In [1]: package_path = '../input/pytorch-image-models/pytorch-image-models-master' #'../input/efficientnet-pyto rch-07/efficientnet pytorch-0.7.0' import sys; sys.path.append(package path) In [2]: **from glob import** glob from sklearn.model selection import GroupKFold, StratifiedKFold import cv2 from skimage import io import torch from torch import nn import os from datetime import datetime import time import random import cv2 import torchvision from torchvision import transforms import pandas as pd import numpy as np from tqdm import tqdm import matplotlib.pyplot as plt from torch.utils.data import Dataset, DataLoader from torch.utils.data.sampler import SequentialSampler, RandomSampler from torch.cuda.amp import autocast, GradScaler import sklearn import warnings import joblib from sklearn.metrics import roc auc score, log loss from sklearn import metrics import warnings import cv2 import pydicom import timm #from efficientnet pytorch import EfficientNet from scipy.ndimage.interpolation import zoom from sklearn.metrics import log_loss In $[3]: | CFG = {$ 'fold num': 5, 'seed': 719, 'model arch': 'tf efficientnet b4 ns', 'img size': 512, 'epochs': 10, 'train bs': 32, 'valid bs': 32, 'lr': 1e-4, 'num workers': 4, 'accum iter': 1, # suppoprt to do batch accumulation for backprop with effectively larger batch siz 'verbose step': 1, 'device': 'cuda:0', 'tta': 3, 'used_epochs': [6,7,8,9], 'weights': [1,1,1,1] In [4]: | train = pd.read csv('../input/cassava-leaf-disease-classification/train.csv') train.head() Out[4]: image_id label **0** 1000015157.jpg 1 1000201771.jpg 100042118.jpg 3 1000723321.jpg 4 1000812911.jpg train.label.value counts() In [5]: Out[5]: 3 13158 2577 2 2386 2189 1087 Name: label, dtype: int64 We could do stratified validation split in each fold to make each fold's train and validation set looks like the whole train set in target distributions. In [6]: submission = pd.read csv('../input/cassava-leaf-disease-classification/sample submission.csv') submission.head() Out[6]: image_id label o 2216849948.jpg **Helper Functions** In [7]: def seed everything(seed): random.seed(seed) os.environ['PYTHONHASHSEED'] = str(seed) np.random.seed(seed) torch.manual seed(seed) torch.cuda.manual seed(seed) torch.backends.cudnn.deterministic = True torch.backends.cudnn.benchmark = True def get img(path): im bgr = cv2.imread(path) im_rgb = im_bgr[:, :, ::-1] #print(im rgb) return im rgb img = get img('../input/cassava-leaf-disease-classification/train images/1000015157.jpg') plt.imshow(img) plt.show() 300 400 500 **Dataset** In [8]: class CassavaDataset(Dataset): def init (self, df, data root, transforms=None, output label=True super().__init__() self.df = df.reset_index(drop=True).copy() self.transforms = transforms self.data root = data root self.output label = output label def len (self): return self.df.shape[0] def getitem (self, index: int): # get labels if self.output label: target = self.df.iloc[index]['label'] path = "{}/{}".format(self.data root, self.df.iloc[index]['image id']) img = get_img(path) if self.transforms: img = self.transforms(image=img)['image'] # do label smoothing if self.output label == True: return img, target else: return img **Define Train\Validation Image Augmentations** In [9]: from albumentations import (HorizontalFlip, VerticalFlip, IAAPerspective, ShiftScaleRotate, CLAHE, RandomRotate90, Transpose, ShiftScaleRotate, Blur, OpticalDistortion, GridDistortion, HueSaturationValue, IAAAdditiveGaussianNoise, GaussNoise, MotionBlur, MedianBlur, IAAPiecewiseAffine, RandomResizedCrop IAASharpen, IAAEmboss, RandomBrightnessContrast, Flip, OneOf, Compose, Normalize, Cutout, CoarseDro pout, ShiftScaleRotate, CenterCrop, Resize from albumentations.pytorch import ToTensorV2 from albumentations import (HorizontalFlip, VerticalFlip, IAAPerspective, ShiftScaleRotate, CLAHE, RandomRotate90, Transpose, ShiftScaleRotate, Blur, OpticalDistortion, GridDistortion, HueSaturationValue, IAAAdditiveGaussianNoise, GaussNoise, MotionBlur, MedianBlur, IAAPiecewiseAffine, RandomResizedCrop IAASharpen, IAAEmboss, RandomBrightnessContrast, Flip, OneOf, Compose, Normalize, Cutout, CoarseDro pout, ShiftScaleRotate, CenterCrop, Resize from albumentations.pytorch import ToTensorV2 def get train transforms(): return Compose([RandomResizedCrop(CFG['img_size'], CFG['img_size']), Transpose (p=0.5), HorizontalFlip (p=0.5), VerticalFlip(p=0.5), ShiftScaleRotate (p=0.5), HueSaturationValue(hue_shift_limit=0.2, sat_shift_limit=0.2, val_shift_limit=0.2, p=0.5), RandomBrightnessContrast(brightness limit=(-0.1, 0.1), contrast limit=(-0.1, 0.1), p=0.5), Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.225], max pixel value=255.0, p= 1.0), CoarseDropout (p=0.5), Cutout (p=0.5), ToTensorV2 (p=1.0),], p=1.) def get valid transforms(): return Compose([CenterCrop(CFG['img size'], CFG['img size'], p=1.), Resize(CFG['img size'], CFG['img size']), Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.225], max pixel value=255.0, p= 1.0), ToTensorV2 (p=1.0),], p=1.)def get_inference_transforms(): return Compose([RandomResizedCrop(CFG['img size'], CFG['img size']), Transpose (p=0.5), HorizontalFlip (p=0.5), VerticalFlip(p=0.5), HueSaturationValue(hue shift limit=0.2, sat shift limit=0.2, val shift limit=0.2, p=0.5), RandomBrightnessContrast(brightness limit=(-0.1, 0.1), contrast limit=(-0.1, 0.1), p=0.5), Normalize (mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.225], max pixel value=255.0, p= 1.0), ToTensorV2 (p=1.0),], p=1.)Model In [10]: class CassvaImgClassifier(nn.Module): def init (self, model arch, n class, pretrained=False): super(). init () self.model = timm.create model(model arch, pretrained=pretrained) n features = self.model.classifier.in features self.model.classifier = nn.Linear(n features, n class) def forward(self, x): x = self.model(x)return x **Main Loop** In [11]: def inference one epoch (model, data loader, device): model.eval() image preds all = [] pbar = tqdm(enumerate(data loader), total=len(data loader)) for step, (imgs) in pbar: imgs = imgs.to(device).float() image preds = model(imgs) #output = model(input) image preds all += [torch.softmax(image preds, 1).detach().cpu().numpy()] image preds all = np.concatenate(image preds all, axis=0) return image preds all In [12]: if name == ' main ': # for training only, need nightly build pytorch seed everything(CFG['seed']) folds = StratifiedKFold(n splits=CFG['fold num']).split(np.arange(train.shape[0]), train.label.valu es) for fold, (trn idx, val idx) in enumerate(folds): # we'll train fold 0 first **if** fold > 0: break print('Inference fold {} started'.format(fold)) valid_ = train.loc[val_idx,:].reset_index(drop=True) valid ds = CassavaDataset(valid , '../input/cassava-leaf-disease-classification/train images/', transforms=get inference_transforms(), output_label=False) test = pd.DataFrame() test['image id'] = list(os.listdir('../input/cassava-leaf-disease-classification/test images/')) test ds = CassavaDataset(test, '../input/cassava-leaf-disease-classification/test images/', tra nsforms=get inference transforms(), output label=False) val loader = torch.utils.data.DataLoader(valid ds, batch size=CFG['valid bs'], num workers=CFG['num workers'], shuffle=False, pin_memory=False, tst loader = torch.utils.data.DataLoader(test ds, batch size=CFG['valid bs'], num workers=CFG['num workers'], shuffle=False, pin_memory=False, device = torch.device(CFG['device']) model = CassvaImgClassifier(CFG['model_arch'], train.label.nunique()).to(device) val preds = [] tst preds = [] #for epoch in range(CFG['epochs']-3): for i, epoch in enumerate(CFG['used epochs']): model.load state dict(torch.load('../input/pytorch-efficientnet-baseline-train-amp-aug/{} f old_{}_{}'.format(CFG['model_arch'], fold, epoch))) with torch.no grad(): for in range(CFG['tta']): val_preds += [CFG['weights'][i]/sum(CFG['weights'])/CFG['tta']*inference_one_epoch(model, val loader, device)] tst preds += [CFG['weights'][i]/sum(CFG['weights'])/CFG['tta']*inference one epoch(model, tst loader, device)] val preds = np.mean(val preds, axis=0) tst preds = np.mean(tst preds, axis=0) print('fold {} validation loss = {:.5f}'.format(fold, log loss(valid .label.values, val preds)))) print('fold {} validation accuracy = {:.5f}'.format(fold, (valid .label.values==np.argmax(val p reds, axis=1)).mean())) del model torch.cuda.empty cache() Inference fold 0 started | 134/134 [01:41<00:00, 1.32it/s] 100%| | 1/1 [00:00<00:00, 1.55it/s] 100%| | 134/134 [01:36<00:00, 1.38it/s] 100%| | 1/1 [00:00<00:00, 4.78it/s] | 134/134 [01:37<00:00, 1.38it/s] 100%| 100%| 1/1 [00:00<00:00, 7.41it/s] 134/134 [01:35<00:00, 1.40it/s] 100%| 100%| 1/1 [00:00<00:00, 7.76it/s] 134/134 [01:33<00:00, 1.43it/s] 100%| 100%| 1/1 [00:00<00:00, 8.10it/s] 134/134 [01:31<00:00, 1.46it/s] 100%| 100%| 1/1 [00:00<00:00, 8.41it/s] 100%| 134/134 [01:32<00:00, 1.45it/s] 100%| 1/1 [00:00<00:00, 5.15it/s] 134/134 [01:37<00:00, 1.37it/s] 100%| 100%| 1/1 [00:00<00:00, 4.62it/s] 134/134 [01:36<00:00, 1.38it/s] 100%| 1/1 [00:00<00:00, 7.59it/s] 100%| 100%| 134/134 [01:35<00:00, 1.40it/s] 100%| 1/1 [00:00<00:00, 8.43it/s] 100%| 134/134 [01:36<00:00, 1.39it/s] 100%| 1/1 [00:00<00:00, 7.31it/s] 134/134 [01:39<00:00, 1.34it/s] 100%| | 1/1 [00:00<00:00, 7.79it/s] fold 0 validation loss = 0.21773fold 0 validation accuracy = 0.93224 In [13]: test['label'] = np.argmax(tst preds, axis=1) test.head() Out[13]: image_id label **0** 2216849948.jpg In [14]: test.to_csv('submission.csv', index=False) Train part is here: https://www.kaggle.com/khyeh0719/pytorch- efficientnet-baseline-train-amp-aug In []: