Superresolution using an efficient subpixel convolutional neural network

This example illustrates how to use the efficient sub-pixel convolution layer described in "Real-Time Single Image and Video Super-Resolution Using an Efficient Sub-Pixel Convolutional Neural Network" - Shi et al. for increasing spatial resolution within your network for tasks such as superresolution.

PyTorch Super Res Example

optional arguments:

```
-h, --help show this help message and exit
--upscale_factor super resolution upscale factor
--batchSize training batch size
--testBatchSize testing batch size
--nEpochs number of epochs to train for
--lr Learning Rate. Default=0.01
--cuda
```

--threads number of threads for data loader to use Default random seed to use. Default=123

This example trains a super-resolution network on the <u>BSD300 dataset</u>, using crops from the 200 training images, and evaluating on crops of the 100 test images. A snapshot of the model after every epoch with filename model_epoch_.pth

Example Usage:

Train

```
python main.py --upscale_factor 3 --batchSize 4 --testBatchSize
100 --nEpochs 30 --lr 0.001
```

Super Resolve

```
python super_resolve.py --input_image dataset/BSDS300/images/
test/16077.jpg --model model_epoch_500.pth --output_filename
out.png
```

```
from os.path import exists, join, basename
from os import makedirs, remove
from six.moves import urllib
import tarfile
from torchvision.transforms import Compose,
CenterCrop, ToTensor, Resize
from dataset import DatasetFromFolder
def download bsd300(dest="dataset"):
    output_image_dir = join(dest, "BSDS300/images")
    if not exists(output image dir):
        makedirs(dest)
        url = "http://www2.eecs.berkeley.edu/
Research/Projects/CS/vision/bsds/BSDS300-images.tgz"
        print("downloading url ", url)
        data = urllib.request.urlopen(url)
        file_path = join(dest, basename(url))
        with open(file path, 'wb') as f:
            f.write(data.read())
        print("Extracting data")
        with tarfile.open(file_path) as tar:
            for item in tar:
                tar.extract(item, dest)
        remove(file path)
    return output_image dir
def calculate valid crop size(crop size,
upscale factor):
```

```
return crop size - (crop size % upscale factor)
def input transform(crop size, upscale factor):
    return Compose([
        CenterCrop(crop size),
        Resize(crop_size // upscale_factor),
        ToTensor(),
    ])
def target transform(crop size):
    return Compose([
        CenterCrop(crop_size),
        ToTensor(),
    ])
def get_training_set(upscale_factor):
    root dir = download bsd300()
    train dir = join(root dir, "train")
    crop size = calculate valid crop size(256,
upscale factor)
    return DatasetFromFolder(train dir,
input transform=input transform(crop size,
upscale factor),
target transform=target transform(crop size))
def get test set(upscale factor):
    root dir = download bsd300()
    test dir = join(root dir, "test")
    crop_size = calculate_valid_crop_size(256,
upscale_factor)
```

```
return DatasetFromFolder(test dir,
input_transform=input transform(crop size,
upscale_factor),
target transform=target transform(crop size))
import torch.utils.data as data
from os import listdir
from os.path import join
from PIL import Image
def is image file(filename):
    return any(filename.endswith(extension) for
extension in [".png", ".jpg", ".jpeg"])
def load img(filepath):
    img = Image.open(filepath).convert('YCbCr')
    y, _, _ = img.split()
    return y
class DatasetFromFolder(data.Dataset):
    def init (self, image dir,
input_transform=None, target_transform=None):
        super(DatasetFromFolder, self). init ()
        self.image filenames = [join(image dir, x)
for x in listdir(image dir) if is image file(x)]
        self.input transform = input transform
        self.target_transform = target transform
    def getitem (self, index):
        input =
```

```
load img(self.image filenames[index])
        target = input.copy()
        if self.input transform:
            input = self.input transform(input)
        if self.target_transform:
            target = self.target transform(target)
        return input, target
    def len (self):
        return len(self.image filenames)
from future import print function
import argparse
from math import log10
import torch
import torch.nn as nn
import torch.optim as optim
from torch.utils.data import DataLoader
from model import Net
from data import get_training_set, get_test_set
# Training settings
parser =
argparse.ArgumentParser(description='PyTorch Super
Res Example')
parser.add argument('--upscale factor', type=int,
required=True, help="super resolution upscale
factor")
parser.add argument('--batchSize', type=int,
default=64, help='training batch size')
parser.add argument('--testBatchSize', type=int,
default=10, help='testing batch size')
parser.add argument('--nEpochs', type=int,
default=2, help='number of epochs to train for')
parser.add argument('--lr', type=float,
default=0.\overline{0}1, help='Learning Rate. Default=0.01')
```

```
parser.add argument('--cuda', action='store true',
help='use cuda?')
parser.add argument('--threads', type=int,
default=4, help='number of threads for data loader
to use')
parser.add argument('--seed', type=int,
default=123, help='random seed to use. Default=123')
opt = parser.parse args()
print(opt)
if opt.cuda and not torch.cuda.is available():
    raise Exception("No GPU found, please run
without --cuda")
torch.manual_seed(opt.seed)
device = torch.device("cuda" if opt.cuda else "cpu")
print('===> Loading datasets')
train set = get training set(opt.upscale factor)
test set = get test set(opt.upscale factor)
training data loader =
DataLoader(dataset=train_set,
num workers=opt.threads, batch size=opt.batchSize,
shuffle=True)
testing data loader = DataLoader(dataset=test set,
num workers=opt.threads,
batch size=opt.testBatchSize, shuffle=False)
print('===> Building model')
model =
Net(upscale factor=opt.upscale factor).to(device)
criterion = nn.MSELoss()
optimizer = optim.Adam(model.parameters(),
lr=opt.lr)
```

```
def train(epoch):
    epoch loss = 0
    for iteration, batch in
enumerate(training data loader, 1):
        input, target = batch[0].to(device),
batch[1].to(device)
        optimizer.zero grad()
        loss = criterion(model(input), target)
        epoch loss += loss.item()
        loss.backward()
        optimizer.step()
        print("===> Epoch[{}]({}/{}): Loss: {:.
4f}".format(epoch, iteration,
len(training_data_loader), loss.item()))
    print("===> Epoch {} Complete: Avg. Loss: {:.
4f}".format(epoch, epoch loss /
len(training data loader)))
def test():
    avg_psnr = 0
    with torch.no grad():
        for batch in testing_data_loader:
            input, target = batch[0].to(device),
batch[1].to(device)
            prediction = model(input)
            mse = criterion(prediction, target)
            psnr = 10 * log10(1 / mse.item())
            avg_psnr += psnr
    print("===> Avg. PSNR: {:.4f}
dB".format(avg psnr / len(testing data loader)))
```

```
def checkpoint(epoch):
    model out path =
"model epoch {}.pth".format(epoch)
    torch.save(model, model out path)
    print("Checkpoint saved to
{}".format(model out path))
for epoch in range(1, opt.nEpochs + 1):
    train(epoch)
    test()
    checkpoint(epoch)
import torch
import torch.nn as nn
import torch.nn.init as init
class Net(nn.Module):
    def init (self, upscale_factor):
        super(Net, self).__init__()
        self.relu = nn.ReLU()
        self.conv1 = nn.Conv2d(1, 64, (5, 5), (1,
1), (2, 2))
        self.conv2 = nn.Conv2d(64, 64, (3, 3), (1,
1), (1, 1))
        self.conv3 = nn.Conv2d(64, 32, (3, 3), (1,
1), (1, 1))
        self.conv4 = nn.Conv2d(32, upscale factor
** 2, (3, 3), (1, 1), (1, 1))
        self.pixel shuffle =
nn.PixelShuffle(upscale factor)
        self. initialize weights()
    def forward(self, x):
```

```
x = self.relu(self.conv1(x))
        x = self.relu(self.conv2(x))
        x = self.relu(self.conv3(x))
        x = self.pixel shuffle(self.conv4(x))
        return x
    def initialize weights(self):
        init.orthogonal (self.conv1.weight,
init.calculate gain('relu'))
        init.orthogonal (self.conv2.weight,
init.calculate gain('relu'))
        init.orthogonal (self.conv3.weight,
init.calculate gain('relu'))
        init.orthogonal (self.conv4.weight)
from future import print function
import argparse
import torch
from PIL import Image
from torchvision.transforms import ToTensor
import numpy as np
# Training settings
parser =
argparse.ArgumentParser(description='PyTorch Super
Res Example')
parser.add argument('--input image', type=str,
required=True, help='input image to use')
parser.add argument('--model', type=str,
required=True, help='model file to use')
parser.add_argument('--output_filename', type=str,
help='where to save the output image')
parser.add argument('--cuda', action='store true',
help='use cuda')
opt = parser.parse args()
print(opt)
```

```
img = Image.open(opt.input_image).convert('YCbCr')
y, cb, cr = img.split()
model = torch.load(opt.model)
img to tensor = ToTensor()
input = img to tensor(y).view(\frac{1}{1}, \frac{1}{2}, y.size[\frac{1}{2}],
y.size[0])
if opt.cuda:
    model = model.cuda()
    input = input.cuda()
out = model(input)
out = out.cpu()
out img y = out[0].detach().numpy()
out img y *= 255.0
out img y = out img y.clip(0, 255)
out_img_y = Image.fromarray(np.uint8(out_img_y[0]),
mode='L')
out img cb = cb.resize(out img y.size,
Image.BICUBIC)
out img cr = cr.resize(out img y.size,
Image.BICUBIC)
out img = Image.merge('YCbCr', [out img y,
out img cb, out_img_cr]).convert('RGB')
out img.save(opt.output filename)
print('output image saved to ', opt.output filename)
```