```
import torch
from torch import nn
import torch.optim as optim
from torch.utils.data import Dataset
from torchvision import datasets
from torchvision.transforms import ToTensor
import time
from sklearn.model_selection import KFold
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import torch.nn.functional as F
```

Problem 1

```
train path = r"C:\Users\ccm51\OneDrive\Desktop\ECGR 4106\fashion-mnist train.csv"
In [53]:
          test path = r"C:\Users\ccm51\OneDrive\Desktop\ECGR 4106\fashion-mnist test.csv"
          train_numpy = np.loadtxt(train_path, dtype = np.float32, delimiter = ",", skiprows=1)
          test_numpy = np.loadtxt(test_path, dtype = np.float32, delimiter = ",", skiprows=1)
           print(train numpy.shape)
          (60000, 785)
In [75]:
          def try_gpu(i=0):
              if torch.cuda.device count() >= i+1:
                   return torch.device(f'cuda:{i}')
               return torch.device('cpu')
In [226...
          def train model(model, total train, total test, loss fn, optim, epochs, update freq):
              train loss hist = []
              train_acc_hist = []
              test_acc_hist = []
              main tic = time.perf counter()
              for epoch in range(1, epochs + 1):
                   tic = time.perf_counter()
                   loss train = 0
                   correct_train = 0
                   for imgs in train loader:
                       labels = imgs[:,-1].long()
                       imgs = imgs[:,:-1]
                       batch size = imgs.shape[0]
                       outputs = model(imgs.view(batch size, -1))
                       _, predicted = torch.max(outputs, dim = 1)
                       correct_train += int((predicted == labels).sum())
                       loss = loss_fn(outputs, labels)
                       optim.zero_grad()
                       loss.backward()
                       optim.step()
                       loss train += loss.item()
                   toc = time.perf counter()
```

```
correct val = 0
                  with torch.no_grad():
                      for imgs in val_loader:
                           labels = imgs[:,-1].long()
                           imgs = imgs[:,:-1]
                           batch size = imgs.shape[0]
                          outputs = model(imgs.view(batch_size, -1))
                           _, predicted = torch.max(outputs, dim = 1)
                          correct val += int((predicted == labels).sum())
                   if epoch == 1 or epoch == epochs or epoch % update freq == 0:
                      print(f"Epoch {epoch}:\n\tDuration = {round(toc - tic, 3)} seconds\n\tTrai
                  train loss hist.append(round(loss train / len(train loader), 5))
                  train acc hist.append(round(correct train/total train, 3))
                  test_acc_hist.append(round(correct_val/total_test, 3))
              main toc = time.perf counter()
              print(f"\nTotal Training Time = {round(main toc - main tic, 3)} seconds\nAverage 1
              return train_loss_hist, train_acc_hist, test_acc_hist
In [216...
          def plot_model(title, fig_num, loss_hist, train_hist, test_hist, leg_loc):
              plot = plt.figure(fig num)
              x = range(1, len(loss hist)+1)
              plt.plot(x, loss_hist)
              plt.plot(x, train hist)
              plt.plot(x, test_hist)
              plt.legend(["Training Loss", "Training Accuracy", "Testing Accuracy"], loc=leg_loc
              plt.xlabel('Epoch')
              plt.ylabel('Error/Accuracy')
              plt.title(title)
          train tensor = torch.from numpy(train numpy)
In [228...
          test tensor = torch.from numpy(test numpy)
          train features = train tensor[:,1:]
          test_features = test_tensor[:,1:]
          train label = torch.unsqueeze(train tensor[:,0], dim=-1)
          test_label = torch.unsqueeze(test_tensor[:,0], dim=-1)
          train_set = torch.cat((train_features, train_label), dim=1).to(device=try_gpu())
          test set = torch.cat((test features, test label), dim=1).to(device=try gpu())
          train_loader = torch.utils.data.DataLoader(train_set, batch_size=600, shuffle=True)
          val_loader = torch.utils.data.DataLoader(test_set, batch_size=test_set.shape[0], shuf-
In [229...
          model 1 = nn.Sequential(
                      nn.Identity(),
                                       # Using Identity Layers as place-holders for Dropout Laye
                      nn.Linear(784, 1024), # First Hidden Layer
                      nn.ReLU(),
                      nn.Identity(),
                      nn.Linear(1024, 512), # Second Hidden Layer
                      nn.ReLU(),
                      nn.Identity(),
                      nn.Linear(512, 256), # Third Hidden Layer
                      nn.ReLU(),
                      nn.Linear(256, 10)).to(device=try_gpu()) # Output Layer
```

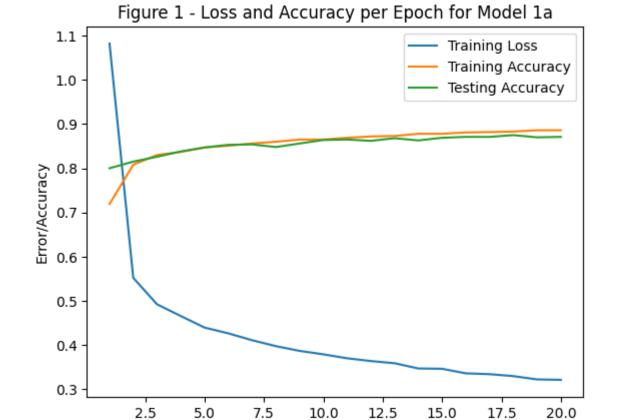
```
optimizer 1 = optim.SGD(model 1.parameters(), lr=1e-3)
          model_1.eval()
          Sequential(
Out[229]:
             (0): Identity()
             (1): Linear(in features=784, out features=1024, bias=True)
             (2): ReLU()
             (3): Identity()
             (4): Linear(in features=1024, out features=512, bias=True)
             (5): ReLU()
             (6): Identity()
             (7): Linear(in features=512, out features=256, bias=True)
             (8): ReLU()
             (9): Linear(in_features=256, out_features=10, bias=True)
  In [ ]: # class Module_1a(nn.Module):
                 def __init__(self):
                     super().__init__()
           #
                     self.fc1 = nn.Linear(784, 1024)
                     self.act1 = nn.ReLU()
           #
           #
                     self.fc2 = nn.Linear(1024, 512)
           #
                     self.act2 = nn.ReLU()
                     self.fc3 = nn.Linear(512, 256)
                     self.act3 = nn.ReLU()
           #
                     self.fc4 = nn.Linear(256, 10)
           #
                 def forward(self, x):
           #
                     out = self.act1(self.fc1(x))
                     out = self.act2(self.fc2(out))
           #
                     out = self.act3(self.fc3(out))
                     out = self.fc4(out)
                     return out
          t_loss_hist_a, t_acc_hist_a, v_acc_hist_a = train_model(model_1,
In [230...
                                                                    train_set.shape[0],
                                                                    test set.shape[0],
                                                                    nn.CrossEntropyLoss(),
                                                                    optimizer_1,
                                                                    20,
                                                                     5)
```

```
Epoch 1:
        Duration = 1.662 seconds
        Training Loss: 1.08243
        Training Accuracy: 0.719
        Validation Accuracy: 0.8
Epoch 5:
        Duration = 1.523 seconds
        Training Loss: 0.43935
        Training Accuracy: 0.847
        Validation Accuracy: 0.847
Epoch 10:
        Duration = 1.544 seconds
        Training Loss: 0.37876
        Training Accuracy: 0.865
        Validation Accuracy: 0.864
Epoch 15:
        Duration = 1.58 seconds
        Training Loss: 0.34601
        Training Accuracy: 0.878
        Validation Accuracy: 0.869
Epoch 20:
        Duration = 1.563 seconds
        Training Loss: 0.321
        Training Accuracy: 0.886
        Validation Accuracy: 0.871
Total Training Time = 34.137 seconds
```

Average Training Time per Epoch = 1.707 seconds

In [231... title_1 = "Figure 1 - Loss and Accuracy per Epoch for Model 1a"

plot_model(title_1, 1, t_loss_hist_a, t_acc_hist_a, v_acc_hist_a, 'upper right')



Epoch

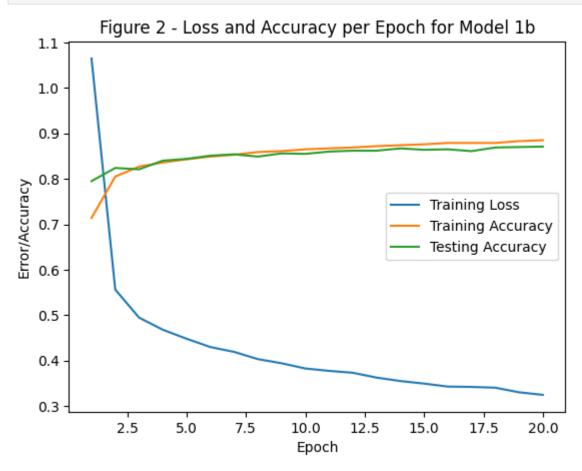
```
torch.save(model_1.state_dict(), "homework_0-model_1.params")
In [232...
          model 2 = nn.Sequential(
In [234...
                       nn.Identity(),
                       nn.Linear(784, 1024), # First Hidden Layer
                       nn.ReLU(),
                       nn.Identity(),
                       nn.Linear(1024, 512), # Second Hidden Layer
                       nn.ReLU(),
                       nn.Identity(),
                       nn.Linear(512, 256), # Third Hidden Layer
                       nn.ReLU(),
                       nn.Linear(256, 10)).to(device=try gpu()) # Output Layer
           optimizer 2 = optim.SGD(model 2.parameters(), lr = 1e-3, weight decay=1e-4)
          model 2.eval()
          Sequential(
Out[234]:
             (0): Identity()
             (1): Linear(in features=784, out features=1024, bias=True)
             (2): ReLU()
             (3): Identity()
             (4): Linear(in_features=1024, out_features=512, bias=True)
             (5): ReLU()
             (6): Identity()
             (7): Linear(in features=512, out features=256, bias=True)
             (8): ReLU()
             (9): Linear(in features=256, out features=10, bias=True)
In [235...
          t loss hist b, t acc hist b, v acc hist b = train model(model 2,
                                                                    train_set.shape[0],
                                                                    test set.shape[0],
                                                                    nn.CrossEntropyLoss(),
                                                                    optimizer_2,
                                                                    20,
                                                                    5)
```

```
Epoch 1:
        Duration = 1.694 seconds
        Training Loss: 1.06484
        Training Accuracy: 0.714
        Validation Accuracy: 0.795
Epoch 5:
        Duration = 1.852 seconds
        Training Loss: 0.44831
        Training Accuracy: 0.843
        Validation Accuracy: 0.844
Epoch 10:
        Duration = 1.772 seconds
        Training Loss: 0.3828
        Training Accuracy: 0.865
        Validation Accuracy: 0.855
Epoch 15:
        Duration = 1.693 seconds
        Training Loss: 0.34954
        Training Accuracy: 0.876
        Validation Accuracy: 0.864
Epoch 20:
        Duration = 1.829 seconds
        Training Loss: 0.32481
        Training Accuracy: 0.885
        Validation Accuracy: 0.871
Total Training Time = 37.19 seconds
```

Total Training Time = 37.19 seconds

Average Training Time per Epoch = 1.86 seconds

In [236...
title_2 = "Figure 2 - Loss and Accuracy per Epoch for Model 1b"
plot_model(title_2, 2, t_loss_hist_b, t_acc_hist_b, v_acc_hist_b, 'right')



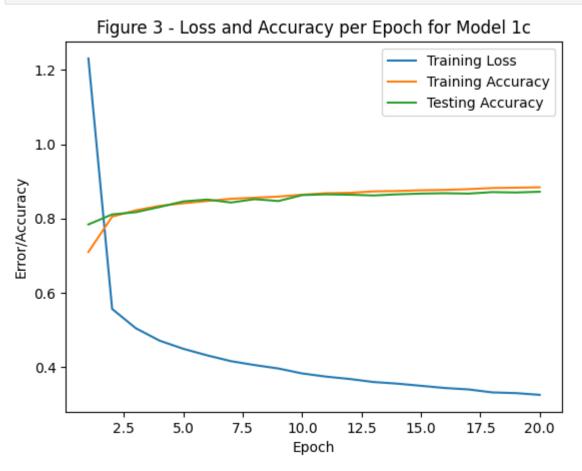
```
model_3 = nn.Sequential(
In [237...
                       nn.Dropout(p=0.3),
                       nn.Linear(784, 1024), # First Hidden Layer
                       nn.ReLU(),
                       nn.Dropout(p=0.3),
                       nn.Linear(1024, 512), # Second Hidden Layer
                       nn.ReLU(),
                       nn.Dropout(p=0.3),
                       nn.Linear(512, 256), # Third Hidden Layer
                       nn.ReLU(),
                       nn.Linear(256, 10)).to(device=try_gpu()) # Output Layer
          optimizer 3 = optim.SGD(model 3.parameters(), lr = 1e-3)
          model_3.eval()
          Sequential(
Out[237]:
            (0): Dropout(p=0.3, inplace=False)
            (1): Linear(in_features=784, out_features=1024, bias=True)
            (2): ReLU()
            (3): Dropout(p=0.3, inplace=False)
             (4): Linear(in features=1024, out features=512, bias=True)
            (5): ReLU()
             (6): Dropout(p=0.3, inplace=False)
            (7): Linear(in features=512, out features=256, bias=True)
            (8): ReLU()
            (9): Linear(in_features=256, out_features=10, bias=True)
          t loss hist c, t acc hist c, v acc hist c = train model(model 3,
In [238...
                                                                   train_set.shape[0],
                                                                    test set.shape[0],
                                                                    nn.CrossEntropyLoss(),
                                                                    optimizer 3,
                                                                    20,
                                                                    5)
```

```
Epoch 1:
        Duration = 2.05 seconds
        Training Loss: 1.23049
        Training Accuracy: 0.71
        Validation Accuracy: 0.784
Epoch 5:
        Duration = 1.537 seconds
        Training Loss: 0.44936
        Training Accuracy: 0.841
        Validation Accuracy: 0.846
Epoch 10:
        Duration = 1.564 seconds
        Training Loss: 0.38318
        Training Accuracy: 0.864
        Validation Accuracy: 0.863
Epoch 15:
        Duration = 1.516 seconds
        Training Loss: 0.34995
        Training Accuracy: 0.876
        Validation Accuracy: 0.867
Epoch 20:
        Duration = 1.525 seconds
        Training Loss: 0.32561
        Training Accuracy: 0.884
        Validation Accuracy: 0.872
Total Training Time = 34.5 seconds
```

Average Training Time = 34.5 Seconds

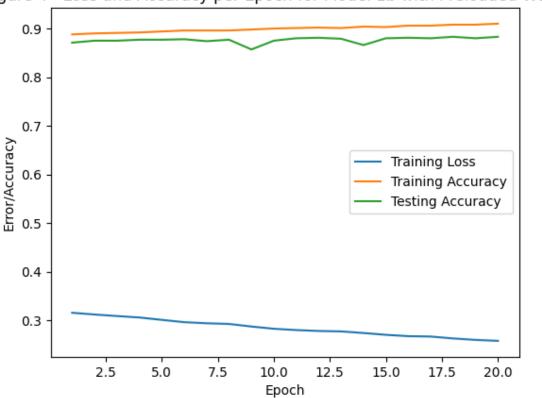
Average Training Time per Epoch = 1.725 seconds

In [246...
title_3 = "Figure 3 - Loss and Accuracy per Epoch for Model 1c"
plot_model(title_3, 3, t_loss_hist_c, t_acc_hist_c, v_acc_hist_c, 'upper right')



```
model_4 = model_2
In [240...
          model_4.load_state_dict(torch.load("homework_0-model_1.params"))
          <All keys matched successfully>
Out[240]:
          t_loss_hist_d, t_acc_hist_d, v_acc_hist_d = train_model(model_4,
In [241...
                                                                    train set.shape[0],
                                                                    test set.shape[0],
                                                                    nn.CrossEntropyLoss(),
                                                                    optimizer_2,
                                                                    20,
                                                                    5)
          Epoch 1:
                  Duration = 1.678 seconds
                  Training Loss: 0.31526
                  Training Accuracy: 0.888
                  Validation Accuracy: 0.871
          Epoch 5:
                  Duration = 1.532 seconds
                  Training Loss: 0.30065
                  Training Accuracy: 0.894
                  Validation Accuracy: 0.877
          Epoch 10:
                  Duration = 1.607 seconds
                  Training Loss: 0.28236
                  Training Accuracy: 0.9
                  Validation Accuracy: 0.875
          Epoch 15:
                  Duration = 1.687 seconds
                  Training Loss: 0.26986
                  Training Accuracy: 0.903
                  Validation Accuracy: 0.88
          Epoch 20:
                  Duration = 1.579 seconds
                  Training Loss: 0.25747
                  Training Accuracy: 0.91
                  Validation Accuracy: 0.883
          Total Training Time = 35.706 seconds
          Average Training Time per Epoch = 1.785 seconds
          title 4 = "Figure 4 - Loss and Accuracy per Epoch for Model 1b with Preloaded Weights"
In [242...
           plot_model(title_4, 4, t_loss_hist_d, t_acc_hist_d, v_acc_hist_d, 'right')
```

Figure 4 - Loss and Accuracy per Epoch for Model 1b with Preloaded Weights

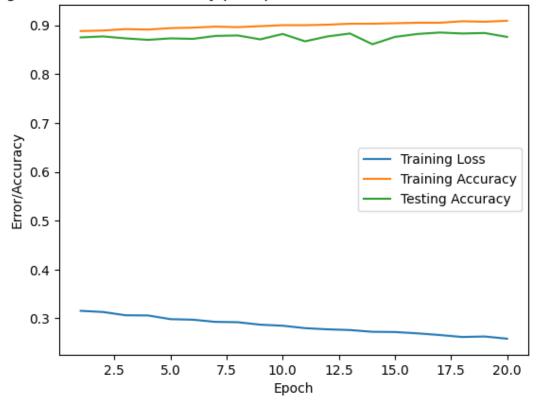


```
Epoch 1:
        Duration = 1.554 seconds
        Training Loss: 0.31538
        Training Accuracy: 0.888
        Validation Accuracy: 0.875
Epoch 5:
        Duration = 1.441 seconds
        Training Loss: 0.29839
        Training Accuracy: 0.894
        Validation Accuracy: 0.873
Epoch 10:
        Duration = 1.567 seconds
        Training Loss: 0.28498
        Training Accuracy: 0.9
        Validation Accuracy: 0.882
Epoch 15:
        Duration = 1.555 seconds
        Training Loss: 0.2721
        Training Accuracy: 0.904
        Validation Accuracy: 0.876
Epoch 20:
        Duration = 1.506 seconds
        Training Loss: 0.25842
        Training Accuracy: 0.909
        Validation Accuracy: 0.876
Total Training Time = 34.356 seconds
```

Average Training Time per Epoch = 1.718 seconds

In [245...
title_5 = "Figure 5 - Loss and Accuracy per Epoch for Model 1c with Preloaded Weights"
plot_model(title_5, 5, t_loss_hist_e, t_acc_hist_e, v_acc_hist_e, 'right')

Figure 5 - Loss and Accuracy per Epoch for Model 1c with Preloaded Weights



Problem 2

```
house_train_path = r"C:\Users\ccm51\OneDrive\Desktop\ECGR 4106\housing_train.csv"
In [271...
          house test path = r"C:\Users\ccm51\OneDrive\Desktop\ECGR 4106\housing test.csv"
          house_raw_train = pd.read_csv(house_train_path, delimiter = ",", header=0)
          house_raw_test = pd.read_csv(house_test_path, delimiter = ",", header=0)
          label = 'SalePrice'
In [272...
          features = pd.concat([house raw train.drop(columns=['Id',label]),
                                house raw test.drop(columns=['Id'])])
          numeric features = features.dtypes[features.dtypes != 'object'].index
          features[numeric features] = features[numeric features].apply(
              lambda x: (x - x.mean()) / (x.std()))
          features[numeric_features] = features[numeric_features].fillna(0)
          features = pd.get_dummies(features, dummy_na=True)
          house_train_dataframe = features[:house_raw_train.shape[0]].copy()
          house train dataframe[label] = house raw train[label]
          house_test_dataframe = features[house_raw_train.shape[0]:].copy()
          def loss_1(predictions, labels):
In [544...
              outputs = torch.square(torch.log10(labels) - torch.log10(torch.squeeze(predictions
              output = torch.sqrt(torch.mean(outputs))
              # print(predictions.shape, labels.shape, predictions, labels)
              # print(torch.squeeze(predictions).shape, outputs, output)
              return output.requires grad ()
          def plot model kfold(title, fig num, train loss, test loss, leg loc):
In [550...
              plot = plt.figure(fig_num)
              x = range(1, len(train loss)+1)
              plt.plot(x, train_loss)
              plt.plot(x, test_loss)
              plt.legend(["Training Loss", "Testing Loss"], loc=leg_loc)
              plt.xlabel('Epoch')
              plt.ylabel('Error')
              plt.title(title)
          house train tensor = torch.tensor(house train dataframe.values).to(device=try gpu()).1
In [311...
          house test tensor = torch.tensor(house test dataframe.values).to(device=try gpu()).to(
          def train_model_kfold(model, train_tensor, batch, loss_fn, optim, epochs, update_freq
In [545...
              kf = KFold(n splits=k folds, shuffle=True, random state=0)
              main_tic = time.perf_counter()
              train loss hist = []
              test_loss_hist = []
              for epoch in range(1, epochs + 1):
                  train loss epoch = []
                  test loss epoch = []
                  tic = time.perf_counter()
                  for i, (train index, test index) in enumerate(kf.split(train tensor)):
                      train, test = train_tensor[train_index], train_tensor[test_index]
```

```
t loader = torch.utils.data.DataLoader(train, batch size=batch, shuffle=Tr
                      v loader = torch.utils.data.DataLoader(test, batch size=test.shape[0], shu
                      loss train = 0
                      loss_val = 0
                      for imgs in t_loader:
                           labels = imgs[:,-1]
                           imgs=imgs[:,:-1]
                           batch_size = imgs.shape[0]
                           outputs = model(imgs.view(batch size, -1))
                           loss t = loss fn(outputs, labels)
                           optim.zero grad()
                           loss_t.backward()
                           optim.step()
                           loss train += loss t.item()
                      with torch.no grad():
                           for imgs in v_loader:
                              labels = imgs[:,-1]
                              imgs = imgs[:,:-1]
                              batch size = imgs.shape[0]
                              outputs = model(imgs.view(batch_size, -1))
                              loss v = loss fn(outputs, labels)
                              loss val += loss v.item()
                      train_loss_epoch.append(round(4*loss_train/len(train), 5))
                      test loss epoch.append(round(loss val, 5))
                  toc = time.perf counter()
                  train loss hist.append(sum(train loss epoch)/len(train loss epoch))
                  test_loss_hist.append(sum(test_loss_epoch)/len(test_loss_epoch))
                  if epoch == 1 or epoch == epochs or epoch % update freq == 0:
                      print(f"Epoch {epoch}:\n\tDuration = {round(toc - tic, 3)} seconds\n\tAver
              main_toc = time.perf_counter()
              print(f"\nTotal Training Time = {round(main toc - main tic, 3)} seconds\nAverage 1
              return train loss hist, test loss hist
          model_house_1 = nn.Sequential(
In [548...
                      nn.Identity(),
                      nn.LazyLinear(512), # First Hidden Layer
                      nn.ReLU(),
                      nn.Identity(),
                      nn.Linear(512, 256), # Second Hidden Layer
                      nn.ReLU(),
                      nn.Identity(),
                      nn.Linear(256, 128), # Third Hidden Layer
                      nn.ReLU(),
                      nn.Linear(128, 1)).to(device=try_gpu()) # Output Layer
          optimizer house 1 = optim.SGD(model house 1.parameters(), lr = 1e-3)
          model_house_1.eval()
```

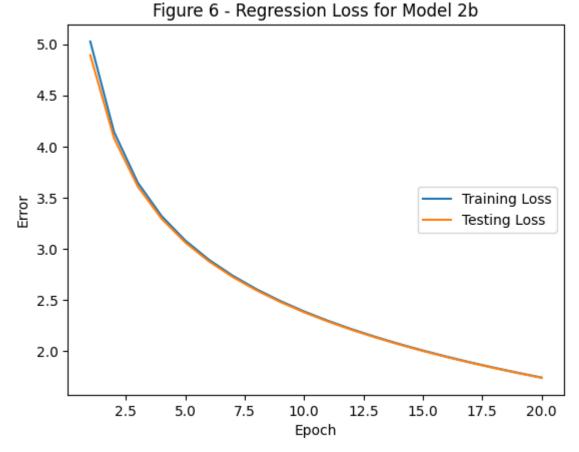
```
Sequential(
Out[548]:
             (0): Identity()
             (1): LazyLinear(in_features=0, out_features=512, bias=True)
             (2): ReLU()
             (3): Identity()
             (4): Linear(in_features=512, out_features=256, bias=True)
             (5): ReLU()
             (6): Identity()
             (7): Linear(in_features=256, out_features=128, bias=True)
             (8): ReLU()
             (9): Linear(in_features=128, out_features=1, bias=True)
          t_loss_hist_house_a, v_loss_hist_house_a = train_model_kfold(model_house_1,
In [549...
                                                                        house train tensor,
                                                                        4,
                                                                        loss_1,
                                                                        optimizer_house_1,
                                                                        20,
                                                                        1,
                                                                        5)
```

```
Epoch 1:
        Duration = 2.776 seconds
        Average Training Loss: 5.02613
        Average Validation Loss: 4.89212
Epoch 2:
        Duration = 2.525 seconds
        Average Training Loss: 4.14203
        Average Validation Loss: 4.08028
Epoch 3:
        Duration = 2.583 seconds
        Average Training Loss: 3.64616
        Average Validation Loss: 3.60849
Epoch 4:
        Duration = 2.533 seconds
        Average Training Loss: 3.32055
        Average Validation Loss: 3.29435
Epoch 5:
        Duration = 2.514 seconds
        Average Training Loss: 3.08113
        Average Validation Loss: 3.06126
Epoch 6:
        Duration = 2.52 seconds
        Average Training Loss: 2.89232
        Average Validation Loss: 2.87653
Epoch 7:
        Duration = 2.831 seconds
        Average Training Loss: 2.7368
        Average Validation Loss: 2.72374
Epoch 8:
        Duration = 2.695 seconds
        Average Training Loss: 2.6046
        Average Validation Loss: 2.59356
Epoch 9:
        Duration = 2.652 seconds
        Average Training Loss: 2.48965
        Average Validation Loss: 2.48022
Epoch 10:
        Duration = 2.669 seconds
        Average Training Loss: 2.38798
        Average Validation Loss: 2.37988
Epoch 11:
        Duration = 2.724 seconds
        Average Training Loss: 2.29704
        Average Validation Loss: 2.28989
Epoch 12:
        Duration = 3.121 seconds
        Average Training Loss: 2.21465
        Average Validation Loss: 2.20833
Epoch 13:
        Duration = 3.249 seconds
        Average Training Loss: 2.13945
        Average Validation Loss: 2.13376
Epoch 14:
        Duration = 2.779 seconds
        Average Training Loss: 2.07004
        Average Validation Loss: 2.06508
Epoch 15:
        Duration = 2.823 seconds
        Average Training Loss: 2.00594
        Average Validation Loss: 2.00144
```

```
Epoch 16:
        Duration = 3.841 seconds
        Average Training Loss: 1.94625
        Average Validation Loss: 1.94216
Epoch 17:
        Duration = 3.622 seconds
        Average Training Loss: 1.89023
        Average Validation Loss: 1.88668
Epoch 18:
        Duration = 3.915 seconds
        Average Training Loss: 1.83785
        Average Validation Loss: 1.83454
Epoch 19:
        Duration = 3.348 seconds
        Average Training Loss: 1.78829
        Average Validation Loss: 1.78537
Epoch 20:
        Duration = 3.083 seconds
        Average Training Loss: 1.74152
        Average Validation Loss: 1.73886
Total Training Time = 58.804 seconds
Average Training Time per Epoch = 2.94 seconds
```

In [559...
title_6 = "Figure 6 - Regression Loss for Model 2b"
plot_model_kfold(title_6, 6, t_loss_hist_house_a, v_loss_hist_house_a, 'right')



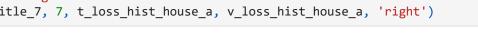


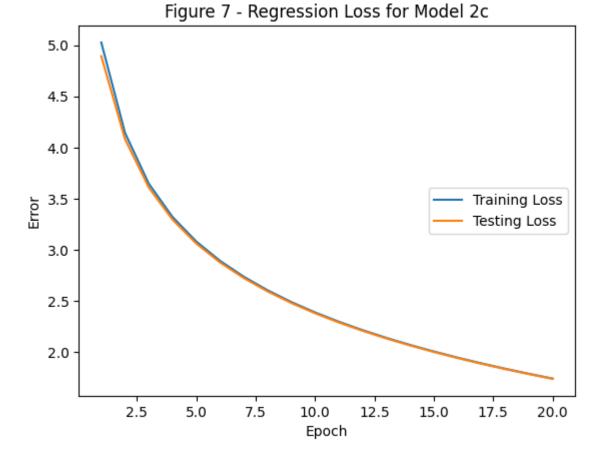
```
nn.Dropout(p=0.3),
                       nn.Linear(512, 256), # Second Hidden Layer
                       nn.ReLU(),
                       nn.Dropout(p=0.3),
                       nn.Linear(256, 128), # Third Hidden Layer
                       nn.ReLU(),
                       nn.Linear(128, 1)).to(device=try_gpu()) # Output Layer
          optimizer_house_2 = optim.SGD(model_house_2.parameters(), lr = 1e-3, weight_decay=1e-4
          model_house_2.eval()
          Sequential(
Out[555]:
             (0): Dropout(p=0.3, inplace=False)
            (1): LazyLinear(in_features=0, out_features=512, bias=True)
            (2): ReLU()
            (3): Dropout(p=0.3, inplace=False)
            (4): Linear(in_features=512, out_features=256, bias=True)
            (5): ReLU()
            (6): Dropout(p=0.3, inplace=False)
            (7): Linear(in_features=256, out_features=128, bias=True)
            (8): ReLU()
            (9): Linear(in_features=128, out_features=1, bias=True)
          t_loss_hist_house_b, v_loss_hist_house_b = train_model_kfold(model_house_2,
In [556...
                                                                       house train tensor,
                                                                       4,
                                                                       loss_1,
                                                                       optimizer_house_2,
                                                                       20,
                                                                       1,
                                                                        5)
```

```
Epoch 1:
        Duration = 2.86 seconds
        Average Training Loss: 4.94739
        Average Validation Loss: 4.80709
Epoch 2:
        Duration = 2.836 seconds
        Average Training Loss: 4.07012
        Average Validation Loss: 4.01156
Epoch 3:
        Duration = 2.81 seconds
        Average Training Loss: 3.59933
        Average Validation Loss: 3.56339
Epoch 4:
        Duration = 3.096 seconds
        Average Training Loss: 3.28708
        Average Validation Loss: 3.26182
Epoch 5:
        Duration = 2.879 seconds
        Average Training Loss: 3.05532
        Average Validation Loss: 3.03613
Epoch 6:
        Duration = 3.168 seconds
        Average Training Loss: 2.87162
        Average Validation Loss: 2.85622
Epoch 7:
        Duration = 3.148 seconds
        Average Training Loss: 2.71951
        Average Validation Loss: 2.70683
Epoch 8:
        Duration = 2.804 seconds
        Average Training Loss: 2.58987
        Average Validation Loss: 2.57918
Epoch 9:
        Duration = 2.976 seconds
        Average Training Loss: 2.47702
        Average Validation Loss: 2.4678
Epoch 10:
        Duration = 2.938 seconds
        Average Training Loss: 2.37706
        Average Validation Loss: 2.36903
Epoch 11:
        Duration = 2.851 seconds
        Average Training Loss: 2.28738
        Average Validation Loss: 2.28034
Epoch 12:
        Duration = 2.778 seconds
        Average Training Loss: 2.20612
        Average Validation Loss: 2.19987
Epoch 13:
        Duration = 2.918 seconds
        Average Training Loss: 2.13176
        Average Validation Loss: 2.12623
Epoch 14:
        Duration = 2.819 seconds
        Average Training Loss: 2.0633
        Average Validation Loss: 2.05837
Epoch 15:
        Duration = 3.014 seconds
        Average Training Loss: 1.9998
        Average Validation Loss: 1.99545
```

```
Epoch 16:
        Duration = 2.861 seconds
        Average Training Loss: 1.94076
        Average Validation Loss: 1.9368
Epoch 17:
        Duration = 2.787 seconds
        Average Training Loss: 1.88549
        Average Validation Loss: 1.8819
Epoch 18:
        Duration = 2.991 seconds
        Average Training Loss: 1.8335
        Average Validation Loss: 1.83029
Epoch 19:
        Duration = 3.275 seconds
        Average Training Loss: 1.78446
        Average Validation Loss: 1.78159
Epoch 20:
        Duration = 3.338 seconds
        Average Training Loss: 1.73803
        Average Validation Loss: 1.73552
Total Training Time = 59.147 seconds
Average Training Time per Epoch = 2.957 seconds
```

title 7 = "Figure 7 - Regression Loss for Model 2c" In [560... plot_model_kfold(title_7, 7, t_loss_hist_house_a, v_loss_hist_house_a, 'right')





```
preds = model house 2(house test tensor)
In [580...
           ensemble preds = torch.exp(torch.squeeze(preds))
           submission = pd.DataFrame({'Id':house_raw_test.Id,
```

```
'SalePrice':ensemble_preds.detach().cpu().numpy()})
submission.to_csv('submission.csv', index=False)

torch.Size([1459, 1])
torch.Size([1459])
```