

In [ ]:

```
import torch
import os
from torchvision import datasets, transforms
from torch.utils.data import Dataset, DataLoader
import pandas as pd
from PIL import Image
```

In [ ]:

```
# 图像
# 指定文件夹路径
folder_path = '/data/xxx/data/GSoC/wikiart'

# 获取文件夹中所有文件的路径
files = os.listdir(folder_path)

# 筛选出所有图片文件路径
image_files = [os.path.join(folder_path, f) for f in files if f.endswith(('.jpg'))]
for img_path in image_files:
    print(img_path)
transform = transforms.Compose([
    transforms.Resize((224, 224)), # 调整图像大小
    transforms.ToTensor(), # 转换为 Tensor 格式
    transforms.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.225]) # 归一化
])
```

In [ ]:

```
# 图像
# dataset = datasets.ImageFolder(root='/data/xxx/data/GSoC/wikiart',
transform=transform)
# img, label = dataset[0] # 获取第一个样本的图像和标签
```

In [ ]:

```
# 表格
genre_train = pd.read_csv('/data/xxx/data/GSoC/label/genre_train.csv', header = None)
genre_train
```

Out[ ]:

	0	1
0	Northern_Renaissance/hieronymus-bosch_st-jacqu...	7
1	Post_Impressionism/vincent-van-gogh_ears-of-wh...	4
2	Symbolism/theodor-severin-kittelsen_kvitebj-rn...	3
3	Expressionism/martiros-saryan_mother-of-the-ar...	6

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**0**

**1**

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	0	1
4	Early_Renaissance/leonardo-da-vinci_study-for-...	8
...	...	...
45498	Post_Impressionism/pablo-picasso_marin-and-stu...	8
45499	Expressionism/ernst-ludwig-kirchner_windswept-...	4
45500	Color_Field_Painting/gene-davis_two-part-blue-...	0
45501	Abstract_Expressionism/frank-bowling_sacha-jas...	0
45502	Ukiyo_e/utagawa-kuniyoshi_the-station-kambara.jpg	2
45503 rows × 2 columns		

In [ ]:

# CSV

```
class CustomDataset(Dataset):
```

```
    def __init__(self, data_dir, label_dir, csv_file, transform=None):
        self.label_dir = label_dir
        self.data_dir = data_dir
        self.transform = transform
        self.csv_data = pd.read_csv(os.path.join(self.label_dir, csv_file),
header=None)
```

```
    def __len__(self):
        return len(self.csv_data)
```

```
    def __getitem__(self, idx):
        file_path = os.path.join(self.data_dir, self.csv_data.iloc[idx, 0])
        c_class = self.csv_data.iloc[idx, 1]
        image_path = file_path
        image = Image.open(image_path)
        preprocess = transforms.Compose([
            transforms.Resize(256),
            transforms.CenterCrop(224),
            transforms.ToTensor(),
            transforms.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224,
0.225])),
        ])
        img_tensor = preprocess(image)
        # Assuming self.transform is a function that processes the data
        if self.transform:
            data = self.transform(img_tensor, c_class)
        else:
            data = (img_tensor, c_class) # Return the file path and class

        return data
```

In [ ]:

```
from torch.utils.data import Subset
from torch.utils.data import random_split
# 假设 dataset 是你的数据集对象，例如一个 Dataset 类的实例
```

In [ ]:

```
data_dir = '/data/xxx/data/GSoC/wikiart'
label_dir = '/data/xxx/data/GSoC/label'
csv_file = 'genre_train.csv'
custom_dataset = CustomDataset(data_dir, label_dir, csv_file)

# Assuming you have a transform function defined
# transform = ...
dataset_size = len(custom_dataset)
train_size = int(0.8 * dataset_size) # 假设 80%的数据用于训练，20%用于验证
val_size = dataset_size - train_size

train_dataset, val_dataset = random_split(custom_dataset, [train_size, val_size])
data_loader = DataLoader(train_dataset, batch_size=32, shuffle=True)

# data_loader = DataLoader(val_dataset, batch_size=32, shuffle=True)
```

In [ ]:

```
custom_dataset[6][0].shape
```

Out [ ]:

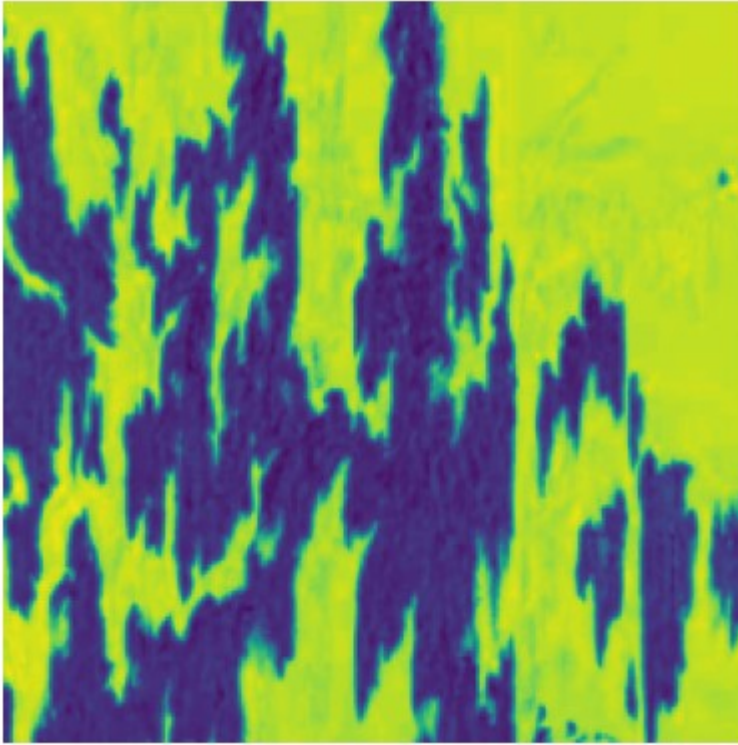
```
torch.Size([3, 224, 224])
```

In [ ]:

```
# 指定已知图片的位置
from matplotlib import pyplot as plt

img_np = custom_dataset[6][0].cpu().numpy() # 如果在 GPU 上，需要先将张量移回 CPU

# 显示图像
plt.imshow(img_np[2])
plt.axis('off') # 关闭坐标轴
plt.show()
```



In [ ]:

```
import timm
```

```
first_network = timm.create_model('resnet18', pretrained=False, num_classes=23)
```

In [ ]:

```
first_network
```

Out[ ]:

```
ResNet(
  (conv1): Conv2d(3, 64, kernel_size=(7, 7), stride=(2, 2), padding=(3, 3), bias=False)
  (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
  (act1): ReLU(inplace=True)
  (maxpool): MaxPool2d(kernel_size=3, stride=2, padding=1, dilation=1, ceil_mode=False)
  (layer1): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (drop_block): Identity()
      (act1): ReLU(inplace=True)
      (aa): Identity()
      (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (act2): ReLU(inplace=True)
    )
```

```

    (1): BasicBlock(
      (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (drop_block): Identity()
      (act1): ReLU(inplace=True)
      (aa): Identity()
      (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (act2): ReLU(inplace=True)
    )
  )
(layer2): Sequential(
  (0): BasicBlock(
    (conv1): Conv2d(64, 128, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1),
bias=False)
    (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (drop_block): Identity()
    (act1): ReLU(inplace=True)
    (aa): Identity()
    (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
    (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (act2): ReLU(inplace=True)
    (downsample): Sequential(
      (0): Conv2d(64, 128, kernel_size=(1, 1), stride=(2, 2), bias=False)
      (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    )
  )
  (1): BasicBlock(
    (conv1): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
    (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (drop_block): Identity()
    (act1): ReLU(inplace=True)
    (aa): Identity()
    (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
    (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (act2): ReLU(inplace=True)
  )
)
(layer3): Sequential(
  (0): BasicBlock(
    (conv1): Conv2d(128, 256, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1),
bias=False)

```

```

        (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
        (drop_block): Identity()
        (act1): ReLU(inplace=True)
        (aa): Identity()
        (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
        (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
        (act2): ReLU(inplace=True)
        (downsample): Sequential(
          (0): Conv2d(128, 256, kernel_size=(1, 1), stride=(2, 2), bias=False)
          (1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
        )
      )
    (1): BasicBlock(
      (conv1): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (drop_block): Identity()
      (act1): ReLU(inplace=True)
      (aa): Identity()
      (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (act2): ReLU(inplace=True)
    )
  )
  (layer4): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(256, 512, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1),
bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (drop_block): Identity()
      (act1): ReLU(inplace=True)
      (aa): Identity()
      (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (act2): ReLU(inplace=True)
      (downsample): Sequential(
        (0): Conv2d(256, 512, kernel_size=(1, 1), stride=(2, 2), bias=False)
        (1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      )
    )
    (1): BasicBlock(
      (conv1): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)

```

```

        (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
        (drop_block): Identity()
        (act1): ReLU(inplace=True)
        (aa): Identity()
        (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
        (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
        (act2): ReLU(inplace=True)
    )
)
(global_pool): SelectAdaptivePool2d (pool_type=avg, flatten=Flatten(start_dim=1,
end_dim=-1))
(fc): Linear(in_features=512, out_features=23, bias=True)
)

```

In [ ]:

```

import torch.nn as nn
import torch.optim as optim
from tqdm.notebook import tqdm
import numpy as np

```

In [ ]:

```

# train_data = ...
# train_labels = ...

# # 创建数据集实例
# train_dataset = CustomDataset(train_data, train_labels, transform=transform)

device = 'cuda:1'

# 定义损失函数和优化器
criterion = nn.CrossEntropyLoss()
optimizer = optim.SGD(first_network.parameters(), lr=0.001, momentum=0.9)

# # 创建数据加载器
# train_loader = DataLoader(train_dataset, batch_size=32, shuffle=True)
first_network = first_network.to(device)
# 训练模型
train_loader = data_loader
try:
    for epoch in range(5): # 假设进行 5 个 epoch 的训练
        running_loss = 0.0
        for inputs, labels in tqdm(train_loader):
            inputs = inputs.to(device)
            labels = labels.to(device)

            optimizer.zero_grad()

            outputs = first_network(inputs)
            loss = criterion(outputs, labels)

```



```

        loss.backward()
        optimizer.step()

        running_loss += loss.item()

        print(f"Epoch {epoch+1} loss: {running_loss/len(train_loader)}")
except Exception as e:
    print(f"Encountered an error in iteration {i}: {e}. Skipping this iteration.")

print('Finished Training')
# 模拟多个 epoch 的训练过程
num_epochs = 5
for epoch in range(num_epochs):
    # 模拟每个 epoch 中多个 batch 的训练过程
    num_batches = 100
    for batch in range(num_batches):
        # 打印当前的 loss 值
        print(f'Epoch [{epoch+1}/{num_epochs}], Batch [{batch+1}/{num_batches}], Loss:
{loss:.4f}')

0%|          | 0/1138 [00:00<?, ?it/s]

```