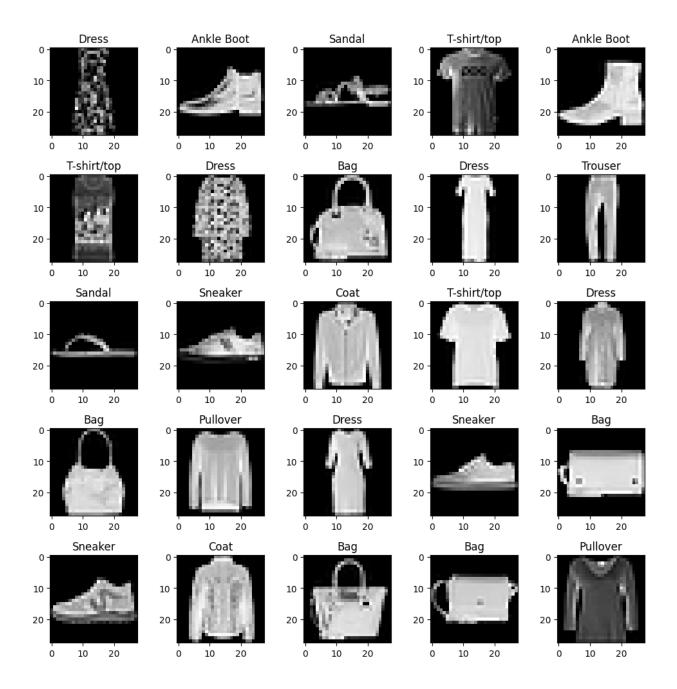
```
import pandas as pd
import matplotlib.pyplot as plt
from tqdm import tqdm

import torch
from torch import nn
import torchvision
import torchvision.transforms as transforms

import numpy as np
import random
np.random.seed(0)
torch.manual_seed(0)
random.seed(0)
```

Data

```
transform = transforms.ToTensor()
training set = torchvision.datasets.FashionMNIST('./data', train=True,
transform=transform, download=True)
validation set = torchvision.datasets.FashionMNIST('./data',
train=False, transform=transform, download=True)
training loader = torch.utils.data.DataLoader(training set,
batch size=32, shuffle=True)
validation loader = torch.utils.data.DataLoader(validation set,
batch size=128, shuffle=False)
100%
              26.4M/26.4M [00:02<00:00, 11.7MB/s]
              29.5k/29.5k [00:00<00:00, 199kB/s]
100%|
              4.42M/4.42M [00:01<00:00, 3.67MB/s]
100%
              || 5.15k/5.15k [00:00<00:00, 10.4MB/s]
100%|
batch = next(iter(training loader))
plt.figure(figsize=(10, 10))
for i, (image, label) in enumerate(zip(*batch)):
   if i > 24:
       break
   plt.subplot(5, 5, i + 1)
   plt.imshow(image[0], cmap="gray")
   plt.title(classes[label])
plt.tight layout()
```



Exercise

You are required to train an MLP on the Fashion MNIST dataset. For this task, you must define the following:

- The objective function
- The model architecture
- The optimizer
- The training loop

You will train three MLP models, each with different hyperparameters. You must vary at least two of the following aspects between the models:

- Number of layers
- Number of neurons
- Activation function
- Optimizer

Deliverables

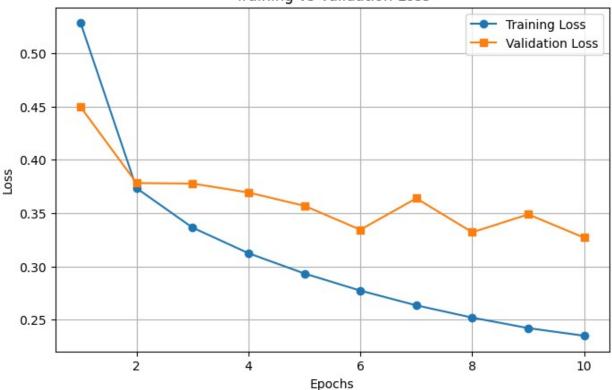
- The complete code
- Learning curves for the three MLP models
- Table summarizing the changes in the hyperparameters and the performance of the models on the train and test sets.
- A write-up analyzing how your choices impacted the results.

```
def train one epoch(model, optimizer, train loader, criterion):
    model.train()
    losses = []
    for images, labels in train loader:
        optimizer.zero grad()
        y pred = model(images)
        loss = criterion(y pred, labels)
        loss.backward()
        optimizer.step()
        losses.append(loss.item())
    return losses
def validate_one_epoch(model, val loader, criterion):
    model.eval()
    losses = []
    with torch.no grad():
        for images, labels in val loader:
            y pred = model(images)
            loss = criterion(y_pred, labels)
            losses.append(loss.item())
    return losses
class Model 1(nn.Module):
    def __init__(self , input_shape , output_shape):
        super(Model_1, self).__init__()
        self.fc1 = nn.Linear(input shape , 256)
        self.fc2 = nn.Linear(256, 128)
```

```
self.fc3 = nn.Linear(128, 64)
        self.out = nn.Linear(64 , output shape)
        self.relu = nn.ReLU()
    def forward(self ,x):
          x = x.view(-1, 28*28)
          x = self.relu(self.fc1(x))
          x = self.relu(self.fc2(x))
          x = self.relu(self.fc3(x))
          x = self.out(x)
          return x
criterion = nn.CrossEntropyLoss()
model = Model 1(28*28, 10)
optimizer = torch.optim.Adam(model.parameters(), lr=0.001)
num epochs = 10
train losses = []
val losses = []
for epoch in range(num epochs):
    losses = train one epoch(model, optimizer, training loader,
criterion)
    avg train loss = np.mean(losses)
    train losses.append(avg train loss)
    val loss = validate one epoch(model, validation loader, criterion)
    avg val loss = np.mean(val loss)
    val losses.append(avg val loss)
    print(f"Epoch {epoch+1}/{num epochs}, Train Loss:
{avg train loss:.4f}, Val Loss: {avg val loss:.4f}")
Epoch 1/10, Train Loss: 0.5283, Val Loss: 0.4495
Epoch 2/10, Train Loss: 0.3734, Val Loss: 0.3783
Epoch 3/10, Train Loss: 0.3365, Val Loss: 0.3778
Epoch 4/10, Train Loss: 0.3124, Val Loss: 0.3694
Epoch 5/10, Train Loss: 0.2934, Val Loss: 0.3569
Epoch 6/10, Train Loss: 0.2774, Val Loss: 0.3345
Epoch 7/10, Train Loss: 0.2636, Val Loss: 0.3639
Epoch 8/10, Train Loss: 0.2521, Val Loss: 0.3321
Epoch 9/10, Train Loss: 0.2422, Val Loss: 0.3489
Epoch 10/10, Train Loss: 0.2350, Val Loss: 0.3270
epochs = range(1, num epochs + 1)
plt.figure(figsize=(8,5))
plt.plot(epochs, train losses, label="Training Loss", marker="o")
plt.plot(epochs, val losses, label="Validation Loss", marker="s")
```

```
plt.xlabel("Epochs")
plt.ylabel("Loss")
plt.title("Training vs Validation Loss")
plt.legend()
plt.grid()
plt.show()
```

Training vs Validation Loss



```
class Model_2(nn.Module):
    def __init__(self , input_shape , output_shape):
        super(Model_2, self).__init__()

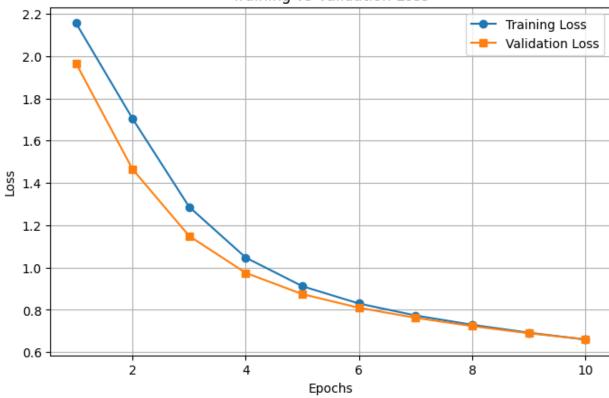
    self.fc1 = nn.Linear(input_shape , 128)
        self.fc2 = nn.Linear(128 , 64)
        self.out = nn.Linear(64 , output_shape)

    self.tanh = nn.Tanh()
    self.relu = nn.ReLU()

def forward(self ,x):
        x = x.view(-1, 28*28)
        x = self.tanh(self.fc1(x))
        x = self.relu(self.fc2(x))
```

```
x = self.out(x)
          return x
model = Model \ 2(28*28, 10)
optimizer = torch.optim.SGD(model.parameters(), lr=0.001)
num epochs = 10
train losses = []
val losses = []
for epoch in range(num epochs):
    losses = train one epoch(model, optimizer, training loader,
criterion)
    avg train loss = np.mean(losses)
    train losses.append(avg train loss)
    val loss = validate one epoch(model, validation loader, criterion)
    avg val loss = np.mean(val loss)
    val losses.append(avg val loss)
    print(f"Epoch {epoch+1}/{num epochs}, Train Loss:
{avg train loss:.4f}, Val Loss: {avg val loss:.4f}")
Epoch 1/10, Train Loss: 2.1211, Val Loss: 1.8950
Epoch 2/10, Train Loss: 1.6245, Val Loss: 1.3966
Epoch 3/10, Train Loss: 1.2417, Val Loss: 1.1308
Epoch 4/10, Train Loss: 1.0430, Val Loss: 0.9860
Epoch 5/10, Train Loss: 0.9248, Val Loss: 0.8918
Epoch 6/10, Train Loss: 0.8431, Val Loss: 0.8233
Epoch 7/10, Train Loss: 0.7818, Val Loss: 0.7699
Epoch 8/10, Train Loss: 0.7337, Val Loss: 0.7279
Epoch 9/10, Train Loss: 0.6948, Val Loss: 0.6938
Epoch 10/10, Train Loss: 0.6623, Val Loss: 0.6645
epochs = range(1, num epochs + 1)
plt.figure(figsize=(8,5))
plt.plot(epochs, train losses, label="Training Loss", marker="o")
plt.plot(epochs, val losses, label="Validation Loss", marker="s")
plt.xlabel("Epochs")
plt.ylabel("Loss")
plt.title("Training vs Validation Loss")
plt.leaend()
plt.grid()
plt.show()
```

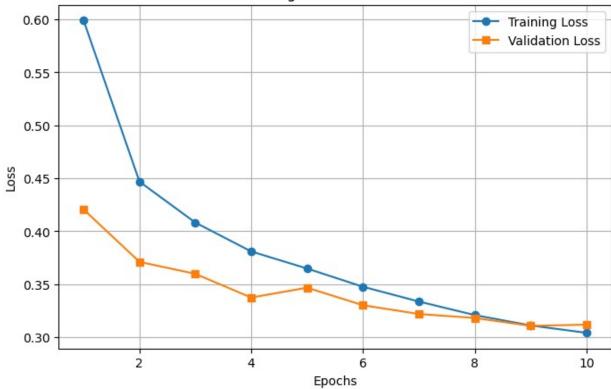
Training vs Validation Loss



```
class Model 3(nn.Module):
    def init (self , input shape , output shape):
        super(Model 3, self). init ()
        self.fc1 = nn.Linear(input shape , 512)
        self.bn1 = nn.BatchNorm1d(\overline{512})
        self.fc2 = nn.Linear(512, 256)
        self.bn2 = nn.BatchNorm1d(256)
        self.fc3 = nn.Linear(256, 128)
        self.bn3 = nn.BatchNorm1d(128)
        self.fc4 = nn.Linear(128, 64)
        self.bn4 = nn.BatchNorm1d(64)
        self.out = nn.Linear(64 , output shape)
        self.relu = nn.ReLU()
        self.dropout = nn.Dropout(0.3)
    def forward(self ,x):
        x = x.view(x.size(0), -1)
        x = self.relu(self.bn1(self.fc1(x)))
        x = self.dropout(x)
        x = self.relu(self.bn2(self.fc2(x)))
        x = self.dropout(x)
        x = self.relu(self.bn3(self.fc3(x)))
```

```
x = self.dropout(x)
        x = self.relu(self.bn4(self.fc4(x)))
        x = self.out(x)
        return x
model = Model 3(28*28, 10)
optimizer = torch.optim.Adam(model.parameters(), lr=0.001)
num epochs = 10
train losses = []
val losses = []
for epoch in range(num epochs):
    losses = train one epoch(model, optimizer, training loader,
criterion)
    avg_train_loss = np.mean(losses)
    train losses.append(avg train loss)
    val loss = validate one epoch(model, validation loader, criterion)
    avg val loss = np.mean(val loss)
    val_losses.append(avg val loss)
    print(f"Epoch {epoch+1}/{num epochs}, Train Loss:
{avg train loss:.4f}, Val Loss: {avg val loss:.4f}")
Epoch 1/10, Train Loss: 0.5971, Val Loss: 0.4329
Epoch 2/10, Train Loss: 0.4461, Val Loss: 0.3798
Epoch 3/10, Train Loss: 0.4077, Val Loss: 0.3618
Epoch 4/10, Train Loss: 0.3812, Val Loss: 0.3488
Epoch 5/10, Train Loss: 0.3618, Val Loss: 0.3384
Epoch 6/10, Train Loss: 0.3488, Val Loss: 0.3211
Epoch 7/10, Train Loss: 0.3306, Val Loss: 0.3152
Epoch 8/10, Train Loss: 0.3222, Val Loss: 0.3079
Epoch 9/10, Train Loss: 0.3142, Val Loss: 0.3115
Epoch 10/10, Train Loss: 0.3048, Val Loss: 0.3139
epochs = range(1, num epochs + 1)
plt.figure(figsize=(8,5))
plt.plot(epochs, train losses, label="Training Loss", marker="o")
plt.plot(epochs, val losses, label="Validation Loss", marker="s")
plt.xlabel("Epochs")
plt.ylabel("Loss")
plt.title("Training vs Validation Loss")
plt.legend()
plt.grid()
plt.show()
```

Training vs Validation Loss



```
data = {
    'train_error' : train_losses,
    'val_error' : val_losses,
    'epochs' : epochs,
    'optimizer' : 'Adam',
    'activation' : 'ReLU'
df1 = pd.DataFrame(data)
df1.head()
{"summary":"{\n \"name\": \"df1\",\n \"rows\": 10,\n \"fields\": [\
    {\n \"column\": \"train_error\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 0.08784346279559255,\n
\"max\": 0.5283446384151776,\n
0.242221633330316543,\n
                              0.37341762903531395,\n
0.27737013767957686\n
                           ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n
                                  },\n {\n \"column\":
                           }\n
\"val_error\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 0.035388592383661883,\n
                                                          \"min\":
                            \"max\": 0.44952993076058884,\n
0.32704490602393693,\n
\"num_unique_values\": 10,\n
0.34886520591717735,\n
0.3
                                 \"samples\": [\n
                              0.3782578267251389,\n
```

```
\"category\",\n \"num_unique_values\": 1,\n \"samples\":
[\n \"Adam\"\n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n {\n \"column\":
\"activation\",\n \"properties\": {\n \"dtype\":
\"category\",\n \"num_unique_values\": 1,\n \"samples\":
[\n \"ReLU\"\n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n ]\
n}","type":"dataframe","variable_name":"df1"}
data = {
   'train_error' : train_losses,
   'val error' : val_losses,
   'epochs': epochs,
   'optimizer' : 'SGD',
   'activation' : 'ReLU and Tanh',
    'no of layers' : '2 hidden layers',
}
df1 = pd.DataFrame(data)
df1.head()
{"summary":"{\n \model{"}: \model{"}: \model{"}: \model{"}: \model{"}: \model{"}: \model{"}} } 
n {\n \"column\": \"train_error\",\n \"properties\": {\n
\"dtype\": \"number\",\n \\\\"std\\\": 0.4735401069200405,\n
0.6644730556614792,\n\\"max\": 1.8949569448640076,\n
```

```
\"category\",\n \"num_unique_values\": 1,\n \"samples\":
 [\n \"SGD\"\n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n {\n \"column\":
\"activation\",\n \"properties\": {\n \"dtype\":
\"category\",\n \"num_unique_values\": 1,\n \"semantic_type\":
[\n \"ReLU and Tanh\"\n ],\n \"semantic_type\":
\"\",\n \"description\": \"\"\n }\n },\n {\n
\"column\": \"no_of_layers\",\n \"properties\": {\n
\"dtype\": \"category\",\n \"num_unique_values\": 1,\n
\"semplos\": [\n \"]
 \"samples\": [\n \"2 hidden layers\"\n ],\n
 \"semantic_type\": \"\",\n \"description\": \"\"\n
                                                                                                                                                                                                                    }\
                 }\n ]\n}","type":"dataframe","variable_name":"df1"}
 data = {
                'train error' : train losses,
               'val_error' : val_losses,
'epochs' : epochs,
               'optimizer' : 'Adam',
               'activation' : 'ReLU',
               'no of layers' : '4 hidden layers',
 }
 df2 = pd.DataFrame(data)
 df2.head()
 {"summary":"{\n \"name\": \"df2\",\n \"rows\": 10,\n \"fields\": [\
 n {\n \"column\": \"train_error\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 0.08774863291683539,\n
\"dtype\": \"number\",\n\\"std\": 0.08774863291683539,\n\\"min\": 0.30482813845674195,\n\\"max\": 0.5970827991088231,\n\\"num_unique_values\": 10,\n\\"samples\": [\n\\
0.3142120253900687,\n\\
0.3488374009847641\n\\],\n\\"semantic_type\": \"\",\n\\"description\": \"\"\n\\\"h\\\"num\": \\"number\",\n\\\"samples\": \\"\"min\": \\"min\": 
 \"num_unique_values\": 10,\n \"samples\": [\n
[\n \"Adam\"\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n {\n \"column\": \"activation\",\n \"properties\": {\n \"dtype\":
```

```
\"category\",\n \"num_unique_values\": 1,\n \"samples\":
[\n \"ReLU\"\n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n {\n \"column\":
\"no_of_layers\",\n \"properties\": {\n \"dtype\":
\"category\",\n \"num_unique_values\": 1,\n \"samples\":
[\n \"4 hidden layers\"\n ],\n
\"semantic_type\": \"\",\n \"description\": \"\"\n }\\n ]\n]\n]\","type":"dataframe","variable_name":"df2"}
```