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In [ ]: import os
        import torch
        import torch.nn as nn
        import torch.optim as optim
        from torch.utils.data import DataLoader, Dataset
        from torchvision import transforms, models
        from torchvision.io import read_image
        from sklearn.model_selection import train_test_split
        from sklearn.metrics import classification_report, confusion_matrix
        import pandas as pd
        import numpy as np
        # Set random seed
        torch.manual_seed(42)
        np.random.seed(42)
        # Step 1: Load Preprocessed Dataset
        base_dir = '/Users/Dell/Documents/TASK1/WikiArt'
        labels_df = pd.read_csv(os.path.join(base_dir, 'labels.csv'))
        print('First few rows of labels.csv:')
        print(labels_df.head())
        # Step 2: Define Dataset
        class WikiArtDataset(Dataset):
            def __init__(self, labels_df, transform=None):
                self.valid_data = [(row['image_path'], row['label'])
                                  for _, row in labels_df.iterrows()
                                  if os.path.exists(row['image_path'])]
                print(f'Number of valid images: {len(self.valid_data)}')
                if not self.valid data:
                    raise ValueError('No valid images found. Run prepare_dataset.py first and check paths.')
                self.transform = transform
            def __len__(self):
               return len(self.valid_data)
            def getitem (self, idx):
                img_path, label = self.valid_data[idx]
               image = read_image(img_path).float() / 255.0
               if self.transform:
                    image = self.transform(image)
                return image, label
        # Step 3: Define Model
        class ConvRecurrentModel(nn.Module):
           def __init__(self, num_classes):
               super(ConvRecurrentModel, self).__init__()
               self.cnn = models.resnet18(weights=models.ResNet18_Weights.DEFAULT)
               self.cnn.fc = nn.Identity()
               self.rnn = nn.GRU(512, 256, num_layers=2, batch_first=True)
               self.fc = nn.Linear(256, num_classes)
            def forward(self, x):
               batch size = x.size(0)
               x = self.cnn(x)
               x = x.unsqueeze(1) # [batch, 1, 512]
               _{n} h_n = self.rnn(x)
               x = self.fc(h_n[-1])
               return x
        # Step 4: Load and Split Data
        label_to_idx = {label: idx for idx, label in enumerate(labels_df['label'].unique())}
        labels_df['label'] = labels_df['label'].map(label_to_idx)
        train_df, test_df = train_test_split(labels_df, test_size=0.2, random_state=42)
        # Transformations
        transform = transforms.Compose([
            transforms.Resize((224, 224)),
            transforms.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.225])
        # Datasets and Loaders
        train_dataset = WikiArtDataset(train_df, transform)
        test_dataset = WikiArtDataset(test_df, transform)
        train_loader = DataLoader(train_dataset, batch_size=32, shuffle=True)
        test_loader = DataLoader(test_dataset, batch_size=32, shuffle=False)
        # Step 5: Model Setup
        device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')
        model = ConvRecurrentModel(num_classes=len(label_to_idx)).to(device)
        criterion = nn.CrossEntropyLoss()
        optimizer = optim.Adam(model.parameters(), lr=0.001)
        # Step 6: Training
        num_epochs = 5
        for epoch in range(num_epochs):
           model.train()
            running_loss = 0.0
            for images, labels in train_loader:
                images, labels = images.to(device), labels.to(device)
                optimizer.zero_grad()
               outputs = model(images)
               loss = criterion(outputs, labels)
               loss.backward()
               optimizer.step()
               running_loss += loss.item()
           print(f'Epoch [{epoch+1}/{num_epochs}], Loss: {running_loss/len(train_loader):.4f}')
        # Step 7: Evaluation
        model.eval()
        y_true, y_pred, y_probs = [], [], []
        with torch.no_grad():
            for images, labels in test_loader:
               images, labels = images.to(device), labels.to(device)
               outputs = model(images)
               probs = torch.softmax(outputs, dim=1)
                _, predicted = torch.max(outputs, 1)
               y_true.extend(labels.cpu().numpy())
               y_pred.extend(predicted.cpu().numpy())
               y_probs.extend(probs.cpu().numpy())
        # Metrics
        print('Classification Report:')
        print(classification_report(y_true, y_pred, target_names=[str(k) for k in label_to_idx.keys()]))
        print('Confusion Matrix:')
        print(confusion_matrix(y_true, y_pred))
        # Outlier Detection
        outliers = [(i, y_true[i], y_pred[i], max(y_probs[i]))
                    for i in range(len(y_true)) if y_true[i] == y_pred[i] and max(y_probs[i]) < 0.5]</pre>
        print(f'Potential Outliers (Correct but Low Confidence < 0.5): {len(outliers)}')</pre>
        for idx, true, pred, prob in outliers[:5]:
           print(f'Index: {idx}, True: {true}, Predicted: {pred}, Confidence: {prob:.4f}')
        # Save model
        torch.save(model.state_dict(), 'conv_recurrent_model.pth')
        print('Model saved as conv_recurrent_model.pth')
      First few rows of labels.csv:
                                                image_path label
       0 /Users/Dell/Documents/TASK1/WikiArt/wikiart/Re...
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1 /Users/Dell/Documents/TASK1/WikiArt/wikiart/Ba...
2 /Users/Dell/Documents/TASK1/WikiArt/wikiart/Po...
3 /Users/Dell/Documents/TASK1/WikiArt/wikiart/Im...
4 /Users/Dell/Documents/TASK1/WikiArt/wikiart/Ro...

Number of valid images: 10676 Number of valid images: 2670