## resnet50-classifyleaves (1)

## March 3, 2024

```
[1]: %matplotlib inline
   import matplotlib.pyplot as plt
   import torch
   from torchvision import transforms
   from torch.utils.data import Dataset, DataLoader
   from PIL import Image
   from torchvision import models
   from torch import nn
   import torch.utils.data as data
   import torchvision.transforms as transforms
   import os
   import csv
   from sklearn.preprocessing import LabelEncoder
   import time
```

```
[2]: #
     os.chdir("/kaggle/input/classify-leaves")
     encoder = LabelEncoder()
     class MyDataset(data.Dataset):
         def __init__(self, csv_path, is_test=False):
             self.is_test = is_test
             self.filenames = []
             self.labels = []
             current_directory = os.getcwd()
             with open(csv_path, 'r') as f:
                 reader = csv.reader(f)
                 for row in reader:
                      if row[0] == 'image':
                          continue
                      elif is_test:
                          self.filenames.append(os.path.join(current_directory,__
      \rightarrowrow[0]))
                      else:
                          self.labels.append(row[1])
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```
self.filenames.append(os.path.join(current_directory,_
 →row[0]))
            print(len(set(self.filenames)))
            # LabelEncoder
            global encoder
            if not is_test:
                labels_encoded = encoder.fit(self.labels)
                LabelEncoder()
                labels_encoded = encoder.transform(self.labels)
                # numpy
                self.labels = torch.tensor(labels_encoded)
   def __getitem__(self, index ):
        image = Image.open(self.filenames[index])
        data = self.preprocess(image)
        if self.is_test:
            return data
        else:
            label = self.labels[index]
            return data ,label
   def __len__(self):
        return len(self.filenames)
   def preprocess(self, data):
        transform_train_list = [
            # transforms.Resize((self.opt.h, self.opt.w), interpolation=3),
            # transforms.Pad(self.opt.pad, padding_mode='edge'),
            # transforms.RandomCrop((self.opt.h, self.opt.w)),
            # transforms.RandomHorizontalFlip(),
            transforms.ToTensor(),
            # transforms.Normalize([0.485, 0.456, 0.406], [0.229, 0.224, 0.225])
       return transforms.Compose(transform_train_list)(data)
def togpu(x):
   device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
   return x.to(device)
def load_model(net, path):
   device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
   net.load_state_dict(torch.load(path, map_location=device))
   return net
def save_model(net, path):
```

```
torch.save(net.state_dict(), path)
         print("saved:", path)
[3]: batch_size = 32
     lr, num_epochs = 0.005, 14
     train_dataset = MyDataset("train.csv")
     train_loader = torch.utils.data.DataLoader(
         train_dataset, batch_size=batch_size, shuffle=True, num_workers=4)
     test dataset = MyDataset('test.csv', is test=True)
     test_loader = torch.utils.data.DataLoader(
         test_dataset, batch_size=1, shuffle=False, num_workers=4)
    18353
    8800
[4]: resnet50 = models.resnet50(pretrained=True)
     num_ftrs = resnet50.fc.in_features
     for param in resnet50.parameters():
         param.requires_grad = True
     resnet50.fc = nn.Sequential(nn.Linear(num_ftrs, 176),
                                 nn.LogSoftmax(dim=1))
    /opt/conda/lib/python3.10/site-packages/torchvision/models/_utils.py:208:
    UserWarning: The parameter 'pretrained' is deprecated since 0.13 and may be
    removed in the future, please use 'weights' instead.
      warnings.warn(
    /opt/conda/lib/python3.10/site-packages/torchvision/models/_utils.py:223:
    UserWarning: Arguments other than a weight enum or `None` for 'weights' are
    deprecated since 0.13 and may be removed in the future. The current behavior is
    equivalent to passing `weights=ResNet50_Weights.IMAGENET1K_V1`. You can also use
    `weights=ResNet50_Weights.DEFAULT` to get the most up-to-date weights.
      warnings.warn(msg)
    Downloading: "https://download.pytorch.org/models/resnet50-0676ba61.pth" to
    /root/.cache/torch/hub/checkpoints/resnet50-0676ba61.pth
    100%|
              | 97.8M/97.8M [00:00<00:00, 172MB/s]
[5]: def train(net): #
         train_loss = [] #
         acc = [] #
         x = list(range(0, num_epochs)) # epoch array
         plt.ion()
         if torch.cuda.device_count() > 1:
```

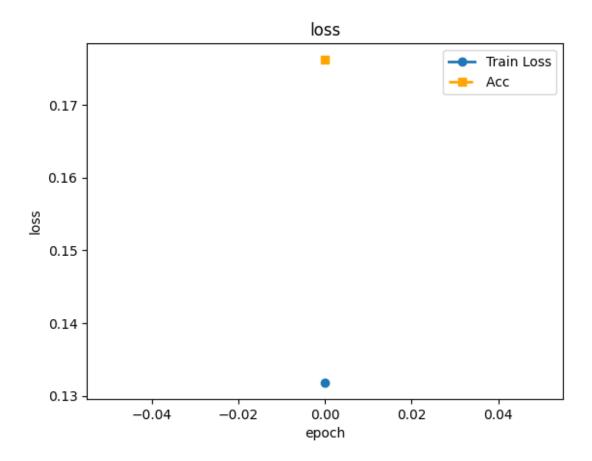
```
print("useing", torch.cuda.device_count(), "GPUs!")
      net = nn.DataParallel(net)
  device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
  net = net.to(device)
  print("training on", device) #
  optimizer = torch.optim.SGD(
      net.parameters(), lr=lr
  ) # SGD
  loss = nn.CrossEntropyLoss() #
  for epoch in range(num_epochs): #
      train_l_sum, train_acc_sum, n, start = (
          0.0,
          0,
          time.time(),
      for X, y in train_loader: #
          X, y = togpu(X), togpu(y) # GPU
          y_hat = net(X) #
          l = loss(y hat, y) #
          optimizer.zero_grad() #
          1.backward() #
          optimizer.step() #
          train_l_sum += l.cpu().item() #
          train_acc_sum += (y_hat.argmax(dim=1) ==
                            y).sum().cpu().item() #
          n += y.shape[0] #
      train_loss.append(train_l_sum / n) #
      acc.append(train_acc_sum / n) #
      ix = x[:epoch +1]
      train_iy = train_loss
      valid_iy = acc
      plt.cla()
      plt.title("loss")
      plt.plot(ix, train_iy, label='Train Loss', linewidth=2, linestyle='-', u
→marker='o')
      plt.plot(ix, valid_iy, label='Acc ', color="orange", linewidth=2,__
⇔linestyle='--', marker='s')
      plt.xlabel("epoch")
      plt.ylabel("loss")
```

```
plt.legend()
        plt.pause(0.5)
          save_model(resnet50 , "/kaggle/working/" + str(epoch + 1) +".pth")
        save_model(resnet50 , "/kaggle/working/re.pth")
        print(
            "epoch %d, loss %.4f, train acc %.3f, time %.1f sec"
            \% (epoch + 1, train_l_sum / n, train_acc_sum / n, time.time() -_{\sqcup}
 ⇔start)
        ) #
    plt.ioff()
    plt.show()
#
# pretrained_path = "/kaggle/working/103.pth"
# if os.path.exists(pretrained_path):
     net = load_model(resnet50, pretrained_path)
      print("Loaded pretrained model")
```

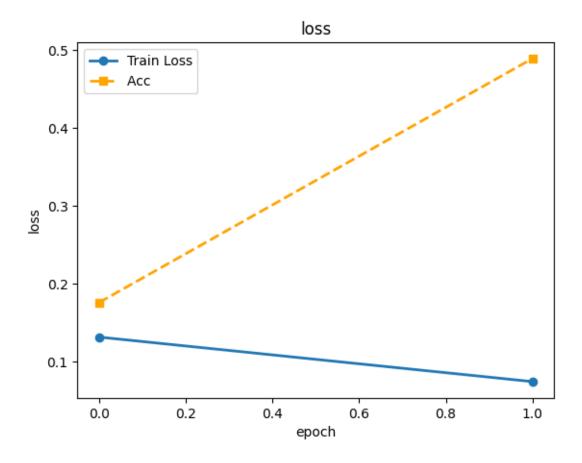
```
[6]: train(resnet50)

# save_model(resnet50 , "/kaggle/working/")
```

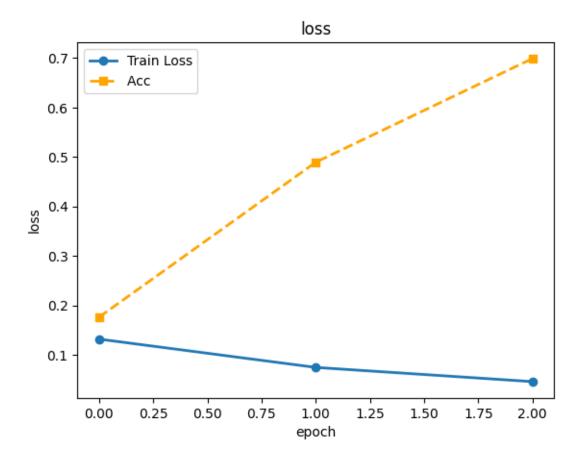
training on cuda



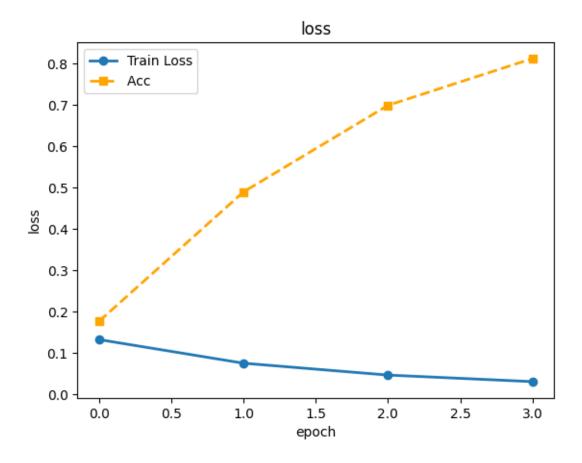
saved: /kaggle/working/re.pth
epoch 1, loss 0.1318, train acc 0.176, time 100.9 sec



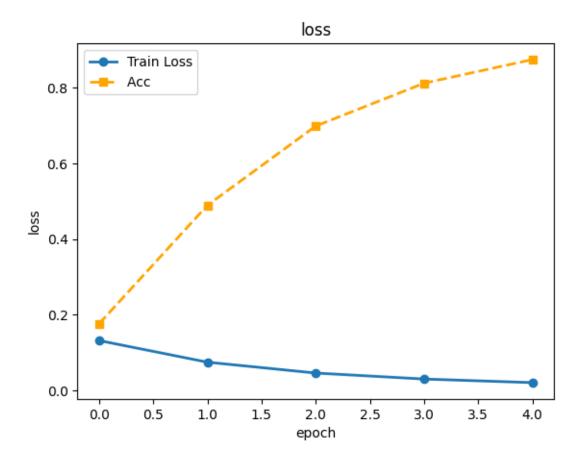
saved: /kaggle/working/re.pth
epoch 2, loss 0.0746, train acc 0.490, time 99.8 sec



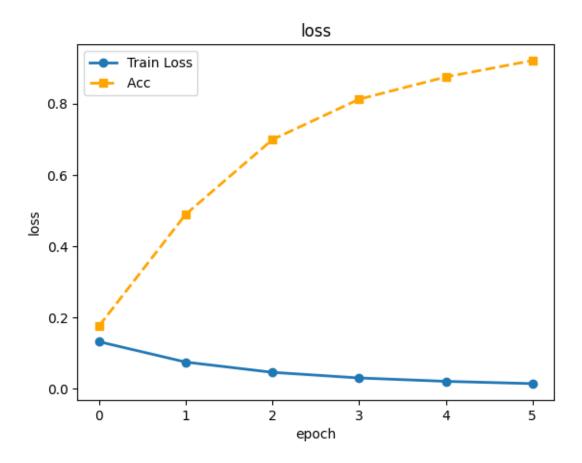
saved: /kaggle/working/re.pth
epoch 3, loss 0.0459, train acc 0.699, time 99.8 sec



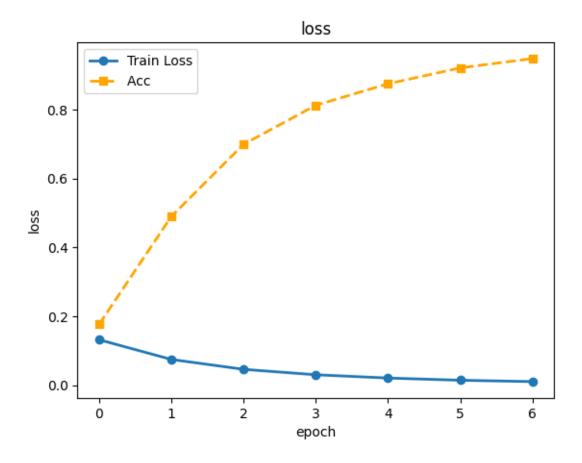
saved: /kaggle/working/re.pth
epoch 4, loss 0.0300, train acc 0.812, time 99.7 sec



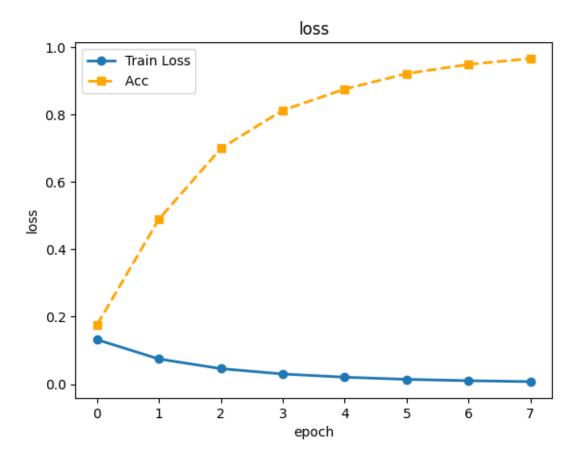
saved: /kaggle/working/re.pth
epoch 5, loss 0.0206, train acc 0.875, time 99.9 sec



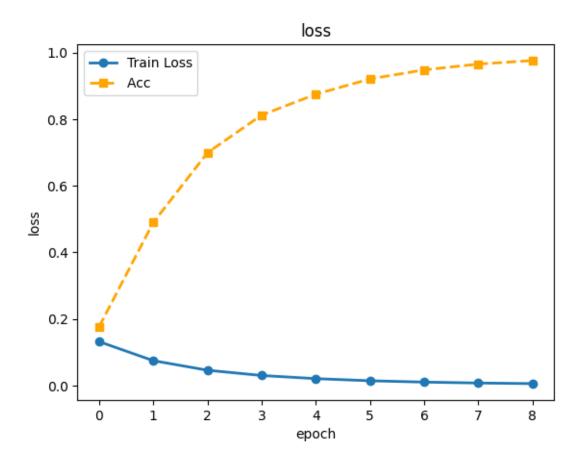
saved: /kaggle/working/re.pth
epoch 6, loss 0.0142, train acc 0.921, time 99.8 sec



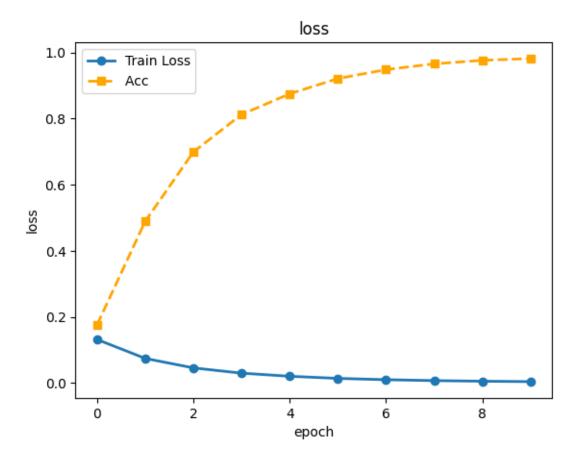
saved: /kaggle/working/re.pth
epoch 7, loss 0.0102, train acc 0.948, time 99.8 sec



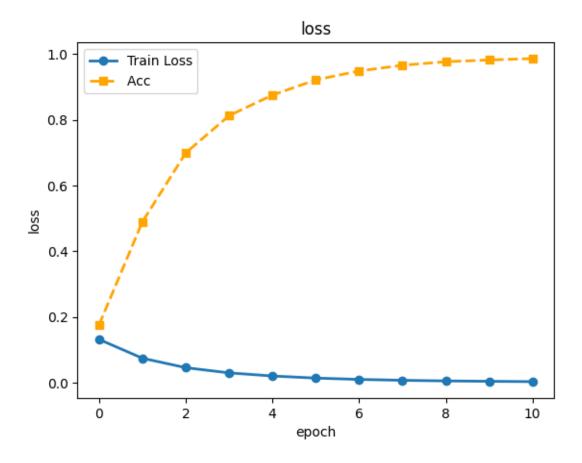
saved: /kaggle/working/re.pth
epoch 8, loss 0.0074, train acc 0.966, time 99.7 sec



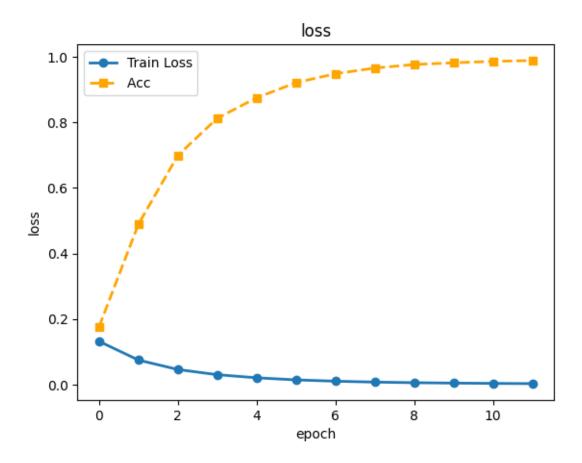
saved: /kaggle/working/re.pth
epoch 9, loss 0.0056, train acc 0.976, time 99.8 sec



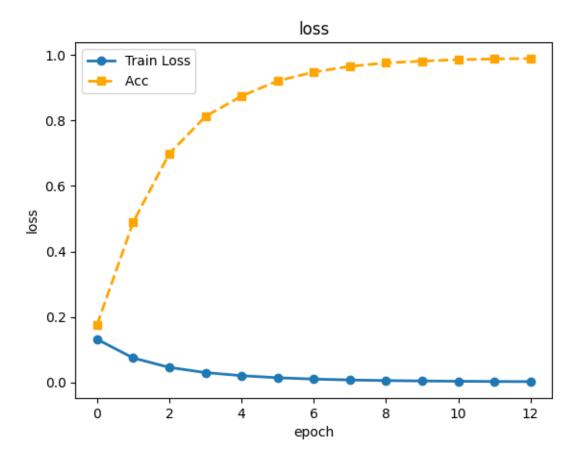
saved: /kaggle/working/re.pth
epoch 10, loss 0.0044, train acc 0.982, time 99.7 sec



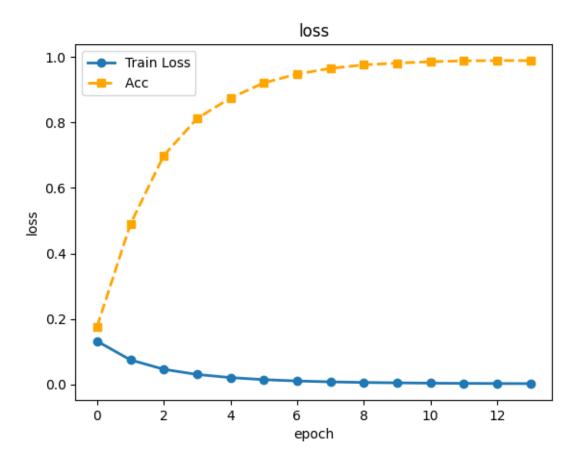
saved: /kaggle/working/re.pth
epoch 11, loss 0.0035, train acc 0.986, time 99.7 sec



saved: /kaggle/working/re.pth
epoch 12, loss 0.0028, train acc 0.988, time 99.7 sec



saved: /kaggle/working/re.pth
epoch 13, loss 0.0024, train acc 0.989, time 100.0 sec



saved: /kaggle/working/re.pth
epoch 14, loss 0.0021, train acc 0.989, time 99.8 sec

```
import pandas as pd
import numpy as np
test_labels = []
i = 0

for i in range(len(test_dataset)):
    test_data = test_dataset.__getitem__(i )
    test_data = test_data.unsqueeze(0)
    test_data = togpu(test_data)
    resnet50 = togpu(resnet50)
    resnet50.eval()
    output = resnet50(test_data)
    predictions = output.argmax(dim=1)
    test_labels.append(predictions.cpu().numpy())

# Convert the list of numpy arrays to a single numpy array
test_labels = np.concatenate(test_labels)
```

```
# Now you can call encoder.inverse_transform
test_labels = encoder.inverse_transform(test_labels)

test_df = pd.read_csv('test.csv')

test_df['label'] = test_labels

test_df.to_csv('/kaggle/working/submission.csv', index=False)
print("##########################")
```

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