Assignment 3 - Part 2 In this assignment, you are to architect a neural network to perform text classification. Grading scheme: Model: 10 • Training: 10 • Test accuracy: 20 >= 70%: 5/10 >= 75%: 10/20 >= 80%: 15/20 >= 85%: 20/20 Total: 40 In [22]: "=" import torch from torch import (nn, optim) from torch.utils.data import (Dataset, DataLoader, random_split) from torchsummaryX import summary import pandas as pd import numpy as np import warnings import test_lib from importlib import reload reload(test_lib) warnings.filterwarnings('ignore') In [23]: **from** torch.optim **import** Adam import time from torch.utils.data import random_split from torchmetrics import Accuracy Load dataset and vocabulary from file You are given two dataset files for training and testing. In [24]: "-" train_dataset = torch.load('./train_dataset.npz') test_dataset = torch.load('./test_dataset.npz') print("Training dataset: %d" % len(train_dataset)) print("Test dataset: %d" % len(test_dataset)) Training dataset: 20000 Test dataset: 5000 Load the vocabulary You are given the vocabulary file. This is used **only** for decoding the integers in the dataset. It is not used for training, nor testing. In [25]: "🔒" vocab = torch.load('./vocab.pt') print("There are %d tokens in vocabulary." % len(vocab)) There are 2000 tokens in vocabulary. Check data In [26]: "-" # @check # @title: data integrity $x, y = train_dataset[100]$ print("Review:", " ".join(vocab.lookup_tokens(x.numpy().tolist()))) print("Label:", y.item()) Review: one of the very best three <unk> <unk> ever . a <unk> the <unk> the <unk> is in top form in the famous in the dark scene . <unk> cunk> cunk> provides excellen t support in his mr . <unk> role as the <unk> of a murder plot . before it s over <unk> s <unk> little <unk> is <unk> to great effect . this minute gem moves about as fast as any <unk> s short and <unk> twice the <unk> . highly r ecommended . <pad> <pad> <pad> Label: 1 Model Construct a model that can process and learn from the samples from dataset. **Hints:** Learn advanced architectural layers: • LSTM: https://pytorch.org/docs/stable/generated/torch.nn.LSTM.html • Conv1D: https://pytorch.org/docs/stable/generated/torch.nn.Conv1d.html • MaxPool1d: https://pytorch.org/docs/stable/generated/torch.nn.MaxPool1d.html • Dropout: https://pytorch.org/docs/stable/generated/torch.nn.Dropout.html In [81]: "🍆" # @workUnit class MyModel(nn.Module): def __init__(self): super().__init__() self.embedding = nn.Embedding(2000,100) self.lstm = nn.LSTM(input_size = 100, hidden_size = 100, $num_layers = 4,$ batch_first = True, dropout=0.2) self.mlp = nn.Sequential(nn.Linear(100,100), nn.ReLU(), nn.Linear(100,2)) def forward(self, tokens): x = self.embedding(tokens) y, (s, c) = self.lstm(x)y = self.mlp(c[0])return y In [82]: "-" # @check # @title: verify model output model = MyModel() dataloader = DataLoader(train_dataset, batch_size=128, shuffle=True) xs, targets = next(iter(dataloader)) model(xs).shape Out[82]: torch.Size([128, 2]) Training Implement a function train that will train the model. Inputs are: • model: an instance of the MyModel • train_dataset: a dataset to be trained on. • epochs: the number of epochs • max batches: optional integer that will limit the number of batches per epoch. Returns a Pandas DataFrame will columns: train_loss and train_acc which are the training loss and accuracy per epoch. Hint: • Start with a simple model, and make sure that you can get a decent performance. • Start with a small number of max_batches to make sure you get a decent training accuracy. • Output debugging message with timing information, so you can estimate the training duration. • For good test accuracy, you need max_matches ~ 500 and 20 epochs or more. In [83]: "🍆" # @workUnit Out[83]: In [84]: def train(model: MyModel,train_dataset: Dataset,epochs: int, max_batches=None) -> pd.DataFrame: device = torch.device('cuda' if torch.cuda.is_available() else 'cpu') dataloader = DataLoader(train_dataset, batch_size=max_batches, shuffle=True) history = { 'train_loss': [], 'train_acc': [], optimizer = torch.optim.Adam(model.parameters()) loss = torch.nn.CrossEntropyLoss() for epoch in range(epochs): l_list **=** [] acc_list = [] accuracy = Accuracy(task='multiclass', num_classes=2) for (xs, targets) in dataloader: xs.to(device) targets.to(device) optimizer.zero_grad() pred = model(xs) 1 = loss(pred, targets) 1.backward() optimizer.step() 1_list.append(l.item()) acc_list.append(accuracy(pred, targets).item()) train_loss, train_acc = np.mean(l_list), np.mean(acc_list) history['train_loss'].append(train_loss) history['train_acc'].append(train_acc) print("train_loss=%.2f train_acc=%.2f" % (train_loss, train_acc)) return pd.DataFrame(history) In [85]: "🍆" # @workUnit # train the model model = MyModel() hist = train(model, train_dataset, epochs=30, max_batches=500) train_loss=0.69 train_acc=0.53 train_loss=0.68 train_acc=0.57 train_loss=0.63 train_acc=0.65 train_loss=0.56 train_acc=0.72 train_loss=0.62 train_acc=0.68 train_loss=0.64 train_acc=0.65 train_loss=0.60 train_acc=0.69 train_loss=0.54 train_acc=0.74 train_loss=0.49 train_acc=0.78 train_loss=0.54 train_acc=0.73 train_loss=0.47 train_acc=0.79 train_loss=0.44 train_acc=0.80 train_loss=0.41 train_acc=0.82 train_loss=0.46 train_acc=0.80 train_loss=0.47 train_acc=0.78 train_loss=0.40 train_acc=0.82 train_loss=0.38 train_acc=0.84 train_loss=0.38 train_acc=0.84 train_loss=0.35 train_acc=0.85 train_loss=0.34 train_acc=0.86 train_loss=0.35 train_acc=0.85 train_loss=0.35 train_acc=0.85 train_loss=0.30 train_acc=0.88 train_loss=0.28 train_acc=0.88 train_loss=0.27 train_acc=0.89 train_loss=0.25 train_acc=0.90 train_loss=0.24 train_acc=0.90 train_loss=0.26 train_acc=0.89 train_loss=0.21 train_acc=0.92 train_loss=0.19 train_acc=0.93 In [86]: "-" # @check # @title: verify dataframe columns hist.columns Index(['train_loss', 'train_acc'], dtype='object') In [87]: "🔒" # Plot the training accuracy hist.train_acc.plot.line(); 0.90 0.85 0.80 0.75 0.70 0.65 0.60 0.55 10 25 In [88]: "=" # Save the entire model to disk torch.save(model, 'mymodel.pt') **Testing** The following code evaluates your model using test_dataset. In [89]: "<mark>-</mark>" # Test your current model in memory test_lib.test_saved_model(model) Saved model has test accuracy = 78.64 In [90]: "-" # Test your saved model on disk test_lib.test_saved_model() Loading from mymodel.pt Saved model has test accuracy = 78.64