```
In [1]: import torch
    import torch.nn as nn
    from torchvision.transforms import transforms
    import numpy as np
    from torch.autograd import Variable
    from torchvision.models import squeezenet1_1
    import torch.functional as F
    from io import open
    import os
    from PIL import Image
    import pathlib
    import glob
    import cv2
```

```
In [2]: train_path='E:\CNN for Bone fracture Image\Fracture detection\seg_train\seg_tra
test_path='E:\CNN for Bone fracture Image\Fracture detection\seg_test\seg_test
pred_path='E:\CNN for Bone fracture Image\Fracture detection\seg_pred\seg_pred
```

```
In [3]: #categories
    root=pathlib.Path(train_path)
    classes=sorted([j.name.split('/')[-1] for j in root.iterdir()])
```

```
In [4]: #CNN Network
        class ConvNet(nn.Module):
            def __init__(self,num_classes=6):
                super(ConvNet,self).__init__()
                #Output size after convolution filter
                \#((w-f+2P)/s) +1
                #Input shape= (256,3,150,150)
                self.conv1=nn.Conv2d(in_channels=3,out_channels=12,kernel_size=3,strid
                #Shape= (256,12,150,150)
                self.bn1=nn.BatchNorm2d(num_features=12)
                #Shape= (256,12,150,150)
                self.relu1=nn.ReLU()
                #Shape= (256,12,150,150)
                self.pool=nn.MaxPool2d(kernel size=2)
                #Reduce the image size be factor 2
                #Shape= (256,12,75,75)
                self.conv2=nn.Conv2d(in_channels=12,out_channels=20,kernel_size=3,stri
                #Shape= (256,20,75,75)
                self.bn2=nn.BatchNorm2d(num_features=20)
                #Shape= (256,20,75,75)
                self.relu2=nn.ReLU()
                #Shape= (256,20,75,75)
                self.pool2=nn.MaxPool2d(kernel_size=2)
                #Reduce the image size be factor 2
                #Shape= (256,20,75,75)
                self.conv3=nn.Conv2d(in_channels=20,out_channels=32,kernel_size=3,stri
                #Shape= (256,32,75,75)
                self.bn3=nn.BatchNorm2d(num_features=32)
                #Shape= (256,32,75,75)
                self.relu3=nn.ReLU()
                #Shape= (256,32,75,75)
                self.pool3=nn.MaxPool2d(kernel_size=2)
                #Reduce the image size be factor 2
                #Shape= (256,32,75,75)
                self.fc=nn.Linear(in_features=75 * 75 * 32,out_features=num_classes)
                #Feed forwad function
            def forward(self,input):
                output=self.conv1(input)
                output=self.bn1(output)
                output=self.relu1(output)
```

```
output=self.pool(output)
                output=self.conv2(output)
                output=self.bn2(output)
                output=self.relu2(output)
                output=self.conv3(output)
                output=self.bn3(output)
                output=self.relu3(output)
                    #Above output will be in matrix form, with shape (256,32,75,75)
                output=output.view(-1,32*75*75)
                output=self.fc(output)
                return output
In [5]: | checkpoint=torch.load('best checkpoint.model')
        model=ConvNet(num classes=2)
        model.load_state_dict(checkpoint)
        model.eval()
Out[5]: ConvNet(
          (conv1): Conv2d(3, 12, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
          (bn1): BatchNorm2d(12, eps=1e-05, momentum=0.1, affine=True, track_running_
        stats=True)
          (relu1): ReLU()
          (pool): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1, ceil_mode
        =False)
          (conv2): Conv2d(12, 20, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
          (bn2): BatchNorm2d(20, eps=1e-05, momentum=0.1, affine=True, track_running_
        stats=True)
          (relu2): ReLU()
          (pool2): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1, ceil_mod
```

(conv3): Conv2d(20, 32, kernel\_size=(3, 3), stride=(1, 1), padding=(1, 1))
(bn3): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track\_running\_

(pool3): MaxPool2d(kernel\_size=2, stride=2, padding=0, dilation=1, ceil\_mod

(fc): Linear(in\_features=180000, out\_features=2, bias=True)

e=False)

e=False)

)

stats=True)

(relu3): ReLU()

```
In [7]: #prediction function
def prediction(img_path,transformer):
    image=Image.open(img_path)
    image_tensor=transformer(image).float()

    image_tensor=image_tensor.unsqueeze_(0)

    if torch.cuda.is_available():
        image_tensor.cuda()

    input=Variable(image_tensor)

    output=model(input)
    index=output.data.numpy().argmax()

    pred=classes[index]
    return pred
```

```
In [21]: images_path=glob.glob(pred_path+'/*.jpg')
```

```
In [22]: pred_dict={}
         idx = 0
         for imgPath in images_path:
             print("processing, ", imgPath)
             pred_dict[idx]=prediction(imgPath,transformer)
             print("well done")
             idx = idx + 1
         processing, E:\CNN for Bone fracture image\Fracture detection\seg_pred\seg
         _pred\85_jpg.rf.8ea2a0ee640bcf6ecbe162a905cbdbaf.jpg
         processing, E:\CNN for Bone fracture Image\Fracture detection\seg_pred\seg
         _pred\87_jpg.rf.0fe6c38fe536952979272ec167087cd6.jpg
         well done
         processing, E:\CNN for Bone fracture Image\Fracture detection\seg_pred\seg
         _pred\89_jpg.rf.31dc58444ebae3a21b38822f8d0b75b6.jpg
         well done
         processing, E:\CNN for Bone fracture Image\Fracture detection\seg pred\seg
         _pred\8_jpg.rf.1a2c67587b5266449e319b6682d062a4.jpg
         well done
         processing, E:\CNN for Bone fracture Image\Fracture detection\seg pred\seg
         _pred\8_jpg.rf.7f03f15f9a90dfca6712a551d6fb5602 - Copy.jpg
         well done
         processing, E:\CNN for Bone fracture Image\Fracture detection\seg pred\seg
         _pred\91_jpg.rf.c1b726bceb3d0a2e6b2450fe88fac3d7.jpg
         well done
         processing, E:\CNN for Bone fracture Image\Fracture detection\seg pred\seg
         _pred\94_jpg.rf.60be4f10d9cafc5ab5aa2199a938584d.jpg
In [23]: pred_dict
Out[23]: {0: 'Non-fracture Image',
          1: 'Non-fracture Image',
          2: 'Non-fracture Image',
          3: 'Non-fracture Image',
          4: 'Fracture Image',
          5: 'Non-fracture Image',
          6: 'Non-fracture Image',
          7: 'Non-fracture Image',
          8: 'Fracture Image',
          9: 'Non-fracture Image',
          10: 'Non-fracture Image',
          11: 'Non-fracture Image',
```

12: 'Non-fracture Image',
13: 'Non-fracture Image',
14: 'Non-fracture Image',
15: 'Fracture Image',
16: 'Fracture Image',
17: 'Non-fracture Image',
18: 'Non-fracture Image',