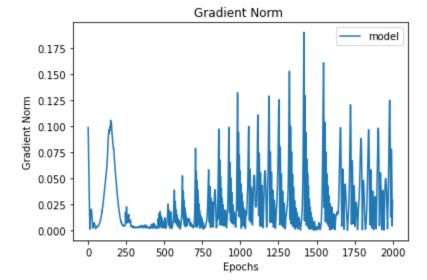
```
In [1]:
            import torch
            import torch.nn as nn
            import torch.nn.functional as F
            from torch.utils.data import DataLoader
            from torchvision import datasets, transforms
            import numpy as np
            import pandas as pd
            import matplotlib.pyplot as plt
            %matplotlib inline
  In [2]:
            X = torch.linspace(-1,1,200).reshape(-1,1)
  In [3]:
            y=(np.sin(5*(np.pi*X)))/((5*(np.pi*X)))
  In [4]:
            class Model(nn.Module):
                def __init__(self, inp=1, h2=5, h3=10, h4=10, h5=10, h6=10, h7=10, h8=5, outp=1):
                    super().__init__()
                    self.fc1 = nn.Linear(inp,h2)
                    self.fc2 = nn.Linear(h2, h3)
                    self.fc3 = nn.Linear(h3, h4)
                    self.fc4 = nn.Linear(h4, h5)
                    self.fc5 = nn.Linear(h5, h6)
                    self.fc6 = nn.Linear(h6, h7)
                    self.fc7 = nn.Linear(h7, h8)
                    self.out = nn.Linear(h8, outp)
                def forward(self, x):
                    x = F.relu(self.fcl(x))
                    x = F.relu(self.fc2(x))
                    x = F.relu(self.fc3(x))
                    x = F.relu(self.fc4(x))
                    x = F.relu(self.fc5(x))
                    x = F.relu(self.fc6(x))
                    x = F.relu(self.fc7(x))
                    x = self.out(x)
                    return x
  In [5]:
            model = Model()
  In [6]:
            criterion = nn.MSELoss()
  In [7]:
            optimizer = torch.optim.Adam(model.parameters(), lr=0.001)
  In [8]:
            epochs = 2000
            losses = []
            grad_norm_list=[]
            for i in range(epochs):
                i+=1
                # feed forwarding
Loading [MathJax]/extensions/Safe.js
```

```
y_pred = model.forward(X)
              # calculate loss
              tloss = criterion(y pred, y)
              losses.append(tloss)
              optimizer.zero grad()
              tloss.backward()
              optimizer.step()
              # Get gradient norm (From slides)
              grad all = 0.0
              for p in model.parameters():
                  grad = 0.0
                  if p.grad is not None:
                       grad = (p.grad.cpu().data.numpy() ** 2).sum()
                  grad all += grad
              grad norm = grad all ** 0.5
              grad norm list.append(grad norm)
 In [9]:
          losses=torch.tensor(losses)
In [10]:
          plt.plot(range(epochs), losses.numpy(), 'r', label='model')
          plt.title("Loss")
          plt.xlabel("Epochs")
          plt.ylabel("Loss")
          plt.legend(loc="upper right")
          plt.show()
                                    Loss
                                                       model
           0.08
           0.06
         S 0.04
            0.02
           0.00
                     250
                          500
                               750
                                    1000 1250
                                             1500 1750 2000
                                   Epochs
In [11]:
          #Gradient norm to iterations
          plt.plot(range(epochs), grad_norm_list,label='model')
          plt.title("Gradient Norm ")
          plt.xlabel("Epochs")
          plt.ylabel("Gradient Norm")
          plt.legend(loc="upper right")
          plt.show()
```



In [ ]: