

Project Milestone 3 - Neural Networks

```
import torch
import numpy as np
import pandas as pd
import torch.nn as nn
import torch.nn.functional as F
import matplotlib.pyplot as plt

random_seed = 1
np.random.seed(random_seed)
torch.manual_seed(random_seed)
```

<torch._C.Generator at 0x113492750>

Dataset Import

```
# Glove
glove_df_train = pd.read_csv("../..project_data/data/glove/glove.train.csv")
glove_df_test = pd.read_csv("../..project_data/data/glove/glove.test.csv")
glove_df_eval = pd.read_csv("../..project_data/data/glove/glove.eval.anon.csv")

# Bag of words
bow_df_train = pd.read_csv("../..project_data/data/bag-of-words/bow.train.csv")
bow_df_test = pd.read_csv("../..project_data/data/bag-of-words/bow.test.csv")
bow_df_eval = pd.read_csv("../..project_data/data/bag-of-words/bow.eval.anon.csv")

# TFIDF
tfidf_df_train = pd.read_csv("../..project_data/data/tfidf/tfidf.train.csv")
tfidf_df_test = pd.read_csv("../..project_data/data/tfidf/tfidf.test.csv")
tfidf_df_eval = pd.read_csv("../..project_data/data/tfidf/tfidf.eval.anon.csv")

print("Glove Dataset")
glove_df_train

print("BOW Dataset")
bow_df_train

print("TFIDF Dataset")
tfidf_df_train
```

Glove Dataset
BOW Dataset
TFIDF Dataset

	label	x0	x1	x2	x3	x4	x5	x6	x7
0	1	0.177057	0.063437	0.083603	0.407604	0.119183	0.096883	0.047931	0.117364
1	1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.020774	0.000000
2	0	0.000000	0.109398	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
3	1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
4	1	0.000000	0.000000	0.044455	0.000000	0.000000	0.000000	0.000000	0.000000
...
17495	1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
17496	1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
17497	0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
17498	0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
17499	0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000

17500 rows × 10001 columns

```

class Model(nn.Module):
    def __init__(
        self,
        num_of_features=10000,
        # hidden_layer_1_size=7500,
        hidden_layer_2_size=5000,
        # hidden_layer_3_size=2500,
        hidden_layer_4_size=500,
        num_of_outputs=2,
    ):
        super().__init__()
        # self.forward_connection_1 = nn.Linear(num_of_features, hidden_layer_1_size)
        # self.forward_connection_2 = nn.Linear(hidden_layer_1_size, hidden_layer_2_size)
        # self.forward_connection_3 = nn.Linear(hidden_layer_2_size, hidden_layer_3_size)
        # self.forward_connection_4 = nn.Linear(hidden_layer_3_size, hidden_layer_4_size)
        self.forward_connection_1 = nn.Linear(num_of_features, hidden_layer_2_size)
        self.dropout = nn.Dropout(p=0.25)
        self.forward_connection_3 = nn.Linear(hidden_layer_2_size, hidden_layer_4_size)
        self.dropout = nn.Dropout(p=0.25)
        self.output_connection = nn.Linear(hidden_layer_4_size, num_of_outputs)

    def forward(self, example):
        example = F.tanh(self.forward_connection_1(example))
        # example = F.relu(self.forward_connection_2(example))
        example = F.tanh(self.forward_connection_3(example))
        # example = F.relu(self.forward_connection_4(example))
        example = F.tanh(self.output_connection(example))

        return example

```

```

nn_relu_bow_model = Model()
nn_relu_tfidf_model = Model()
nn_relu_glove_model = Model(
    num_of_features=300,
    # hidden_layer_1_size=200,
    hidden_layer_2_size=200,
    # hidden_layer_3_size=50,
    hidden_layer_4_size=50,
    num_of_outputs=2,
)

```

```

def plot_loss(loss, epochs, label):
    if not loss or type(loss) is not list:
        print(f"Can't plot loss curve. Invalid loss: {loss}")
        return

    fig = plt.figure()

    # Plot the data
    plt.plot(epochs, loss, label="Loss Epoch Curve")

    # plt.xticks(epochs)
    # plt.yticks([50, 60, 70, 80, 90])

    # Label the x-axis & y-axis
    plt.xlabel("Epochs")
    plt.ylabel("Loss Values")

    plt.legend()

    # Add title to graph
    plt.title(f"{label}")

    # Save the figure
    fig.savefig(f"figs/{0}.png".format(label))

```

```

epochs = 400

```

```

glove_dataset_setup_dict = {
    "Glove": {"Train": glove_df_train, "Test": glove_df_test, "Eval": glove_df_eval},
    # "BOW": {"Train": bow_df_train, "Test": bow_df_test, "Eval": bow_df_eval},
    # "TFIDF": {"Train": tfidf_df_train, "Test": tfidf_df_test, "Eval": tfidf_df_eval}
}

bow_dataset_setup_dict = {
    # "Glove": {"Train": glove_df_train, "Test": glove_df_test, "Eval": glove_df_eval},
    "BOW": {"Train": bow_df_train, "Test": bow_df_test, "Eval": bow_df_eval},
    # "TFIDF": {"Train": tfidf_df_train, "Test": tfidf_df_test, "Eval": tfidf_df_eval}
}

tfidf_dataset_setup_dict = {

```

```

# "Glove": {"Train": glove_df_train, "Test": glove_df_test, "Eval": glove_df_eval},
# "BOW": {"Train": bow_df_train, "Test": bow_df_test, "Eval": bow_df_eval},
"TFIDF": {"Train": tfidf_df_train, "Test": tfidf_df_test, "Eval": tfidf_df_eval},
}

optimizers = [adam_optimizer]
loss_functions = [cross_entropy_loss_func]
models = [nn_relu_model, nn_relu_glove_model]

def train_setup(dataset_setup_dict, model):
    for name, dataset in dataset_setup_dict.items():
        print(f"\nNN on Dataset {name}")
        X_Train = dataset["Train"].drop("label", axis=1)
        Y_Train = dataset["Train"]["label"]

        X_Train = X_Train.values
        Y_Train = Y_Train.values

        X_Train = torch.FloatTensor(X_Train)
        Y_Train = torch.LongTensor(Y_Train)

        # Function to be used to measure error or how far the model is from prediction
        cross_entropy_loss_func = nn.CrossEntropyLoss()

        # Control learning rate throughout the epochs for optimizing learning
        adam_optimizer = torch.optim.Adam(model.parameters(), lr=0.0001)

        losses = []
        for epoch in range(1, epochs + 1):
            predict = model.forward(X_Train)
            loss = cross_entropy_loss_func(predict, Y_Train)

            losses.append(loss.detach().numpy())
            if epoch % 10 == 0:
                print(f" Epoch {epoch} Loss {loss}")

            # Back propagation to fine tune the weights
            adam_optimizer.zero_grad()
            loss.backward()
            adam_optimizer.step()

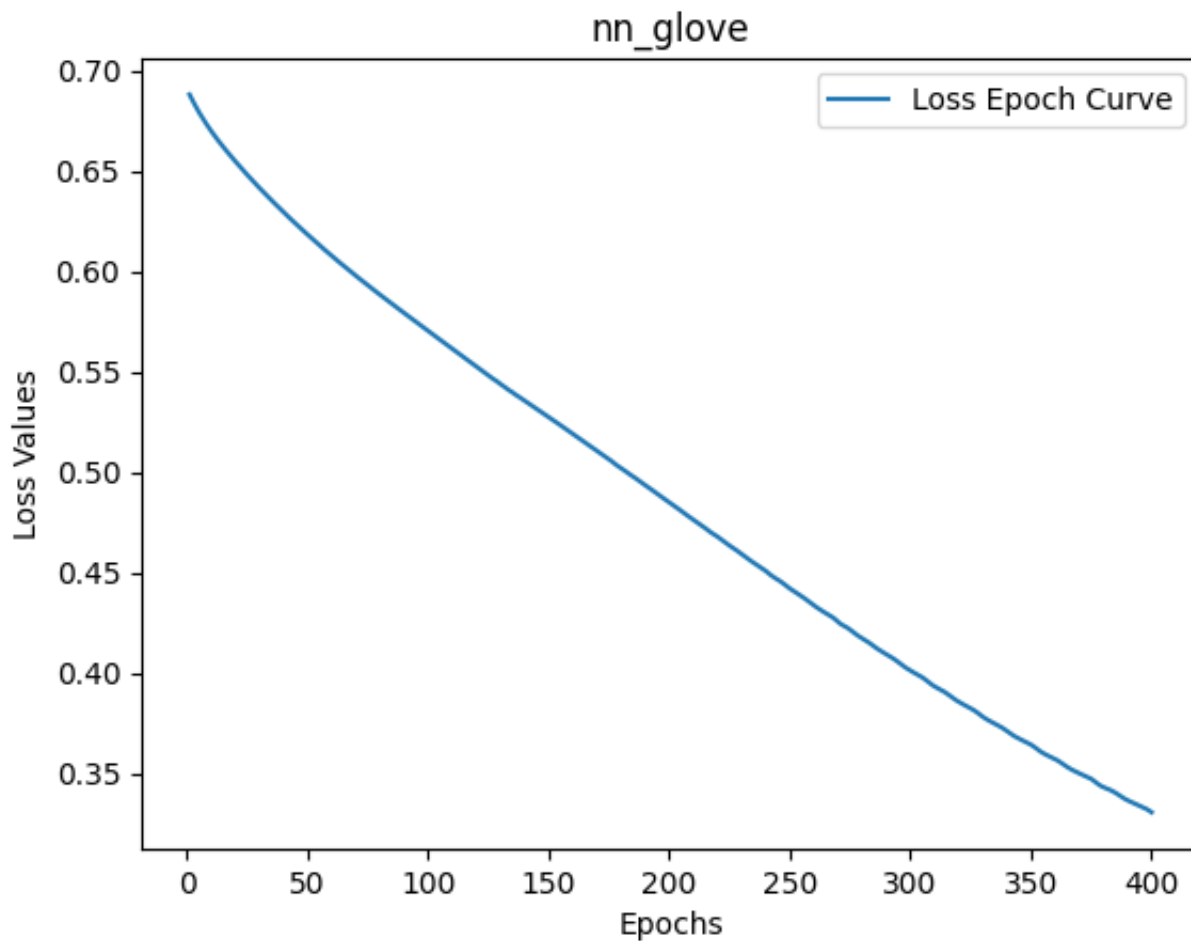
        plot_loss(loss=losses, epochs=range(1, epochs + 1), label=f"nn_{name.lower()}")

train_setup(glove_dataset_setup_dict, nn_relu_glove_model)
# train_setup(bow_dataset_setup_dict, nn_relu_bow_model)
# train_setup(tfidf_dataset_setup_dict, nn_relu_tfidf_model)

```

NN on Dataset Glove

Epoch 10	Loss	0.6697897911071777
Epoch 20	Loss	0.6547130346298218
Epoch 30	Loss	0.641475260257721
Epoch 40	Loss	0.6294039487838745
Epoch 50	Loss	0.6182059049606323
Epoch 60	Loss	0.6076284646987915
Epoch 70	Loss	0.5977135896682739
Epoch 80	Loss	0.5882777571678162
Epoch 90	Loss	0.5791840553283691
Epoch 100	Loss	0.5702342987060547
Epoch 110	Loss	0.5613484978675842
Epoch 120	Loss	0.5524449348449707
Epoch 130	Loss	0.5436106324195862
Epoch 140	Loss	0.5354931950569153
Epoch 150	Loss	0.5273266434669495
Epoch 160	Loss	0.5189746022224426
Epoch 170	Loss	0.5104275941848755
Epoch 180	Loss	0.5019893646240234
Epoch 190	Loss	0.4934263527393341
Epoch 200	Loss	0.4849076569080353
Epoch 210	Loss	0.4762338101863861
Epoch 220	Loss	0.46776145696640015
Epoch 230	Loss	0.45904406905174255
Epoch 240	Loss	0.45071637630462646
Epoch 250	Loss	0.4419417083263397
Epoch 260	Loss	0.43351784348487854
Epoch 270	Loss	0.4253121316432953
Epoch 280	Loss	0.41722992062568665
Epoch 290	Loss	0.40921419858932495
Epoch 300	Loss	0.4010772705078125
Epoch 310	Loss	0.39319464564323425
Epoch 320	Loss	0.38555556535720825
Epoch 330	Loss	0.3780970275402069
Epoch 340	Loss	0.3709421455860138
Epoch 350	Loss	0.3642127215862274
Epoch 360	Loss	0.35705217719078064
Epoch 370	Loss	0.3499416708946228
Epoch 380	Loss	0.3430417776107788
Epoch 390	Loss	0.3365381360054016
Epoch 400	Loss	0.3305492401123047



```
def test_setup(name, dataset_setup_dict, model):
    for name, dataset in dataset_setup_dict.items():
        # Turn off back propagation to only get the prediction.
        with torch.no_grad():
            X_Test = dataset["Test"].drop("label", axis=1)
            Y_Test = dataset["Test"]["label"]

            X_Test = X_Test.values
            Y_Test = Y_Test.values

            X_Test = torch.FloatTensor(X_Test)
            Y_Test = torch.LongTensor(Y_Test)

            # Test Loss
            Y_Test_Eval = model.forward(X_Test)
            loss = cross_entropy_loss_func(Y_Test_Eval, Y_Test)
            print(f" {name} Test dataset Loss {loss}")

            correct_predictions = 0
            total = X_Test.shape[0]
            for index, example in enumerate(X_Test):
                prediction = model.forward(example)
                if prediction.argmax().item() == Y_Test[index]:
                    correct_predictions += 1

            print(f" {name} Accuracy: {correct_predictions / total}\n")
```

```
test_setup(name="NN-Glove", dataset_setup_dict=glove_dataset_setup_dict, model=nn_relu)
# test_setup(name="NN-BOW", dataset_setup_dict=bow_dataset_setup_dict, model=nn_relu)
# test_setup(name="NN-TFIDF", dataset_setup_dict=tfidf_dataset_setup_dict, model=nn_relu)
```

Glove Test dataset Loss 0.6063986420631409

Glove Accuracy: 0.6746666666666666

```
def export_prediction_to_csv(name, prediction_list):
    df = pd.DataFrame(prediction_list)
    df.to_csv(f"results/{name.lower()}.csv", index=True, index_label="example_id", header=True)
    print(f" {name} Predictions Saved.\n")
```

```
def eval_setup(name, dataset_setup_dict, model):
    for name, dataset in dataset_setup_dict.items():
        # Turn off back propagation to only get the prediction.
        with torch.no_grad():
            X_Eval = dataset["Eval"].drop("label", axis=1)
            Y_Eval = dataset["Eval"]["label"]

            X_Eval = X_Eval.values
            Y_Eval = Y_Eval.values

            X_Eval = torch.FloatTensor(X_Eval)
            Y_Eval = torch.LongTensor(Y_Eval)

            prediction_list = []
            for index, example in enumerate(X_Eval):
                prediction = model.forward(example)
                prediction_list.append(prediction.argmax().item())

    return prediction_list
```

```
glove_prediction_list = eval_setup(
    name="NN-Glove-tanh", dataset_setup_dict=glove_dataset_setup_dict, model=nn_relu_tanh
)
export_prediction_to_csv(name="NN-Glove-tanh", prediction_list=glove_prediction_list)

# bow_prediction_list = eval_setup(name="NN-BOW", dataset_setup_dict=bow_dataset_setup_dict, model=nn_relu_bow)
# export_prediction_to_csv(name="NN-BOW", prediction_list=bow_prediction_list)

# tfidf_prediction_list = eval_setup(
#     name="NN-TFIDF", dataset_setup_dict=tfidf_dataset_setup_dict, model=nn_relu_tfidf
# )
# export_prediction_to_csv(name="NN-TFIDF", prediction_list=tfidf_prediction_list)
```

NN-Glove-tanh Predictions Saved.