

ECE695DL: Homework 4

Leann Demorest

15 March 2021

model.py

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#####
# BME/ECE 695 Deep Learning
# Homework 4 - Leann Demorest
# model.py
# March 15, 2021
#####
import torch
import torch.nn as nn
import torch.nn.functional as F

class Net(nn.Module):
    def __init__(self, Convo_layers, Pad = False):
        super(Net, self).__init__()
        self.Convo_layers = Convo_layers #Set number of
        convolutional layers
        self.Pad = Pad #Padding True or False

        #Set up convolutions and max pool
        self.conv1 = nn.Conv2d(3, 128, 3)
        self.conv2 = nn.Conv2d(128, 128, 3)
        self.pool = nn.MaxPool2d(2, 2)

        #Net 1: One layer
        if self.Convo_layers == 1 and self.Pad == False:
            self.fc1 = nn.Linear(128*31*31, 1000)
        #Net 2: Two layers
        elif self.Convo_layers == 2 and self.Pad == False:
            self.fc1 = nn.Linear(128*14*14, 1000)
        #Net 3: Two layers and padding
        elif self.Pad == True:
            self.conv1 = nn.Conv2d(3, 128, 3, padding = 1)
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        self.fc1 = nn.Linear(128*15*15, 1000)
        self.fc2 = nn.Linear(1000, 10)
    def forward(self, x):
        #First convoltional layer
        x = self.pool(F.relu(self.conv1(x)))
        #Send through 2nd convolution (if applicable) and reshape
        tensor
        if self.Convo_layers == 1 and self.Pad == False:
            x = x.view(-1, 128*31*31)
        elif self.Convo_layers == 2 and self.Pad == False:
            x = self.pool(F.relu(self.conv2(x)))
            x = x.view(-1, 128*14*14)
        elif self.Pad == True:
            x = self.pool(F.relu(self.conv2(x)))
            x = x.view(-1, 128*15*15)
        #send through fully connected first layer
        x = F.relu(self.fc1(x))
        #send through second fully conneced later to output
        x = self.fc2(x)
        return x

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dataloader.py

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#####
# BME/ECE 695 Deep Learning
# Homework 4 - Leann Demorest
# dataloader.py
# March 15, 2021
#####
import torch
import torchvision
import torchvision.transforms as tv
from torch.utils.data import DataLoader, Dataset

import glob
import os
from PIL import Image

class my_dataset_class(Dataset):
    def __init__(self, args, transform):
        self.FolderPath = []

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self.RootFolder = args.root_path
self.class_list = args.class_list
labels = []
for items in self.class_list:
    self.FolderPath.append(os.path.join(self.RootFolder,
    items))
    labels.append([items, self.class_list.index(items)])
self.ImageList = []
for path in self.FolderPath:
    ClassType = path.split("/")[-1]
    #Get the class type from the folder name
    EncodedClass = self.class_list.index(ClassType)
    #Encode it based on class_list
    for image in glob.glob1(path, '*'):
        self.ImageList.append([EncodedClass,
        os.path.join(path, image)])
self.transform = transform
self.Count = {}
for label in labels:
    self.Count[label[1]] = [label[0],
    [0]*len(self.class_list)]

def __len__(self): #get length by looking at total contents of
each folder.
    length = 0
    for path in self.FolderPath:
        length += len(glob.glob1(path, '*'))
    return length

def __getitem__(self, index): #return one transformed item and
label
    image = self.ImageList[index][1]
    image = Image.open(image)
    H, W = image.size
    image = self.transform(image).to(dtype=torch.float64)
    Label = self.ImageList[index][0]
    return image, Label

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hw04_coco_downloader.py

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#####
# BME/ECE 695 Deep Learning
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# hw04_coco_downloader.py

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# March 15, 2021
#####
import argparse
import json
import ast
import requests
import os
from PIL import Image
from requests.exceptions import ConnectionError, ReadTimeout,
TooManyRedirects, MissingSchema, InvalidURL
import logging
from pycocotools.coco import COCO
import numpy as np
import matplotlib.pyplot as plt

import argparse
import json
import ast
import requests
import os
from PIL import Image
from requests.exceptions import ConnectionError, ReadTimeout,
TooManyRedirects, MissingSchema, InvalidURL
import logging

class CocoDownloader():
    def __init__(self, args):
        #set up coco and class variables
        self.json = args.coco_json_path
        self.classes = args.class_list
        self.images_per_class = args.images_per_class
        self.coco = COCO(self.json)
        self.root_path = args.root_path
        self.class_folder = dict.fromkeys(self.classes)
        #make folders if they don't exist
        for category in self.classes:
            self.class_folder[category]=self.root_path+category
            if not os.path.exists(self.class_folder[category]):
                os.makedirs(self.class_folder[category])
        #get image urls for each class
    def get_image_urls(self):
        self.class_urls = dict.fromkeys(self.classes)
        for key in self.class_urls:
            ID = self.coco.getCatIds(key)
            ImgId = self.coco.getImgIds(catIds = ID)
            img_url = self.coco.loadImgs(ImgId)

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        self.class_urls[key] = [i['coco_url'] for i in img_url]
#download and resize images while i<self.images_per_class
def save_resize_images(self):
    for key in self.class_urls:
        i = 0; j = 0 #i = images downloaded, j = list index
        folder = self.class_folder[key]
        while i < self.images_per_class:
            self.current_url = self.class_urls[key][j]
            Result = self.get_image(folder)
            if Result == True: i += 1
            j+=1
#get image (ignore duplicates)
def get_image(self, folder):
    if len(self.current_url) <= 1:
        return False
    try: img_response = requests.get(self.current_url, timeout
    = 1)
    except Exception as e:
        return False
    img_name = self.current_url.split('/')[-1]
    img_name = img_name.split("?")[0]
    img_file_path = os.path.join(folder, img_name)
    if os.path.exists(img_file_path):
        print('duplicate')
        return False
    with open(img_file_path, 'wb') as img_f:
        img_f.write(img_response.content)
    im = Image.open(img_file_path)
    if im.mode != "RGB":
        im = im.convert(mode = "RGB")
    im_resized = im.resize((64, 64), Image.BOX)
    im_resized.save(img_file_path)
    return True

#parse arguments from commandline
parser = argparse.ArgumentParser(description = 'HW04 COCO
downloader')
parser.add_argument('--root_path', required = True, type = str)
parser.add_argument('--coco_json_path', required = True, type =
str)
parser.add_argument('--class_list', required = True, nargs='*',
type=str,)
parser.add_argument('--images_per_class', required=True, type=int)
args, args_other = parser.parse_known_args()

#set up instance and call functions

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CocoDownload = CocoDownloader(args)
CocoDownload.get_image_urls()
CocoDownload.save_resize_images()
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hw04_training.py

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#####
# BME/ECE 695 Deep Learning
# Homework 4 - Leann Demorest
# hw04_training.py
# March 15, 2021
#####
import argparse
import torch
import torch.nn as nn
import torch.nn.functional as F
import torchvision
import torchvision.transforms as tv
import torch.optim as optim
from torchsummary import summary
from torch.utils.data import DataLoader, Dataset

import copy
import glob
import os
import random
import numpy as np
from PIL import Image
import matplotlib.pyplot as plt

from dataloader import my_dataset_class
from model import Net

def run_training(net, path):
    net = copy.deepcopy(net)
    net = net.to(device) #set up CPU/GPU
    CrossEntropy = torch.nn.CrossEntropyLoss() #Set loss function
    optimizer = torch.optim.SGD(net.parameters(), lr = 1e-3,
    momentum = 0.95) #set optimization params
    epochs = 10 #ran 10 epochs on google colab. Only have CPU on PC
    AverageLossPlot = []; IterationLossPlot = [] #Initialize plot
    lists
    for epoch in range(epochs):
        summed_loss = 0.0
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    for batch, data in enumerate(TrainDataLoader):
        #run through dataset with a batch at a time
        images, ground_truth = data
        images = images.to(device).float()
        ground_truth = ground_truth.to(device)
        #forward propogate
        optimizer.zero_grad()
        outputs = net(images)
        #back-propogate
        loss = CrossEntropy(outputs, ground_truth)
        loss.backward()
        optimizer.step()
        summed_loss += loss.item() #convert to float
        if (batch+1)%500 == 0: #append data at batch 250.
            print("\n[epoch:%d, batch:%d] loss:%.2f"%(epoch+1,
                batch+1, summed_loss/float(250)))
            AverageLossPlot.append(summed_loss/float(500))
            IterationLossPlot.append(len(AverageLossPlot))
            summed_loss = 0.0
    torch.save(net.state_dict(), path) #save output to be used for
    validation
    return AverageLossPlot, IterationLossPlot

if __name__ == '__main__':
    #runs CPU on my personal computer. Ran GPU on Google Colab
    if torch.cuda.is_available() == True: device =
        torch.device("cuda:0")
    else: device = torch.device("cpu")

    # initialization reference:
    https://engineering.purdue.edu/DeepLearn/pdf-kak/week3.pdf
    seed = 0
    random.seed(seed)
    torch.manual_seed(seed)
    torch.cuda.manual_seed(seed)
    np.random.seed(seed)
    torch.backends.cudnn.deterministic = True
    torch.backends.cudnn.benchmark = False
    os.environ['PYTHONHASHSEED'] = str(seed)

    #parse arguments from commandline
    parser = argparse.ArgumentParser(description =
        'HW04_Training/Validation')
    parser.add_argument('--root_path', type=str, required=True)
    parser.add_argument('--class_list', nargs='*', type=str,
        required = True)

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args, args_other = parser.parse_known_args()

#Initialize transformation
transform = tvtn.Compose([tvtn.ToTensor(), tvtn.Normalize((0.5,
0.5, 0.5), (0.5, 0.5, 0.5))])
batch = 10

#Call dataloader and initialize instance
TrainingClass = my_dataset_class(args, transform)
TrainDataLoader = torch.utils.data.DataLoader(dataset =
TrainingClass, batch_size = batch, shuffle = True, num_workers
= 0)

#run net 1 and plot
model1 = Net(1, False)
AverageLossPlot, IterationLossPlot = run_training(model1,
"net1.pth")
plt.plot(IterationLossPlot, AverageLossPlot, label = 'Net 1')
#run net 2 and plot
model2 = Net(2, False)
AverageLossPlot, IterationLossPlot = run_training(model2,
"net2.pth")
plt.plot(IterationLossPlot, AverageLossPlot, label = 'Net 2')
#run net 3 and plot
model3 = Net(2, True)
AverageLossPlot, IterationLossPlot = run_training(model3,
"net3.pth")
plt.plot(IterationLossPlot, AverageLossPlot, label='Net 3')
#format plot and save
plt.legend()
plt.title('Comparison of CNN Architectures')
plt.xlabel('Iterations')
plt.ylabel('Loss')
plt.savefig('train_loss.jpg')

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hw04_validation.py

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# BME/ECE 695 Deep Learning
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# hw04_validation.py

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# March 15, 2021
#####
import argparse
import torch
import torch.nn as nn
import torch.nn.functional as F
import torchvision
import torchvision.transforms as tv
import torch.optim as optim
from torchsummary import summary
from torch.utils.data import DataLoader, Dataset
import seaborn as sn
import pandas as pd
import matplotlib.pyplot as plt
import copy
import glob
import os
import random
import numpy as np
from PIL import Image
import matplotlib.pyplot as plt

from dataloader import my_dataset_class
from model import Net

def run_validation(net, path, Validation = False):
    net = copy.deepcopy(net)
    NetValidationClassCount = copy.deepcopy(ValidationClass.Count)
    net = net.to(device)
    CrossEntropy = torch.nn.CrossEntropyLoss()
    TotalRight = 0
    Total = 0
    for batch, data in enumerate(ValidationDataLoader): #run
    through dataset with a batch at a time
        images, ground_truth = data
        images = images.to(device).float()
        ground_truth = ground_truth.to(device)
        outputs = net(images)
        #Determine which label was selected for each image
        prediction = [torch.argmax(output) for output in outputs]
        #Compare with ground truth to get number right
        NumberRight = [prediction[k] == ground_truth[k] for k in
        prediction]
        #Increment total right and total
        TotalRight += sum(bool(l) for l in NumberRight)
        Total += len(ground_truth)

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        #Create dictionary to store predictions for confusion
        matrix
        for j in range(len(ground_truth)):
            NetValidationClassCount[int(ground_truth[j])][1][int(pr
            ediction[j])] += 1
    #Calculate accuracy
    Accuracy = TotalRight/Total

    #Create Confusion Matrix Array
    List = []
    for key in NetValidationClassCount:
        List.append(NetValidationClassCount[key][1])
    Array = pd.DataFrame(List, columns =
    ValidationClass.class_list, index = ValidationClass.class_list)
    plt.figure(figsize = (10,7))
    sn.heatmap(Array, annot=True, fmt = 'd', linewidths=.5)
    NetNumber = path.strip(".pth").strip("net")
    #Plot confusion matrix and save figure
    plt.title("Net "+str(NetNumber)+"\n"+r'Accuracy:
    '+str(Accuracy))
    plt.xlabel('Prediction Label')
    plt.ylabel('Ground Truth Label')
    plt.tight_layout()
    plt.savefig("net"+str(NetNumber)+"_confusion_matrix.jpg")

if __name__ == '__main__':
    if torch.cuda.is_available()== True: device =
    torch.device("cuda:0")
    else: device = torch.device("cpu")

    # initialization reference:
    https://engineering.purdue.edu/DeepLearn/pdf-kak/week3.pdf
    seed = 0
    random.seed(seed)
    torch.manual_seed(seed)
    torch.cuda.manual_seed(seed)
    np.random.seed(seed)
    torch.backends.cudnn.deterministic = True
    torch.backends.cudnn.benchmarks = False
    os.environ['PYTHONHASHSEED'] = str(seed)

    #parse arguments from commandline
    parser = argparse.ArgumentParser(description =
    'HW04_Training/Validation')
    parser.add_argument('--root_path', type=str, required=True)
    parser.add_argument('--class_list', nargs='*', type=str,

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```

required = True)
args, args_other = parser.parse_known_args()

#Initialize transformation
transform = tvn.Compose([tvn.ToTensor(), tvn.Normalize((0.5,
0.5, 0.5), (0.5, 0.5, 0.5))])
batch = 10

#Call dataloader and initialize instance
ValidationClass = my_dataset_class(args, transform)
ValidationDataLoader = torch.utils.data.DataLoader(dataset =
ValidationClass, batch_size = batch, shuffle = True,
num_workers = 0)

#run nets for validation
model1 = Net(1, False)
model1.load_state_dict(torch.load("net1.pth",
map_location=device))
model1.eval()
run_validation(model1, "net1.pth", True)

model2 = Net(2, False)
model2.load_state_dict(torch.load("net2.pth",
map_location=device))
model2.eval()
run_validation(model2, "net2.pth", True)

model3 = Net(2, True)
model3.load_state_dict(torch.load("net3.pth",
map_location=device))
model3.eval()
run_validation(model3, "net3.pth", True)

```

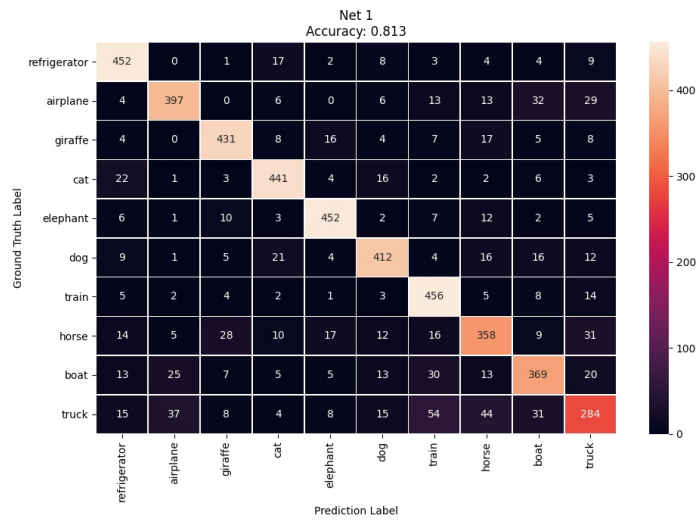


Figure 1: net1 confusion matrix

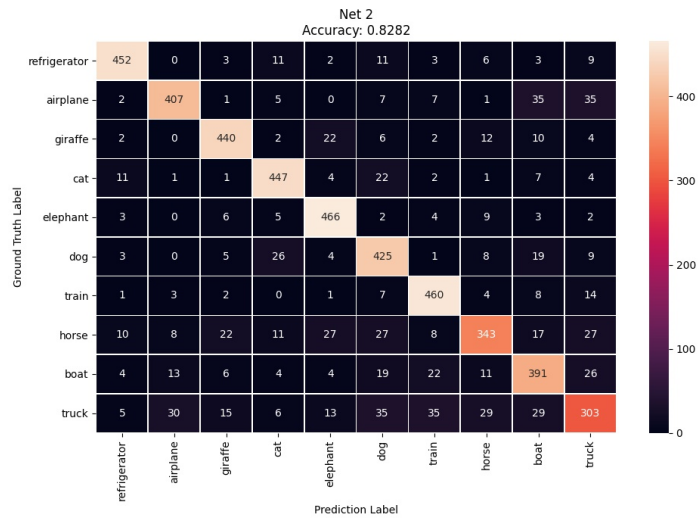


Figure 2: net2 confusion matrix

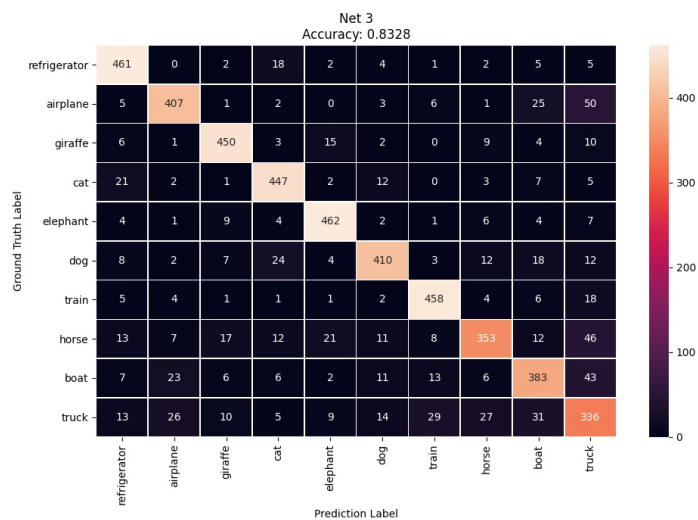


Figure 3: net3 confusion matrix

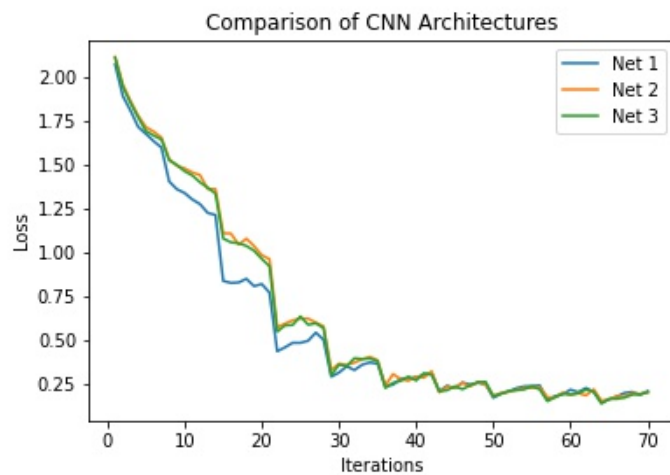


Figure 4: train loss