## mark\_goldstein\_mg3479\_A2\_code

## February 24, 2020

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In [ ]: # Import dependencies
        import torch
        import torch.nn as nn
        from plot_lib import set_default, show_scatterplot, plot_bases
        import matplotlib.pyplot as plt
        import random
        import numpy as np
In [ ]: # Set up your device
        cuda = torch.cuda.is_available()
        device = torch.device("cuda:0" if cuda else "cpu")
In [ ]: # Set up random seed to 1008. Do not change the random seed.
        # Yes, these are all necessary when you run experiments!
        seed = 1008
        random.seed(seed)
        np.random.seed(seed)
        torch.manual_seed(seed)
        if cuda:
            torch.cuda.manual_seed(seed)
            torch.cuda.manual_seed_all(seed)
            torch.backends.cudnn.benchmark = False
            torch.backends.cudnn.deterministic = True
In [ ]: # Define data generating functions
        def quadratic_data_generator(data_size):
            \# f(x) = y = x^2 + 4x - 3
            # generate an input tensor of size data_size with torch.randn
            x = torch.randn(data_size, 1) - 2
            x = x.to(device)
            # TODO
            111
            y = \dots
            111
            # placeholder
            y = torch.ones(data_size,1)
            return x,y
```

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def cubic_data_generator(data_size=100):
            # f(x) = y = x^3 + 4x^2 - 3
            # generate an input tensor of size data_size with torch.randn
            x = torch.randn(data_size, 1) - 2
            x = x.to(device)
            # TODO
            111
            y = \dots
            111
            # placeholder
            y = torch.ones(data_size,1)
            return x, y
In [ ]: # Generate the data with 128 datapoints
        x, y = quadratic_data_generator(128)
        plt.scatter(x,y)
        plt.show()
In [ ]: # Define a Linear Classifier with a single linear layer and no non-linearity
        # (no hidden layer)
        class Linear_OH(nn.Module):
            def __init__(self):
                super().__init__()
                # TODO
                self.classifer = None
            def forward(self, x):
                return self.classifier(x)
In []: # Define a Linear Classifier with a single hidden layer of size 5 and ReLU non-linearity
        class Linear_1H(nn.Module):
            def __init__(self):
                super().__init__()
                # TODO
                self.classifer = None
            def forward(self, x):
                return self.classifier(x)
In [ ]: # Define a Linear Classifier with a two hidden layers of size 5 and ReLU non-linearity
        class Linear_2H(nn.Module):
            def __init__(self):
                super().__init__()
                self.classifer = None
```

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def forward(self, x):
                return self.classifier(x)
In []: '''
        TODO: Training function
        Hint: look at some example pytorch tutorials to learn how to
            - initialize optimizers
            - zero gradient
            - backprop the loss
            - step the gradient
        Note: This is full batch. We compute forward on whole x,y.
        No need for dataloaders nor loop over batches.
        Just pass all of x to model's forward pass.
        def train(model, epochs, x, y):
            # Set model to training mode
            model.train()
            # Define MSE loss function
            criterion = None
            # TODO: Define the SGD optimizer with learning rate 0.01
            optimizer = None
            for epoch in range(epochs):
                # TODO: Forward data through model to predict y
                y_pred = None
                # TODO: Compute loss in terms of predicted and true y
                loss = None
                # TODO: Zero gradient
                # TODO: call backward on loss
                # TODO: step the optimizer
                # every 100 epochs, print
                if (epoch+1) \% 100 == 0:
                    print('Epoch {} loss: {}'.format(epoch+1, loss.item()))
            \# return y\_pred without gradient information, for plotting
            return y_pred.detach()
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In [ ]: # OH model on quadratic data
        model = Linear_OH()
        y_pred = train(model, epochs=1000, x=x, y=y)
        # Plot predictions vs actual data
        plt.scatter(x, y)
        plt.scatter(x, y_pred)
        plt.show()
In []: # 1H model on quadratic data
        model = Linear_1H()
        y_pred = train(model, epochs=1000, x=x, y=y)
        plt.scatter(x, y)
        plt.scatter(x, y_pred)
        plt.show()
In [ ]: # 2H model on quadratic data
        model = Linear_2H()
        y_pred = train(model, epochs=1000, x=x, y=y)
        plt.scatter(x, y)
        plt.scatter(x, y_pred)
        plt.show()
In []: # Generate cubic data with 128 data points
        x, y = cubic_data_generator(128)
In []: # OH model on cubic data
        model = Linear_OH()
        y_pred = train(model, epochs=1000, x=x, y=y)
        plt.scatter(x, y)
        plt.scatter(x, y_pred)
        plt.show()
In [ ]: # 1H model on cubic data
        model = Linear_1H()
        y_pred = train(model, epochs=1000, x=x, y=y)
        plt.scatter(x, y)
        plt.scatter(x, y_pred)
        plt.show()
In [ ]: # 2H model on cubic data
        model = Linear_2H()
        y_pred = train(model, epochs=1000, x=x, y=y)
        plt.scatter(x, y)
        plt.scatter(x, y_pred)
        plt.show()
```