

```
1 import torch
2 from torch import nn
3
4 import torchvision
5 from torchvision import datasets
6 from torchvision.transforms import ToTensor
7 import torch.nn.functional as F
8 import matplotlib.pyplot as plt
9 import numpy as np
10 import pandas as pd
11 import matplotlib.pyplot as plt
12 import zipfile
13 import os
14
15 import numpy as np
16 import pandas as pd
17 import torch.optim as optim
18 from torch.utils.data import Dataset
19 import time
20
21 class CustomDataset(Dataset):
22     def __init__(self, dataframe, transform=None):
23         self.dataframe = dataframe
24         self.transform = transform
25
26     def __len__(self):
27         return len(self.dataframe)
28
29     def __getitem__(self, idx):
30         img_values = self.dataframe.iloc[idx, 1:].
31         values
32         img = img_values.reshape(28, 28).astype('
33         float32')
34         label = int(self.dataframe.iloc[idx, 0])
35
36         if self.transform:
37             img = self.transform(img)
38
39         return img, label
```

```
40 train_data = pd.read_csv('sign_mnist_train.csv')
41 test_data = pd.read_csv('sign_mnist_test.csv')
42 print(train_data)
43
44 print(train_data.shape)
45
46 print(train_data.head())
47
48 column_names = train_data.columns
49
50 print(column_names)
51
52 # Access the first row using iloc (index location)
53 first_row = train_data.iloc[0]
54
55 # Extract the pixel values (excluding the label
   column)
56 img_values = first_row[1:].values
57
58 # Reshape the pixel values into a 28x28 image (
   assuming it's an image)
59 img = img_values.reshape(28, 28)
60
61 # Access the label (assuming the label is in the
   first column)
62 label = first_row[0]
63
64 # Display the image and label
65 print(img)
66
67 print(label)
68
69 print(img.shape)
70
71 print(len(train_data))
72
73 print(len(test_data))
74
75 print(type(train_data))
76
77 classes = train_data['label'].unique()
```

```
78
79 print(classes)
80
81 device = "cuda" if torch.cuda.is_available() else "
    cpu"
82
83 print(device)
84
85 plt.imshow(img.squeeze())
86
87 plt.show()
88
89 plt.imshow(img.squeeze(), cmap="gray")
90
91 plt.show()
92
93 import torch
94 import numpy as np
95 import matplotlib.pyplot as plt
96
97 class_names = train_data['label'].unique()
98 fig = plt.figure(figsize=(6, 8))
99 rows = 4
100 cols = 4
101
102 for i in range(1, rows * cols + 1):
103     random_index = torch.randint(0, len(train_data
    ), size=[1]).item()
104
105     # Ensure that the random_index is within the
    valid range
106     random_index = random_index % len(train_data)
107
108     img, target = train_data.iloc[random_index, 1
    :], train_data.iloc[random_index, 0]
109     target = target - 1
110     fig.add_subplot(rows, cols, i)
111     plt.imshow(img.values.reshape(28, 28), cmap="
    gray")
112     plt.title(class_names[target])
113     plt.axis(False)
```

```
114
115 plt.tight_layout()
116 plt.show()
117
118
119
120 def accuracy(y_true, y_pred):
121     correct_classification=torch.eq(y_true,y_pred).sum
122     ().item()
123     acc=(correct_classification/len(y_pred))*100
124     return acc
125
126 import torch
127 import torch.nn as nn
128 import torch.optim as optim
129 from torch.utils.data import DataLoader
130 from tqdm import tqdm
131
132
133 class Model(nn.Module):
134     def __init__(self, input_shape, hidden_units,
135                  output_shape):
136         super(Model, self).__init__()
137         self.input_shape = input_shape
138         self.layer_stack = nn.Sequential(
139             nn.Flatten(),
140             nn.Linear(in_features=input_shape,
141                      out_features=hidden_units),
142             nn.ReLU(),
143             nn.Linear(in_features=hidden_units,
144                      out_features=output_shape)
145         )
146
147     def forward(self, x):
148         return self.layer_stack(x)
149
150 print(train_data['label'].unique())
151
152 train_data['label'] = train_data['label'] - 1
153 # Instantiate the model
```

```
151 model = Model(input_shape=28 * 28, hidden_units=64,
    output_shape=len(classes))
152
153 optimizer = optim.Adam(model.parameters(), lr=0.01)
154
155
156 # Choose device
157 device = "cuda" if torch.cuda.is_available() else "
    cpu"
158
159 model.to(device)
160
161 # Loss function and optimizer
162 loss_function = nn.CrossEntropyLoss()
163 optimizer = optim.Adam(model.parameters(), lr=0.01)
164 transform = ToTensor()
165 train_dataset = CustomDataset(dataframe=train_data,
    transform=transform)
166 train_dataLoader = DataLoader(train_dataset,
    batch_size=32, shuffle=True)
167
168 # Training function
169 def train_step(model, data_loader, loss_fn,
    optimizer, accuracy, device):
170     train_loss, train_acc = 0, 0
171     model.train()
172
173     for batch, (X, y) in enumerate(data_loader):
174         X, y = X.to(device), y.to(device)
175
176         optimizer.zero_grad()
177         y_pred = model(X)
178
179         loss = loss_fn(y_pred, y)
180         train_loss += loss.item()
181         train_acc += accuracy(y, y_pred.argmax(dim=1
    ))
182
183         loss.backward()
184         optimizer.step()
185
```

```

186         if batch % 100 == 0:
187             print(f"Batch: {batch}/{len(data_loader)}
188             }, Loss: {loss.item():.4f}")
189
190
191         train_loss /= len(data_loader)
192         train_acc /= len(data_loader)
193
194         print(f"Train acc :{train_acc:.3f} | Train Loss:
195         {train_loss:.3f}\n")
196
197 # Testing function (similar to the training loop)
198 def test_step(model, data_loader, loss_fn, accuracy
199             , device):
200     test_loss, test_acc = 0, 0
201     model.eval()
202
203     with torch.no_grad():
204         for X, y in data_loader:
205             X, y = X.to(device), y.to(device)
206
207             y_pred = model(X)
208             loss = loss_fn(y_pred, y)
209             test_loss += loss.item()
210             test_acc += accuracy(y, y_pred.argmax(
211             dim=1))
212
213     test_loss /= len(data_loader)
214     test_acc /= len(data_loader)
215
216     print(f"Test acc :{test_acc:.3f} | Test Loss:{
217     test_loss:.3f}\n")
218
219 # Training loop
220 epochs = 5
221
222 for epoch in tqdm(range(epochs)):
223     print(f"Epoch : {epoch}\n")

```

```
222     train_step(model, train_dataLoader,
    loss_function, optimizer, accuracy, device)
223
224 from cvzone.ClassificationModule import Classifier
225
226 class ClassificationModule:
227     def __init__(self, model, data_loader, device):
228         self.model = model
229         self.data_loader = data_loader
230         self.device = device
231
232     def classify(self):
233         predictions = []
234         ground_truth = []
235
236         self.model.eval()
237         self.model.to(self.device)
238
239         with torch.no_grad():
240             for X, y in self.data_loader:
241                 X, y = X.to(self.device), y.to(self.
device)
242
243                 y_pred = self.model(X)
244                 predictions.extend(y_pred.argmax(dim
=1).tolist())
245                 ground_truth.extend(y.tolist())
246
247         return predictions, ground_truth
248
249 # Usage
250 classifier = ClassificationModule(model,
    test_dataLoader, device)
251 predicted, actual = classifier.classify()
252
253 prediction, index = classifier.classify()
254 print(prediction, index)
255
256 import torch
257 import torchvision
258 import torch.nn as nn
```

```
259 from torchvision import transforms
260 import cv2
261 import os
262
263
264
265 # Function to classify images
266 def classify_image(img_path):
267     img = cv2.imread(img_path, 0) # Read the image
268     in grayscale
269     img = cv2.resize(img, (28, 28)) # Resize the
270     image to match the MNIST dataset
271     img = transforms.ToTensor()(img).unsqueeze(0)
272     # Convert image to tensor
273     model.eval() # Set model to evaluation mode
274
275     with torch.no_grad():
276         output = model(img)
277         prediction = output.argmax(dim=1).item()
278     return prediction
279
280
281 import cv2
282 import time
283 from cvzone.HandTrackingModule import HandDetector
284 cap = cv2.VideoCapture(0)
285 detector = HandDetector(maxHands=1)
286 folder = "Data/A"
287 counter = 0
288 #labels = ["A", "B", "C"]
289 while True:
290     success, img = cap.read()
291     hands, img = detector.findHands(img)
292     if hands:
293         hand = hands[0]
294         x, y, w, h = hand['bbox']
295         img_crop = img[y:y + h, x:x + w]
296         if img_crop.shape[0] > 0 and img_crop.shape[
297             1] > 0:
298             cv2.imshow("ImageCrop", img_crop)
```



```
296         predicted_label = classify_image(img_crop
297     )
298     print(f"Predicted Label: {predicted_label
299         }")
300
301     cv2.imshow("Image", img)
302     key = cv2.waitKey(1)
303     if key == ord("q"):
304         counter += 1
305         cv2.imwrite(f'{folder}/Image_{time.time()}.
306         jpg', img_crop)
307         print(counter)
308
309
310
311 # Classify images in the folder
312
313
314 folder_path = 'C:\\Users\\91965\\PycharmProjects\\
315 Signlanguagettranslation\\Data\\A'
316
317
318 for filename in os.listdir(folder_path):
319     if filename.endswith('.jpg') or filename.
320     endsuffix('.png'):
321         image_path = os.path.join(folder_path,
322         filename)
323         predicted_label = classify_image(image_path)
324         print(f"Image: {filename}, Predicted Label:
325         {predicted_label}")
326 cap.release()
```