

1 Dataset Preparation

Organize your dataset in the following structure: dataset/ train/ class_1/ image1.jpg
image2.jpg class_2/ image3.jpg image4.jpg

2 Load the Dataset

Use torchvision.datasets.ImageFolder to load the training and testing datasets. Apply the following transformations: Resize the images to 28×28 . Convert the images to PyTorch tensors. Normalize the images with mean (0.5,0.5,0.5) and standard deviation (0.5,0.5,0.5).

3.Create DataLoaders Batch size: 4 Shuffle: True for training and False for testing.

4.Visualize a Batch

Write a function imshow to display a batch of images. Use torchvision.utils.make_grid to create a grid of images.

```
In [1]: import torch
import os
import torchvision
import torchvision.transforms as transforms
from torchvision import datasets, transforms
import matplotlib.pyplot as plt
import numpy as np
```

```
In [2]: # 1. Organize your dataset
# Assume the dataset is organized in 'dataset/train' and 'dataset/test' director

# 2. Load Dataset
# Define the transformations (ToTensor and Normalize)
# BEGIN SOLUTION
transform = transforms.Compose([
    transforms.ToTensor(),
    transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5)) # Normalize with
])
# END SOLUTION
```

```
In [4]: # Path to your dataset folder
dataset_dir = 'data/shapes'
```

```
In [6]: # Set the path for your dataset
train_dir = './data/shapes/'
#test_dir = './dataset/test'
```

```
In [8]: # Use ImageFolder to Load the dataset from the directory
#root=train_dir: Specifies the directory containing training images.
trainset = torchvision.datasets.ImageFolder(root=train_dir, transform=transform)
#testset = torchvision.datasets.ImageFolder(root=test_dir, transform=transform)
```

```

dataset/
├── train/
│   ├── class_0/  # Folder for class 0 images
│   │   ├── img_1.jpg
│   │   ├── img_2.jpg
│   ├── class_1/  # Folder for class 1 images
│   │   ├── img_3.jpg
│   │   ├── img_4.jpg
├── test/
│   ├── class_0/
│   └── class_1/

```

```

In [10]: # 3. Create Train and Test DataLoader
# BEGIN SOLUTION
batch_size = 4
trainloader = torch.utils.data.DataLoader(trainset, batch_size=batch_size, shuffle=True)
# END SOLUTION

```

```

In [12]: # 4. Display Sample Data
# Function to display an image
def imshow(img):
    img = img / 2 + 0.5 # Unnormalize the image (since we normalized it with mean)
    npimg = img.numpy()
    plt.imshow(np.transpose(npimg, (1, 2, 0))) # Transpose to (H, W, C) for display
    plt.show()

# is used to create an iterator from the trainloader,
# which is a PyTorch DataLoader.
dataiter = iter(trainloader) # Converts trainloader into an iterator.
images, labels = next(dataiter) # Retrieves the next batch of images and labels.

```

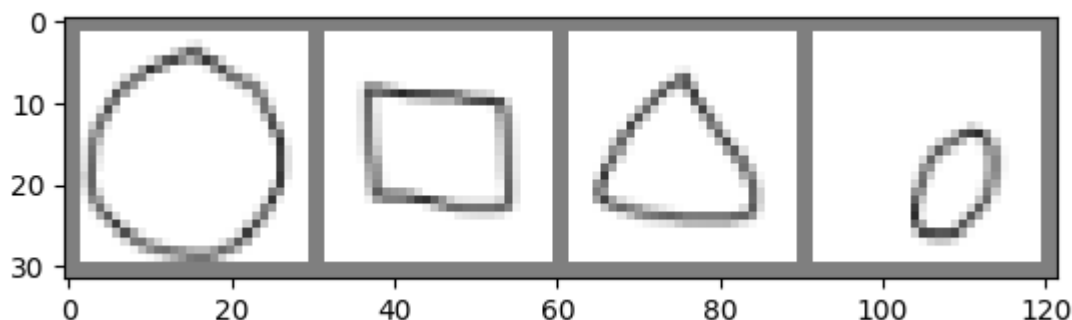
```

In [13]: # Assertions for data batch
assert isinstance(images, torch.Tensor), "Images should be a PyTorch tensor."
assert isinstance(labels, torch.Tensor), "Labels should be a PyTorch tensor."
assert images.ndim == 4, "Images should have 4 dimensions (batch_size, channels, height, width)."
assert labels.ndim == 1, "Labels should have 1 dimension (batch_size)."
assert images.shape[0] == labels.shape[0], "Number of images should match the number of labels."

# Display images with labels
print("Labels: ", labels) # Labels correspond to class indices
imshow(torchvision.utils.make_grid(images))

Labels:  tensor([0, 1, 2, 0])

```



```
In [14]: # Get a batch of training data
# Display images
print("Labels: ", labels)
#BEGIN SOLUTION
#arranges multiple images into a single image grid.
# imshow() displays the images after unnormalization.
imshow(torchvision.utils.make_grid(images))
#END SOLUTION
```

Labels: tensor([0, 1, 2, 0])



```
In [16]: # 5. Device Check (optional)
device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
print(f"Using device: {device}")
```

Using device: cuda

[NVSHARE][WARN]: Couldn't open file /var/run/secrets/kubernetes.io/serviceaccount/namespace to read Pod namespace
 [NVSHARE][INFO]: Successfully initialized nvshare GPU
 [NVSHARE][INFO]: Client ID = 461b12af2a865081

```
In [17]: # Print Dataset Statistics
print(f"Total images in the dataset: {len(trainset)}")
print(f"Number of classes: {len(trainset.classes)}")
print(f"Classes: {trainset.classes}")
```

Total images in the dataset: 301

Number of classes: 3

Classes: ['circles', 'squares', 'triangles']