Libraries import pandas as pd In [1]: import numpy as np import re In [2]: import spacy from sentence transformers import SentenceTransformer, util In [4]: import torch from PIL import Image import torchvision.transforms as transforms from sklearn.preprocessing import LabelEncoder import torch.nn as nn import torch.nn.functional as F import torch.optim as optim Preprocessing labels = pd.read csv('./labels.csv') In [5]: Loading dataset labels In [6]: Out[6]: **Unnamed:** image name motivational overall sentiment text_ocr text_corrected humour sarcasm offensive 0 LOOK THERE LOOK THERE MY FRIEND MY FRIEND 0 0 image_1.jpg LIGHTYEAR LIGHTYEAR hilarious not_offensive not_motivational general very_positive **NOW ALL NOW ALL** SOHALIK... SOHALIK... The best of The best of #10 #10 not_offensive YearChallenge! YearChallenge! not_funny image_2.jpeg motivational general very_positive Completed in Completed in le... le.. Sam Thorne Sam Thorne @Strippin (@Strippin (2 2 image_3.JPG very_funny not_offensive not_motivational positive not sarcastic Follow Follow Follow Follow Saw every... Saw every... 10 Year 10 Year Challenge -Challenge -3 3 positive image_4.png very_funny twisted_meaning very_offensive motivational Sweet Dee Sweet Dee Edition Edition 10 YEAR 10 YEAR **CHALLENGE CHALLENGE** 4 WITH NO WITH NO image_5.png hilarious very_twisted very_offensive not_motivational neutral FILTER 47 FILTER 47 Hilarious ... Hilarious ... Tuesday is Tuesday is Mardi Gras Mardi Gras 6987 6987 very_funny twisted_meaning very_offensive image_6988.jpg neutral motivational Wednesday is Wednesday is Valentine's... Valentine's... **MUST WATCH MUST WATCH MOVIES OF MOVIES OF** 2017 ITI 6988 6988 image_6989.jpg 2017 ITI funny twisted_meaning not_offensive not_motivational neutral Chennai Chennai memes MA... memes MA... LESS MORE LESS MORE TALKING TALKING 6989 image_6990.png 6989 PLANNING PLANNING funny general slight not_motivational positive SODA JUNK SODA JUNK FOOD COMP... FOOD COMP... When I VERY When I have have time is a time is a 6990 6990 image_6991.jpg not_funny twisted_meaning not_offensive very_positive motivational fantasy No fantasy. no one one has ... has time... The starting The starting point for every point for every not_funny 6991 image_6992.jpg positive not_sarcastic not_offensive motivational good idea is good idea is "Wha... "Wha... 6992 rows × 9 columns Mapping the labels ('overall_sentiment'). 'very_positive' and 'positive' will be merged to 1 label, same goes for negative labels. In [7]: print(labels['overall_sentiment'].unique()) labels['overall_sentiment'] = labels['overall_sentiment'].replace({'very_positive':'positive', 'very_negative': labels['humour'] = labels['humour'].replace({'hilarious':1, 'funny':1, 'very_funny':1, 'not_funny':0}) ['very positive' 'positive' 'neutral' 'negative' 'very negative'] labels['offensive'] = labels['offensive'].replace({'very_offensive':1, 'slight':1, 'hateful_offensive':1, 'not_ In [8]: labels['motivational'] = labels['motivational'].replace({'motivational':1,'not motivational':0}) In [9]: labels['sarcasm'] = labels['sarcasm'].replace({'general':1, 'twisted meaning':1, 'very twisted':1,'not sarcasti In [10]: labels['sarcasm'].unique() In [11]: array([1, 0], dtype=int64) Out[11]: Function preprocessor is used to remove punctuation in the captions and convert the text to lower case. In [12]: def preprocessor(text): text = re.sub($r'[^\w\s]', '', text$) text = text.lower() return text Removing rows that have null values (there were not many). In [13]: labels = labels.dropna() labels = labels.reset index() In [14]: for i in range(len(labels)): labels['text_corrected'][i] = preprocessor(labels['text_corrected'][i]) <ipython-input-14-c51c6e5353a3>:2: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#ret urning-a-view-versus-a-copy labels['text_corrected'][i] = preprocessor(labels['text_corrected'][i]) Selecting the columns needed (images, captions, and label). data used = labels[['image name','text corrected', 'humour', 'sarcasm', 'offensive','motivational', 'overall se In [15]: data used In [16]: Out[16]: text corrected humour sarcasm offensive motivational overall sentiment image_name image_1.jpg look there my friend lightyear now all sohalik... positive positive image_2.jpeg the best of 10 yearchallenge completed in less... 2 image_3.JPG sam thorne strippin follow follow saw everyon... 0 positive 3 10 year challenge sweet dee edition image_4.png positive 4 image_5.png 10 year challenge with no filter 47 hilarious ... 1 0 neutral 6825 image_6988.jpg tuesday is mardi gras wednesday is valentines ... 1 1 neutral must watch movies of 2017 iti chennai memes ma... 6826 image_6989.jpg neutral less more talking planning soda junk food comp... 1 0 image_6990.png positive when i have time is a fantasy no one has time ... 6828 image_6991.jpg positive image_6992.jpg 6829 the starting point for every good idea is what... 0 0 positive 6830 rows × 7 columns In [17]: le = LabelEncoder() data used['overall sentiment'] = le.fit transform(data used['overall sentiment']) <ipython-input-17-648562a388ae>:2: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#ret urning-a-view-versus-a-copy data_used['overall_sentiment'] = le.fit_transform(data_used['overall_sentiment']) Checking the maximum length of captions for padding In [18]: lens = []for i in data used['text corrected']: lens.append(len(i.split())) max(lens) Out[18]: **DataLoader** In [19]: import os from skimage import io from torch.utils.data import Dataset from torch.utils.data import DataLoader Using transformer. In [20]: sentence model = SentenceTransformer('all-MiniLM-L6-v2') Function for padding text. def pad tensor(t): In [21]: t = torch.tensor(t) padding = max(lens) - t.size()[0] t = torch.nn.functional.pad(t, (0, padding)) return t Load data using custom dataloader class. It returns image int form of tensor, transformed text, and transformed label. ${\tt \#reference: https://github.com/aladdinpersson/Machine-Learning-Collection/blob/master/ML/Pytorch/Basics/custom}$ In [22]: class MemeSentimentDataset(Dataset): __init__(self, csv_file, root_dir, transform = None, freq_threshold = 5): self.df = csv file.copy() self.img = self.df['image_name'] self.captions = self.df['text_corrected'] self.root dir = root dir self.transform = transform def __len__(self): return len(self.df) def __getitem__(self, index): image = Image.open(os.path.join(self.root dir, self.df.iloc[index, 0])).convert("RGB") y label1 = torch.tensor(int(self.df.iloc[index, 2])) y label2 = torch.tensor(int(self.df.iloc[index, 3])) y_label3 = torch.tensor(int(self.df.iloc[index, 4])) y label4 = torch.tensor(int(self.df.iloc[index, 5])) y label5 = torch.tensor(int(self.df.iloc[index, 6])) caption = self.captions[index] if self.transform: image = self.transform(image) padded_text_tensor = pad_tensor(sentence_model.encode(caption, convert_to_tensor=True)) return (image, padded_text_tensor, y_label1, y_label2, y_label3, y_label4, y_label5) Transforming the read data. In [23]: mean = [0.4, 0.5, 0.5]std = [0.4, 0.5, 0.5]transform = transforms.Compose([transforms.Resize((100, 100)), transforms.ToTensor(), transforms.Normalize(torch.Tensor(mean), torch.Tensor(std)) In [24]: dataset = MemeSentimentDataset(csv file = data used, root dir = './images', transform = transform) Data split into train and test parts. In [25]: train set, test set = torch.utils.data.random split(dataset, [5000, 1830]) train loader = DataLoader(dataset = train set, batch size = 32, shuffle=True) test loader = DataLoader(dataset = test set, batch size = 32, shuffle=True) In [26]: # for data in train loader: image,text,lab1,lab2,lab3,lab4,lab5 = data print(image.size()) In []: In []: In [27]: # for i, data in enumerate(train loader): i, t, l = dataprint(t.shape) **Neural Networks** Neural Networks for images. In [28]: class NN(nn.Module): def init__(self): super(NN, self).__init__() self.flatten = nn.Flatten() self.linear relu stack = nn.Sequential(nn.Linear(100*100*3, 256),nn.ReLU(inplace=False), nn.Linear(256, 64), nn.ReLU(inplace=False), nn.Linear(64, 32), nn.ReLU(inplace=False), nn.Linear(32, 3)def forward(self, x): x = self.flatten(x)logits = self.linear relu stack(x) return logits Neural Network for text class NN text(nn.Module): def __init__(self): super(NN text, self). init () self.flatten = nn.Flatten() self.linear relu stack = nn.Sequential(nn.Linear(187, 256), nn.ReLU(inplace=False), nn.Linear(256, 64), nn.ReLU(inplace=False), nn.Linear(64, 32), nn.ReLU(inplace=False), nn.Linear(32, 3)def forward(self, x): x = self.flatten(x)logits = self.linear_relu_stack(x) return logits Combining both image and text Neural Networks. In [30]: class Combined model(nn.Module): def __init__(self, modelA, modelB): super(Combined_model, self).__init__() self.modelA = modelA self.modelB = modelB self.classifier = nn.Linear(6, 128) self.classifier1 = nn.Linear(128, 64) self.classifier2 = nn.Linear(64, 32)self.classifier3 = nn.Linear(32, 16)self.out1 = nn.Linear(16, 2,)self.out2 = nn.Linear(16, 2)self.out3 = nn.Linear(16, 2)self.out4 = nn.Linear(16, 2)self.out5 = nn.Linear(16, 3)def forward(self, x1, x2): x1 = self.modelA(x1)x2 = self.modelB(x2)x = torch.cat((x1, x2), dim=1)x = self.classifier(x)x = self.classifier1(x)x = self.classifier2(x)x = self.classifier3(x)out1 = self.out1(x)out2 = self.out2(x)out3 = self.out3(x)out4 = self.out4(x)out5 = self.out5(x)In [31]: net = NN()net_text = NN_text() In [32]: combined_model = Combined_model(net, net_text) In [33]: image = torch.rand(1,3,100,100) text = torch.rand(1,187)out1,out2,out3,out4,out5 = combined_model(image,text) In []: Data training and testing. This also shows losses and accuracies after each epoch. In [39]: #Reference: AI lab task def train_loop(dataloader, model, loss_fn1, optimizer1): size = len(dataloader.dataset) test loss = 0 correct1 = 0 correct2 = 0 correct3 = 0correct4 = 0correct5 = 0for batch, (X, T, y1, y2, y3, y4, y5) in enumerate(dataloader): # Compute prediction and loss pred1, pred2, pred3, pred4, pred5 = model(X,T) loss1 = loss_fn1(pred1, y1) loss1 += loss_fn1(pred2, y2) loss1 += loss fn1(pred3, y3) loss1 += loss_fn1(pred4, y4) loss1 += loss_fn1(pred5, y5) # Backpropagation with torch.autograd.set_detect_anomaly(True): optimizer1.zero_grad() loss1.backward() optimizer1.step() **if** batch % 10 == 0: loss1, current = loss1.item(), batch * len(X) print(f"loss: {loss1:>7f} [{current:>5d}/{size:>5d}]") In [40]: def test_loop(dataloader, model, loss fn1): size = len(dataloader.dataset) num_batches = len(dataloader) test loss = 0 correct1 = 0correct2 = 0correct3 = 0correct4 = 0correct5 = 0with torch.no_grad(): for X, T, y1, y2, y3, y4, y5 in dataloader: pred1, pred2, pred3, pred4, pred5 = model(X,T) test_loss += loss_fn1(pred1, y1).item() test_loss += loss_fn1(pred2, y2).item() test_loss += loss_fn1(pred3, y3).item() test_loss += loss_fn1(pred4, y4).item() test_loss += loss_fn1(pred5, y5).item() correct1 += (pred1.argmax(1) == y1).type(torch.float).sum().item() correct2 += (pred2.argmax(1) == y2).type(torch.float).sum().item()correct3 += (pred3.argmax(1) == y3).type(torch.float).sum().item() correct4 += (pred4.argmax(1) == y4).type(torch.float).sum().item() correct5 += (pred5.argmax(1) == y5).type(torch.float).sum().item() test_loss /= num_batches correct1 /= size correct2 /= size correct3 /= size correct4 /= size correct5 /= size print(f"\n Humour Accuracy: {(100*correct1):>0.1f}%") print(f"\n Sarcasm Accuracy: {(100*correct2):>0.1f}%") print(f"\n Offensive Accuracy: {(100*correct3):>0.1f}%") print(f"\n Motivational Accuracy: {(100*correct4):>0.1f}%") print(f"\n Overall sentiment Accuracy: {(100*correct5):>0.1f}%") I used Adam optimizer and cross entropy loss. In [41]: #model=Combined_model(M1,M2) loss fn1 = nn.CrossEntropyLoss() optimizer1 = torch.optim.Adam(combined_model.parameters(), lr=0.001) In [42]: epochs = 1 for t in range(epochs): print(f"Epoch {t+1}\n----") train_loop(train_loader, combined_model, loss_fn1, optimizer1) test_loop(test_loader, combined_model, loss_fn1) print("Done!") Epoch 1 <ipython-input-21-03565fe0dd97>:2: UserWarning: To copy construct from a tensor, it is recommended to use sourc eTensor.clone().detach() or sourceTensor.clone().detach().requires_grad_(True), rather than torch.tensor(source t = torch.tensor(t) loss: 3.331372 [0/5000] loss: 2.956993 [320/5000] loss: 3.459740 [640/ 5000] loss: 3.223244 [960/5000] loss: 3.061769 [1280/ 5000] loss: 3.064447 [1600/ 5000] loss: 3.354167 [1920/ 5000] loss: 3.184566 [2240/ 5000] loss: 3.313868 [2560/ 5000] loss: 3.441355 [2880/ 5000] loss: 3.434478 [3200/ 5000] loss: 3.630612 [3520/ 5000] loss: 3.178700 [3840/ 5000] loss: 3.067171 [4160/ 5000] loss: 3.031541 [4480/ 5000] loss: 3.217144 [4800/ 5000] Humour Accuracy: 76.0% Sarcasm Accuracy: 78.1% Offensive Accuracy: 59.8% Motivational Accuracy: 66.6% Overall sentiment Accuracy: 59.8% Done! Visualization In [43]: from torchviz import make_dot import os In [44]: os.environ["PATH"] += os.pathsep + 'C:/Program Files/Graphviz/bin/' In [45]: network = next(iter(train loader)) graph=combined_model(network[0],network[1]) make_dot(graph) <ipython-input-21-03565fe0dd97>:2: UserWarning: To copy construct from a tensor, it is recommended to use sourc eTensor.clone().detach() or sourceTensor.clone().detach().requires_grad_(True), rather than torch.tensor(source Tensor). t = torch.tensor(t) Out[45]: d0 Accum **Text reading**