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import torch
import torchvision
import matplotlib.pyplot as plt
import os
from PIL import Image
from torch.utils.data import Dataset, DataLoader
from torchvision import transforms
import torch.nn.functional as F
from torch import nn
import numpy as np
import math
# Define transformations
IMG SIZE = 64
data_transforms = transforms.Compose([
   transforms.Resize((IMG_SIZE, IMG_SIZE)),
   transforms.RandomHorizontalFlip(),
   transforms.ToTensor(),
   transforms.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.225])
])
# Custom dataset class
class CustomDataset(Dataset):
   def __init__(self, root_dir, transform=None):
       self.root_dir = root_dir
        self.transform = transform
        self.classes = os.listdir(root_dir)
        self.image_paths = []
        self.labels = []
        for label, cls in enumerate(self.classes):
            cls_path = os.path.join(root_dir, cls)
            for img in os.listdir(cls_path):
                if os.path.isfile(os.path.join(cls_path, img)):
                    self.image_paths.append(os.path.join(cls_path, img))
                    self.labels.append(label)
   def __len__(self):
        return len(self.image_paths)
   def __getitem__(self, idx):
        img_path = self.image_paths[idx]
       label = self.labels[idx]
        image = Image.open(img_path).convert("RGB")
        if self.transform:
            image = self.transform(image)
        return image, label
# Define your dataset path
root_dir = 'dataset_new_diff' # Replace with your actual dataset path
# Load the custom dataset
custom_dataset = CustomDataset(root_dir=root_dir, transform=data_transforms)
dataloader = DataLoader(custom_dataset, batch_size=32, shuffle=True)
# Function to show some images from the dataset
def show_images(dataset, num_samples=20, cols=4):
   plt.figure(figsize=(15,15))
    for i, (img, label) in enumerate(dataset):
        if i == num_samples:
       plt.subplot(int(num_samples/cols) + 1, cols, i + 1)
       plt.imshow(img.permute(1, 2, 0)) # Convert CHW to HWC
       plt.title(label)
# Show some images from the custom dataset
show_images(custom_dataset)
```



[#] from google.colab import drive

[#] drive.mount('/content/drive')

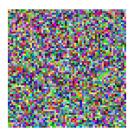
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# Define beta schedule
T = 1000
betas = torch.linspace(0.0001, 0.02, T)
alphas = 1. - betas
alphas_cumprod = torch.cumprod(alphas, axis=0)
alphas_cumprod_prev = F.pad(alphas_cumprod[:-1], (1, 0), value=1.0)
sqrt recip alphas = torch.sqrt(1.0 / alphas)
sqrt_alphas_cumprod = torch.sqrt(alphas_cumprod)
sqrt_one_minus_alphas_cumprod = torch.sqrt(1. - alphas_cumprod)
posterior variance = betas * (1. - alphas cumprod prev) / (1. - alphas cumprod)
def get_index_from_list(vals, t, x_shape):
       batch_size = t.shape[0]
       out = vals.gather(-1, t.cpu())
       return out.reshape(batch_size, *((1,) * (len(x_shape) - 1))).to(t.device)
def forward_diffusion_sample(x_0, t, device="cuda:1"):
       noise = torch.randn_like(x_0)
       sqrt_alphas_cumprod_t = get_index_from_list(sqrt_alphas_cumprod, t, x_0.shape)
       sqrt_one_minus_alphas_cumprod_t = get_index_from_list(sqrt_one_minus_alphas_cumprod, t, x_0.shape)
       return \ sqrt\_alphas\_cumprod\_t.to(device) \ * \ x\_0.to(device) \ + \ sqrt\_one\_minus\_alphas\_cumprod\_t.to(device) \ * \ noise.to(device), \ noise.to(device) \ + \ sqrt\_one\_minus\_alphas\_cumprod\_t.to(device) \ * \ noise.to(device), \ noise.to(device) \ + \ sqrt\_one\_minus\_alphas\_cumprod\_t.to(device) \ * \ noise.to(device), \ noise.to(device) \ + \ sqrt\_one\_minus\_alphas\_cumprod\_t.to(device) \ * \ noise.to(device), \ noise.to(device) \ + \ sqrt\_one\_minus\_alphas\_cumprod\_t.to(device) \ * \ noise.to(device) \ + \ sqrt\_one\_minus\_alphas\_cumprod\_t.to(device) \ + \ sqrt\_one\_minus\_alphas\_cumprod\_t.to(device) \ * \ noise.to(device) \ + \ sqrt\_one\_minus\_alphas\_cumprod\_t.to(device) \ + \ sqrt\_one\_minus\_alphas\_cumprod\_t.to(device) \ * \ noise.to(device) \ + \ sqrt\_one\_minus\_alphas\_cumprod\_t.to(device) \ * \ noise.to(device) \ + \ sqrt\_one\_minus\_alphas\_cumprod\_t.to(device) \ + \ sq
class Block(nn.Module):
       def __init__(self, in_ch, out_ch, time_emb_dim, up=False):
              super().__init__()
              self.time_mlp = nn.Linear(time_emb_dim, out_ch)
                      self.conv1 = nn.Conv2d(2*in_ch, out_ch, 3, padding=1)
                     self.transform = nn.ConvTranspose2d(out_ch, out_ch, 4, 2, 1)
              else:
                     self.conv1 = nn.Conv2d(in_ch, out_ch, 3, padding=1)
                     self.transform = nn.Conv2d(out_ch, out_ch, 4, 2, 1)
              self.conv2 = nn.Conv2d(out_ch, out_ch, 3, padding=1)
              self.bnorm1 = nn.BatchNorm2d(out_ch)
              self.bnorm2 = nn.BatchNorm2d(out_ch)
              self.relu = nn.ReLU()
       def forward(self, x, t):
              h = self.bnorm1(self.relu(self.conv1(x)))
              time emb = self.relu(self.time mlp(t))
              time\_emb = time\_emb[(..., ) + (None, ) * 2]
              h = h + time emb
              h = self.bnorm2(self.relu(self.conv2(h)))
              return self.transform(h)
class SinusoidalPositionEmbeddings(nn.Module):
       def __init__(self, dim):
             super().__init__()
              self.dim = dim
       def forward(self, time):
              device = time.device
              half dim = self.dim // 2
              embeddings = math.log(10000) / (half_dim - 1)
              embeddings = torch.exp(torch.arange(half_dim, device=device) * -embeddings)
              embeddings = time[:, None] * embeddings[None, :]
              embeddings = torch.cat((embeddings.sin(), embeddings.cos()), dim=-1)
              return embeddings
class SimpleUnet(nn.Module):
       def __init__(self):
              super().__init__()
              image channels = 3
              down_channels = (64, 128, 256, 512, 1024)
              up_channels = (1024, 512, 256, 128, 64)
              out dim = 3
              time_emb_dim = 32
              self.time_mlp = nn.Sequential(
                            {\tt SinusoidalPositionEmbeddings(time\_emb\_dim),}
                            nn.Linear(time_emb_dim, time_emb_dim),
                            nn.ReLU()
                     )
              self.conv0 = nn.Conv2d(image_channels, down_channels[0], 3, padding=1)
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self.downs = nn.ModuleList([Block(down_channels[i], down_channels[i+1], time_emb_dim) for i in range(len(down_channels)-1)])
        self.ups = nn.ModuleList([Block(up\_channels[i], up\_channels[i+1], time\_emb\_dim, up=True) \ for \ i \ in \ range(len(up\_channels)-1)])
        self.output = nn.Conv2d(up_channels[-1], out_dim, 1)
    def forward(self, x, timestep):
        t = self.time_mlp(timestep)
        x = self.conv0(x)
        residual_inputs = []
        for down in self.downs:
            x = down(x, t)
           residual_inputs.append(x)
        for up in self.ups:
           residual_x = residual_inputs.pop()
           x = torch.cat((x, residual_x), dim=1)
           x = up(x, t)
        return self.output(x)
device = 'cuda:1'
model = SimpleUnet().to(device)
print("Num params: ", sum(p.numel() for p in model.parameters()))
→ Num params: 62438883
def get_loss(model, x_0, t):
   x_noisy, noise = forward_diffusion_sample(x_0, t, device)
   noise_pred = model(x_noisy, t)
   return F.l1_loss(noise, noise_pred)
def show_tensor_image(image):
   reverse_transform = transforms.Compose([
        transforms.Normalize(mean=[-0.485 / 0.229, -0.456 / 0.224, -0.406 / 0.225],
                             std=[1 / 0.229, 1 / 0.224, 1 / 0.225]),
        transforms.ToPILImage()
   1)
   image = reverse_transform(image.squeeze())
   plt.imshow(image)
   plt.axis('off')
   plt.show()
# device = "cuda:1" if torch.cuda.is_available() else "cpu"
device = 'cuda:1'
@torch.no_grad()
def sample_timestep(x, t):
   betas_t = get_index_from_list(betas, t, x.shape)
   # print(betas t.device)
   sqrt_one_minus_alphas_cumprod_t = get_index_from_list(sqrt_one_minus_alphas_cumprod, t, x.shape)
   # print(sqrt_one_minus_alphas_cumprod_t.device)
   sqrt recip alphas t = get index from list(sqrt recip alphas, t, x.shape)
   # print(sqrt_recip_alphas_t.device)
   \verb|model_mean| = \verb|sqrt_recip_alphas_t| * (x - betas_t * model(x, t) / sqrt_one_minus_alphas_cumprod_t)|
   # print("hi", model mean.device)
   posterior_variance_t = get_index_from_list(posterior_variance, t, x.shape)
   # Generate noise
   noise = torch.randn_like(x)
   # Check for each element if it's at timestep 0
   is_t_0 = t == 0
   # Where t is 0, return model mean directly, else add noise
   result = torch.where(is_t_0.unsqueeze(-1).unsqueeze(-1), model_mean, model_mean + torch.sqrt(posterior_variance_t) * noise
   return result
@torch.no_grad()
def sample_plot_image(category, num_images=1):
   img_size = IMG_SIZE
    img = torch.randn((num_images, 3, img_size, img_size), device=device)
   plt.figure(figsize=(15,15))
   plt.axis('off')
   category_idx = custom_dataset.classes.index(category)
   t = torch.tensor([category_idx] * num_images, device=device)
   for i in range(num images):
        for timestep in reversed(range(T)):
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# print(torch.full((num_images,), timestep, device=device, dtype=torch.long).device)
    img = sample_timestep(img, torch.full((num_images,), timestep, device=device, dtype=torch.long))
    plt.subplot(1, num_images, i + 1)
    show_tensor_image(img[i])
    plt.show()

# Example usage to generate images of a specific category
sample_plot_image("mahesh_babu", num_images=5) # Replace "Prabha
```







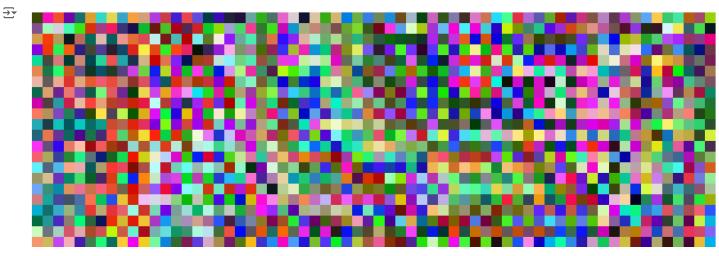


/home/student/2020/cs20btech11028/anaconda3/envs/diff/lib/python3.10/site-packages/torchvision/transforms/functional.py:282: RuntimeWarn npimg = (npimg * 255).astype(np.uint8)





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# Training loop
device = "cuda:1" if torch.cuda.is_available() else "cpu"
model.to(device)
optimizer = torch.optim.Adam(model.parameters(), lr=0.001)
epochs = 1000
for epoch in range(epochs):
    for i, (images, _) in enumerate(dataloader):
        optimizer.zero_grad()
        batch_size = images.shape[0]
        t = torch.randint(0, T, (batch_size,), device=device).long()
        loss = get_loss(model, images.to(device), t)
        loss.backward()
        optimizer.step()
        if i % 100 == 0:
            print(f"Epoch {epoch} Batch {i} Loss: {loss.item()}", end='\r')
            # sample_plot_image("mahesh_babu")
₹
sample_plot_image("mahesh_babu")
```



sample_plot_image("prabhas")

