## ECGR\_4105\_HW5\_Problem\_1\_Source\_Code

## November 16, 2024

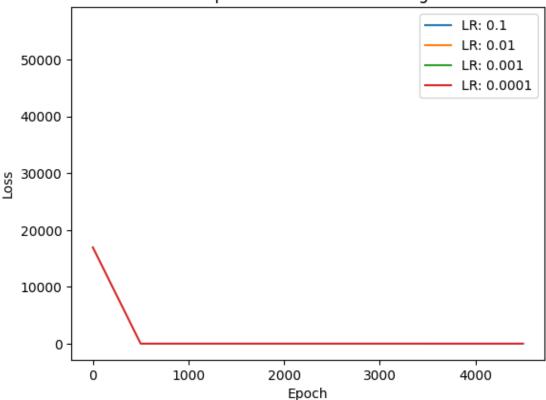
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[]: # Link to Google Colab: https://colab.research.google.com/drive/
      → 1uuVb2mNPxP1AjOUYkQ3Zse1hhXMfgxj1?usp=sharing
[]: import torch
     import torch.nn as nn
     import torch.optim as optim
     import matplotlib.pyplot as plt
[]: t_u = torch.linspace(0, 10, 100) # input values
     y_actual = 3.5 * t_u ** 2 + 2 * t_u + 0.5
[]: # linear model to non-linear model
     class NonLinearModel(nn.Module):
         def __init__(self):
             super(NonLinearModel, self).__init__()
             self.w2 = nn.Parameter(torch.randn(()))
             self.w1 = nn.Parameter(torch.randn(()))
             self.b = nn.Parameter(torch.randn(()))
         def forward(self, t_u):
             return self.w2 * t_u ** 2 + self.w1 * t_u + self.b
[]: | # function to train non-linear
     def train_model(learning_rate, epochs=5000, log_interval=500):
         model = NonLinearModel()
         criterion = nn.MSELoss()
         optimizer = optim.SGD(model.parameters(), lr=learning_rate)
         losses = []
         for epoch in range(epochs):
             optimizer.zero_grad()
             output = model(t_u)
             loss = criterion(output, y_actual)
             loss.backward()
             optimizer.step()
             if epoch % log_interval == 0:
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losses.append(loss.item())
                 print(f"Epoch {epoch}, Loss: {loss.item()}")
         return model, losses
[]: class LinearModel(nn.Module):
         def __init__(self):
             super(LinearModel, self).__init__()
             self.w = nn.Parameter(torch.randn(()))
             self.b = nn.Parameter(torch.randn(()))
         def forward(self, t_u):
             return self.w * t_u + self.b
[]: # function to train linear for comparison
     def train_linear_model(learning_rate=0.01, epochs=5000):
         model = LinearModel()
         criterion = nn.MSELoss()
         optimizer = optim.SGD(model.parameters(), lr=learning_rate)
         for epoch in range(epochs):
             optimizer.zero_grad()
             output = model(t_u)
             loss = criterion(output, y_actual)
             loss.backward()
             optimizer.step()
         return model
[]: # Training with different learning rates
     learning_rates = [0.1, 0.01, 0.001, 0.0001]
     all_losses = {}
[]: for lr in learning_rates:
         print(f"\nTraining with learning rate: {lr}")
         model, losses = train_model(lr)
         all_losses[lr] = losses
    Training with learning rate: 0.1
    Epoch 0, Loss: 52230.57421875
    Epoch 500, Loss: nan
    Epoch 1000, Loss: nan
    Epoch 1500, Loss: nan
    Epoch 2000, Loss: nan
    Epoch 2500, Loss: nan
    Epoch 3000, Loss: nan
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Epoch 3500, Loss: nan
    Epoch 4000, Loss: nan
    Epoch 4500, Loss: nan
    Training with learning rate: 0.01
    Epoch 0, Loss: 56430.42578125
    Epoch 500, Loss: nan
    Epoch 1000, Loss: nan
    Epoch 1500, Loss: nan
    Epoch 2000, Loss: nan
    Epoch 2500, Loss: nan
    Epoch 3000, Loss: nan
    Epoch 3500, Loss: nan
    Epoch 4000, Loss: nan
    Epoch 4500, Loss: nan
    Training with learning rate: 0.001
    Epoch 0, Loss: 35192.46875
    Epoch 500, Loss: nan
    Epoch 1000, Loss: nan
    Epoch 1500, Loss: nan
    Epoch 2000, Loss: nan
    Epoch 2500, Loss: nan
    Epoch 3000, Loss: nan
    Epoch 3500, Loss: nan
    Epoch 4000, Loss: nan
    Epoch 4500, Loss: nan
    Training with learning rate: 0.0001
    Epoch 0, Loss: 16955.552734375
    Epoch 500, Loss: 2.676705837249756
    Epoch 1000, Loss: 1.656842827796936
    Epoch 1500, Loss: 1.0265941619873047
    Epoch 2000, Loss: 0.6370910406112671
    Epoch 2500, Loss: 0.3963528871536255
    Epoch 3000, Loss: 0.2475438266992569
    Epoch 3500, Loss: 0.15553589165210724
    Epoch 4000, Loss: 0.09863147884607315
    Epoch 4500, Loss: 0.06342044472694397
[]: # Plotting the losses for each learning rate
     for lr, losses in all_losses.items():
         plt.plot(range(0, 5000, 500), losses, label=f"LR: {lr}")
     plt.xlabel("Epoch")
     plt.ylabel("Loss")
     plt.legend()
```

```
plt.title("Loss vs. Epoch for Different Learning Rates")
plt.show()
```

## Loss vs. Epoch for Different Learning Rates



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[]: linear_model = train_linear_model()

# Plotting the models for comparison
plt.plot(t_u, y_actual, label="Actual Data")
plt.plot(t_u, model(t_u).detach(), label="Non-Linear Model", linestyle="--")
plt.plot(t_u, linear_model(t_u).detach(), label="Linear Model", linestyle="-.")
plt.xlabel("t_u (Input)")
plt.ylabel("Prediction")
plt.legend()
plt.title("Non-Linear vs. Linear Model Predictions")
plt.show()
```



