huynhnguyenthedan-is-lab03

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##

Bài tập Biến hình và Xử lý ảnh

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```
MSSV: 21110256
    Lớp: 21TTH1
    Lab03_IS_Introduction to Deep Learning
[1]: import torch
     import torch.nn as nn
     import torch.nn.functional as F
     import torch.optim as optim
     import torch.utils.data as data
     import torchvision.transforms as transforms
     import torchvision.datasets as datasets
     from sklearn import metrics
     from sklearn import decomposition
     from sklearn import manifold
     from tqdm.notebook import trange, tqdm
     import matplotlib.pyplot as plt
     import numpy as np
     import copy
     import random
     import time
[2]: SEED = 1234
     random.seed(SEED)
     np.random.seed(SEED)
     torch.manual_seed(SEED)
     torch.cuda.manual_seed(SEED)
     torch.backends.cudnn.deterministic = True
```

1 MNIST dataset

```
[4]: mean = train_data.data.float().mean() / 255
std = train_data.data.float().std() / 255
print(f'Calculated mean: {mean}')
print(f'Calculated std: {std}')
```

Calculated mean: 0.13066048920154572 Calculated std: 0.30810779333114624

1.1 Useful Functions

1.1 Oscial Falletions

```
[5]: def plot_images(images):
    n_images = len(images)
    rows = int(np.sqrt(n_images))
    cols = int(np.sqrt(n_images))

fig = plt.figure()
    for i in range(rows*cols):
        ax = fig.add_subplot(rows, cols, i+1)
        ax.imshow(images[i].view(28, 28).cpu().numpy(), cmap='bone')
        ax.axis('off')
```

1.1.1 Define model

```
[6]: class MLP(nn.Module):
    def __init__(self, input_dim, output_dim):
        super().__init__()

    self.input_fc = nn.Linear(input_dim, 250)
    self.hidden_fc = nn.Linear(250, 100)
    self.output_fc = nn.Linear(100, output_dim)

def forward(self, x):

# x = [batch size, height, width]

batch_size = x.shape[0]
```

```
x = x.view(batch_size, -1)
             \# x = [batch size, height * width]
             h_1 = F.relu(self.input_fc(x))
             \# h_1 = [batch size, 250]
             h_2 = F.relu(self.hidden_fc(h_1))
             \# h_2 = [batch \ size, \ 100]
             y_pred = self.output_fc(h_2)
             # y_pred = [batch size, output dim]
             return y_pred, h_2
[7]: def calculate_accuracy(y_pred, y):
         top_pred = y_pred.argmax(1, keepdim=True)
         correct = top_pred.eq(y.view_as(top_pred)).sum()
         acc = correct.float() / y.shape[0]
         return acc
[8]: def train(model, iterator, optimizer, criterion, device):
         epoch_loss = 0
         epoch_acc = 0
         model.train()
         for (x, y) in tqdm(iterator, desc="Training", leave=False):
             x = x.to(device)
             y = y.to(device)
             optimizer.zero_grad()
             y_pred, = model(x)
             loss = criterion(y_pred, y)
             acc = calculate_accuracy(y_pred, y)
             loss.backward()
```

```
optimizer.step()
              epoch_loss += loss.item()
              epoch_acc += acc.item()
          return epoch_loss / len(iterator), epoch_acc / len(iterator)
 [9]: def evaluate(model, iterator, criterion, device):
          epoch_loss = 0
          epoch_acc = 0
          model.eval()
          with torch.no_grad():
              for (x, y) in tqdm(iterator, desc="Evaluating", leave=False):
                  x = x.to(device)
                  y = y.to(device)
                  y_pred, _ = model(x)
                  loss = criterion(y_pred, y)
                  acc = calculate_accuracy(y_pred, y)
                  epoch_loss += loss.item()
                  epoch_acc += acc.item()
          return epoch_loss / len(iterator), epoch_acc / len(iterator)
[10]: def epoch_time(start_time, end_time):
          elapsed_time = end_time - start_time
          elapsed_mins = int(elapsed_time / 60)
          elapsed_secs = int(elapsed_time - (elapsed_mins * 60))
          return elapsed_mins, elapsed_secs
[11]: def process_data(add_transform, show_img = True, mean=mean, std=std):
        train_transforms = transforms.Compose([
                                  transforms.RandomRotation(5, fill=(0,)),
                                  transforms.RandomCrop(28, padding=2),
                                  *add_transform,
```

transforms.Normalize(mean=[mean], std=[std])

transforms.ToTensor(),

```
])
        train_data = datasets.MNIST(root=ROOT,
                                  train=True,
                                  download=True,
                                  transform=train_transforms)
        print(f'Number of training examples: {len(train_data)}')
        test_transforms = transforms.Compose([
                                 transforms.ToTensor(),
                                 transforms.Normalize(mean=[mean], std=[std])
        test_data = datasets.MNIST(root=ROOT,
                                 train=False,
                                 download=True,
                                 transform=test_transforms)
        print(f'Number of testing examples: {len(test_data)}')
        if show_img:
          N IMAGES = 25
          images = [image for image, label in [train_data[i] for i in_
       →range(N_IMAGES)]]
          plot_images(images)
        VALID_RATIO = 0.9
        n_train_examples = int(len(train_data) * VALID_RATIO)
        n_valid_examples = len(train_data) - n_train_examples
        train_data, valid_data = data.random_split(train_data,
                                                  [n_train_examples, n_valid_examples])
        valid_data = copy.deepcopy(valid_data)
        valid_data.dataset.transform = test_transforms
        return train_data, valid_data, test_data
[12]: def load_DataLoader(train_data, valid_data, test_data, BATCH_SIZE = 64):
        train_iterator = data.DataLoader(train_data,
                                       shuffle=True,
                                       batch_size=BATCH_SIZE)
        valid_iterator = data.DataLoader(valid_data,
```

batch_size=BATCH_SIZE)

```
[13]: def train_model(model, train_iterator, valid_iterator, EPOCHS = 10):
        device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')
        best_valid_loss = float('inf')
        optimizer = optim.Adam(model.parameters())
        criterion = nn.CrossEntropyLoss()
        model = model.to(device)
        criterion = criterion.to(device)
        for epoch in trange(EPOCHS):
          start_time = time.monotonic()
          train loss, train acc = train(model, train iterator, optimizer, criterion,
          valid_loss, valid_acc = evaluate(model, valid_iterator, criterion, device)
          if valid_loss < best_valid_loss:</pre>
              best_valid_loss = valid_loss
          end_time = time.monotonic()
          epoch_mins, epoch_secs = epoch_time(start_time, end_time)
          print(f'Epoch: {epoch+1:02} | Epoch Time: {epoch_mins}m {epoch_secs}s')
          print(f'\tTrain Loss: {train_loss:.3f} | Train Acc: {train_acc*100:.2f}%')
          print(f'\t Val. Loss: {valid_loss:.3f} | Val. Acc: {valid_acc*100:.2f}%')
```

```
[14]: def test_model(model, test_iterator):
    device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')
    criterion = nn.CrossEntropyLoss().to(device)
    test_loss, test_acc = evaluate(model, test_iterator, criterion, device)
    print(f'Test Loss: {test_loss:.3f} | Test Acc: {test_acc*100:.2f}%')
    return round(test_loss, 3), round(test_acc*100, 2)
```

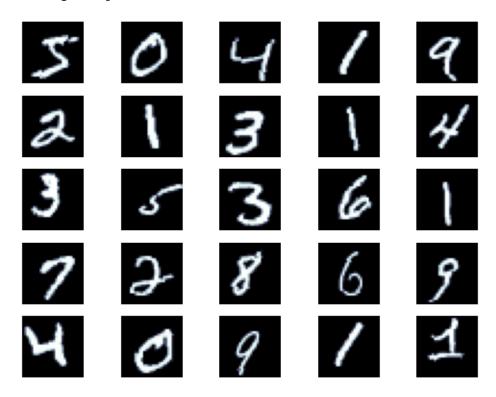
[]:

1.2 Train model với Batch size = 64, Activation: Relu

```
[15]: INPUT_DIM = 28 * 28
OUTPUT_DIM = 10
```

```
[16]: # Augmentation
add_transform = []
```

Number of training examples: 60000 Number of testing examples: 10000



[17]: # Train model

Number of training examples: 60000 Number of testing examples: 10000

0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/844 [00:00<?, ?it/s]

train_model(model, train_iterator, valid_iterator)

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: 0m 26s

Train Loss: 0.413 | Train Acc: 87.13% Val. Loss: 0.142 | Val. Acc: 95.67%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0% | 0/94 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 27s

Train Loss: 0.170 | Train Acc: 94.65% Val. Loss: 0.108 | Val. Acc: 96.46%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0% | 0/94 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 26s

Train Loss: 0.142 | Train Acc: 95.54% Val. Loss: 0.082 | Val. Acc: 97.58%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 26s

Train Loss: 0.119 | Train Acc: 96.20% Val. Loss: 0.082 | Val. Acc: 97.54%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: Om 26s

Train Loss: 0.108 | Train Acc: 96.59% Val. Loss: 0.080 | Val. Acc: 97.76%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 22s

Train Loss: 0.103 | Train Acc: 96.80% Val. Loss: 0.069 | Val. Acc: 98.01%

Training: 0% | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 23s

Train Loss: 0.092 | Train Acc: 97.05% Val. Loss: 0.060 | Val. Acc: 98.30%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 22s

Train Loss: 0.086 | Train Acc: 97.28% Val. Loss: 0.062 | Val. Acc: 98.22%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: 0m 22s

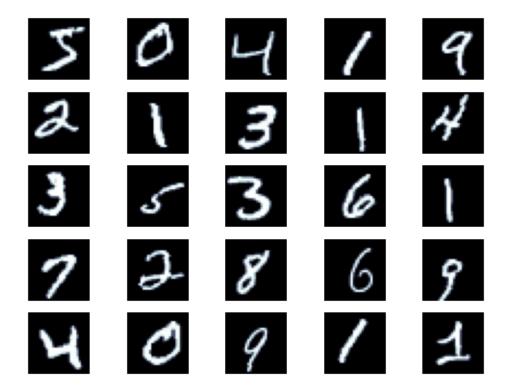
Train Loss: 0.083 | Train Acc: 97.41% Val. Loss: 0.061 | Val. Acc: 98.01%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: Om 22s

Train Loss: 0.081 | Train Acc: 97.44% Val. Loss: 0.057 | Val. Acc: 98.22%



[18]: test_loss, test_acc = test_model(model, test_iterator)

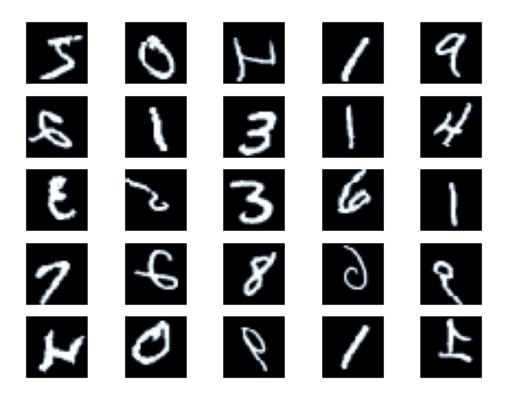
Evaluating: 0%| | 0/157 [00:00<?, ?it/s]

Test Loss: 0.056 | Test Acc: 98.17%

1.3 Thêm phần tạo đa dang data Augmentation với RandomHorizontalFlip

[19]: # Augmentation
add_transform = [transforms.RandomHorizontalFlip()]
train_data_Flip, valid_data_Flip, test_data_Flip = process_data(add_transform)

Number of training examples: 60000 Number of testing examples: 10000



0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/844 [00:00<?, ?it/s]
Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: 0m 24s

Train Loss: 0.594 | Train Acc: 80.58% Val. Loss: 0.269 | Val. Acc: 91.26%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 24s

Train Loss: 0.303 | Train Acc: 90.30% Val. Loss: 0.210 | Val. Acc: 93.15%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0% | 0/94 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 24s

Train Loss: 0.248 | Train Acc: 92.15% Val. Loss: 0.183 | Val. Acc: 94.04%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0% | 0/94 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 24s

Train Loss: 0.221 | Train Acc: 92.95% Val. Loss: 0.174 | Val. Acc: 94.41%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 24s

Train Loss: 0.202 | Train Acc: 93.46% Val. Loss: 0.167 | Val. Acc: 94.34%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 24s

Train Loss: 0.188 | Train Acc: 93.92% Val. Loss: 0.213 | Val. Acc: 92.90%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 24s

Train Loss: 0.180 | Train Acc: 94.22% Val. Loss: 0.149 | Val. Acc: 94.95%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 24s

Train Loss: 0.166 | Train Acc: 94.69% Val. Loss: 0.128 | Val. Acc: 96.09%

Training: 0% | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: 0m 24s

Train Loss: 0.163 | Train Acc: 94.85% Val. Loss: 0.144 | Val. Acc: 95.28%

Training: 0% | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: 0m 24s

Train Loss: 0.158 | Train Acc: 94.89% Val. Loss: 0.129 | Val. Acc: 95.81%

[21]: test_loss_Flip, test_acc_Flip = test_model(model, test_iterator_Flip)

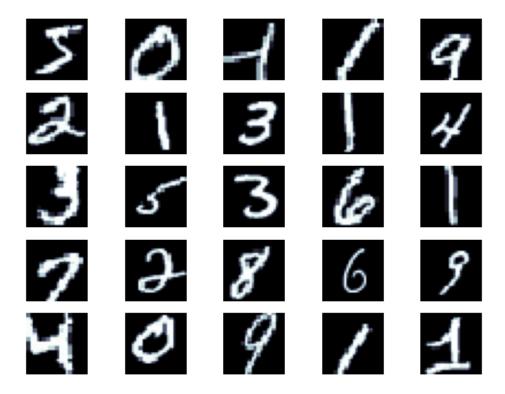
Evaluating: 0%| | 0/157 [00:00<?, ?it/s]

Test Loss: 0.114 | Test Acc: 96.24%

1.4 Thêm phần tạo đa dạng data Augmentation với RandomZoomOut

[22]: # Augmentation add_transform = [transforms.RandomAffine(degrees=0, scale=(1.0, 1.5))] train_data_ZoomOut, valid_data_ZoomOut, test_data_ZoomOut = →process_data(add_transform)

Number of training examples: 60000 Number of testing examples: 10000



train_model(model, train_iterator_ZoomOut, valid_iterator_ZoomOut)

0%1 | 0/10 [00:00<?, ?it/s] Training: 0%1 | 0/844 [00:00<?, ?it/s] 0%1 | 0/94 [00:00<?, ?it/s] Evaluating: Epoch: 01 | Epoch Time: Om 26s Train Loss: 0.433 | Train Acc: 86.59% Val. Loss: 0.197 | Val. Acc: 93.95% 0%| | 0/844 [00:00<?, ?it/s] Training: Evaluating: 0%1 | 0/94 [00:00<?, ?it/s] Epoch: 02 | Epoch Time: Om 26s Train Loss: 0.193 | Train Acc: 94.09% Val. Loss: 0.141 | Val. Acc: 95.89% 0%| | 0/844 [00:00<?, ?it/s] Training: Evaluating: 0%1 | 0/94 [00:00<?, ?it/s] Epoch: 03 | Epoch Time: 0m 25s Train Loss: 0.158 | Train Acc: 95.08% Val. Loss: 0.111 | Val. Acc: 96.35% 0%1 | 0/844 [00:00<?, ?it/s] Training: Evaluating: 0%1 | 0/94 [00:00<?, ?it/s] Epoch: 04 | Epoch Time: 0m 26s Train Loss: 0.139 | Train Acc: 95.70% Val. Loss: 0.111 | Val. Acc: 96.55% 0%1 | 0/844 [00:00<?, ?it/s] Training: Evaluating: 0%1 | 0/94 [00:00<?, ?it/s] Epoch: 05 | Epoch Time: 0m 26s Train Loss: 0.127 | Train Acc: 95.98% Val. Loss: 0.110 | Val. Acc: 96.38% 0%1 | 0/844 [00:00<?, ?it/s] Training: | 0/94 [00:00<?, ?it/s] Evaluating: 0%1 Epoch: 06 | Epoch Time: 0m 26s Train Loss: 0.120 | Train Acc: 96.21% Val. Loss: 0.096 | Val. Acc: 97.00% | 0/844 [00:00<?, ?it/s] Training: 0%| | 0/94 [00:00<?, ?it/s]

Evaluating:

0%1

Epoch: 07 | Epoch Time: 0m 25s

Train Loss: 0.110 | Train Acc: 96.62% Val. Loss: 0.085 | Val. Acc: 97.29%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 25s

Train Loss: 0.108 | Train Acc: 96.64% Val. Loss: 0.085 | Val. Acc: 97.46%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: 0m 26s

Train Loss: 0.102 | Train Acc: 96.86% Val. Loss: 0.072 | Val. Acc: 97.85%

Training: 0% | | 0/844 [00:00<?, ?it/s]

Evaluating: 0% | 0/94 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: Om 25s

Train Loss: 0.101 | Train Acc: 96.92% Val. Loss: 0.086 | Val. Acc: 97.26%

[24]: test_loss_ZoomOut, test_acc_ZoomOut = test_model(model, test_iterator_ZoomOut)

Evaluating: 0%| | 0/157 [00:00<?, ?it/s]

Test Loss: 0.083 | Test Acc: 97.36%

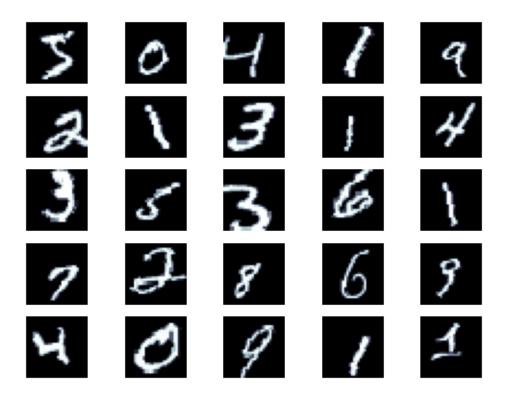
1.5 Thêm phần tạo đa dang data Augmentation với RandomAffine

[25]: # Augmentation
add_transform = [transforms.RandomAffine(degrees=0, translate=(0.1, 0.1),

→scale=(0.8, 1.2), shear=10)]
train_data_Affine, valid_data_Affine, test_data_Affine =

→process_data(add_transform)

Number of training examples: 60000 Number of testing examples: 10000



[26]: # Train model

0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/844 [00:00<?, ?it/s]
Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: 0m 26s

Train Loss: 0.625 | Train Acc: 80.05% Val. Loss: 0.214 | Val. Acc: 94.13%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 26s

Train Loss: 0.293 | Train Acc: 90.85% Val. Loss: 0.143 | Val. Acc: 95.72%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 26s

Train Loss: 0.235 | Train Acc: 92.55% Val. Loss: 0.118 | Val. Acc: 96.42%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0% | 0/94 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 27s

Train Loss: 0.206 | Train Acc: 93.50% Val. Loss: 0.102 | Val. Acc: 96.89%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 27s

Train Loss: 0.189 | Train Acc: 94.05% Val. Loss: 0.089 | Val. Acc: 97.20%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 26s

Train Loss: 0.177 | Train Acc: 94.42% Val. Loss: 0.087 | Val. Acc: 97.33%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 27s

Train Loss: 0.166 | Train Acc: 94.76% Val. Loss: 0.086 | Val. Acc: 97.33%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 26s

Train Loss: 0.158 | Train Acc: 95.01% Val. Loss: 0.074 | Val. Acc: 97.53%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: 0m 26s

Train Loss: 0.155 | Train Acc: 95.03% Val. Loss: 0.080 | Val. Acc: 97.46%

Training: 0% | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: Om 26s

Train Loss: 0.150 | Train Acc: 95.35% Val. Loss: 0.071 | Val. Acc: 97.92%

[27]: test_loss_Affine, test_acc_Affine = test_model(model, test_iterator_Affine)

Evaluating: 0% | 0/157 [00:00<?, ?it/s]

Test Loss: 0.062 | Test Acc: 97.85%

1.6 Thay đổi Batch Size = 32

[28]: # Train model train_iterator_32, valid_iterator_32, test_iterator_32 = 0 oload_DataLoader(train_data, valid_data, test_data, BATCH_SIZE = 32) model = MLP(INPUT_DIM, OUTPUT_DIM) train_model(model, train_iterator_32, valid_iterator_32)

0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/1688 [00:00<?, ?it/s]

Evaluating: 0%| | 0/188 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: 0m 27s

Train Loss: 0.357 | Train Acc: 88.79% Val. Loss: 0.135 | Val. Acc: 95.74%

Training: 0%| | 0/1688 [00:00<?, ?it/s]

Evaluating: 0% | 0/188 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 27s

Train Loss: 0.168 | Train Acc: 94.84% Val. Loss: 0.098 | Val. Acc: 97.17%

Training: 0%| | 0/1688 [00:00<?, ?it/s]

Evaluating: 0%| | 0/188 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 27s

Train Loss: 0.136 | Train Acc: 95.74% Val. Loss: 0.078 | Val. Acc: 97.47%

Training: 0%| | 0/1688 [00:00<?, ?it/s]

Evaluating: 0%| | 0/188 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 27s

Train Loss: 0.119 | Train Acc: 96.35% Val. Loss: 0.072 | Val. Acc: 97.74%

Training: 0%| | 0/1688 [00:00<?, ?it/s]

Evaluating: 0%| | 0/188 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 26s Train Loss: 0.109 | Train Acc: 96.67% Val. Loss: 0.066 | Val. Acc: 97.96% Training: 0%| | 0/1688 [00:00<?, ?it/s] Evaluating: 0%| | 0/188 [00:00<?, ?it/s] Epoch: 06 | Epoch Time: 0m 26s Train Loss: 0.102 | Train Acc: 96.78% Val. Loss: 0.068 | Val. Acc: 97.97% Training: 0%| | 0/1688 [00:00<?, ?it/s] | 0/188 [00:00<?, ?it/s] Evaluating: 0%1 Epoch: 07 | Epoch Time: 0m 26s Train Loss: 0.094 | Train Acc: 97.09% Val. Loss: 0.069 | Val. Acc: 97.67% 0%1 Training: | 0/1688 [00:00<?, ?it/s] 0%1 | 0/188 [00:00<?, ?it/s] Evaluating: Epoch: 08 | Epoch Time: 0m 26s Train Loss: 0.091 | Train Acc: 97.18% Val. Loss: 0.069 | Val. Acc: 97.87% 0%1 | 0/1688 [00:00<?, ?it/s] Training: 0%| | 0/188 [00:00<?, ?it/s] Evaluating: Epoch: 09 | Epoch Time: 0m 26s Train Loss: 0.089 | Train Acc: 97.25% Val. Loss: 0.062 | Val. Acc: 98.25% | 0/1688 [00:00<?, ?it/s] Training: 0%| Evaluating: 0%| | 0/188 [00:00<?, ?it/s] Epoch: 10 | Epoch Time: Om 26s Train Loss: 0.084 | Train Acc: 97.36% Val. Loss: 0.065 | Val. Acc: 98.11% [29]: test_loss_32, test_acc_32 = test_model(model, test_iterator_32) | 0/313 [00:00<?, ?it/s] Evaluating: 0%| Test Loss: 0.060 | Test Acc: 98.29% 1.7 Thay đổi Batch Size = 128[30]: # Train model

→load_DataLoader(train_data, valid_data, test_data, BATCH_SIZE = 128)

train_iterator_128, valid_iterator_128, test_iterator_128 =_

model = MLP(INPUT_DIM, OUTPUT_DIM) train_model(model, train_iterator_128, valid_iterator_128)

0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/422 [00:00<?, ?it/s]

Evaluating: 0% | 0/47 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: Om 19s

Train Loss: 0.486 | Train Acc: 85.02% Val. Loss: 0.184 | Val. Acc: 94.53%

Training: 0%| | 0/422 [00:00<?, ?it/s]

Evaluating: 0%| | 0/47 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 19s

Train Loss: 0.188 | Train Acc: 94.33% Val. Loss: 0.120 | Val. Acc: 96.30%

Training: 0% | | 0/422 [00:00<?, ?it/s]

Evaluating: 0%| | 0/47 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 20s

Train Loss: 0.143 | Train Acc: 95.62% Val. Loss: 0.089 | Val. Acc: 97.50%

Training: 0%| | 0/422 [00:00<?, ?it/s]

Evaluating: 0% | 0/47 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 27s

Train Loss: 0.124 | Train Acc: 96.12% Val. Loss: 0.083 | Val. Acc: 97.48%

Training: 0% | 0/422 [00:00<?, ?it/s]

Evaluating: 0%| | 0/47 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: Om 21s

Train Loss: 0.110 | Train Acc: 96.55% Val. Loss: 0.074 | Val. Acc: 97.63%

Training: 0%| | 0/422 [00:00<?, ?it/s]

Evaluating: 0%| | 0/47 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 20s

Train Loss: 0.104 | Train Acc: 96.79% Val. Loss: 0.068 | Val. Acc: 97.95%

Training: 0% | 0/422 [00:00<?, ?it/s]

Evaluating: 0%| | 0/47 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 20s

Train Loss: 0.093 | Train Acc: 97.09% Val. Loss: 0.061 | Val. Acc: 98.11%

Training: 0%| | 0/422 [00:00<?, ?it/s]

Evaluating: 0%| | 0/47 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 21s

Train Loss: 0.090 | Train Acc: 97.18% Val. Loss: 0.060 | Val. Acc: 98.23%

Training: 0%| | 0/422 [00:00<?, ?it/s]

Evaluating: 0%| | 0/47 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: 0m 21s

Train Loss: 0.083 | Train Acc: 97.45% Val. Loss: 0.061 | Val. Acc: 98.22%

Training: 0%| | 0/422 [00:00<?, ?it/s]

Evaluating: 0% | 0/47 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: Om 20s

Train Loss: 0.080 | Train Acc: 97.50% Val. Loss: 0.062 | Val. Acc: 98.05%

[31]: test_loss_128, test_acc_128 = test_model(model, test_iterator_128)

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Test Loss: 0.059 | Test Acc: 98.04%

1.8 Thay đổi Batch Size = 256

[32]: # Train model

train_iterator_256, valid_iterator_256, test_iterator_256 = \(\to \) \(\to

0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/211 [00:00<?, ?it/s]
Evaluating: 0%| | 0/24 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: Om 19s

Train Loss: 0.613 | Train Acc: 81.22% Val. Loss: 0.189 | Val. Acc: 94.34%

Training: 0% | 0/211 [00:00<?, ?it/s]
Evaluating: 0% | 0/24 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 19s

Train Loss: 0.212 | Train Acc: 93.66% Val. Loss: 0.126 | Val. Acc: 96.05%

Training: 0%| | 0/211 [00:00<?, ?it/s]

Evaluating: 0% | 0/24 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 19s

Train Loss: 0.160 | Train Acc: 95.21% Val. Loss: 0.112 | Val. Acc: 96.64%

Training: 0%| | 0/211 [00:00<?, ?it/s]

Evaluating: 0%| | 0/24 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 19s

Train Loss: 0.136 | Train Acc: 95.83% Val. Loss: 0.104 | Val. Acc: 96.80%

Training: 0% | 0/211 [00:00<?, ?it/s]

Evaluating: 0%| | 0/24 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: Om 19s

Train Loss: 0.121 | Train Acc: 96.24% Val. Loss: 0.079 | Val. Acc: 97.41%

Training: 0%| | 0/211 [00:00<?, ?it/s]

Evaluating: 0%| | 0/24 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 19s

Train Loss: 0.110 | Train Acc: 96.58% Val. Loss: 0.077 | Val. Acc: 97.70%

Training: 0%| | 0/211 [00:00<?, ?it/s]

Evaluating: 0%| | 0/24 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 19s

Train Loss: 0.098 | Train Acc: 96.89% Val. Loss: 0.077 | Val. Acc: 97.64%

Training: 0% | 0/211 [00:00<?, ?it/s]

Evaluating: 0%| | 0/24 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 19s

Train Loss: 0.094 | Train Acc: 97.07% Val. Loss: 0.059 | Val. Acc: 98.07%

Training: 0% | 0/211 [00:00<?, ?it/s]

Evaluating: 0%| | 0/24 [00:00<?, ?it/s]

```
Epoch: 09 | Epoch Time: 0m 19s
             Train Loss: 0.088 | Train Acc: 97.24%
              Val. Loss: 0.059 | Val. Acc: 98.01%
     Training:
                 0%|
                              | 0/211 [00:00<?, ?it/s]
                   0%|
                                | 0/24 [00:00<?, ?it/s]
     Evaluating:
     Epoch: 10 | Epoch Time: Om 19s
             Train Loss: 0.083 | Train Acc: 97.43%
              Val. Loss: 0.058 | Val. Acc: 98.29%
[33]: test_loss_256, test_acc_256 = test_model(model, test_iterator_256)
     Evaluating:
                   0%|
                                | 0/40 [00:00<?, ?it/s]
     Test Loss: 0.053 | Test Acc: 98.27%
     1.9 Đổi hành Activation: LeakyRelu
[34]: class MLP_LeakyRelu(nn.Module):
          def __init__(self, input_dim, output_dim):
              super().__init__()
              self.input_fc = nn.Linear(input_dim, 250)
              self.hidden_fc = nn.Linear(250, 100)
              self.output_fc = nn.Linear(100, output_dim)
          def forward(self, x):
              batch_size = x.shape[0]
              x = x.view(batch_size, -1)
              h_1 = F.leaky_relu(self.input_fc(x), negative_slope=0.2)
              h_2 = F.leaky_relu(self.hidden_fc(h_1), negative_slope=0.2)
              y_pred = self.output_fc(h_2)
              return y_pred, h_2
[35]: # Train model
      train_iterator_LeakyRelu, valid_iterator_LeakyRelu, test_iterator_LeakyRelu =_u
       ⇔load_DataLoader(train_data, valid_data, test_data)
      model_LeakyRelu = MLP_LeakyRelu(INPUT_DIM, OUTPUT_DIM)
      train model (model LeakyRelu, train iterator LeakyRelu, valid iterator LeakyRelu)
       0%1
                    | 0/10 [00:00<?, ?it/s]
     Training:
                 0%1
                              | 0/844 [00:00<?, ?it/s]
     Evaluating:
                                | 0/94 [00:00<?, ?it/s]
                   0%|
     Epoch: 01 | Epoch Time: 0m 23s
             Train Loss: 0.457 | Train Acc: 85.73%
```

Val. Loss: 0.172 | Val. Acc: 94.54%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 23s

Train Loss: 0.199 | Train Acc: 93.91% Val. Loss: 0.113 | Val. Acc: 96.88%

Training: 0% | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 23s

Train Loss: 0.165 | Train Acc: 94.98% Val. Loss: 0.106 | Val. Acc: 96.66%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0% | 0/94 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 23s

Train Loss: 0.147 | Train Acc: 95.53% Val. Loss: 0.083 | Val. Acc: 97.39%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 23s

Train Loss: 0.127 | Train Acc: 96.08% Val. Loss: 0.089 | Val. Acc: 97.33%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 22s

Train Loss: 0.126 | Train Acc: 96.14% Val. Loss: 0.076 | Val. Acc: 97.73%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 22s

Train Loss: 0.115 | Train Acc: 96.41% Val. Loss: 0.079 | Val. Acc: 97.64%

Training: 0% | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 22s

Train Loss: 0.110 | Train Acc: 96.58% Val. Loss: 0.091 | Val. Acc: 97.24%

Training: 0%| | 0/844 [00:00<?, ?it/s]

```
Evaluating:
                   0%1
                                | 0/94 [00:00<?, ?it/s]
     Epoch: 09 | Epoch Time: 0m 22s
             Train Loss: 0.105 | Train Acc: 96.77%
              Val. Loss: 0.082 | Val. Acc: 97.32%
     Training:
                 0%|
                              | 0/844 [00:00<?, ?it/s]
     Evaluating: 0%|
                                | 0/94 [00:00<?, ?it/s]
     Epoch: 10 | Epoch Time: Om 22s
             Train Loss: 0.102 | Train Acc: 96.88%
              Val. Loss: 0.078 | Val. Acc: 97.65%
[36]: test_loss_LeakyRelu, test_acc_LeakyRelu = test_model(model_LeakyRelu,_u
       →test_iterator_LeakyRelu)
                                | 0/157 [00:00<?, ?it/s]
     Evaluating:
                   0%|
     Test Loss: 0.070 | Test Acc: 97.68%
     1.10 Đổi hành Activation: ELU
[37]: class MLP_ELU(nn.Module):
          def __init__(self, input_dim, output_dim):
              super().__init__()
              self.input_fc = nn.Linear(input_dim, 250)
              self.hidden_fc = nn.Linear(250, 100)
              self.output_fc = nn.Linear(100, output_dim)
          def forward(self, x):
              batch_size = x.shape[0]
              x = x.view(batch size, -1)
              h 1 = F.elu(self.input fc(x), alpha=1.0)
              h_2 = F.elu(self.hidden_fc(h_1), alpha=1.0)
              y_pred = self.output_fc(h_2)
              return y_pred, h_2
[38]: # Train model
      train_iterator_ELU, valid_iterator_ELU, test_iterator_ELU =_
      →load_DataLoader(train_data, valid_data, test_data)
      model_ELU = MLP_ELU(INPUT_DIM, OUTPUT_DIM)
      train_model(model_ELU, train_iterator_ELU, valid_iterator_ELU)
       0%1
                    | 0/10 [00:00<?, ?it/s]
                 0%1
                              | 0/844 [00:00<?, ?it/s]
     Training:
                                | 0/94 [00:00<?, ?it/s]
                   0%1
     Evaluating:
```

Epoch: 01 | Epoch Time: 0m 22s

Train Loss: 0.422 | Train Acc: 86.70% Val. Loss: 0.133 | Val. Acc: 95.90%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0% | 0/94 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 22s

Train Loss: 0.165 | Train Acc: 94.82% Val. Loss: 0.114 | Val. Acc: 96.65%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 22s

Train Loss: 0.126 | Train Acc: 96.02% Val. Loss: 0.082 | Val. Acc: 97.50%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 22s

Train Loss: 0.112 | Train Acc: 96.47% Val. Loss: 0.081 | Val. Acc: 97.41%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 22s

Train Loss: 0.100 | Train Acc: 96.90% Val. Loss: 0.065 | Val. Acc: 98.08%

Training: 0%| | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 22s

Train Loss: 0.095 | Train Acc: 97.01% Val. Loss: 0.065 | Val. Acc: 97.95%

Training: 0% | 0/844 [00:00<?, ?it/s]

Evaluating: 0%| | 0/94 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 22s

Train Loss: 0.084 | Train Acc: 97.34% Val. Loss: 0.063 | Val. Acc: 97.96%

Training: 0% | 0/844 [00:00<?, ?it/s]

Evaluating: 0% | 0/94 [00:00<?, ?it/s]

```
Epoch: 08 | Epoch Time: 0m 23s
             Train Loss: 0.084 | Train Acc: 97.37%
              Val. Loss: 0.065 | Val. Acc: 97.98%
     Training:
                 0%|
                              | 0/844 [00:00<?, ?it/s]
     Evaluating:
                   0%|
                                 | 0/94 [00:00<?, ?it/s]
     Epoch: 09 | Epoch Time: 0m 23s
             Train Loss: 0.078 | Train Acc: 97.48%
              Val. Loss: 0.050 | Val. Acc: 98.30%
                              | 0/844 [00:00<?, ?it/s]
     Training:
                 0%|
     Evaluating:
                   0%1
                                 | 0/94 [00:00<?, ?it/s]
     Epoch: 10 | Epoch Time: Om 23s
             Train Loss: 0.074 | Train Acc: 97.70%
              Val. Loss: 0.053 | Val. Acc: 98.45%
[39]: test_loss_ELU, test_acc_ELU = test_model(model_ELU, test_iterator_ELU)
     Evaluating:
                   0%1
                                 | 0/157 [00:00<?, ?it/s]
     Test Loss: 0.059 | Test Acc: 97.99%
     1.11 Tạo bảng so sánh (MNIST dataset)
[40]: from tabulate import tabulate
      data_header = ['Model with', 'Test_loss', 'Test_acc']
      data_values = [
            ['Batch size = 64 + RELU', test_loss, test_acc],
            ['RandomHorizontalFlip', test_loss_Flip, test_acc_Flip],
            ['RandomZoomOut', test_loss_ZoomOut, test_acc_ZoomOut],
            ['RandomAffine', test_loss_Affine, test_acc_Affine],
            ['Batch size = 32', test_loss_32, test_acc_32],
            ['Batch size = 128', test_loss_128, test_acc_128],
            ['Batch size = 256', test_loss_256, test_acc_256],
            ['LeakyRelu', test loss LeakyRelu, test acc LeakyRelu],
            ['ELU', test_loss_ELU, test_acc_ELU]
      ]
[41]: # Generate the table
      result = tabulate(
      data_values,
      headers=data_header,
      tablefmt='fancy_grid',
      # Print the resulting table
```

print(result)

Model with	Test_loss	Test_acc
Batch size = 64 + RELU	0.056	98.17
RandomHorizontalFlip	0.114	96.24
RandomZoomOut	0.083	97.36
RandomAffine	0.062	97.85
Batch size = 32	0.06	98.29
Batch size = 128	0.059	98.04
Batch size = 256	0.053	98.27
LeakyRelu	0.07	97.68
ELU	0.059	97.99

2 CIFAR10 dataset

Downloading https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz to data\cifar-10-python.tar.gz

100% | 170498071/170498071 [00:20<00:00, 8247125.15it/s]

Extracting data\cifar-10-python.tar.gz to data

```
[43]: mean_cifar10 = train_data_cifar10.data.mean() / 255
std_cifar10 = train_data_cifar10.data.std() / 255
print(f'Calculated mean: {mean_cifar10}')
print(f'Calculated std: {std_cifar10}')
```

Calculated mean: 0.4733630004850899 Calculated std: 0.2515689250632208

```
[44]: def process_data_cifar10(add_transform, mean=mean_cifar10, std=std_cifar10):
        train_transforms = transforms.Compose([
                                  transforms.Grayscale(num_output_channels=1),
                                  transforms.RandomRotation(5, fill=(0,)),
                                  transforms.RandomCrop(28, padding=2),
                                  *add_transform,
                                  transforms.Resize((28, 28)),
                                  transforms.ToTensor(),
                                  transforms.Normalize(mean=[mean], std=[std])
                                            1)
        train_data = datasets.CIFAR10(root=ROOT,
                                  train=True,
                                  download=True,
                                  transform=train_transforms)
        print(f'Number of training examples: {len(train_data)}')
        test_transforms = transforms.Compose([
                                 transforms.Grayscale(num_output_channels=1),
                                 transforms.Resize((28, 28)),
                                 transforms.ToTensor(),
                                 transforms.Normalize(mean=[mean], std=[std])
                                           ])
        test_data = datasets.CIFAR10(root=ROOT,
                                 train=False.
                                 download=True,
                                 transform=test_transforms)
        print(f'Number of testing examples: {len(test_data)}')
        N IMAGES = 25
        images = [image for image, label in [train_data[i] for i in range(N_IMAGES)]]
        plot_images(images)
        VALID RATIO = 0.9
        n_train_examples = int(len(train_data) * VALID_RATIO)
        n_valid_examples = len(train_data) - n_train_examples
        train_data, valid_data = data.random_split(train_data,
                                                  [n_train_examples, n_valid_examples])
        valid_data = copy.deepcopy(valid_data)
        valid_data.dataset.transform = test_transforms
```

return train_data, valid_data, test_data

2.1 Train model với Batch size = 64, Activation: Relu

[45]: INPUT_DIM = 28 * 28 OUTPUT_DIM = 10

[46]: # Augmentation
add_transform = []
train_data, valid_data, test_data = process_data_cifar10(add_transform)

Files already downloaded and verified Number of training examples: 50000 Files already downloaded and verified Number of testing examples: 10000



0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: 0m 20s

Train Loss: 2.011 | Train Acc: 27.29% Val. Loss: 1.902 | Val. Acc: 31.98%

Training: 0% | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 20s

Train Loss: 1.881 | Train Acc: 32.69% Val. Loss: 1.814 | Val. Acc: 36.51%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: Om 20s

Train Loss: 1.825 | Train Acc: 34.94% Val. Loss: 1.791 | Val. Acc: 36.73%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 20s

Train Loss: 1.784 | Train Acc: 36.58% Val. Loss: 1.765 | Val. Acc: 37.56%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: Om 20s

Train Loss: 1.760 | Train Acc: 37.38% Val. Loss: 1.731 | Val. Acc: 38.27%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 19s

Train Loss: 1.743 | Train Acc: 37.78% Val. Loss: 1.711 | Val. Acc: 39.38%

Training: 0% | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 20s

Train Loss: 1.722 | Train Acc: 38.54% Val. Loss: 1.690 | Val. Acc: 39.85%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 20s

Train Loss: 1.706 | Train Acc: 39.02% Val. Loss: 1.708 | Val. Acc: 38.88%

Training: 0%| | 0/704 [00:00<?, ?it/s]
Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: Om 20s

Train Loss: 1.692 | Train Acc: 39.96% Val. Loss: 1.710 | Val. Acc: 40.05%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: Om 20s

Train Loss: 1.682 | Train Acc: 40.04% Val. Loss: 1.699 | Val. Acc: 40.25%

[48]: test_loss, test_acc = test_model(model, test_iterator)

Evaluating: 0%| | 0/157 [00:00<?, ?it/s]

Test Loss: 1.685 | Test Acc: 40.60%

2.2 Thêm phần tạo đa dang data Augmentation với RandomHorizontalFlip

Files already downloaded and verified Number of training examples: 50000 Files already downloaded and verified Number of testing examples: 10000



[50]: # Train model

0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/704 [00:00<?, ?it/s]
Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: 0m 21s

Train Loss: 2.010 | Train Acc: 27.74% Val. Loss: 1.887 | Val. Acc: 32.46%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 21s

Train Loss: 1.881 | Train Acc: 32.86% Val. Loss: 1.831 | Val. Acc: 35.70%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0% | 0/79 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 21s

Train Loss: 1.828 | Train Acc: 34.54% Val. Loss: 1.766 | Val. Acc: 38.29%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0% | 0/79 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 21s

Train Loss: 1.794 | Train Acc: 35.69% Val. Loss: 1.750 | Val. Acc: 37.92%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 21s

Train Loss: 1.771 | Train Acc: 36.40% Val. Loss: 1.742 | Val. Acc: 38.98%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 21s

Train Loss: 1.750 | Train Acc: 37.70% Val. Loss: 1.709 | Val. Acc: 39.72%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 21s

Train Loss: 1.733 | Train Acc: 38.17% Val. Loss: 1.718 | Val. Acc: 40.84%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 21s

Train Loss: 1.715 | Train Acc: 38.85% Val. Loss: 1.694 | Val. Acc: 39.24%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: 0m 22s

Train Loss: 1.705 | Train Acc: 39.04% Val. Loss: 1.667 | Val. Acc: 40.72%

Training: 0% | 0/704 [00:00<?, ?it/s]

Evaluating: 0% | 0/79 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: Om 22s

Train Loss: 1.698 | Train Acc: 39.28% Val. Loss: 1.666 | Val. Acc: 41.69%

[51]: test_loss_Flip, test_acc_Flip = test_model(model, test_iterator_Flip)

Evaluating: 0%| | 0/157 [00:00<?, ?it/s]

Test Loss: 1.662 | Test Acc: 40.96%

2.3 Thêm phần tạo đa dang data Augmentation với RandomZoomOut

[52]: # Augmentation

add_transform = [transforms.RandomAffine(degrees=0, scale=(1.0, 1.5))]

train_data_ZoomOut, valid_data_ZoomOut, test_data_ZoomOut = □

→process_data_cifar10(add_transform)

Files already downloaded and verified Number of training examples: 50000 Files already downloaded and verified Number of testing examples: 10000



[53]: # Train model

0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/704 [00:00<?, ?it/s]
Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: Om 22s

Train Loss: 2.063 | Train Acc: 25.33% Val. Loss: 1.983 | Val. Acc: 28.90%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 22s

Train Loss: 1.968 | Train Acc: 29.57% Val. Loss: 1.924 | Val. Acc: 31.47%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 23s

Train Loss: 1.918 | Train Acc: 31.04% Val. Loss: 1.869 | Val. Acc: 32.52%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 23s

Train Loss: 1.888 | Train Acc: 32.46% Val. Loss: 1.866 | Val. Acc: 34.26%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 23s

Train Loss: 1.862 | Train Acc: 33.16% Val. Loss: 1.843 | Val. Acc: 34.20%

Training: 0% | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 23s

Train Loss: 1.843 | Train Acc: 34.02% Val. Loss: 1.829 | Val. Acc: 34.06%

Training: 0% | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 23s

Train Loss: 1.823 | Train Acc: 34.89% Val. Loss: 1.811 | Val. Acc: 35.34%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 23s

Train Loss: 1.810 | Train Acc: 35.15% Val. Loss: 1.784 | Val. Acc: 36.43%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: 0m 23s

Train Loss: 1.800 | Train Acc: 35.49% Val. Loss: 1.763 | Val. Acc: 38.43%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: 0m 23s

Train Loss: 1.783 | Train Acc: 36.21% Val. Loss: 1.752 | Val. Acc: 38.25%

[54]: test_loss_ZoomOut, test_acc_ZoomOut = test_model(model, test_iterator_ZoomOut)

Evaluating: 0% | 0/157 [00:00<?, ?it/s]

Test Loss: 1.762 | Test Acc: 38.25%

2.4 Thêm phần tạo đa dang data Augmentation với RandomAffine

Files already downloaded and verified Number of training examples: 50000 Files already downloaded and verified Number of testing examples: 10000



[56]: # Train model

0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/704 [00:00<?, ?it/s]
Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: 0m 24s

Train Loss: 2.111 | Train Acc: 22.80% Val. Loss: 1.997 | Val. Acc: 28.66%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 24s

Train Loss: 2.014 | Train Acc: 26.95% Val. Loss: 1.918 | Val. Acc: 32.08%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0% | 0/79 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 32s

Train Loss: 1.970 | Train Acc: 28.69% Val. Loss: 1.909 | Val. Acc: 31.96%

Training: 0% | 0/704 [00:00<?, ?it/s]

Evaluating: 0% | 0/79 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 28s

Train Loss: 1.938 | Train Acc: 30.04% Val. Loss: 1.874 | Val. Acc: 34.06%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 29s

Train Loss: 1.912 | Train Acc: 30.98% Val. Loss: 1.864 | Val. Acc: 34.47%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 28s

Train Loss: 1.898 | Train Acc: 31.41% Val. Loss: 1.825 | Val. Acc: 35.30%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 28s

Train Loss: 1.882 | Train Acc: 31.92% Val. Loss: 1.890 | Val. Acc: 33.54%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 28s

Train Loss: 1.871 | Train Acc: 32.53% Val. Loss: 1.792 | Val. Acc: 36.89%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: 0m 28s

Train Loss: 1.860 | Train Acc: 33.05% Val. Loss: 1.848 | Val. Acc: 34.77%

Training: 0% | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: Om 29s Train Loss: 1.854 | Train Loss: 1.8

Train Loss: 1.854 | Train Acc: 33.16% Val. Loss: 1.796 | Val. Acc: 36.85%

[57]: test_loss_Affine, test_acc_Affine = test_model(model, test_iterator_Affine)

Evaluating: 0% | 0/157 [00:00<?, ?it/s]

Test Loss: 1.808 | Test Acc: 36.18%

2.5 Thay đổi Batch Size = 32

[58]: # Train model train_iterator_32, valid_iterator_32, test_iterator_32 = 0 oload_DataLoader(train_data, valid_data, test_data, BATCH_SIZE = 32) model = MLP(INPUT_DIM, OUTPUT_DIM) train_model(model, train_iterator_32, valid_iterator_32)

0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/1407 [00:00<?, ?it/s]

Evaluating: 0%| | 0/157 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: 0m 29s

Train Loss: 2.011 | Train Acc: 27.50% Val. Loss: 1.902 | Val. Acc: 32.70%

Training: 0%| | 0/1407 [00:00<?, ?it/s]

Evaluating: 0%| | 0/157 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 28s

Train Loss: 1.889 | Train Acc: 32.49% Val. Loss: 1.823 | Val. Acc: 34.95%

Training: 0%| | 0/1407 [00:00<?, ?it/s]

Evaluating: 0%| | 0/157 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 28s

Train Loss: 1.842 | Train Acc: 34.44% Val. Loss: 1.802 | Val. Acc: 36.13%

Training: 0%| | 0/1407 [00:00<?, ?it/s]

Evaluating: 0%| | 0/157 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 28s

Train Loss: 1.809 | Train Acc: 35.39% Val. Loss: 1.783 | Val. Acc: 36.58%

Training: 0%| | 0/1407 [00:00<?, ?it/s]

Evaluating: 0%| | 0/157 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 28s Train Loss: 1.785 | Train Acc: 36.23% Val. Loss: 1.780 | Val. Acc: 36.98% Training: 0%| | 0/1407 [00:00<?, ?it/s] Evaluating: 0%| | 0/157 [00:00<?, ?it/s] Epoch: 06 | Epoch Time: 0m 28s Train Loss: 1.770 | Train Acc: 36.79% Val. Loss: 1.743 | Val. Acc: 37.60% Training: 0%| | 0/1407 [00:00<?, ?it/s] | 0/157 [00:00<?, ?it/s] Evaluating: 0%1 Epoch: 07 | Epoch Time: 0m 28s Train Loss: 1.752 | Train Acc: 37.58% Val. Loss: 1.727 | Val. Acc: 38.91% 0%1 Training: | 0/1407 [00:00<?, ?it/s] 0%1 | 0/157 [00:00<?, ?it/s] Evaluating: Epoch: 08 | Epoch Time: 0m 28s Train Loss: 1.737 | Train Acc: 37.90% Val. Loss: 1.731 | Val. Acc: 38.91% 0%| | 0/1407 [00:00<?, ?it/s] Training: | 0/157 [00:00<?, ?it/s] 0%| Evaluating: Epoch: 09 | Epoch Time: 0m 28s Train Loss: 1.722 | Train Acc: 38.58% Val. Loss: 1.696 | Val. Acc: 40.09% | 0/1407 [00:00<?, ?it/s] Training: 0%| 0%| Evaluating: | 0/157 [00:00<?, ?it/s] Epoch: 10 | Epoch Time: 0m 28s Train Loss: 1.716 | Train Acc: 38.91% Val. Loss: 1.697 | Val. Acc: 39.83% [59]: test_loss_32, test_acc_32 = test_model(model, test_iterator_32) | 0/313 [00:00<?, ?it/s] Evaluating: 0%| Test Loss: 1.689 | Test Acc: 40.25% 2.6 Thay đổi Batch Size = 128[60]: # Train model train_iterator_128, valid_iterator_128, test_iterator_128 =_ →load_DataLoader(train_data, valid_data, test_data, BATCH_SIZE = 128)

model = MLP(INPUT_DIM, OUTPUT_DIM) train_model(model, train_iterator_128, valid_iterator_128)

0%| | 0/10 [00:00<?, ?it/s]

Training: 0% | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: 0m 24s

Train Loss: 2.019 | Train Acc: 27.23% Val. Loss: 1.875 | Val. Acc: 33.01%

Training: 0%| | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 24s

Train Loss: 1.884 | Train Acc: 32.75% Val. Loss: 1.805 | Val. Acc: 35.70%

Training: 0% | | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 21s

Train Loss: 1.820 | Train Acc: 34.89% Val. Loss: 1.785 | Val. Acc: 36.13%

Training: 0%| | 0/352 [00:00<?, ?it/s]

Evaluating: 0% | 0/40 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 21s

Train Loss: 1.783 | Train Acc: 36.28% Val. Loss: 1.739 | Val. Acc: 38.34%

Training: 0% | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: Om 21s

Train Loss: 1.755 | Train Acc: 37.71% Val. Loss: 1.723 | Val. Acc: 38.69%

Training: 0%| | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 19s

Train Loss: 1.734 | Train Acc: 38.28% Val. Loss: 1.699 | Val. Acc: 39.71%

Training: 0% | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 18s Train Loss: 1.712 | Train Acc: 39.20% Val. Loss: 1.671 | Val. Acc: 40.35% Training: 0%| | 0/352 [00:00<?, ?it/s] 0%| | 0/40 [00:00<?, ?it/s] Evaluating: Epoch: 08 | Epoch Time: 0m 18s Train Loss: 1.702 | Train Acc: 39.33% Val. Loss: 1.663 | Val. Acc: 41.11% | 0/352 [00:00<?, ?it/s] Training: 0%| Evaluating: 0%| | 0/40 [00:00<?, ?it/s] Epoch: 09 | Epoch Time: 0m 18s Train Loss: 1.688 | Train Acc: 39.83% Val. Loss: 1.654 | Val. Acc: 41.70% 0%1 Training: | 0/352 [00:00<?, ?it/s] 0%| | 0/40 [00:00<?, ?it/s] Evaluating: Epoch: 10 | Epoch Time: Om 18s Train Loss: 1.672 | Train Acc: 40.70% Val. Loss: 1.646 | Val. Acc: 42.05% [61]: test_loss_128, test_acc_128 = test_model(model, test_iterator_128) Evaluating: 0%1 | 0/79 [00:00<?, ?it/s] Test Loss: 1.634 | Test Acc: 42.24% 2.7 Thay đổi Batch Size = 256[62]: # Train model train_iterator_256, valid_iterator_256, test_iterator_256 =_ aload_DataLoader(train_data, valid_data, test_data, BATCH_SIZE = 256) model = MLP(INPUT_DIM, OUTPUT_DIM) train_model(model, train_iterator_256, valid_iterator_256) 0%1 | 0/10 [00:00<?, ?it/s] 0%1 | 0/176 [00:00<?, ?it/s] Training: 0%| | 0/20 [00:00<?, ?it/s] Evaluating: Epoch: 01 | Epoch Time: 0m 17s Train Loss: 2.033 | Train Acc: 26.47% Val. Loss: 1.900 | Val. Acc: 31.66%

| 0/176 [00:00<?, ?it/s]

| 0/20 [00:00<?, ?it/s]

Training:

Evaluating:

0%|

Epoch: 02 | Epoch Time: 0m 17s

Train Loss: 1.895 | Train Acc: 32.53% Val. Loss: 1.839 | Val. Acc: 34.78%

Training: 0%| | 0/176 [00:00<?, ?it/s]

Evaluating: 0% | 0/20 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 16s

Train Loss: 1.833 | Train Acc: 34.95% Val. Loss: 1.795 | Val. Acc: 36.72%

Training: 0%| | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 17s

Train Loss: 1.792 | Train Acc: 36.42% Val. Loss: 1.781 | Val. Acc: 36.43%

Training: 0%| | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 18s

Train Loss: 1.763 | Train Acc: 37.31% Val. Loss: 1.721 | Val. Acc: 38.51%

Training: 0%| | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 19s

Train Loss: 1.738 | Train Acc: 38.30% Val. Loss: 1.713 | Val. Acc: 39.63%

Training: 0%| | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 21s

Train Loss: 1.716 | Train Acc: 38.91% Val. Loss: 1.705 | Val. Acc: 39.14%

Training: 0%| | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 19s

Train Loss: 1.705 | Train Acc: 39.31% Val. Loss: 1.658 | Val. Acc: 40.88%

Training: 0% | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: Om 19s

Train Loss: 1.687 | Train Acc: 39.88% Val. Loss: 1.643 | Val. Acc: 41.89%

Training: 0% | 0/176 [00:00<?, ?it/s]

Evaluating: 0% | 0/20 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: Om 20s

Train Loss: 1.669 | Train Acc: 40.68% Val. Loss: 1.649 | Val. Acc: 41.28%

[63]: test_loss_256, test_acc_256 = test_model(model, test_iterator_256)

Evaluating: 0% | 0/40 [00:00<?, ?it/s]

Test Loss: 1.648 | Test Acc: 41.12%

2.8 Đổi hành Activation: LeakyRelu

[64]: # Train model

0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/704 [00:00<?, ?it/s]
Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: 0m 27s

Train Loss: 2.034 | Train Acc: 26.67% Val. Loss: 1.925 | Val. Acc: 30.85%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0% | 0/79 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 24s

Train Loss: 1.907 | Train Acc: 31.82% Val. Loss: 1.838 | Val. Acc: 34.32%

Training: 0% | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 36s

Train Loss: 1.854 | Train Acc: 33.89% Val. Loss: 1.791 | Val. Acc: 37.36%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 28s Train Loss: 1.823 | Train Acc: 35.09%

Val. Loss: 1.755 | Val. Acc: 38.01%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 27s

Train Loss: 1.801 | Train Acc: 35.84% Val. Loss: 1.762 | Val. Acc: 37.22%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: Om 39s

Train Loss: 1.781 | Train Acc: 36.49% Val. Loss: 1.762 | Val. Acc: 37.80%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 34s

Train Loss: 1.766 | Train Acc: 36.95% Val. Loss: 1.702 | Val. Acc: 39.56%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 31s

Train Loss: 1.744 | Train Acc: 37.67% Val. Loss: 1.721 | Val. Acc: 38.47%

Training: 0%| | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: 0m 28s

Train Loss: 1.739 | Train Acc: 38.17% Val. Loss: 1.710 | Val. Acc: 40.39%

Training: 0% | 0/704 [00:00<?, ?it/s]

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: 0m 32s

Train Loss: 1.726 | Train Acc: 38.18% Val. Loss: 1.662 | Val. Acc: 40.92%

[65]: test_loss_LeakyRelu, test_acc_LeakyRelu = test_model(model_LeakyRelu,_u + test_iterator_LeakyRelu)

Evaluating: 0%| | 0/157 [00:00<?, ?it/s]

Test Loss: 1.648 | Test Acc: 41.31%

2.9 Đổi hành Activation: ELU

```
[66]: # Train model
      train_iterator_ELU, valid_iterator_ELU, test_iterator_ELU =_
       →load_DataLoader(train_data, valid_data, test_data)
      model ELU = MLP ELU(INPUT DIM, OUTPUT DIM)
      train_model(model_ELU, train_iterator_ELU, valid_iterator_ELU)
                    | 0/10 [00:00<?, ?it/s]
       0%1
                               | 0/704 [00:00<?, ?it/s]
     Training:
                 0%1
                                 | 0/79 [00:00<?, ?it/s]
     Evaluating:
                   0%1
     Epoch: 01 | Epoch Time: 0m 34s
             Train Loss: 2.016 | Train Acc: 27.22%
              Val. Loss: 1.863 | Val. Acc: 32.50%
     Training:
                 0%1
                              | 0/704 [00:00<?, ?it/s]
     Evaluating:
                   0%1
                                 | 0/79 [00:00<?, ?it/s]
     Epoch: 02 | Epoch Time: 0m 32s
             Train Loss: 1.862 | Train Acc: 33.04%
              Val. Loss: 1.763 | Val. Acc: 37.30%
                 0%|
                              | 0/704 [00:00<?, ?it/s]
     Training:
                   0%|
                                | 0/79 [00:00<?, ?it/s]
     Evaluating:
     Epoch: 03 | Epoch Time: 0m 34s
             Train Loss: 1.802 | Train Acc: 35.31%
              Val. Loss: 1.744 | Val. Acc: 37.80%
     Training:
                 0%1
                              | 0/704 [00:00<?, ?it/s]
                   0%1
                                 | 0/79 [00:00<?, ?it/s]
     Evaluating:
     Epoch: 04 | Epoch Time: 0m 34s
             Train Loss: 1.769 | Train Acc: 36.74%
              Val. Loss: 1.699 | Val. Acc: 39.16%
     Training:
                 0%1
                              | 0/704 [00:00<?, ?it/s]
                   0%1
                                 | 0/79 [00:00<?, ?it/s]
     Evaluating:
     Epoch: 05 | Epoch Time: 0m 34s
             Train Loss: 1.744 | Train Acc: 37.45%
              Val. Loss: 1.684 | Val. Acc: 39.72%
                 0%|
                              | 0/704 [00:00<?, ?it/s]
     Training:
     Evaluating: 0%|
                                | 0/79 [00:00<?, ?it/s]
```

Epoch: 06 | Epoch Time: 0m 32s Train Loss: 1.721 | Train Acc: 38.30% Val. Loss: 1.674 | Val. Acc: 41.00% Training: 0%| | 0/704 [00:00<?, ?it/s] 0%| | 0/79 [00:00<?, ?it/s] Evaluating: Epoch: 07 | Epoch Time: 0m 32s Train Loss: 1.701 | Train Acc: 38.82% Val. Loss: 1.676 | Val. Acc: 40.82% | 0/704 [00:00<?, ?it/s] Training: 0%| Evaluating: 0%| | 0/79 [00:00<?, ?it/s] Epoch: 08 | Epoch Time: 0m 32s Train Loss: 1.685 | Train Acc: 39.69% Val. Loss: 1.650 | Val. Acc: 41.57% Training: 0%1 | 0/704 [00:00<?, ?it/s] 0%| | 0/79 [00:00<?, ?it/s] Evaluating: Epoch: 09 | Epoch Time: 0m 32s Train Loss: 1.676 | Train Acc: 39.87% Val. Loss: 1.635 | Val. Acc: 42.17% 0%| | 0/704 [00:00<?, ?it/s] Training: Evaluating: 0%| | 0/79 [00:00<?, ?it/s] Epoch: 10 | Epoch Time: Om 31s Train Loss: 1.660 | Train Acc: 40.66% Val. Loss: 1.614 | Val. Acc: 41.38% [67]: | test_loss_ELU, test_acc_ELU = test_model(model_ELU, test_iterator_ELU) 0%1 | 0/157 [00:00<?, ?it/s] Evaluating: Test Loss: 1.631 | Test Acc: 42.42% 2.9.1 Combine Batch Size 128 + ELU [68]: # Train model train_iterator_128_ELU, valid_iterator_128_ELU, test_iterator_128_ELU =__ aload_DataLoader(train_data, valid_data, test_data, BATCH_SIZE = 128) model ELU = MLP ELU(INPUT DIM, OUTPUT DIM) train_model(model_ELU, train_iterator_128_ELU, valid_iterator_128_ELU) 0%1 | 0/10 [00:00<?, ?it/s]

| 0/352 [00:00<?, ?it/s]

| 0/40 [00:00<?, ?it/s]

Training:

Evaluating:

0%1

0%|

Epoch: 01 | Epoch Time: 0m 30s

Train Loss: 2.028 | Train Acc: 26.95% Val. Loss: 1.882 | Val. Acc: 33.46%

Training: 0%| | 0/352 [00:00<?, ?it/s]

Evaluating: 0% | 0/40 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 30s

Train Loss: 1.885 | Train Acc: 32.21% Val. Loss: 1.813 | Val. Acc: 34.55%

Training: 0%| | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 29s

Train Loss: 1.812 | Train Acc: 35.28% Val. Loss: 1.730 | Val. Acc: 37.87%

Training: 0%| | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: Om 30s

Train Loss: 1.772 | Train Acc: 36.33% Val. Loss: 1.706 | Val. Acc: 38.26%

Training: 0%| | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: Om 30s

Train Loss: 1.740 | Train Acc: 37.59% Val. Loss: 1.691 | Val. Acc: 39.49%

Training: 0%| | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 30s

Train Loss: 1.724 | Train Acc: 38.31% Val. Loss: 1.671 | Val. Acc: 39.75%

Training: 0% | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 29s

Train Loss: 1.703 | Train Acc: 38.97% Val. Loss: 1.660 | Val. Acc: 41.23%

Training: 0% | 0/352 [00:00<?, ?it/s]

Evaluating: 0% | 0/40 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 29s

Train Loss: 1.686 | Train Acc: 39.74% Val. Loss: 1.638 | Val. Acc: 41.68%

Training: 0%| | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: 0m 29s

Train Loss: 1.675 | Train Acc: 40.23% Val. Loss: 1.650 | Val. Acc: 41.78%

Training: 0%| | 0/352 [00:00<?, ?it/s]

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Epoch: 10 | Epoch Time: Om 30s

Train Loss: 1.660 | Train Acc: 40.39% Val. Loss: 1.610 | Val. Acc: 42.46%

[69]: test_loss_128_ELU, test_acc_128_ELU = test_model(model_ELU,_u + test_iterator_128_ELU)

Evaluating: 0%| | 0/79 [00:00<?, ?it/s]

Test Loss: 1.615 | Test Acc: 42.31%

2.9.2 Combine Batch Size 256 + ELU

[70]: # Train model

train_iterator_256_ELU, valid_iterator_256_ELU, test_iterator_256_ELU = load_DataLoader(train_data, valid_data, test_data, BATCH_SIZE = 256)
model_ELU = MLP_ELU(INPUT_DIM, OUTPUT_DIM)
train_model(model_ELU, train_iterator_256_ELU, valid_iterator_256_ELU)

0%| | 0/10 [00:00<?, ?it/s]

Training: 0%| | 0/176 [00:00<?, ?it/s]
Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 01 | Epoch Time: 0m 28s

Train Loss: 2.061 | Train Acc: 25.57% Val. Loss: 1.910 | Val. Acc: 31.49%

Training: 0%| | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 02 | Epoch Time: 0m 28s

Train Loss: 1.909 | Train Acc: 31.40% Val. Loss: 1.839 | Val. Acc: 34.09%

Training: 0% | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 03 | Epoch Time: 0m 28s

Train Loss: 1.833 | Train Acc: 34.41% Val. Loss: 1.752 | Val. Acc: 38.10%

Training: 0%| | 0/176 [00:00<?, ?it/s]

Evaluating: 0% | 0/20 [00:00<?, ?it/s]

Epoch: 04 | Epoch Time: 0m 23s

Train Loss: 1.788 | Train Acc: 36.26% Val. Loss: 1.724 | Val. Acc: 38.72%

Training: 0%| | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 05 | Epoch Time: 0m 23s

Train Loss: 1.756 | Train Acc: 37.24% Val. Loss: 1.698 | Val. Acc: 38.92%

Training: 0%| | 0/176 [00:00<?, ?it/s]

Evaluating: 0% | 0/20 [00:00<?, ?it/s]

Epoch: 06 | Epoch Time: 0m 28s

Train Loss: 1.738 | Train Acc: 37.76% Val. Loss: 1.685 | Val. Acc: 40.04%

Training: 0% | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 07 | Epoch Time: 0m 28s

Train Loss: 1.710 | Train Acc: 38.82% Val. Loss: 1.657 | Val. Acc: 40.05%

Training: 0% | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 08 | Epoch Time: 0m 28s

Train Loss: 1.699 | Train Acc: 39.42% Val. Loss: 1.645 | Val. Acc: 41.83%

Training: 0% | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

Epoch: 09 | Epoch Time: 0m 29s

Train Loss: 1.679 | Train Acc: 40.25% Val. Loss: 1.628 | Val. Acc: 41.61%

Training: 0%| | 0/176 [00:00<?, ?it/s]

Evaluating: 0%| | 0/20 [00:00<?, ?it/s]

```
Epoch: 10 | Epoch Time: 0m 29s
Train Loss: 1.667 | Train Acc: 40.58%
```

Val. Loss: 1.626 | Val. Acc: 42.23%

Evaluating: 0%| | 0/40 [00:00<?, ?it/s]

Test Loss: 1.627 | Test Acc: 42.31%

2.10 Tạo bảng so sánh (CIFAR10 dataset)

```
[73]: # Generate the table
  result = tabulate(
  data_values,
  headers=data_header,
  tablefmt='fancy_grid',
  )
  # Print the resulting table
  print(result)
```

Model with	Test_loss	Test_acc
Batch size = 64 + RELU	1.685	40.6
RandomHorizontalFlip	1.662	40.96
RandomZoomOut	1.762	38.25

RandomAffine	1.808	36.18
Batch size = 32	1.689	40.25
Batch size = 128	1.634	42.24
Batch size = 256	1.648	41.12
LeakyRelu	1.648	41.31
ELU	1.631	42.42
Batch Size 128 + ELU	1.615	42.31
Batch Size 256 + ELU	1.627	42.31