

sandip2_and_nb17_project_output

April 30, 2022

1 Training and testing Multi-label Image Classifier and using pre-trained scene classifier and a heat map generator

```
[ ]: from google.colab import drive
drive.mount('/content/gdrive')
import os
os.chdir("gdrive/My Drive/Project")
```

Mounted at /content/gdrive

1.1 Downloading and formatting SUN datasets

```
[ ]: # downloading data from SUN dataset
import os
synset_url = 'http://cs.brown.edu/~gmpatter/Attributes/SUNAttributeDB.tar.gz'
os.system('wget ' + synset_url)
synset_url = 'http://cs.brown.edu/~gmpatter/Attributes/SUNAttributeDB_Images.
↳tar.gz'
os.system('wget ' + synset_url)
```

```
[ ]: 2
```

```
[ ]: import tarfile
def unpack_gz(filename):
    # open file
    file = tarfile.open(filename)

    # print file names
    print(file.getnames())

    # extract files
    file.extractall('.')
    # file.extract('sample.txt', './Destination_FolderName')

    # close file
    file.close()
```

```
#unpacking data from SUN DATASET
unpack_gz('SUNAttributeDB_Images.tar.gz')

#unpacking images from SUN DATASET
unpack_gz('SUNAttributeDB.tar.gz')
```

1.2 Importing necessary libraries

```
[ ]: import scipy.io
import os
import numpy as np
import torch
import torch.nn as nn
import torchvision

from torchvision import transforms
from sklearn.metrics import average_precision_score
from PIL import Image, ImageDraw
import matplotlib.pyplot as plt
from sun_dataloader import SunDataset, SUN_CLASSES

%matplotlib inline
%load_ext autoreload
%autoreload 2
```

```
[ ]: # declare what device to use: gpu/cpu
device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
```

1.3 Reading SUN Dataset

1.3.1 Loading Training Data

In the following cell we will load the training data and also apply some transforms to the data.

```
[ ]: # Transforms applied to the training data
normalize = transforms.Normalize(mean=[0.485, 0.456, 0.406],
                                std= [0.229, 0.224, 0.225])

train_transform = transforms.Compose([
    transforms.Resize(227),
    transforms.CenterCrop(227),
    transforms.ToTensor(),
    normalize
])

[ ]: ds_train = SunDataset('images/', 'train', train_transform)
```

1.3.2 Loading Validation Data

Loading the validation data for the SUN dataset.

```
[ ]: # Transforms applied to the testing data
test_transform = transforms.Compose([
    transforms.Resize(227),
    transforms.CenterCrop(227),
    transforms.ToTensor(),
    normalize,
])

[ ]: ds_val = SunDataset('images/', 'val', test_transform)
```

1.3.3 Visualizing the Data

1.3.4 Displaying the sorted distribution of image number per attribute in SUN Database

```
[ ]: labels_mat = scipy.io.loadmat('./SUNAttributeDB/attributeLabels_continuous.mat')
attrib_presence = labels_mat['labels_cv'] # 14340 images x 102 attributes
attrib_presence = (attrib_presence > 1/3).astype(int) # if two out of three
↳ voted that there is an attribute in the image, we will consider that this is
↳ true

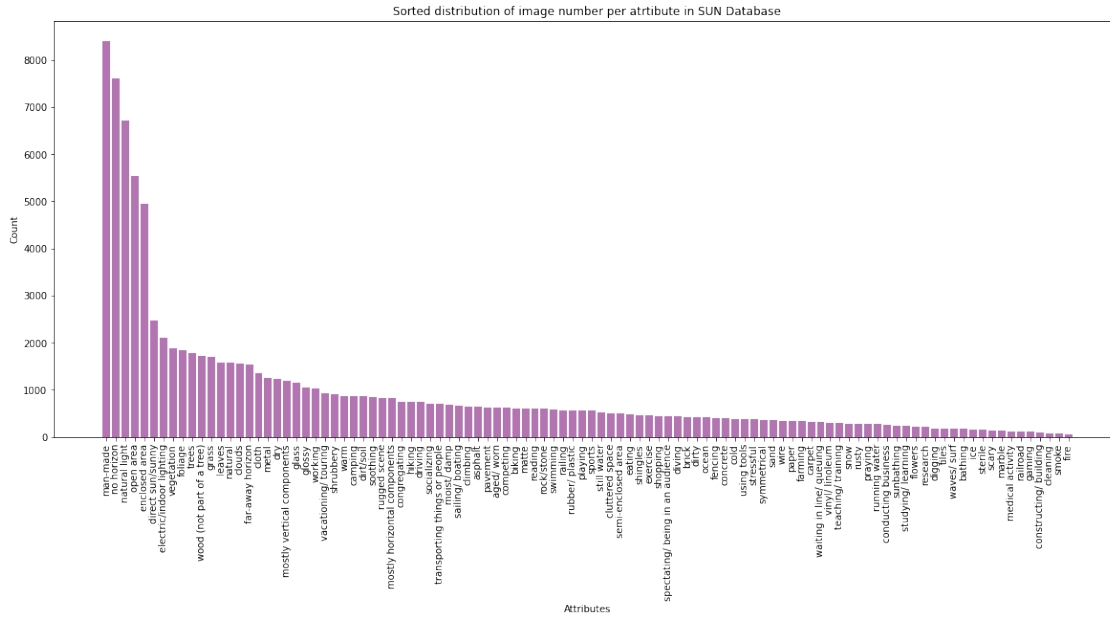
[ ]: attribute_numbers = np.sum(attrib_presence, axis = 0)
indices = np.argsort(attribute_numbers)[::-1]
height = [attribute_numbers[i] for i in indices]
bars = [SUN_CLASSES[i] for i in indices]
x_pos = np.arange(len(bars))

# Create bars and choose color
plt.figure(figsize=(20, 8))
plt.bar(x_pos, height, color = (0.5,0.1,0.5,0.6))

# Add title and axis names
plt.title('Sorted distribution of image number per attribute in SUN Database')
plt.xlabel('Attributes')
plt.ylabel('Count')

# Create names on the x axis
plt.xticks(x_pos, bars, rotation='vertical')

# Show graph
plt.show()
```



1.3.5 Displaying some of the images and the attributes that are present in these images

```
[ ]: for i in range(5):
    idx = np.random.randint(0, len(ds_train.names)+1)
    _imgpath = os.path.join('images/', ds_train.names[idx])
    img = Image.open(_imgpath).convert('RGB')

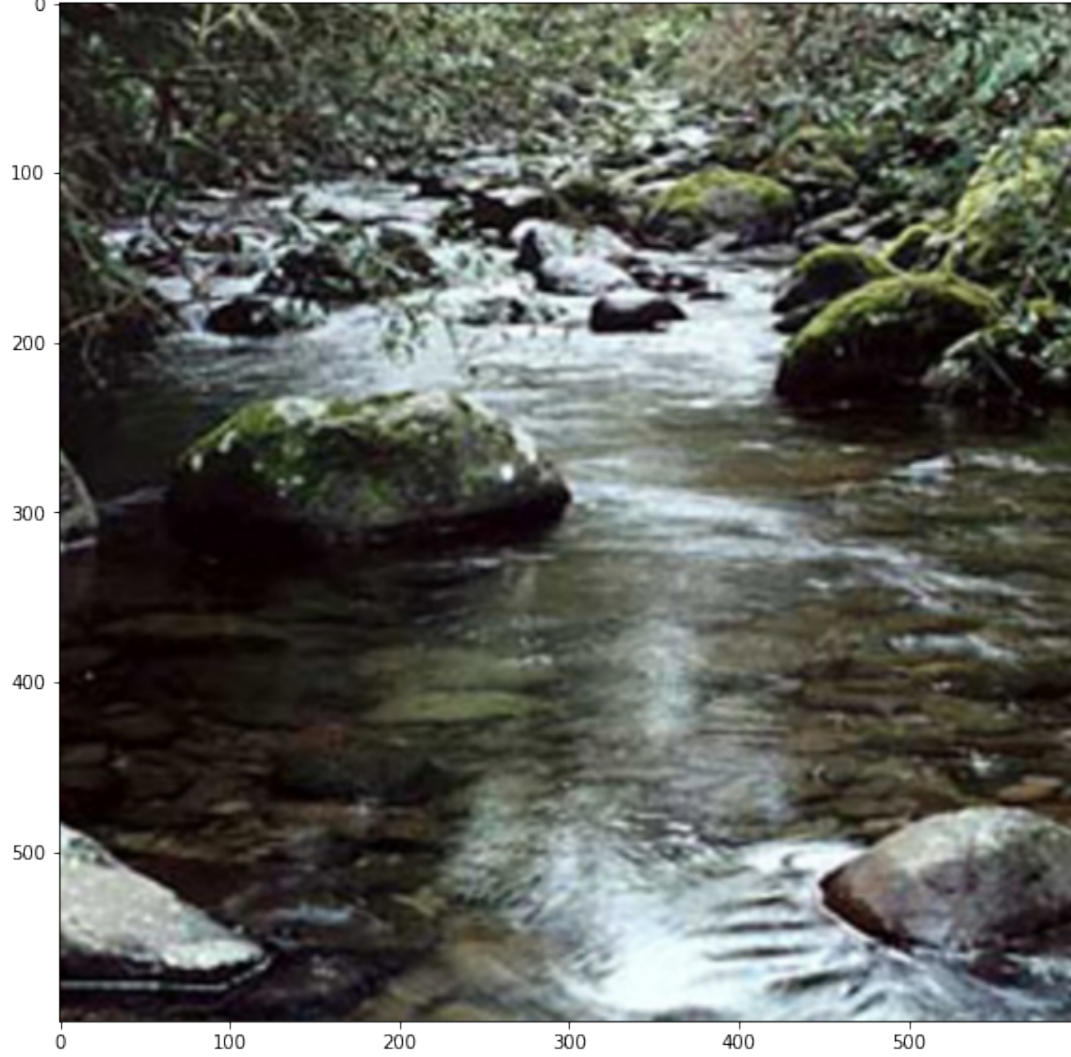
    draw = ImageDraw.Draw(img)
    plt.figure(figsize = (10,10))
    plt.imshow(np.array(img))

    attributes = ds_train.labels[idx]
    present_indices = []
    for i, is_present in enumerate(ds_train.labels[idx]):
        if is_present > 0:
            present_indices.append(i)
    plt.title("Attributes for the image below are: \n" + ", ".
    ↪join([SUN_CLASSES[idx] for idx in present_indices]), pad = 5,
    ↪fontdict={'fontsize': 14})
```

Attributes for the image below are:
wire, pavement, clouds, natural light, open area, far-away horizon



Attributes for the image below are:
foliage, running water, natural light, natural, open area, no horizon, rugged scene



Attributes for the image below are:
shopping, no horizon



Attributes for the image below are:
paper, electric/indoor lighting, matte, man-made, enclosed area, no horizon, cluttered space



Attributes for the image below are:
biking, sports, playing, vegetation, shrubbery, foliage, sand, dirt/soil, natural light, direct sun/sunny, dirty, warm, man-made, open area, far-away horizon



2 Classification

```
[ ]: # declare what device to use: gpu/cpu
device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
```



```
[ ]: train_loader = torch.utils.data.DataLoader(dataset=ds_train,
                                                batch_size=50,
                                                shuffle=True,
                                                num_workers=1)
```

```
[ ]: val_loader = torch.utils.data.DataLoader(dataset=ds_val,
                                              batch_size=50,
                                              shuffle=True,
                                              num_workers=1)
```

```
[ ]: def train_classifier(train_loader, classifier, criterion, optimizer):
    classifier.train()
    loss_ = 0.0
    losses = []
    for i, (images, labels) in enumerate(train_loader):
        images, labels = images.to(device), labels.to(device)
        optimizer.zero_grad()
        logits = classifier(images)
        loss = criterion(logits, labels)
        loss.backward()
        optimizer.step()
        losses.append(loss)
    return torch.stack(losses).mean().item()
```

```
[ ]: def test_classifier(test_loader, classifier, criterion, print_ind_classes=True,
    ↪ print_total=True):
    classifier.eval()
    losses = []
    with torch.no_grad():
        y_true = np.zeros((0,102))
        y_score = np.zeros((0,102))
        for i, (images, labels) in enumerate(test_loader):
            images, labels = images.to(device), labels.to(device)
            logits = classifier(images)
            y_true = np.concatenate((y_true, labels.cpu().numpy()), axis=0)
            y_score = np.concatenate((y_score, logits.cpu().numpy()), axis=0)
            loss = criterion(logits, labels)
            losses.append(loss.item())
    aps = []
    for i in range(y_true.shape[1]):
        ap = average_precision_score(y_true[:, i], y_score[:, i])
        if print_ind_classes:
            print('----- Class: {:<12}      AP: {:>8.4f} -----'.
    ↪ format(SUN_CLASSES[i], ap))
        aps.append(ap)

    mAP = np.mean(aps)
```

```

        test_loss = np.mean(losses)
        if print_total:
            print('mAP: {0:.4f}'.format(mAP))
            print('Avg loss: {}'.format(test_loss))

    return mAP, test_loss, aps

```

```

[ ]: # plot functions
def plot_losses(train, val, test_frequency, num_epochs):
    plt.plot(train, label="train")
    indices = [i for i in range(num_epochs) if ((i+1)%test_frequency == 0 or i_
    ↪==0)]
    plt.plot(indices, val, label="val")
    plt.title("Loss Plot")
    plt.ylabel("Loss")
    plt.xlabel("Epoch")
    plt.legend()
    plt.show()

def plot_mAP(train, val, test_frequency, num_epochs):
    indices = [i for i in range(num_epochs) if ((i+1)%test_frequency == 0 or i_
    ↪==0)]
    plt.plot(indices, train, label="train")
    plt.plot(indices, val, label="val")
    plt.title("mAP Plot")
    plt.ylabel("mAP")
    plt.xlabel("Epoch")
    plt.legend()
    plt.show()

```

2.1 Training the network

```

[ ]: def train(classifier, num_epochs, train_loader, val_loader, criterion,
    ↪optimizer, test_frequency=5):
    train_losses = []
    train_mAPs = []
    val_losses = []
    val_mAPs = []

    for epoch in range(1, num_epochs+1):
        print("Starting epoch number " + str(epoch))
        train_loss = train_classifier(train_loader, classifier, criterion,
        ↪optimizer)
        train_losses.append(train_loss)
        print("Loss for Training on Epoch " + str(epoch) + " is " +
        ↪str(train_loss))

```

```

        if(epoch%test_frequency==0 or epoch==1):
            mAP_train, _, _ = test_classifier(train_loader, classifier,
            ↪criterion, False, False)
            train_mAPs.append(mAP_train)
            mAP_val, val_loss, _ = test_classifier(val_loader, classifier,
            ↪criterion)
            print('Evaluating classifier')
            print("Mean Precision Score for Testing on Epoch " +str(epoch) + "
            ↪is "+ str(mAP_val))
            val_losses.append(val_loss)
            val_mAPs.append(mAP_val)
            scheduler.step()

    return classifier, train_losses, val_losses, train_mAPs, val_mAPs

```

2.2 Training with pretrained DenseNet with 201 layers

```

[ ]: num_epochs = 40
    test_frequency = 5

    # Load Pretrained DenseNet
    classifier = torchvision.models.densenet201(pretrained=True)
    classifier.classifier = nn.Linear(1920, 102)
    classifier = classifier.to(device)
    optimizer = torch.optim.SGD(classifier.parameters(), lr=0.1, momentum=0.9)
    decay_rate = 0.97
    scheduler = torch.optim.lr_scheduler.ExponentialLR(optimizer=optimizer,
    ↪gamma=decay_rate)
    criterion = nn.MultiLabelSoftMarginLoss()

```

Downloading: "<https://download.pytorch.org/models/densenet201-c1103571.pth>" to
/root/.cache/torch/hub/checkpoints/densenet201-c1103571.pth

0%| | 0.00/77.4M [00:00<?, ?B/s]

```

[ ]: classifier, train_losses, val_losses, train_mAPs, val_mAPs = train(classifier,
    ↪num_epochs, train_loader, val_loader, criterion, optimizer, test_frequency)

```

Starting epoch number 1

Loss for Training on Epoch 1 is 0.14774060249328613

-----	Class: driving	AP: 0.7346	-----
-----	Class: biking	AP: 0.6716	-----
-----	Class: transporting things or people	AP: 0.6077	-----
-----	Class: sunbathing	AP: 0.2489	-----
-----	Class: vacationing/ touring	AP: 0.3861	-----
-----	Class: hiking	AP: 0.6272	-----
-----	Class: climbing	AP: 0.6765	-----
-----	Class: camping	AP: 0.7205	-----

-----	Class: reading	AP:	0.4759	-----
-----	Class: studying/ learning	AP:	0.5221	-----
-----	Class: teaching/ training	AP:	0.3831	-----
-----	Class: research	AP:	0.3135	-----
-----	Class: diving	AP:	0.4898	-----
-----	Class: swimming	AP:	0.7288	-----
-----	Class: bathing	AP:	0.3474	-----
-----	Class: eating	AP:	0.4326	-----
-----	Class: cleaning	AP:	0.5294	-----
-----	Class: socializing	AP:	0.5335	-----
-----	Class: congregating	AP:	0.5500	-----
-----	Class: waiting in line/ queuing	AP:	0.2028	-----
-----	Class: competing	AP:	0.6593	-----
-----	Class: sports	AP:	0.8049	-----
-----	Class: exercise	AP:	0.5664	-----
-----	Class: playing	AP:	0.3964	-----
-----	Class: gaming	AP:	0.0922	-----
-----	Class: spectating/ being in an audience	AP:	0.5305	-----
-----	Class: farming	AP:	0.3594	-----
-----	Class: constructing/ building	AP:	0.2049	-----
-----	Class: shopping	AP:	0.6769	-----
-----	Class: medical activity	AP:	0.1243	-----
-----	Class: working	AP:	0.6204	-----
-----	Class: using tools	AP:	0.4033	-----
-----	Class: digging	AP:	0.2202	-----
-----	Class: conducting business	AP:	0.2451	-----
-----	Class: praying	AP:	0.5746	-----
-----	Class: fencing	AP:	0.1062	-----
-----	Class: railing	AP:	0.3184	-----
-----	Class: wire	AP:	0.2069	-----
-----	Class: railroad	AP:	0.2188	-----
-----	Class: trees	AP:	0.8505	-----
-----	Class: grass	AP:	0.8431	-----
-----	Class: vegetation	AP:	0.7852	-----
-----	Class: shrubbery	AP:	0.4842	-----
-----	Class: foliage	AP:	0.8460	-----
-----	Class: leaves	AP:	0.7054	-----
-----	Class: flowers	AP:	0.5863	-----
-----	Class: asphalt	AP:	0.7567	-----
-----	Class: pavement	AP:	0.3747	-----
-----	Class: shingles	AP:	0.5623	-----
-----	Class: carpet	AP:	0.2586	-----
-----	Class: brick	AP:	0.5231	-----
-----	Class: tiles	AP:	0.0360	-----
-----	Class: concrete	AP:	0.1540	-----
-----	Class: metal	AP:	0.5456	-----
-----	Class: paper	AP:	0.4810	-----
-----	Class: wood (not part of a tree)	AP:	0.6448	-----

-----	Class: vinyl/ linoleum	AP: 0.1712	-----
-----	Class: rubber/ plastic	AP: 0.3583	-----
-----	Class: cloth	AP: 0.7459	-----
-----	Class: sand	AP: 0.5588	-----
-----	Class: rock/stone	AP: 0.5887	-----
-----	Class: dirt/soil	AP: 0.6467	-----
-----	Class: marble	AP: 0.0693	-----
-----	Class: glass	AP: 0.4315	-----
-----	Class: waves/ surf	AP: 0.5542	-----
-----	Class: ocean	AP: 0.6972	-----
-----	Class: running water	AP: 0.5889	-----
-----	Class: still water	AP: 0.4204	-----
-----	Class: ice	AP: 0.4536	-----
-----	Class: snow	AP: 0.7785	-----
-----	Class: clouds	AP: 0.8693	-----
-----	Class: smoke	AP: 0.0352	-----
-----	Class: fire	AP: 0.0701	-----
-----	Class: natural light	AP: 0.8527	-----
-----	Class: direct sun/sunny	AP: 0.5790	-----
-----	Class: electric/indoor lighting	AP: 0.7176	-----
-----	Class: aged/ worn	AP: 0.3565	-----
-----	Class: glossy	AP: 0.4345	-----
-----	Class: matte	AP: 0.2320	-----
-----	Class: sterile	AP: 0.1390	-----
-----	Class: moist/ damp	AP: 0.3540	-----
-----	Class: dry	AP: 0.6248	-----
-----	Class: dirty	AP: 0.5714	-----
-----	Class: rusty	AP: 0.3162	-----
-----	Class: warm	AP: 0.3168	-----
-----	Class: cold	AP: 0.6433	-----
-----	Class: natural	AP: 0.8861	-----
-----	Class: man-made	AP: 0.8818	-----
-----	Class: open area	AP: 0.9222	-----
-----	Class: semi-enclosed area	AP: 0.2096	-----
-----	Class: enclosed area	AP: 0.9381	-----
-----	Class: far-away horizon	AP: 0.8413	-----
-----	Class: no horizon	AP: 0.8554	-----
-----	Class: rugged scene	AP: 0.7391	-----
-----	Class: mostly vertical components	AP: 0.4756	-----
-----	Class: mostly horizontal components	AP: 0.2845	-----
-----	Class: symmetrical	AP: 0.2231	-----
-----	Class: cluttered space	AP: 0.4256	-----
-----	Class: scary	AP: 0.2851	-----
-----	Class: soothing	AP: 0.2767	-----
-----	Class: stressful	AP: 0.2988	-----

mAP: 0.4918

Avg loss: 0.10844768660849538

Evaluating classifier

Mean Precision Score for Testing on Epoch 1 is 0.4917565609063136

Starting epoch number 2

Loss for Training on Epoch 2 is 0.10917959362268448

Starting epoch number 3

Loss for Training on Epoch 3 is 0.09801068156957626

Starting epoch number 4

Loss for Training on Epoch 4 is 0.08920326083898544

Starting epoch number 5

Loss for Training on Epoch 5 is 0.08024616539478302

-----	Class: driving	AP:	0.8489	-----
-----	Class: biking	AP:	0.7812	-----
-----	Class: transporting things or people	AP:	0.8865	-----
-----	Class: sunbathing	AP:	0.4251	-----
-----	Class: vacationing/ touring	AP:	0.5732	-----
-----	Class: hiking	AP:	0.7973	-----
-----	Class: climbing	AP:	0.7817	-----
-----	Class: camping	AP:	0.8549	-----
-----	Class: reading	AP:	0.7613	-----
-----	Class: studying/ learning	AP:	0.6200	-----
-----	Class: teaching/ training	AP:	0.6499	-----
-----	Class: research	AP:	0.5928	-----
-----	Class: diving	AP:	0.6734	-----
-----	Class: swimming	AP:	0.8642	-----
-----	Class: bathing	AP:	0.5830	-----
-----	Class: eating	AP:	0.8299	-----
-----	Class: cleaning	AP:	0.6429	-----
-----	Class: socializing	AP:	0.7270	-----
-----	Class: congregating	AP:	0.7814	-----
-----	Class: waiting in line/ queuing	AP:	0.4998	-----
-----	Class: competing	AP:	0.8411	-----
-----	Class: sports	AP:	0.9189	-----
-----	Class: exercise	AP:	0.7180	-----
-----	Class: playing	AP:	0.6711	-----
-----	Class: gaming	AP:	0.5256	-----
-----	Class: spectating/ being in an audience	AP:	0.8828	-----
-----	Class: farming	AP:	0.6970	-----
-----	Class: constructing/ building	AP:	0.4033	-----
-----	Class: shopping	AP:	0.9228	-----
-----	Class: medical activity	AP:	0.5854	-----
-----	Class: working	AP:	0.8747	-----
-----	Class: using tools	AP:	0.7907	-----
-----	Class: digging	AP:	0.6172	-----
-----	Class: conducting business	AP:	0.5732	-----
-----	Class: praying	AP:	0.8030	-----
-----	Class: fencing	AP:	0.4275	-----
-----	Class: railing	AP:	0.7005	-----
-----	Class: wire	AP:	0.6292	-----
-----	Class: railroad	AP:	0.7969	-----

-----	Class: trees	AP:	0.9398	-----
-----	Class: grass	AP:	0.9383	-----
-----	Class: vegetation	AP:	0.8461	-----
-----	Class: shrubbery	AP:	0.7161	-----
-----	Class: foliage	AP:	0.9429	-----
-----	Class: leaves	AP:	0.7941	-----
-----	Class: flowers	AP:	0.9490	-----
-----	Class: asphalt	AP:	0.8517	-----
-----	Class: pavement	AP:	0.5975	-----
-----	Class: shingles	AP:	0.8356	-----
-----	Class: carpet	AP:	0.5631	-----
-----	Class: brick	AP:	0.7334	-----
-----	Class: tiles	AP:	0.3264	-----
-----	Class: concrete	AP:	0.4674	-----
-----	Class: metal	AP:	0.7856	-----
-----	Class: paper	AP:	0.6800	-----
-----	Class: wood (not part of a tree)	AP:	0.8291	-----
-----	Class: vinyl/ linoleum	AP:	0.3766	-----
-----	Class: rubber/ plastic	AP:	0.6008	-----
-----	Class: cloth	AP:	0.9055	-----
-----	Class: sand	AP:	0.7515	-----
-----	Class: rock/stone	AP:	0.8150	-----
-----	Class: dirt/soil	AP:	0.7936	-----
-----	Class: marble	AP:	0.1366	-----
-----	Class: glass	AP:	0.7540	-----
-----	Class: waves/ surf	AP:	0.7413	-----
-----	Class: ocean	AP:	0.8881	-----
-----	Class: running water	AP:	0.8282	-----
-----	Class: still water	AP:	0.7196	-----
-----	Class: ice	AP:	0.7088	-----
-----	Class: snow	AP:	0.9945	-----
-----	Class: clouds	AP:	0.9655	-----
-----	Class: smoke	AP:	0.8333	-----
-----	Class: fire	AP:	0.9429	-----
-----	Class: natural light	AP:	0.9214	-----
-----	Class: direct sun/sunny	AP:	0.7806	-----
-----	Class: electric/indoor lighting	AP:	0.9131	-----
-----	Class: aged/ worn	AP:	0.5983	-----
-----	Class: glossy	AP:	0.6775	-----
-----	Class: matte	AP:	0.4483	-----
-----	Class: sterile	AP:	0.4631	-----
-----	Class: moist/ damp	AP:	0.6340	-----
-----	Class: dry	AP:	0.7246	-----
-----	Class: dirty	AP:	0.7934	-----
-----	Class: rusty	AP:	0.6297	-----
-----	Class: warm	AP:	0.5557	-----
-----	Class: cold	AP:	0.8823	-----
-----	Class: natural	AP:	0.9574	-----

```

----- Class: man-made          AP:  0.9449  -----
----- Class: open area         AP:  0.9558  -----
----- Class: semi-enclosed area AP:  0.5720  -----
----- Class: enclosed area     AP:  0.9725  -----
----- Class: far-away horizon  AP:  0.9432  -----
----- Class: no horizon        AP:  0.9239  -----
----- Class: rugged scene      AP:  0.8452  -----
----- Class: mostly vertical components AP:  0.7119  -----
----- Class: mostly horizontal components AP:  0.5627  -----
----- Class: symmetrical       AP:  0.5587  -----
----- Class: cluttered space   AP:  0.7025  -----
----- Class: scary             AP:  0.5453  -----
----- Class: soothing          AP:  0.6147  -----
----- Class: stressful         AP:  0.5353  -----

```

mAP: 0.7275

Avg loss: 0.07482510278451032

Evaluating classifier

Mean Precision Score for Testing on Epoch 5 is 0.7274553257253299

Starting epoch number 6

Loss for Training on Epoch 6 is 0.0714421272277832

Starting epoch number 7

Loss for Training on Epoch 7 is 0.06306614726781845

Starting epoch number 8

Loss for Training on Epoch 8 is 0.05546502023935318

Starting epoch number 9

Loss for Training on Epoch 9 is 0.048265427350997925

Starting epoch number 10

Loss for Training on Epoch 10 is 0.04157853126525879

```

----- Class: driving          AP:  0.8712  -----
----- Class: biking           AP:  0.8727  -----
----- Class: transporting things or people AP:  0.9211  -----
----- Class: sunbathing       AP:  0.5685  -----
----- Class: vacationing/ touring AP:  0.7748  -----
----- Class: hiking           AP:  0.9057  -----
----- Class: climbing         AP:  0.8787  -----
----- Class: camping          AP:  0.8867  -----
----- Class: reading          AP:  0.8753  -----
----- Class: studying/ learning AP:  0.6971  -----
----- Class: teaching/ training AP:  0.8141  -----
----- Class: research         AP:  0.7937  -----
----- Class: diving           AP:  0.6989  -----
----- Class: swimming        AP:  0.9211  -----
----- Class: bathing          AP:  0.6248  -----
----- Class: eating           AP:  0.9191  -----
----- Class: cleaning         AP:  0.7500  -----
----- Class: socializing      AP:  0.8598  -----
----- Class: congregating     AP:  0.8577  -----
----- Class: waiting in line/ queuing AP:  0.6977  -----

```

-----	Class: competing	AP: 0.8806	-----
-----	Class: sports	AP: 0.9459	-----
-----	Class: exercise	AP: 0.8177	-----
-----	Class: playing	AP: 0.8366	-----
-----	Class: gaming	AP: 0.6864	-----
-----	Class: spectating/ being in an audience	AP: 0.9244	-----
-----	Class: farming	AP: 0.8312	-----
-----	Class: constructing/ building	AP: 0.7145	-----
-----	Class: shopping	AP: 0.9527	-----
-----	Class: medical activity	AP: 0.6938	-----
-----	Class: working	AP: 0.9208	-----
-----	Class: using tools	AP: 0.8970	-----
-----	Class: digging	AP: 0.6738	-----
-----	Class: conducting business	AP: 0.7619	-----
-----	Class: praying	AP: 0.9247	-----
-----	Class: fencing	AP: 0.5466	-----
-----	Class: railing	AP: 0.8681	-----
-----	Class: wire	AP: 0.8739	-----
-----	Class: railroad	AP: 0.7987	-----
-----	Class: trees	AP: 0.9654	-----
-----	Class: grass	AP: 0.9716	-----
-----	Class: vegetation	AP: 0.9376	-----
-----	Class: shrubbery	AP: 0.8694	-----
-----	Class: foliage	AP: 0.9604	-----
-----	Class: leaves	AP: 0.9045	-----
-----	Class: flowers	AP: 1.0000	-----
-----	Class: asphalt	AP: 0.9380	-----
-----	Class: pavement	AP: 0.7898	-----
-----	Class: shingles	AP: 0.9035	-----
-----	Class: carpet	AP: 0.7645	-----
-----	Class: brick	AP: 0.8853	-----
-----	Class: tiles	AP: 0.6592	-----
-----	Class: concrete	AP: 0.7764	-----
-----	Class: metal	AP: 0.9330	-----
-----	Class: paper	AP: 0.8142	-----
-----	Class: wood (not part of a tree)	AP: 0.8915	-----
-----	Class: vinyl/ linoleum	AP: 0.7021	-----
-----	Class: rubber/ plastic	AP: 0.8889	-----
-----	Class: cloth	AP: 0.9712	-----
-----	Class: sand	AP: 0.7985	-----
-----	Class: rock/stone	AP: 0.9168	-----
-----	Class: dirt/soil	AP: 0.9020	-----
-----	Class: marble	AP: 0.4038	-----
-----	Class: glass	AP: 0.9254	-----
-----	Class: waves/ surf	AP: 0.8602	-----
-----	Class: ocean	AP: 0.9110	-----
-----	Class: running water	AP: 0.9354	-----
-----	Class: still water	AP: 0.8128	-----

```

----- Class: ice                AP: 0.7737 -----
----- Class: snow                AP: 0.9561 -----
----- Class: clouds              AP: 0.9904 -----
----- Class: smoke               AP: 1.0000 -----
----- Class: fire                AP: 0.9250 -----
----- Class: natural light       AP: 0.9675 -----
----- Class: direct sun/sunny    AP: 0.9158 -----
----- Class: electric/indoor lighting AP: 0.9589 -----
----- Class: aged/ worn          AP: 0.7729 -----
----- Class: glossy              AP: 0.8332 -----
----- Class: matte               AP: 0.7139 -----
----- Class: sterile             AP: 0.8321 -----
----- Class: moist/ damp         AP: 0.8131 -----
----- Class: dry                 AP: 0.8962 -----
----- Class: dirty               AP: 0.8199 -----
----- Class: rusty               AP: 0.7482 -----
----- Class: warm                AP: 0.8165 -----
----- Class: cold                AP: 0.9385 -----
----- Class: natural             AP: 0.9760 -----
----- Class: man-made            AP: 0.9762 -----
----- Class: open area           AP: 0.9784 -----
----- Class: semi-enclosed area  AP: 0.8820 -----
----- Class: enclosed area       AP: 0.9786 -----
----- Class: far-away horizon    AP: 0.9715 -----
----- Class: no horizon          AP: 0.9602 -----
----- Class: rugged scene        AP: 0.8974 -----
----- Class: mostly vertical components AP: 0.8949 -----
----- Class: mostly horizontal components AP: 0.8267 -----
----- Class: symmetrical         AP: 0.7883 -----
----- Class: cluttered space     AP: 0.8964 -----
----- Class: scary               AP: 0.6709 -----
----- Class: soothing            AP: 0.8574 -----
----- Class: stressful           AP: 0.7713 -----

```

mAP: 0.8488

Avg loss: 0.05157194800417999

Evaluating classifier

Mean Precision Score for Testing on Epoch 10 is 0.8488024385133978

Starting epoch number 11

Loss for Training on Epoch 11 is 0.0361127108335495

Starting epoch number 12

Loss for Training on Epoch 12 is 0.03120662085711956

Starting epoch number 13

Loss for Training on Epoch 13 is 0.027179595082998276

Starting epoch number 14

Loss for Training on Epoch 14 is 0.023117564618587494

Starting epoch number 15

Loss for Training on Epoch 15 is 0.020076945424079895

```

----- Class: driving            AP: 0.9181 -----

```

-----	Class: biking	AP: 0.9207	-----
-----	Class: transporting things or people	AP: 0.9249	-----
-----	Class: sunbathing	AP: 0.7588	-----
-----	Class: vacationing/ touring	AP: 0.8249	-----
-----	Class: hiking	AP: 0.9376	-----
-----	Class: climbing	AP: 0.9050	-----
-----	Class: camping	AP: 0.9241	-----
-----	Class: reading	AP: 0.8637	-----
-----	Class: studying/ learning	AP: 0.8178	-----
-----	Class: teaching/ training	AP: 0.9465	-----
-----	Class: research	AP: 0.8971	-----
-----	Class: diving	AP: 0.8536	-----
-----	Class: swimming	AP: 0.9488	-----
-----	Class: bathing	AP: 0.6691	-----
-----	Class: eating	AP: 0.9353	-----
-----	Class: cleaning	AP: 1.0000	-----
-----	Class: socializing	AP: 0.8873	-----
-----	Class: congregating	AP: 0.9167	-----
-----	Class: waiting in line/ queuing	AP: 0.8363	-----
-----	Class: competing	AP: 0.9358	-----
-----	Class: sports	AP: 0.9621	-----
-----	Class: exercise	AP: 0.8672	-----
-----	Class: playing	AP: 0.9038	-----
-----	Class: gaming	AP: 0.8230	-----
-----	Class: spectating/ being in an audience	AP: 0.9436	-----
-----	Class: farming	AP: 0.8680	-----
-----	Class: constructing/ building	AP: 0.8200	-----
-----	Class: shopping	AP: 0.9707	-----
-----	Class: medical activity	AP: 0.8036	-----
-----	Class: working	AP: 0.9440	-----
-----	Class: using tools	AP: 0.9184	-----
-----	Class: digging	AP: 0.8282	-----
-----	Class: conducting business	AP: 0.8489	-----
-----	Class: praying	AP: 0.8777	-----
-----	Class: fencing	AP: 0.6532	-----
-----	Class: railing	AP: 0.8538	-----
-----	Class: wire	AP: 0.9709	-----
-----	Class: railroad	AP: 0.8818	-----
-----	Class: trees	AP: 0.9832	-----
-----	Class: grass	AP: 0.9707	-----
-----	Class: vegetation	AP: 0.9667	-----
-----	Class: shrubbery	AP: 0.9313	-----
-----	Class: foliage	AP: 0.9795	-----
-----	Class: leaves	AP: 0.9493	-----
-----	Class: flowers	AP: 1.0000	-----
-----	Class: asphalt	AP: 0.9610	-----
-----	Class: pavement	AP: 0.8824	-----
-----	Class: shingles	AP: 0.9209	-----

-----	Class: carpet	AP: 0.8577	-----
-----	Class: brick	AP: 0.9302	-----
-----	Class: tiles	AP: 0.6853	-----
-----	Class: concrete	AP: 0.8484	-----
-----	Class: metal	AP: 0.9563	-----
-----	Class: paper	AP: 0.8831	-----
-----	Class: wood (not part of a tree)	AP: 0.9007	-----
-----	Class: vinyl/ linoleum	AP: 0.8429	-----
-----	Class: rubber/ plastic	AP: 0.9333	-----
-----	Class: cloth	AP: 0.9814	-----
-----	Class: sand	AP: 0.9231	-----
-----	Class: rock/stone	AP: 0.9662	-----
-----	Class: dirt/soil	AP: 0.9405	-----
-----	Class: marble	AP: 0.7589	-----
-----	Class: glass	AP: 0.9405	-----
-----	Class: waves/ surf	AP: 0.8557	-----
-----	Class: ocean	AP: 0.9345	-----
-----	Class: running water	AP: 0.9717	-----
-----	Class: still water	AP: 0.8655	-----
-----	Class: ice	AP: 0.7482	-----
-----	Class: snow	AP: 1.0000	-----
-----	Class: clouds	AP: 0.9954	-----
-----	Class: smoke	AP: 1.0000	-----
-----	Class: fire	AP: 1.0000	-----
-----	Class: natural light	AP: 0.9706	-----
-----	Class: direct sun/sunny	AP: 0.9541	-----
-----	Class: electric/indoor lighting	AP: 0.9794	-----
-----	Class: aged/ worn	AP: 0.8205	-----
-----	Class: glossy	AP: 0.8885	-----
-----	Class: matte	AP: 0.8050	-----
-----	Class: sterile	AP: 0.7978	-----
-----	Class: moist/ damp	AP: 0.9239	-----
-----	Class: dry	AP: 0.9311	-----
-----	Class: dirty	AP: 0.8991	-----
-----	Class: rusty	AP: 0.7963	-----
-----	Class: warm	AP: 0.8891	-----
-----	Class: cold	AP: 0.9792	-----
-----	Class: natural	AP: 0.9859	-----
-----	Class: man-made	AP: 0.9730	-----
-----	Class: open area	AP: 0.9819	-----
-----	Class: semi-enclosed area	AP: 0.8579	-----
-----	Class: enclosed area	AP: 0.9792	-----
-----	Class: far-away horizon	AP: 0.9656	-----
-----	Class: no horizon	AP: 0.9711	-----
-----	Class: rugged scene	AP: 0.9344	-----
-----	Class: mostly vertical components	AP: 0.9127	-----
-----	Class: mostly horizontal components	AP: 0.8717	-----
-----	Class: symmetrical	AP: 0.8839	-----


```

----- Class: cluttered space      AP:  0.9194 -----
----- Class: scary                AP:  0.8924 -----
----- Class: soothing             AP:  0.9035 -----
----- Class: stressful            AP:  0.8709 -----
mAP: 0.9026
Avg loss: 0.040300142212674535
Evaluating classifier
Mean Precision Score for Testing on Epoch 15 is 0.9026054632505955
Starting epoch number 16
Loss for Training on Epoch 16 is 0.016973763704299927
Starting epoch number 17
Loss for Training on Epoch 17 is 0.014170930720865726
Starting epoch number 18
Loss for Training on Epoch 18 is 0.012336528860032558
Starting epoch number 19
Loss for Training on Epoch 19 is 0.010685698129236698
Starting epoch number 20
Loss for Training on Epoch 20 is 0.008981846272945404
----- Class: driving              AP:  0.9343 -----
----- Class: biking              AP:  0.9344 -----
----- Class: transporting things or people      AP:  0.9284 -----
----- Class: sunbathing          AP:  0.9481 -----
----- Class: vacationing/ touring      AP:  0.8469 -----
----- Class: hiking              AP:  0.9575 -----
----- Class: climbing            AP:  0.9221 -----
----- Class: camping             AP:  0.9480 -----
----- Class: reading             AP:  0.8973 -----
----- Class: studying/ learning      AP:  0.8544 -----
----- Class: teaching/ training      AP:  0.9460 -----
----- Class: research            AP:  0.9403 -----
----- Class: diving              AP:  0.8970 -----
----- Class: swimming           AP:  0.9621 -----
----- Class: bathing            AP:  0.7781 -----
----- Class: eating             AP:  0.9433 -----
----- Class: cleaning           AP:  1.0000 -----
----- Class: socializing         AP:  0.9265 -----
----- Class: congregating        AP:  0.9253 -----
----- Class: waiting in line/ queuing      AP:  0.8491 -----
----- Class: competing          AP:  0.9322 -----
----- Class: sports             AP:  0.9748 -----
----- Class: exercise           AP:  0.8930 -----
----- Class: playing            AP:  0.9300 -----
----- Class: gaming             AP:  0.8650 -----
----- Class: spectating/ being in an audience      AP:  0.9516 -----
----- Class: farming            AP:  0.9031 -----
----- Class: constructing/ building      AP:  0.7990 -----
----- Class: shopping           AP:  0.9540 -----
----- Class: medical activity      AP:  0.8545 -----

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-----	Class: working	AP:	0.9493	-----
-----	Class: using tools	AP:	0.9260	-----
-----	Class: digging	AP:	0.8467	-----
-----	Class: conducting business	AP:	0.9003	-----
-----	Class: praying	AP:	0.9197	-----
-----	Class: fencing	AP:	0.7281	-----
-----	Class: railing	AP:	0.8332	-----
-----	Class: wire	AP:	0.9766	-----
-----	Class: railroad	AP:	0.8920	-----
-----	Class: trees	AP:	0.9813	-----
-----	Class: grass	AP:	0.9764	-----
-----	Class: vegetation	AP:	0.9781	-----
-----	Class: shrubbery	AP:	0.9671	-----
-----	Class: foliage	AP:	0.9794	-----
-----	Class: leaves	AP:	0.9607	-----
-----	Class: flowers	AP:	1.0000	-----
-----	Class: asphalt	AP:	0.9643	-----
-----	Class: pavement	AP:	0.8944	-----
-----	Class: shingles	AP:	0.9356	-----
-----	Class: carpet	AP:	0.9060	-----
-----	Class: brick	AP:	0.9173	-----
-----	Class: tiles	AP:	0.7546	-----
-----	Class: concrete	AP:	0.8533	-----
-----	Class: metal	AP:	0.9662	-----
-----	Class: paper	AP:	0.9096	-----
-----	Class: wood (not part of a tree)	AP:	0.9286	-----
-----	Class: vinyl/ linoleum	AP:	0.9108	-----
-----	Class: rubber/ plastic	AP:	0.9376	-----
-----	Class: cloth	AP:	0.9837	-----
-----	Class: sand	AP:	0.9363	-----
-----	Class: rock/stone	AP:	0.9869	-----
-----	Class: dirt/soil	AP:	0.9394	-----
-----	Class: marble	AP:	0.7744	-----
-----	Class: glass	AP:	0.9461	-----
-----	Class: waves/ surf	AP:	0.9531	-----
-----	Class: ocean	AP:	0.9572	-----
-----	Class: running water	AP:	0.9910	-----
-----	Class: still water	AP:	0.9360	-----
-----	Class: ice	AP:	0.8034	-----
-----	Class: snow	AP:	0.9952	-----
-----	Class: clouds	AP:	0.9943	-----
-----	Class: smoke	AP:	1.0000	-----
-----	Class: fire	AP:	0.9667	-----
-----	Class: natural light	AP:	0.9732	-----
-----	Class: direct sun/sunny	AP:	0.9560	-----
-----	Class: electric/indoor lighting	AP:	0.9790	-----
-----	Class: aged/ worn	AP:	0.8792	-----
-----	Class: glossy	AP:	0.9069	-----

-----	Class: matte	AP:	0.8290	-----
-----	Class: sterile	AP:	0.8624	-----
-----	Class: moist/ damp	AP:	0.9203	-----
-----	Class: dry	AP:	0.9393	-----
-----	Class: dirty	AP:	0.9238	-----
-----	Class: rusty	AP:	0.8239	-----
-----	Class: warm	AP:	0.9406	-----
-----	Class: cold	AP:	0.9857	-----
-----	Class: natural	AP:	0.9854	-----
-----	Class: man-made	AP:	0.9810	-----
-----	Class: open area	AP:	0.9874	-----
-----	Class: semi-enclosed area	AP:	0.9279	-----
-----	Class: enclosed area	AP:	0.9838	-----
-----	Class: far-away horizon	AP:	0.9685	-----
-----	Class: no horizon	AP:	0.9713	-----
-----	Class: rugged scene	AP:	0.9425	-----
-----	Class: mostly vertical components	AP:	0.9367	-----
-----	Class: mostly horizontal components	AP:	0.8730	-----
-----	Class: symmetrical	AP:	0.9101	-----
-----	Class: cluttered space	AP:	0.9294	-----
-----	Class: scary	AP:	0.9075	-----
-----	Class: soothing	AP:	0.9377	-----
-----	Class: stressful	AP:	0.8915	-----

mAP: 0.9238

Avg loss: 0.03712121859706681

Evaluating classifier

Mean Precision Score for Testing on Epoch 20 is 0.9238226933886258

Starting epoch number 21

Loss for Training on Epoch 21 is 0.007741714362055063

Starting epoch number 22

Loss for Training on Epoch 22 is 0.006721013225615025

Starting epoch number 23

Loss for Training on Epoch 23 is 0.006086838431656361

Starting epoch number 24

Loss for Training on Epoch 24 is 0.005285778548568487

Starting epoch number 25

Loss for Training on Epoch 25 is 0.004733097739517689

-----	Class: driving	AP:	0.9485	-----
-----	Class: biking	AP:	0.9352	-----
-----	Class: transporting things or people	AP:	0.9308	-----
-----	Class: sunbathing	AP:	0.9712	-----
-----	Class: vacationing/ touring	AP:	0.8729	-----
-----	Class: hiking	AP:	0.9606	-----
-----	Class: climbing	AP:	0.9220	-----
-----	Class: camping	AP:	0.9595	-----
-----	Class: reading	AP:	0.9010	-----
-----	Class: studying/ learning	AP:	0.8754	-----
-----	Class: teaching/ training	AP:	0.9677	-----

-----	Class: research	AP: 0.9502	-----
-----	Class: diving	AP: 0.9160	-----
-----	Class: swimming	AP: 0.9689	-----
-----	Class: bathing	AP: 0.8263	-----
-----	Class: eating	AP: 0.9559	-----
-----	Class: cleaning	AP: 1.0000	-----
-----	Class: socializing	AP: 0.9253	-----
-----	Class: congregating	AP: 0.9279	-----
-----	Class: waiting in line/ queuing	AP: 0.8802	-----
-----	Class: competing	AP: 0.9488	-----
-----	Class: sports	AP: 0.9708	-----
-----	Class: exercise	AP: 0.9077	-----
-----	Class: playing	AP: 0.9256	-----
-----	Class: gaming	AP: 0.8979	-----
-----	Class: spectating/ being in an audience	AP: 0.9543	-----
-----	Class: farming	AP: 0.9030	-----
-----	Class: constructing/ building	AP: 0.7978	-----
-----	Class: shopping	AP: 0.9590	-----
-----	Class: medical activity	AP: 0.8684	-----
-----	Class: working	AP: 0.9568	-----
-----	Class: using tools	AP: 0.9191	-----
-----	Class: digging	AP: 0.8576	-----
-----	Class: conducting business	AP: 0.9082	-----
-----	Class: praying	AP: 0.9312	-----
-----	Class: fencing	AP: 0.7834	-----
-----	Class: railing	AP: 0.8716	-----
-----	Class: wire	AP: 0.9861	-----
-----	Class: railroad	AP: 0.8932	-----
-----	Class: trees	AP: 0.9850	-----
-----	Class: grass	AP: 0.9843	-----
-----	Class: vegetation	AP: 0.9824	-----
-----	Class: shrubbery	AP: 0.9629	-----
-----	Class: foliage	AP: 0.9831	-----
-----	Class: leaves	AP: 0.9732	-----
-----	Class: flowers	AP: 1.0000	-----
-----	Class: asphalt	AP: 0.9718	-----
-----	Class: pavement	AP: 0.9066	-----
-----	Class: shingles	AP: 0.9492	-----
-----	Class: carpet	AP: 0.9125	-----
-----	Class: brick	AP: 0.9745	-----
-----	Class: tiles	AP: 0.7381	-----
-----	Class: concrete	AP: 0.8660	-----
-----	Class: metal	AP: 0.9663	-----
-----	Class: paper	AP: 0.9104	-----
-----	Class: wood (not part of a tree)	AP: 0.9118	-----
-----	Class: vinyl/ linoleum	AP: 0.8981	-----
-----	Class: rubber/ plastic	AP: 0.9343	-----
-----	Class: cloth	AP: 0.9802	-----

-----	Class: sand	AP: 0.9593	-----
-----	Class: rock/stone	AP: 0.9915	-----
-----	Class: dirt/soil	AP: 0.9486	-----
-----	Class: marble	AP: 0.8016	-----
-----	Class: glass	AP: 0.9481	-----
-----	Class: waves/ surf	AP: 0.9764	-----
-----	Class: ocean	AP: 0.9606	-----
-----	Class: running water	AP: 0.9689	-----
-----	Class: still water	AP: 0.9443	-----
-----	Class: ice	AP: 0.7911	-----
-----	Class: snow	AP: 0.9933	-----
-----	Class: clouds	AP: 0.9933	-----
-----	Class: smoke	AP: 1.0000	-----
-----	Class: fire	AP: 1.0000	-----
-----	Class: natural light	AP: 0.9758	-----
-----	Class: direct sun/sunny	AP: 0.9615	-----
-----	Class: electric/indoor lighting	AP: 0.9748	-----
-----	Class: aged/ worn	AP: 0.8862	-----
-----	Class: glossy	AP: 0.9042	-----
-----	Class: matte	AP: 0.8343	-----
-----	Class: sterile	AP: 0.8620	-----
-----	Class: moist/ damp	AP: 0.9409	-----
-----	Class: dry	AP: 0.9393	-----
-----	Class: dirty	AP: 0.9126	-----
-----	Class: rusty	AP: 0.8384	-----
-----	Class: warm	AP: 0.9497	-----
-----	Class: cold	AP: 0.9938	-----
-----	Class: natural	AP: 0.9831	-----
-----	Class: man-made	AP: 0.9862	-----
-----	Class: open area	AP: 0.9870	-----
-----	Class: semi-enclosed area	AP: 0.9312	-----
-----	Class: enclosed area	AP: 0.9841	-----
-----	Class: far-away horizon	AP: 0.9667	-----
-----	Class: no horizon	AP: 0.9700	-----
-----	Class: rugged scene	AP: 0.9526	-----
-----	Class: mostly vertical components	AP: 0.9465	-----
-----	Class: mostly horizontal components	AP: 0.8714	-----
-----	Class: symmetrical	AP: 0.9101	-----
-----	Class: cluttered space	AP: 0.9556	-----
-----	Class: scary	AP: 0.8988	-----
-----	Class: soothing	AP: 0.9472	-----
-----	Class: stressful	AP: 0.8966	-----

mAP: 0.9313

Avg loss: 0.036338181395469044

Evaluating classifier

Mean Precision Score for Testing on Epoch 25 is 0.9313354824456421

Starting epoch number 26

Loss for Training on Epoch 26 is 0.004400004167109728

Starting epoch number 27
Loss for Training on Epoch 27 is 0.0038972629699856043
Starting epoch number 28
Loss for Training on Epoch 28 is 0.003554541151970625
Starting epoch number 29
Loss for Training on Epoch 29 is 0.0032744365744292736
Starting epoch number 30
Loss for Training on Epoch 30 is 0.00294415932148695

-----	Class: driving	AP: 0.9571	-----
-----	Class: biking	AP: 0.9408	-----
-----	Class: transporting things or people	AP: 0.9338	-----
-----	Class: sunbathing	AP: 0.9826	-----
-----	Class: vacationing/ touring	AP: 0.8805	-----
-----	Class: hiking	AP: 0.9677	-----
-----	Class: climbing	AP: 0.9188	-----
-----	Class: camping	AP: 0.9541	-----
-----	Class: reading	AP: 0.8961	-----
-----	Class: studying/ learning	AP: 0.9009	-----
-----	Class: teaching/ training	AP: 0.9668	-----
-----	Class: research	AP: 0.9480	-----
-----	Class: diving	AP: 0.9111	-----
-----	Class: swimming	AP: 0.9708	-----
-----	Class: bathing	AP: 0.8282	-----
-----	Class: eating	AP: 0.9636	-----
-----	Class: cleaning	AP: 1.0000	-----
-----	Class: socializing	AP: 0.9351	-----
-----	Class: congregating	AP: 0.9302	-----
-----	Class: waiting in line/ queuing	AP: 0.8712	-----
-----	Class: competing	AP: 0.9454	-----
-----	Class: sports	AP: 0.9726	-----
-----	Class: exercise	AP: 0.9271	-----
-----	Class: playing	AP: 0.9141	-----
-----	Class: gaming	AP: 0.9448	-----
-----	Class: spectating/ being in an audience	AP: 0.9504	-----
-----	Class: farming	AP: 0.9121	-----
-----	Class: constructing/ building	AP: 0.8437	-----
-----	Class: shopping	AP: 0.9653	-----
-----	Class: medical activity	AP: 0.8802	-----
-----	Class: working	AP: 0.9566	-----
-----	Class: using tools	AP: 0.9253	-----
-----	Class: digging	AP: 0.8738	-----
-----	Class: conducting business	AP: 0.9217	-----
-----	Class: praying	AP: 0.9155	-----
-----	Class: fencing	AP: 0.7935	-----
-----	Class: railing	AP: 0.8736	-----
-----	Class: wire	AP: 0.9821	-----
-----	Class: railroad	AP: 0.8842	-----
-----	Class: trees	AP: 0.9864	-----

-----	Class: grass	AP: 0.9829	-----
-----	Class: vegetation	AP: 0.9852	-----
-----	Class: shrubbery	AP: 0.9703	-----
-----	Class: foliage	AP: 0.9885	-----
-----	Class: leaves	AP: 0.9675	-----
-----	Class: flowers	AP: 1.0000	-----
-----	Class: asphalt	AP: 0.9765	-----
-----	Class: pavement	AP: 0.9237	-----
-----	Class: shingles	AP: 0.9436	-----
-----	Class: carpet	AP: 0.9285	-----
-----	Class: brick	AP: 0.9469	-----
-----	Class: tiles	AP: 0.7345	-----
-----	Class: concrete	AP: 0.8697	-----
-----	Class: metal	AP: 0.9672	-----
-----	Class: paper	AP: 0.9109	-----
-----	Class: wood (not part of a tree)	AP: 0.9239	-----
-----	Class: vinyl/ linoleum	AP: 0.9106	-----
-----	Class: rubber/ plastic	AP: 0.9347	-----
-----	Class: cloth	AP: 0.9845	-----
-----	Class: sand	AP: 0.9617	-----
-----	Class: rock/stone	AP: 0.9938	-----
-----	Class: dirt/soil	AP: 0.9488	-----
-----	Class: marble	AP: 0.7826	-----
-----	Class: glass	AP: 0.9497	-----
-----	Class: waves/ surf	AP: 0.9938	-----
-----	Class: ocean	AP: 0.9670	-----
-----	Class: running water	AP: 0.9803	-----
-----	Class: still water	AP: 0.9538	-----
-----	Class: ice	AP: 0.8041	-----
-----	Class: snow	AP: 0.9957	-----
-----	Class: clouds	AP: 0.9965	-----
-----	Class: smoke	AP: 1.0000	-----
-----	Class: fire	AP: 1.0000	-----
-----	Class: natural light	AP: 0.9795	-----
-----	Class: direct sun/sunny	AP: 0.9606	-----
-----	Class: electric/indoor lighting	AP: 0.9823	-----
-----	Class: aged/ worn	AP: 0.8920	-----
-----	Class: glossy	AP: 0.9184	-----
-----	Class: matte	AP: 0.8318	-----
-----	Class: sterile	AP: 0.8871	-----
-----	Class: moist/ damp	AP: 0.9513	-----
-----	Class: dry	AP: 0.9442	-----
-----	Class: dirty	AP: 0.9207	-----
-----	Class: rusty	AP: 0.8414	-----
-----	Class: warm	AP: 0.9551	-----
-----	Class: cold	AP: 0.9973	-----
-----	Class: natural	AP: 0.9897	-----
-----	Class: man-made	AP: 0.9855	-----

```

----- Class: open area      AP:  0.9873  -----
----- Class: semi-enclosed area  AP:  0.9433  -----
----- Class: enclosed area    AP:  0.9853  -----
----- Class: far-away horizon  AP:  0.9751  -----
----- Class: no horizon       AP:  0.9676  -----
----- Class: rugged scene     AP:  0.9542  -----
----- Class: mostly vertical components  AP:  0.9438  -----
----- Class: mostly horizontal components  AP:  0.8824  -----
----- Class: symmetrical      AP:  0.9065  -----
----- Class: cluttered space   AP:  0.9452  -----
----- Class: scary            AP:  0.9261  -----
----- Class: soothing         AP:  0.9408  -----
----- Class: stressful        AP:  0.8900  -----

```

mAP: 0.9355

Avg loss: 0.036887487011223005

Evaluating classifier

Mean Precision Score for Testing on Epoch 30 is 0.9355186768650677

Starting epoch number 31

Loss for Training on Epoch 31 is 0.002681059530004859

Starting epoch number 32

Loss for Training on Epoch 32 is 0.0024660290218889713

Starting epoch number 33

Loss for Training on Epoch 33 is 0.0023123957216739655

Starting epoch number 34

Loss for Training on Epoch 34 is 0.0022481216583400965

Starting epoch number 35

Loss for Training on Epoch 35 is 0.002160392003133893

```

----- Class: driving      AP:  0.9510  -----
----- Class: biking      AP:  0.9443  -----
----- Class: transporting things or people  AP:  0.9311  -----
----- Class: sunbathing   AP:  0.9812  -----
----- Class: vacationing/ touring  AP:  0.8813  -----
----- Class: hiking       AP:  0.9633  -----
----- Class: climbing     AP:  0.9257  -----
----- Class: camping      AP:  0.9613  -----
----- Class: reading      AP:  0.8997  -----
----- Class: studying/ learning  AP:  0.9099  -----
----- Class: teaching/ training  AP:  0.9652  -----
----- Class: research     AP:  0.9534  -----
----- Class: diving       AP:  0.9005  -----
----- Class: swimming     AP:  0.9890  -----
----- Class: bathing      AP:  0.8267  -----
----- Class: eating       AP:  0.9662  -----
----- Class: cleaning     AP:  1.0000  -----
----- Class: socializing  AP:  0.9401  -----
----- Class: congregating AP:  0.9333  -----
----- Class: waiting in line/ queuing  AP:  0.8713  -----
----- Class: competing    AP:  0.9483  -----

```

-----	Class: sports	AP: 0.9739	-----
-----	Class: exercise	AP: 0.9115	-----
-----	Class: playing	AP: 0.9322	-----
-----	Class: gaming	AP: 0.9746	-----
-----	Class: spectating/ being in an audience	AP: 0.9495	-----
-----	Class: farming	AP: 0.9099	-----
-----	Class: constructing/ building	AP: 0.8534	-----
-----	Class: shopping	AP: 0.9633	-----
-----	Class: medical activity	AP: 0.8710	-----
-----	Class: working	AP: 0.9569	-----
-----	Class: using tools	AP: 0.9205	-----
-----	Class: digging	AP: 0.8661	-----
-----	Class: conducting business	AP: 0.9081	-----
-----	Class: praying	AP: 0.9145	-----
-----	Class: fencing	AP: 0.8049	-----
-----	Class: railing	AP: 0.8632	-----
-----	Class: wire	AP: 0.9787	-----
-----	Class: railroad	AP: 0.8970	-----
-----	Class: trees	AP: 0.9863	-----
-----	Class: grass	AP: 0.9843	-----
-----	Class: vegetation	AP: 0.9828	-----
-----	Class: shrubbery	AP: 0.9654	-----
-----	Class: foliage	AP: 0.9867	-----
-----	Class: leaves	AP: 0.9689	-----
-----	Class: flowers	AP: 1.0000	-----
-----	Class: asphalt	AP: 0.9781	-----
-----	Class: pavement	AP: 0.9282	-----
-----	Class: shingles	AP: 0.9496	-----
-----	Class: carpet	AP: 0.9298	-----
-----	Class: brick	AP: 0.9390	-----
-----	Class: tiles	AP: 0.7663	-----
-----	Class: concrete	AP: 0.8592	-----
-----	Class: metal	AP: 0.9644	-----
-----	Class: paper	AP: 0.9119	-----
-----	Class: wood (not part of a tree)	AP: 0.9345	-----
-----	Class: vinyl/ linoleum	AP: 0.9012	-----
-----	Class: rubber/ plastic	AP: 0.9431	-----
-----	Class: cloth	AP: 0.9850	-----
-----	Class: sand	AP: 0.9632	-----
-----	Class: rock/stone	AP: 0.9902	-----
-----	Class: dirt/soil	AP: 0.9467	-----
-----	Class: marble	AP: 0.7992	-----
-----	Class: glass	AP: 0.9554	-----
-----	Class: waves/ surf	AP: 0.9938	-----
-----	Class: ocean	AP: 0.9692	-----
-----	Class: running water	AP: 0.9754	-----
-----	Class: still water	AP: 0.9602	-----
-----	Class: ice	AP: 0.8184	-----

-----	Class: snow	AP: 0.9985	-----
-----	Class: clouds	AP: 0.9953	-----
-----	Class: smoke	AP: 1.0000	-----
-----	Class: fire	AP: 0.9667	-----
-----	Class: natural light	AP: 0.9832	-----
-----	Class: direct sun/sunny	AP: 0.9574	-----
-----	Class: electric/indoor lighting	AP: 0.9855	-----
-----	Class: aged/ worn	AP: 0.8883	-----
-----	Class: glossy	AP: 0.9105	-----
-----	Class: matte	AP: 0.8481	-----
-----	Class: sterile	AP: 0.8771	-----
-----	Class: moist/ damp	AP: 0.9488	-----
-----	Class: dry	AP: 0.9452	-----
-----	Class: dirty	AP: 0.9276	-----
-----	Class: rusty	AP: 0.8285	-----
-----	Class: warm	AP: 0.9547	-----
-----	Class: cold	AP: 0.9928	-----
-----	Class: natural	AP: 0.9849	-----
-----	Class: man-made	AP: 0.9876	-----
-----	Class: open area	AP: 0.9872	-----
-----	Class: semi-enclosed area	AP: 0.9441	-----
-----	Class: enclosed area	AP: 0.9879	-----
-----	Class: far-away horizon	AP: 0.9702	-----
-----	Class: no horizon	AP: 0.9716	-----
-----	Class: rugged scene	AP: 0.9607	-----
-----	Class: mostly vertical components	AP: 0.9493	-----
-----	Class: mostly horizontal components	AP: 0.8810	-----
-----	Class: symmetrical	AP: 0.9118	-----
-----	Class: cluttered space	AP: 0.9466	-----
-----	Class: scary	AP: 0.9284	-----
-----	Class: soothing	AP: 0.9405	-----
-----	Class: stressful	AP: 0.8906	-----

mAP: 0.9364

Avg loss: 0.03818422883492092

Evaluating classifier

Mean Precision Score for Testing on Epoch 35 is 0.9364261012813871

Starting epoch number 36

Loss for Training on Epoch 36 is 0.002036087913438678

Starting epoch number 37

Loss for Training on Epoch 37 is 0.001957477070391178

Starting epoch number 38

Loss for Training on Epoch 38 is 0.0018339568050578237

Starting epoch number 39

Loss for Training on Epoch 39 is 0.0017476597568020225

Starting epoch number 40

Loss for Training on Epoch 40 is 0.0016411376418545842

-----	Class: driving	AP: 0.9598	-----
-----	Class: biking	AP: 0.9380	-----

-----	Class: transporting things or people	AP: 0.9311	-----
-----	Class: sunbathing	AP: 0.9806	-----
-----	Class: vacationing/ touring	AP: 0.8810	-----
-----	Class: hiking	AP: 0.9657	-----
-----	Class: climbing	AP: 0.9368	-----
-----	Class: camping	AP: 0.9613	-----
-----	Class: reading	AP: 0.8980	-----
-----	Class: studying/ learning	AP: 0.9305	-----
-----	Class: teaching/ training	AP: 0.9687	-----
-----	Class: research	AP: 0.9535	-----
-----	Class: diving	AP: 0.9389	-----
-----	Class: swimming	AP: 0.9844	-----
-----	Class: bathing	AP: 0.8262	-----
-----	Class: eating	AP: 0.9579	-----
-----	Class: cleaning	AP: 1.0000	-----
-----	Class: socializing	AP: 0.9434	-----
-----	Class: congregating	AP: 0.9334	-----
-----	Class: waiting in line/ queuing	AP: 0.8655	-----
-----	Class: competing	AP: 0.9498	-----
-----	Class: sports	AP: 0.9721	-----
-----	Class: exercise	AP: 0.8887	-----
-----	Class: playing	AP: 0.9208	-----
-----	Class: gaming	AP: 0.9690	-----
-----	Class: spectating/ being in an audience	AP: 0.9506	-----
-----	Class: farming	AP: 0.9123	-----
-----	Class: constructing/ building	AP: 0.8288	-----
-----	Class: shopping	AP: 0.9752	-----
-----	Class: medical activity	AP: 0.8785	-----
-----	Class: working	AP: 0.9552	-----
-----	Class: using tools	AP: 0.9223	-----
-----	Class: digging	AP: 0.8469	-----
-----	Class: conducting business	AP: 0.9146	-----
-----	Class: praying	AP: 0.8994	-----
-----	Class: fencing	AP: 0.8104	-----
-----	Class: railing	AP: 0.8722	-----
-----	Class: wire	AP: 0.9775	-----
-----	Class: railroad	AP: 0.8957	-----
-----	Class: trees	AP: 0.9875	-----
-----	Class: grass	AP: 0.9790	-----
-----	Class: vegetation	AP: 0.9863	-----
-----	Class: shrubbery	AP: 0.9686	-----
-----	Class: foliage	AP: 0.9878	-----
-----	Class: leaves	AP: 0.9620	-----
-----	Class: flowers	AP: 1.0000	-----
-----	Class: asphalt	AP: 0.9785	-----
-----	Class: pavement	AP: 0.9175	-----
-----	Class: shingles	AP: 0.9396	-----
-----	Class: carpet	AP: 0.9259	-----

-----	Class: brick	AP: 0.9626	-----
-----	Class: tiles	AP: 0.7603	-----
-----	Class: concrete	AP: 0.8528	-----
-----	Class: metal	AP: 0.9643	-----
-----	Class: paper	AP: 0.9108	-----
-----	Class: wood (not part of a tree)	AP: 0.9348	-----
-----	Class: vinyl/ linoleum	AP: 0.9002	-----
-----	Class: rubber/ plastic	AP: 0.9356	-----
-----	Class: cloth	AP: 0.9836	-----
-----	Class: sand	AP: 0.9642	-----
-----	Class: rock/stone	AP: 0.9929	-----
-----	Class: dirt/soil	AP: 0.9478	-----
-----	Class: marble	AP: 0.8100	-----
-----	Class: glass	AP: 0.9539	-----
-----	Class: waves/ surf	AP: 0.9903	-----
-----	Class: ocean	AP: 0.9633	-----
-----	Class: running water	AP: 0.9813	-----
-----	Class: still water	AP: 0.9605	-----
-----	Class: ice	AP: 0.8119	-----
-----	Class: snow	AP: 0.9933	-----
-----	Class: clouds	AP: 0.9955	-----
-----	Class: smoke	AP: 1.0000	-----
-----	Class: fire	AP: 1.0000	-----
-----	Class: natural light	AP: 0.9809	-----
-----	Class: direct sun/sunny	AP: 0.9645	-----
-----	Class: electric/indoor lighting	AP: 0.9843	-----
-----	Class: aged/ worn	AP: 0.8969	-----
-----	Class: glossy	AP: 0.9117	-----
-----	Class: matte	AP: 0.8486	-----
-----	Class: sterile	AP: 0.8678	-----
-----	Class: moist/ damp	AP: 0.9540	-----
-----	Class: dry	AP: 0.9446	-----
-----	Class: dirty	AP: 0.9378	-----
-----	Class: rusty	AP: 0.8207	-----
-----	Class: warm	AP: 0.9593	-----
-----	Class: cold	AP: 0.9938	-----
-----	Class: natural	AP: 0.9869	-----
-----	Class: man-made	AP: 0.9870	-----
-----	Class: open area	AP: 0.9890	-----
-----	Class: semi-enclosed area	AP: 0.9405	-----
-----	Class: enclosed area	AP: 0.9883	-----
-----	Class: far-away horizon	AP: 0.9704	-----
-----	Class: no horizon	AP: 0.9718	-----
-----	Class: rugged scene	AP: 0.9562	-----
-----	Class: mostly vertical components	AP: 0.9503	-----
-----	Class: mostly horizontal components	AP: 0.8840	-----
-----	Class: symmetrical	AP: 0.9139	-----
-----	Class: cluttered space	AP: 0.9509	-----


```
----- Class: scary          AP:  0.9346 -----  
----- Class: soothing       AP:  0.9432 -----  
----- Class: stressful      AP:  0.8880 -----
```

mAP: 0.9368

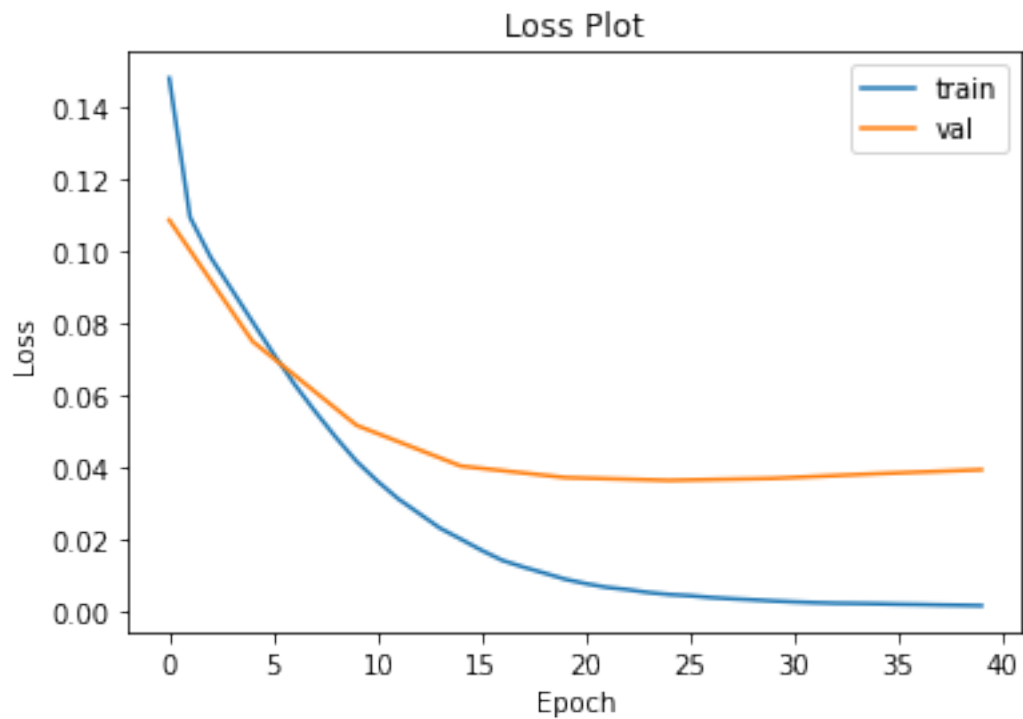
Avg loss: 0.0392803238897488

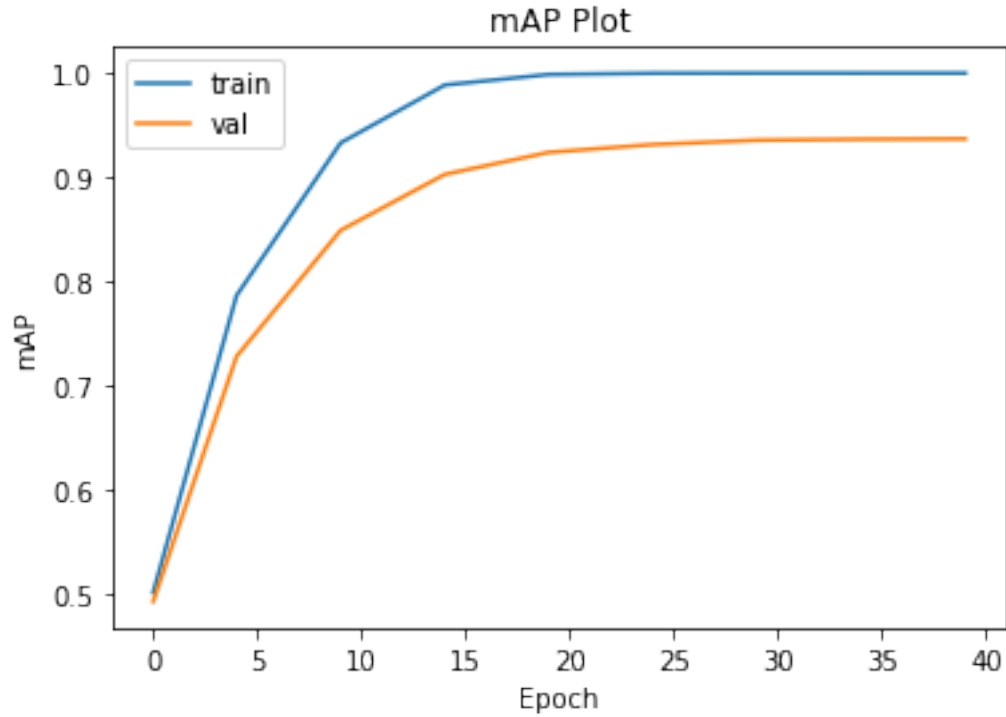
Evaluating classifier

Mean Precision Score for Testing on Epoch 40 is 0.9368053344609433

```
[ ]: torch.save(classifier.state_dict(), './sun_attribute_classifier.pth')
```

```
[ ]: plot_losses(train_losses, val_losses, test_frequency, num_epochs)  
     plot_mAP(train_mAPs, val_mAPs, test_frequency, num_epochs)
```





2.3 Testing

```
[ ]: ds_test = SunDataset('images/', 'test', test_transform)
test_loader = torch.utils.data.DataLoader(dataset=ds_test,
                                           batch_size=50,
                                           shuffle=False,
                                           num_workers=1)

[ ]: mAP_test, test_loss, test_aps = test_classifier(test_loader, classifier,
↪ criterion)
print("Test mAP: ", mAP_test)
```

```
----- Class: driving          AP:  0.9876 -----
----- Class: biking          AP:  0.9742 -----
----- Class: transporting things or people      AP:  0.9460 -----
----- Class: sunbathing       AP:  0.9251 -----
----- Class: vacationing/ touring      AP:  0.9345 -----
----- Class: hiking           AP:  0.9697 -----
----- Class: climbing         AP:  0.9732 -----
----- Class: camping          AP:  0.9915 -----
----- Class: reading          AP:  0.8877 -----
----- Class: studying/ learning      AP:  0.9697 -----
----- Class: teaching/ training      AP:  0.8502 -----
----- Class: research         AP:  0.9530 -----
```

-----	Class: diving	AP: 0.9667	-----
-----	Class: swimming	AP: 0.9456	-----
-----	Class: bathing	AP: 0.9911	-----
-----	Class: eating	AP: 0.9539	-----
-----	Class: cleaning	AP: 1.0000	-----
-----	Class: socializing	AP: 0.9098	-----
-----	Class: congregating	AP: 0.9641	-----
-----	Class: waiting in line/ queuing	AP: 0.7965	-----
-----	Class: competing	AP: 0.9637	-----
-----	Class: sports	AP: 0.9834	-----
-----	Class: exercise	AP: 0.9395	-----
-----	Class: playing	AP: 0.9603	-----
-----	Class: gaming	AP: 0.9809	-----
-----	Class: spectating/ being in an audience	AP: 0.9873	-----
-----	Class: farming	AP: 0.9587	-----
-----	Class: constructing/ building	AP: 1.0000	-----
-----	Class: shopping	AP: 0.9687	-----
-----	Class: medical activity	AP: 0.9272	-----
-----	Class: working	AP: 0.9612	-----
-----	Class: using tools	AP: 0.9831	-----
-----	Class: digging	AP: 0.9298	-----
-----	Class: conducting business	AP: 0.8189	-----
-----	Class: praying	AP: 1.0000	-----
-----	Class: fencing	AP: 0.7758	-----
-----	Class: railing	AP: 0.9561	-----
-----	Class: wire	AP: 0.8876	-----
-----	Class: railroad	AP: 1.0000	-----
-----	Class: trees	AP: 0.9906	-----
-----	Class: grass	AP: 0.9735	-----
-----	Class: vegetation	AP: 0.9871	-----
-----	Class: shrubbery	AP: 0.9333	-----
-----	Class: foliage	AP: 0.9812	-----
-----	Class: leaves	AP: 0.9871	-----
-----	Class: flowers	AP: 0.9971	-----
-----	Class: asphalt	AP: 0.9807	-----
-----	Class: pavement	AP: 0.9439	-----
-----	Class: shingles	AP: 0.9635	-----
-----	Class: carpet	AP: 0.9973	-----
-----	Class: brick	AP: 0.9334	-----
-----	Class: tiles	AP: 0.8407	-----
-----	Class: concrete	AP: 0.7910	-----
-----	Class: metal	AP: 0.9383	-----
-----	Class: paper	AP: 0.9611	-----
-----	Class: wood (not part of a tree)	AP: 0.9833	-----
-----	Class: vinyl/ linoleum	AP: 0.8094	-----
-----	Class: rubber/ plastic	AP: 0.9404	-----
-----	Class: cloth	AP: 0.9232	-----
-----	Class: sand	AP: 0.9589	-----

-----	Class: rock/stone	AP: 0.9552	-----
-----	Class: dirt/soil	AP: 0.9727	-----
-----	Class: marble	AP: 0.9087	-----
-----	Class: glass	AP: 0.9491	-----
-----	Class: waves/ surf	AP: 0.9704	-----
-----	Class: ocean	AP: 0.9586	-----
-----	Class: running water	AP: 0.9334	-----
-----	Class: still water	AP: 0.9519	-----
-----	Class: ice	AP: 0.9862	-----
-----	Class: snow	AP: 0.9902	-----
-----	Class: clouds	AP: 0.9932	-----
-----	Class: smoke	AP: 1.0000	-----
-----	Class: fire	AP: 0.9405	-----
-----	Class: natural light	AP: 0.9803	-----
-----	Class: direct sun/sunny	AP: 0.9885	-----
-----	Class: electric/indoor lighting	AP: 0.9675	-----
-----	Class: aged/ worn	AP: 0.9137	-----
-----	Class: glossy	AP: 0.9318	-----
-----	Class: matte	AP: 0.8767	-----
-----	Class: sterile	AP: 0.9633	-----
-----	Class: moist/ damp	AP: 0.9316	-----
-----	Class: dry	AP: 0.9593	-----
-----	Class: dirty	AP: 0.9202	-----
-----	Class: rusty	AP: 0.8597	-----
-----	Class: warm	AP: 0.9403	-----
-----	Class: cold	AP: 0.9589	-----
-----	Class: natural	AP: 0.9683	-----
-----	Class: man-made	AP: 0.9794	-----
-----	Class: open area	AP: 0.9927	-----
-----	Class: semi-enclosed area	AP: 0.9623	-----
-----	Class: enclosed area	AP: 0.9850	-----
-----	Class: far-away horizon	AP: 0.9877	-----
-----	Class: no horizon	AP: 0.9763	-----
-----	Class: rugged scene	AP: 0.9867	-----
-----	Class: mostly vertical components	AP: 0.9428	-----
-----	Class: mostly horizontal components	AP: 0.9073	-----
-----	Class: symmetrical	AP: 0.9367	-----
-----	Class: cluttered space	AP: 0.9627	-----
-----	Class: scary	AP: 0.9591	-----
-----	Class: soothing	AP: 0.9050	-----
-----	Class: stressful	AP: 0.8950	-----

mAP: 0.9479
Avg loss: 0.0339951959663424
Test mAP: 0.9478817674151206

2.4 Attribute classifier on a sample image

```
[ ]: def format_image(x):
    scale = np.random.rand() * 2 + 0.25
    w = int(x.size[0] * scale)
    h = int(x.size[1] * scale)
    if min(w, h) < 227:
        scale = 227 / min(w, h)
        w = int(x.size[0] * scale)
        h = int(x.size[1] * scale)
    x = test_transform(x)
    x = x.view(1, 3, 227, 227)
    return x

def get_top_10_attributes(x, classifier):
    classifier.eval()
    classifier = classifier.cpu()
    logits = classifier(x)
    sorted_indices = torch.argsort(logits, descending=True)
    top_10_indices = sorted_indices.numpy()[0][:10]
    return [SUN_CLASSES[i] for i in top_10_indices]
```

```
[ ]: load_network_path = './sun_attribute_classifier.pth'

# use to load a previously trained network
if load_network_path is not None:
    print('Loading saved network from {}'.format(load_network_path))
    classifier = torchvision.models.densenet201()
    classifier.classifier = nn.Linear(1920, 102)
    classifier.load_state_dict(torch.load(load_network_path, map_location=torch.
    ↪device('cpu'))))
```

Loading saved network from ./sun_attribute_classifier.pth

```
[ ]: img = Image.open('./university.jfif').convert('RGB')
img_formatted = format_image(img)
top_10_attributes = get_top_10_attributes(img_formatted, classifier)

draw = ImageDraw.Draw(img)
plt.figure(figsize = (10,10))
plt.imshow(img)
plt.title("Predicted attributes for the image below are: \n" + ", ".
    ↪join(top_10_attributes), pad = 5, fontdict={'fontsize': 14});
```

Predicted attributes for the image below are:
 natural light, open area, grass, man-made, clouds, competing, sports, exercise, direct sun/sunny, vegetation



2.5 Predicting the scene of an input image using the pre-trained scene classifier and outputting a heat map for an input image

```
[ ]: from torch.autograd import Variable as V
import torchvision.models as models
from torchvision import transforms as trn
from torch.nn import functional as F
import cv2
from run_all_classifiers import load_labels, load_model, returnTF,
    ↪load_attributes_classifier, returnCAM, features_blobs
```

```
[ ]: # load the test image
img_name = 'university.jfif'
img = Image.open(img_name)
draw = ImageDraw.Draw(img)
plt.figure(figsize = (10,10))
plt.imshow(img)
plt.title("Input image", pad = 5, fontdict={'fontsize': 14});
input_img = V(tf(img).unsqueeze(0))

# load the labels
classes, labels_IO, labels_attribute = load_labels()
# load the model
model = load_model()
# load the transformer
tf = returnTF() # image transformer
```

```

# get the softmax weight
params = list(model.parameters())
weight_softmax = params[-2].data.numpy()
weight_softmax[weight_softmax<0] = 0

# forward pass
logit = model.forward(input_img)
h_x = F.softmax(logit, 1).data.squeeze()
probs, idx = h_x.sort(0, True)
probs = probs.numpy()
idx = idx.numpy()

# output the IO prediction
io_image = np.mean(labels_IO[idx[:10]]) # vote for the indoor or outdoor
if io_image < 0.5:
    print('--TYPE OF ENVIRONMENT: indoor')
else:
    print('--TYPE OF ENVIRONMENT: outdoor')

# output the prediction of scene category
print('--SCENE CATEGORIES:')
for i in range(0, 5):
    print('{:.3f} -> {}'.format(probs[i], classes[idx[i]]))

# generate class activation mapping
CAMs = returnCAM(features_blobs[0], weight_softmax, [idx[0]])
# render the CAM and output
img = cv2.imread(img_name)
height, width, _ = img.shape
heatmap = cv2.applyColorMap(cv2.resize(CAMs[0],(width, height)), cv2.
    ↳COLORMAP_JET)
result = heatmap * 0.4 + img * 0.5
cv2.imwrite('cam.jpg', result)

cam_name = 'cam.jpg'
cam_img = Image.open(cam_name)
draw = ImageDraw.Draw(cam_img)
plt.figure(figsize = (10,10))
plt.imshow(cam_img)
plt.title("Class activation map image", pad = 5, fontdict={'fontsize': 14});

```

```

--TYPE OF ENVIRONMENT: outdoor
--SCENE CATEGORIES:
0.700 -> campus
0.062 -> courthouse
0.052 -> embassy
0.030 -> golf_course
0.022 -> lawn

```

