```

Out[18]:

```
(conv1): BasicConv2d(
 (conv): Conv2d(3, 64, kernel size=(7, 7), stride=(2, 2), padding=(3, 3), bias=False)
 (bn): BatchNorm2d(64, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
(maxpool1): MaxPool2d(kernel_size=3, stride=2, padding=0, dilation=1, ceil_mode=True)
(conv2): BasicConv2d(
 (conv): Conv2d(64, 64, kernel_size=(1, 1), stride=(1, 1), bias=False)
 (bn): BatchNorm2d(64, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
```

```
(conv): Conv2d(64, 192, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
 (bn): BatchNorm2d(192, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
(maxpool2): MaxPool2d(kernel_size=3, stride=2, padding=0, dilation=1, ceil_mode=True)
(inception3a): Inception(
 (branch1): BasicConv2d(
  (conv): Conv2d(192, 64, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn): BatchNorm2d(64, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch2): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(192, 96, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(96, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(96, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(128, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch3): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(192, 16, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(16, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(16, 32, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(32, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch4): Sequential(
  (0): MaxPool2d(kernel_size=3, stride=1, padding=1, dilation=1, ceil_mode=True)
  (1): BasicConv2d(
   (conv): Conv2d(192, 32, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(32, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
(inception3b): Inception(
 (branch1): BasicConv2d(
  (conv): Conv2d(256, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn): BatchNorm2d(128, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch2): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(256, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(128, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(128, 192, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(192, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch3): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(256, 32, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(32, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(32, 96, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(96, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch4): Sequential(
  (0): MaxPool2d(kernel_size=3, stride=1, padding=1, dilation=1, ceil_mode=True)
  (1): BasicConv2d(
   (conv): Conv2d(256, 64, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(64, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
(maxpool3): MaxPool2d(kernel_size=3, stride=2, padding=0, dilation=1, ceil_mode=True)
(inception4a): Inception(
 (branch1): BasicConv2d(
  (conv): Conv2d(480, 192, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn): BatchNorm2d(192, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch2): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(480, 96, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(96, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(96, 208, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d (208, eps=0.001, momentum=0.1, affine=True, track\_running\_stats=True)
 (branch3): Sequential(
```

```
(0): BasicConv2d(
   (conv): Conv2d(480, 16, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(16, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(16, 48, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(48, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch4): Sequential(
  (0): MaxPool2d(kernel_size=3, stride=1, padding=1, dilation=1, ceil_mode=True)
  (1): BasicConv2d(
   (conv): Conv2d(480, 64, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(64, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
(inception4b): Inception(
(branch1): BasicConv2d(
  (conv): Conv2d(512, 160, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn): BatchNorm2d(160, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch2): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(512, 112, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(112, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
   (conv): Conv2d(112, 224, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(224, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch3): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(512, 24, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(24, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(24, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(64, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch4): Sequential(
  (0): MaxPool2d(kernel_size=3, stride=1, padding=1, dilation=1, ceil_mode=True)
  (1): BasicConv2d(
   (conv): Conv2d(512, 64, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(64, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
(inception4c): Inception(
 (branch1): BasicConv2d(
  (conv): Conv2d(512, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn): BatchNorm2d(128, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch2): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(512,\,128,\,kernel\_size=(1,\,1),\,stride=(1,\,1),\,bias=False)
   (bn): BatchNorm2d(128, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(128, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(256, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch3): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(512, 24, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(24, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(24, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(64, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch4): Sequential(
  (0): MaxPool2d(kernel_size=3, stride=1, padding=1, dilation=1, ceil_mode=True)
   (conv): Conv2d(512, 64, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(64, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
(inception4d): Inception(
 (branch1): BasicConv2d(
  (conv): Conv2d(512, 112, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn): BatchNorm2d(112, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
```

```
(branch2): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(512, 144, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(144, eps=0.001, momentum=0.1, affine=True, track running stats=True)
   (conv): Conv2d(144, 288, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(288, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch3): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(512, 32, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(32, eps=0.001, momentum=0.1, affine=True, track running stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(32, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(64, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch4): Sequential(
  (0): MaxPool2d(kernel_size=3, stride=1, padding=1, dilation=1, ceil_mode=True)
  (1): BasicConv2d(
   (conv): Conv2d(512, 64, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(64, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
(inception4e): Inception(
 (branch1): BasicConv2d(
  (conv): Conv2d(528, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn): BatchNorm2d(256, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch2): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(528, 160, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(160, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
   (conv): Conv2d(160, 320, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(320, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch3): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(528, 32, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(32, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(32, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(128, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch4): Sequential(
  (0): MaxPool2d(kernel_size=3, stride=1, padding=1, dilation=1, ceil_mode=True)
   (conv): Conv2d(528, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(128, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
(maxpool4): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1, ceil_mode=True)
(inception5a): Inception(
 (branch1): BasicConv2d(
  (conv): Conv2d(832, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn): BatchNorm2d(256, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch2): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(832, 160, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(160, eps=0.001, momentum=0.1, affine=True, track running stats=True)
  (1): BasicConv2d(
   (conv): Conv2d(160, 320, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(320, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (branch3): Sequential(
  (0): BasicConv2d(
   (conv): Conv2d(832, 32, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn): BatchNorm2d(32, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
   (conv): Conv2d(32, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
   (bn): BatchNorm2d(128, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
```

```
(branch4): Sequential(
    (0): MaxPool2d(kernel_size=3, stride=1, padding=1, dilation=1, ceil_mode=True)
    (1): BasicConv2d(
     (conv): Conv2d(832, 128, kernel size=(1, 1), stride=(1, 1), bias=False)
     (bn): BatchNorm2d(128, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (inception5b): Inception(
  (branch1): BasicConv2d(
    (conv): Conv2d(832, 384, kernel_size=(1, 1), stride=(1, 1), bias=False)
    (bn): BatchNorm2d(384, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (branch2): Sequential(
    (0): BasicConv2d(
     (conv): Conv2d(832, 192, kernel_size=(1, 1), stride=(1, 1), bias=False)
     (bn): BatchNorm2d(192, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
    (1): BasicConv2d(
     (conv): Conv2d(192, 384, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
     (bn): BatchNorm2d(384, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (branch3): Sequential(
    (0): BasicConv2d(
     (conv): Conv2d(832, 48, kernel_size=(1, 1), stride=(1, 1), bias=False)
     (bn): BatchNorm2d(48, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
    (1): BasicConv2d(
     (conv): Conv2d(48, 128, kernel size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
     (bn): BatchNorm2d(128, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
  (branch4): Sequential(
    (0): MaxPool2d(kernel size=3, stride=1, padding=1, dilation=1, ceil mode=True)
    (1): BasicConv2d(
     (conv): Conv2d(832, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
     (bn): BatchNorm2d(128, eps=0.001, momentum=0.1, affine=True, track_running_stats=True)
 (aux1): None
 (aux2): None
 (avgpool): AdaptiveAvgPool2d(output_size=(1, 1))
 (dropout): Dropout(p=0.2, inplace=False)
 (fc): Linear(in_features=1024, out_features=1000, bias=True)
Loss and optimizer
                                                                                                                                                             In [19]:
criterion = nn.CrossEntropyLoss()
optimizer = optim.Adam(model.parameters(), Ir = 1e-3)
                                                                                                                                                             In [20]:
device = torch.device('cuda:0' if torch.cuda.is available() else 'cpu')
                                                                                                                                                             In [21]:
num_epochs = 2
training
                                                                                                                                                             In [25]:
for epoch in range(num_epochs):
  losses = []
  for batch_idx, (data, targets) in enumerate(train_loader):
     data = data.to(device = device)
     targets = targets.to(device = device)
     #forward
     scores = model(data)
     loss = criterion(scores,targets)
     losses.append(loss.item())
     #backward
     optimizer.zero_grad()
     loss.backward()
```

#adam step
optimizer.step()

print(f"Cost at epoch {epoch} is {sum(losses)/len(losses)}")

Cost at epoch 0 is 0.026988541723408633 Cost at epoch 1 is 0.007583270934344961

check accuracy

```
In [30]:
def check_accuracy(loader, model):
  num_correct = 0
  num_samples = 0
  model.eval()
  with torch.no_grad():
     for x,y in loader:
       x = x.to(device = device)
       y = y.to(device = device)
       scores = model(x)
       _, predictions = scores.max(1)
       num_correct += (predictions == y).sum()
       num_samples += predictions.size(0)
     print(f"Got \{num\_correct\} \ / \ \{num\_samples\} \ with \ accuracy \ \{float(num\_correct) \ / \ \{num\_samples\}^*100\}")
  model.train()
                                                                                                                                                                In [31]:
print("Checking accuracy on Training Set")
check_accuracy(train_loader, model)
Checking accuracy on Training Set
Got 26369 / 26378 with accuracy 99.96588065812419
                                                                                                                                                                In [32]:
print("Checking accuracy on Testing Set")
check_accuracy(test_loader, model)
Checking accuracy on Testing Set
Got 11302 / 11305 with accuracy 99.97346306943831
```

In []: