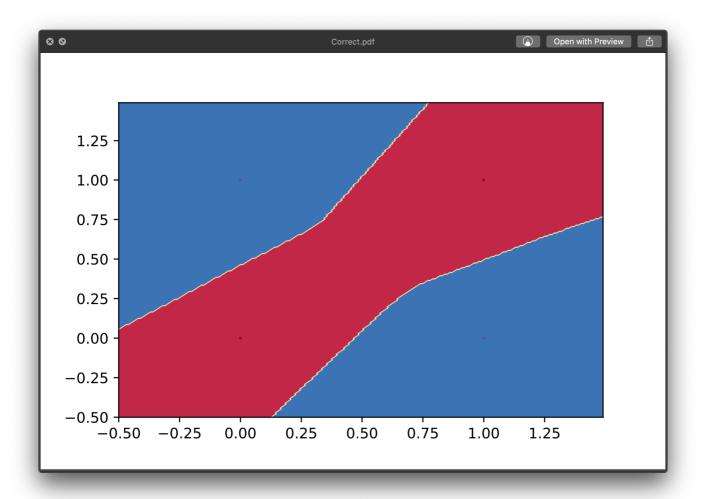
```
In [1]:
         #Question 1
In [2]:
         import torch
         import torch.nn as nn
         import torch.nn.functional as F
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
In [3]:
         class Net(nn.Module):
             def init (self):
                 super(Net, self).__init__()
                 self.fc1 = nn.Linear(2, 20)
                 self.fc2 = nn.Linear(20, 20)
                 self.fc3 = nn.Linear(20, 20)
                 self.fc4 = nn.Linear(20, 20)
                 self.fc5 = nn.Linear(20, 20)
                 self.fc6 = nn.Linear(20, 20)
                 self.fc7 = nn.Linear(20, 2)
             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 = self.fc7(x)
                 return F.log softmax(x)
                 #return F.softmax(x)
```

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```
In [1]:
         #Question 1
In [2]:
         import torch
         import torch.nn as nn
         import torch.nn.functional as F
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
In [3]:
         class Net(nn.Module):
             def init (self):
                 super(Net, self).__init__()
                 self.fc1 = nn.Linear(2, 20)
                 self.fc2 = nn.Linear(20, 20)
                 self.fc3 = nn.Linear(20, 20)
                 self.fc4 = nn.Linear(20, 20)
                 self.fc5 = nn.Linear(20, 20)
                 self.fc6 = nn.Linear(20, 20)
                 self.fc7 = nn.Linear(20, 2)
             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 = self.fc7(x)
                 return F.log softmax(x)
                 #return F.softmax(x)
```

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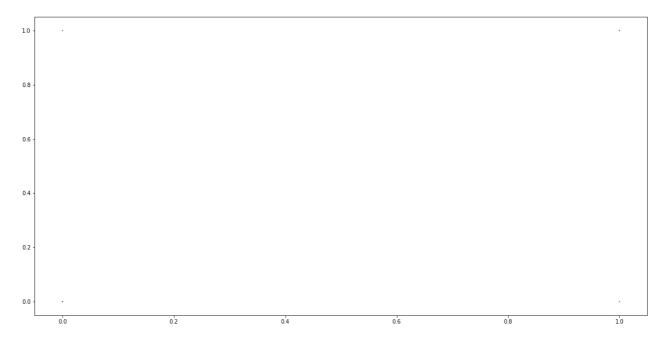
```
In [4]:
                           #%% plot function
                             def plot_data(X, y, filename):
                                         plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Spectral, s = 1)
                                         plt.savefig(filename)
                                         plt.close()
                             def plot decision boundary(clf, X, y, filename):
                                          # Set min and max values and give it some padding
                                          \#x \min, x \max = X[:, 0].\min() - .1, X[:, 0].\max() + .1
                                          \#y \min_{x \in X} y \max_{x \in X} = X[:, 1].\min_{x \in X} (y - .1, X[:, 1].\max_{x \in X} (y + .1)
                                         x_{min}, x_{max} = -0.5, 1.5
                                         y_{min}, y_{max} = -0.5, 1.5
                                         h = 0.01
                                         # Generate a grid of points with distance h between them
                                         xx, yy = np.meshgrid(np.arange(x_min, x_max, h), np.arange(y_min, y_max, h), np.arange
                                          # Predict the function value for the whole gid
                                         #Z = clf.predict(np.c [xx.ravel(), yy.ravel()])
                                         X_out = net(torch.tensor(np.c_[xx.ravel(), yy.ravel()], dtype = torch.;
                                         Z = X \text{ out.data.max}(1)[1]
                                         # Z.shape
                                          Z = Z.reshape(xx.shape)
                                          # Plot the contour and training examples
                                         plt.contourf(xx, yy, Z, cmap=plt.cm.Spectral)
                                         plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Spectral, s = 1)
                                         plt.savefig(filename)
                                         plt.close()
```

```
In [5]: #%% read data

data = pd.read_csv("XOR.csv")
X = data.values[:, 0:2] # Take only the first two features.
X = torch.tensor(X, dtype = torch.float)
y = data.values[:, 2]
y = torch.tensor(y, dtype = torch.long)

plot_data(X,y,'data.pdf')
plt.figure(figsize=(20,10))
plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Spectral, s = 1)
plt.show()
```

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```
In [6]: data
```

Out[6]: Input A Input B Output 0 0 0 0 0 1 0 1 1 2 1 0 1 3 1 1 0

```
In [7]: #%% train
   net = Net()

# create a stochastic gradient descent optimizer
   learning_rate = .01
   optimizer = torch.optim.SGD(net.parameters(), lr=learning_rate, momentum=0)

# create a loss function
#criterion = nn.CrossEntropyLoss()
   criterion = nn.NLLLoss()
```

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```
In [8]:
         # nepochs = 10000
         nepochs = 3000 #10000
         data, target = X, y
         # run the main training loop
         for epoch in range(nepochs):
              adjust learning rate if desired
         #
              if epoch % 3000 == 0 and epoch <= 24000:
         #
                  for g in optimizer.param groups:
         #
                      g['lr'] = g['lr']/2
             optimizer.zero_grad()
             # forward propagate
             net_out = net(data)
             # compute loss
             loss = criterion(net_out, target)
             # backpropagate
             loss.backward()
             # update parameters
             optimizer.step()
             # print out report
             if epoch % 10 == 0:
                 print('Epoch ', epoch, 'Loss ', loss.item())
                 net out = net(data)
                 pred = net_out.data.max(1)[1] # get the index of the max log-prob
                 correctidx = pred.eq(target.data)
                 ncorrect = correctidx.sum()
                 accuracy = ncorrect.item()/len(data)
                 if accuracy== 1:
                     break
                 print('Training accuracy is ', accuracy)
```

<ipython-input-3-b68d425386b7>:21: UserWarning: Implicit dimension choice f
or log_softmax has been deprecated. Change the call to include dim=X as an
argument.

```
return F.log_softmax(x)
Epoch 0 Loss 0.6941758394241333
Training accuracy is 0.5
Epoch 10 Loss 0.6936168670654297
Training accuracy is 0.5
Epoch 20 Loss 0.6931634545326233
Training accuracy is 0.5
Epoch 30 Loss 0.6930881142616272
Training accuracy is 0.5
Epoch 40 Loss 0.6930862665176392
Training accuracy is 0.5
Epoch 50 Loss 0.6930666565895081
Training accuracy is 0.5
Epoch 60 Loss 0.6930466890335083
Training accuracy is 0.5
Epoch 70 Loss 0.6930317282676697
Training accuracy is 0.75
Epoch 80 Loss 0.6930174827575684
Training accuracy is 0.5
Epoch 90 Loss 0.6930090188980103
Training accuracy is 0.5
Epoch 100 Loss 0.6930021047592163
```

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Training accuracy is 0.75 Epoch 110 Loss 0.6929948925971985 Training accuracy is 0.75 Epoch 120 Loss 0.6929873824119568 Training accuracy is 0.75 Epoch 130 Loss 0.6929792761802673 Training accuracy is 0.75 Epoch 140 Loss 0.6929664015769958 Training accuracy is 0.75 Epoch 150 Loss 0.6929500102996826 Training accuracy is 0.75 Epoch 160 Loss 0.6929318904876709 Training accuracy is 0.75 Epoch 170 Loss 0.692912757396698 Training accuracy is 0.75 Epoch 180 Loss 0.6928925514221191 Training accuracy is 0.75 Epoch 190 Loss 0.6928713321685791 Training accuracy is 0.75 Epoch 200 Loss 0.6928490996360779 Training accuracy is 0.75 Epoch 210 Loss 0.692834734916687 Training accuracy is 0.75 Epoch 220 Loss 0.69282066822052 Training accuracy is 0.75 Epoch 230 Loss 0.692805826663971 Training accuracy is 0.75 Epoch 240 Loss 0.692790150642395 Training accuracy is 0.75 Epoch 250 Loss 0.6927738189697266 Training accuracy is 0.75 Epoch 260 Loss 0.6927566528320312 Training accuracy is 0.75 Epoch 270 Loss 0.6927385330200195 Training accuracy is 0.75 Epoch 280 Loss 0.6927194595336914 Training accuracy is 0.75 Epoch 290 Loss 0.6926991939544678 Training accuracy is 0.75 Epoch 300 Loss 0.6926789283752441 Training accuracy is 0.75 Epoch 310 Loss 0.6926573514938354 Training accuracy is 0.75 Epoch 320 Loss 0.6926344633102417 Training accuracy is 0.75 Epoch 330 Loss 0.6926104426383972 Training accuracy is 0.75 Epoch 340 Loss 0.6925852298736572 Training accuracy is 0.75 Epoch 350 Loss 0.6925604343414307 Training accuracy is 0.75 Epoch 360 Loss 0.692534327507019 Training accuracy is 0.75 Epoch 370 Loss 0.6925068497657776 Training accuracy is 0.75 Epoch 380 Loss 0.6924779415130615 Training accuracy is 0.75 Epoch 390 Loss 0.692447304725647 Training accuracy is 0.75 Epoch 400 Loss 0.6924149990081787 Training accuracy is 0.75

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Epoch 410 Loss 0.6923807263374329 Training accuracy is 0.75 Epoch 420 Loss 0.6923444271087646 Training accuracy is 0.75 Epoch 430 Loss 0.6923058032989502 Training accuracy is 0.75 Epoch 440 Loss 0.6922647356987 Training accuracy is 0.75 Epoch 450 Loss 0.6922210454940796 Training accuracy is 0.75 Epoch 460 Loss 0.6921743154525757 Training accuracy is 0.75 Epoch 470 Loss 0.6921243667602539 Training accuracy is 0.75 Epoch 480 Loss 0.6920709609985352 Training accuracy is 0.75 Epoch 490 Loss 0.6920138001441956 Training accuracy is 0.75 Epoch 500 Loss 0.6919522881507874 Training accuracy is 0.75 Epoch 510 Loss 0.6918848752975464 Training accuracy is 0.75 Epoch 520 Loss 0.6918103098869324 Training accuracy is 0.75 Epoch 530 Loss 0.6917284727096558 Training accuracy is 0.75 Epoch 540 Loss 0.6916390657424927 Training accuracy is 0.75 Epoch 550 Loss 0.691541314125061 Training accuracy is 0.75 Epoch 560 Loss 0.6914339065551758 Training accuracy is 0.5 Epoch 570 Loss 0.691317081451416 Training accuracy is 0.5 Epoch 580 Loss 0.6911896467208862 Training accuracy is 0.5 Epoch 590 Loss 0.6910485029220581 Training accuracy is 0.5 Epoch 600 Loss 0.6908772587776184 Training accuracy is 0.5 Epoch 610 Loss 0.6906797289848328 Training accuracy is 0.5 Epoch 620 Loss 0.690450131893158 Training accuracy is 0.5 Epoch 630 Loss 0.6902066469192505 Training accuracy is 0.5 Epoch 640 Loss 0.6899532675743103 Training accuracy is 0.5 Epoch 650 Loss 0.6896663308143616 Training accuracy is 0.5 Epoch 660 Loss 0.6893412470817566 Training accuracy is 0.5 Epoch 670 Loss 0.6889669895172119 Training accuracy is 0.75 Epoch 680 Loss 0.6885308027267456 Training accuracy is 0.5 Epoch 690 Loss 0.6880131959915161 Training accuracy is 0.75 Epoch 700 Loss 0.6873992681503296 Training accuracy is 0.75 Epoch 710 Loss 0.6866694688796997

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```
Training accuracy is 0.75
Epoch 720 Loss 0.6856508851051331
Training accuracy is 0.75
Epoch 730 Loss 0.6842676401138306
Training accuracy is 0.75
Epoch 740 Loss 0.6827037930488586
Training accuracy is 0.75
Epoch 750 Loss 0.6807572841644287
Training accuracy is 0.75
Epoch 760 Loss 0.6783814430236816
Training accuracy is 0.75
Epoch 770 Loss 0.6752572059631348
Training accuracy is 0.75
Epoch 780 Loss 0.6711351871490479
Training accuracy is 0.75
Epoch 790 Loss 0.6654714345932007
Training accuracy is 0.75
Epoch 800 Loss 0.6570992469787598
Training accuracy is 0.75
Epoch 810 Loss 0.6444829702377319
Training accuracy is 0.75
Epoch 820 Loss 0.6250220537185669
```

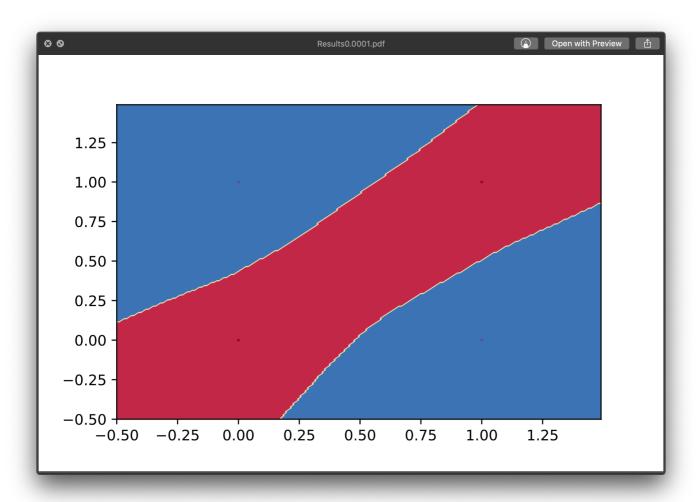
```
In [9]:
         #%% compute accuracy on training data
         net out = net(data)
         pred = net_out.data.max(1)[1] # get the index of the max log-probability
         correctidx = pred.eq(target.data)
         ncorrect = correctidx.sum()
         accuracy = ncorrect.item()/len(data)
         print('Training accuracy is ', accuracy)
         plt.scatter(X[:, 0], X[:, 1], c=pred, cmap=plt.cm.Spectral, s = 1)
         #%% if need to train further
         # for epoch in range(nepochs):
               # resize data from (batch size, 1, 28, 28) to (batch size, 28*28)
         #
               optimizer.zero grad()
         #
               net out = net(data)
         #
               loss = criterion(net out, target)
         #
               loss.backward()
         #
              optimizer.step()
         #
               if epoch % 100 == 0:
                   print('Epoch ', epoch, 'Loss ', loss.item())
         #
         #%% plot outputs
         plot_decision_boundary(net, X, y, 'Results0.0001.pdf')
         #plot decision boundary(net, X[correctidx,:], y[correctidx], 'Correct.pdf'
         #plot decision boundary(net, X[~correctidx,:], y[~correctidx], 'Inorrect.p
         #%% save model
```

Training accuracy is 1.0

<ipython-input-3-b68d425386b7>:21: UserWarning: Implicit dimension choice f
or log_softmax has been deprecated. Change the call to include dim=X as an
argument.

return F.log_softmax(x)

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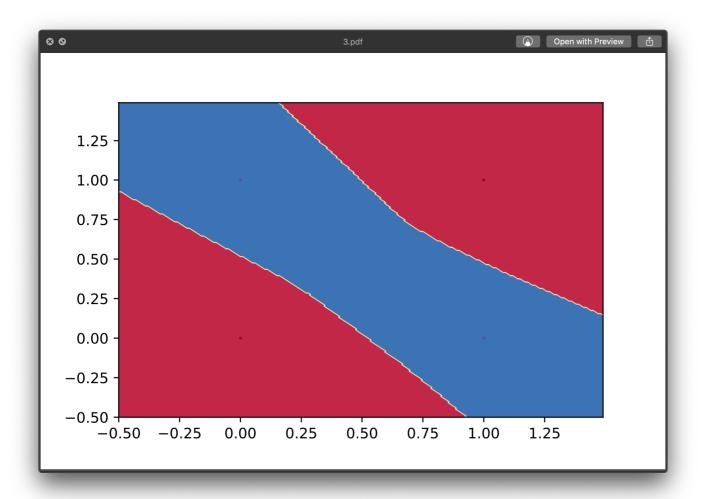
```
In [10]:
          #Question 2:
In [11]:
          # nepochs = 10000
          nepochs = 3000 #10000
          data, target = X, y
          # run the main training loop
          for epoch in range(nepochs):
               adjust learning rate if desired
               if epoch % 3000 == 0 and epoch <= 24000:
          #
          #
                   for g in optimizer.param groups:
          #
                       g['lr'] = g['lr']/2
              optimizer.zero grad()
              # forward propagate
              net_out = net(data)
              # compute loss
              loss = criterion(net_out, target)
              # backpropagate
              loss.backward()
              # update parameters
              optimizer.step()
              # print out report
              if epoch % 10 == 0:
                  print('Epoch ', epoch, 'Loss ', loss.item())
                  net out = net(data)
                  pred = net_out.data.max(1)[1] # get the index of the max log-prob
                  correctidx = pred.eq(target.data)
                  ncorrect = correctidx.sum()
                  accuracy = ncorrect.item()/len(data)
                  if epoch< 0.001:</pre>
                      break
                  print('Training accuracy is ', accuracy)
         Epoch 0 Loss 0.6225759387016296
```

<ipython-input-3-b68d425386b7>:21: UserWarning: Implicit dimension choice f
or log_softmax has been deprecated. Change the call to include dim=X as an
argument.

return F.log_softmax(x)

```
In [12]: #Question 3
```

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```
In [13]:
          class Net(nn.Module):
              def __init__(self):
                  super(Net, self).__init__()
                  self.fc1 = nn.Linear(2, 20)
                  self.fc2 = nn.Linear(20, 20)
                  self.fc3 = nn.Linear(20, 20)
                  \#self.fc4 = nn.Linear(20, 20)
                  \#self.fc5 = nn.Linear(20, 20)
                  \#self.fc6 = nn.Linear(20, 20)
                  \#self.fc7 = nn.Linear(20, 2)
              def forward(self, x):
                  x = F.relu(self.fc1(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 = self.fc3(x)
                  return F.log_softmax(x)
                  #return F.softmax(x)
```

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```
In [14]:
          \# nepochs = 10000
          nepochs = 3000 #10000
          data, target = X, y
          # run the main training loop
          for epoch in range(nepochs):
               adjust learning rate if desired
          #
               if epoch % 3000 == 0 and epoch <= 24000:
          #
                   for g in optimizer.param groups:
          #
                       g['lr'] = g['lr']/2
              optimizer.zero_grad()
              # forward propagate
              net_out = net(data)
              # compute loss
              loss = criterion(net_out, target)
              # backpropagate
              loss.backward()
              # update parameters
              optimizer.step()
              # print out report
              if epoch % 10 == 0:
                  print('Epoch ', epoch, 'Loss ', loss.item())
                  net out = net(data)
                  pred = net_out.data.max(1)[1] # get the index of the max log-prob
                  correctidx = pred.eq(target.data)
                  ncorrect = correctidx.sum()
                  accuracy = ncorrect.item()/len(data)
                  #if epochs < 0.001:
                       break
                  print('Training accuracy is ', accuracy)
```

<ipython-input-3-b68d425386b7>:21: UserWarning: Implicit dimension choice f
or log_softmax has been deprecated. Change the call to include dim=X as an
argument.

```
return F.log_softmax(x)
Epoch 0 Loss 0.6200512647628784
Training accuracy is 1.0
Epoch 10 Loss 0.5861670970916748
Training accuracy is 1.0
Epoch 20 Loss 0.5298529863357544
Training accuracy is 1.0
Epoch 30 Loss 0.43002861738204956
Training accuracy is 1.0
Epoch 40 Loss 0.27282190322875977
Training accuracy is 1.0
Epoch 50 Loss 0.13154783844947815
Training accuracy is 1.0
Epoch 60 Loss 0.05327668413519859
Training accuracy is 1.0
Epoch 70 Loss 0.023709189146757126
Training accuracy is 1.0
Epoch 80 Loss 0.013243664987385273
Training accuracy is 1.0
Epoch 90 Loss 0.008865930140018463
Training accuracy is 1.0
Epoch 100 Loss 0.006658598780632019
```

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Training accuracy is 1.0 Epoch 110 Loss 0.005379956215620041 Training accuracy is 1.0 Epoch 120 Loss 0.0045394496992230415 Training accuracy is 1.0 Epoch 130 Loss 0.003928876481950283 Training accuracy is 1.0 Epoch 140 Loss 0.00346049340441823 Training accuracy is 1.0 Epoch 150 Loss 0.0030872030183672905 Training accuracy is 1.0 Epoch 160 Loss 0.0027822465635836124 Training accuracy is 1.0 Epoch 170 Loss 0.0025281780399382114 Training accuracy is 1.0 Epoch 180 Loss 0.0023130846675485373 Training accuracy is 1.0 Epoch 190 Loss 0.0021286052651703358 Training accuracy is 1.0 Epoch 200 Loss 0.001969016157090664 Training accuracy is 1.0 Epoch 210 Loss 0.0018297492060810328 Training accuracy is 1.0 Epoch 220 Loss 0.001707063871435821 Training accuracy is 1.0 Epoch 230 Loss 0.0015983477933332324 Training accuracy is 1.0 Epoch 240 Loss 0.0015014930395409465 Training accuracy is 1.0 Epoch 250 Loss 0.0014145683962851763 Training accuracy is 1.0 Epoch 260 Loss 0.0013362973695620894 Training accuracy is 1.0 Epoch 270 Loss 0.0012653418816626072 Training accuracy is 1.0 Epoch 280 Loss 0.001200811006128788 Training accuracy is 1.0 Epoch 290 Loss 0.0011419614311307669 Training accuracy is 1.0 Epoch 300 Loss 0.001088079996407032 Training accuracy is 1.0 Epoch 310 Loss 0.0010385422501713037 Training accuracy is 1.0 Epoch 320 Loss 0.000992931891232729 Training accuracy is 1.0 Epoch 330 Loss 0.0009508328512310982 Training accuracy is 1.0 Epoch 340 Loss 0.0009117689915001392 Training accuracy is 1.0 Epoch 350 Loss 0.0008755025337450206 Training accuracy is 1.0 Epoch 360 Loss 0.0008417657809332013 Training accuracy is 1.0 Epoch 370 Loss 0.000810201745480299 Training accuracy is 1.0 Epoch 380 Loss 0.0007807806832715869 Training accuracy is 1.0 Epoch 390 Loss 0.0007531456649303436 Training accuracy is 1.0 Epoch 400 Loss 0.0007272965158335865 Training accuracy is 1.0

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Epoch 410 Loss 0.0007029358530417085 Training accuracy is 1.0 Epoch 420 Loss 0.0006800339324399829 Training accuracy is 1.0 Epoch 430 Loss 0.0006584418006241322 Training accuracy is 1.0 Epoch 440 Loss 0.0006379810511134565 Training accuracy is 1.0 Epoch 450 Loss 0.0006187112303450704 Training accuracy is 1.0 Epoch 460 Loss 0.000600394036155194 Training accuracy is 1.0 Epoch 470 Loss 0.0005830595036968589 Training accuracy is 1.0 Epoch 480 Loss 0.0005665883654728532 Training accuracy is 1.0 Epoch 490 Loss 0.0005509211914613843 Training accuracy is 1.0 Epoch 500 Loss 0.0005360282375477254 Training accuracy is 1.0 Epoch 510 Loss 0.000521879643201828 Training accuracy is 1.0 Epoch 520 Loss 0.0005083564901724458 Training accuracy is 1.0 Epoch 530 Loss 0.0004953990573994815 Training accuracy is 1.0 Epoch 540 Loss 0.0004830968100577593 Training accuracy is 1.0 Epoch 550 Loss 0.000471300765639171 Training accuracy is 1.0 Epoch 560 Loss 0.00045998120913282037 Training accuracy is 1.0 Epoch 570 Loss 0.00044919774518348277 Training accuracy is 1.0 Epoch 580 Loss 0.0004388609668239951 Training accuracy is 1.0 Epoch 590 Loss 0.00042888158350251615 Training accuracy is 1.0 Epoch 600 Loss 0.00041931914165616035 Training accuracy is 1.0 Epoch 610 Loss 0.0004101736412849277 Training accuracy is 1.0 Epoch 620 Loss 0.00040135576273314655 Training accuracy is 1.0 Epoch 630 Loss 0.0003928953083232045 Training accuracy is 1.0 Epoch 640 Loss 0.000384702900191769 Training accuracy is 1.0 Epoch 650 Loss 0.0003768083406612277 Training accuracy is 1.0 Epoch 660 Loss 0.00036927120527252555 Training accuracy is 1.0 Epoch 670 Loss 0.0003619723138399422 Training accuracy is 1.0 Epoch 680 Loss 0.0003549117245711386 Training accuracy is 1.0 Epoch 690 Loss 0.000348119210684672 Training accuracy is 1.0 Epoch 700 Loss 0.0003415053943172097 Training accuracy is 1.0 Epoch 710 Loss 0.0003351596824359149

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Training accuracy is 1.0 Epoch 720 Loss 0.00032902247039601207 Training accuracy is 1.0 Epoch 730 Loss 0.0003230639558751136 Training accuracy is 1.0 Epoch 740 Loss 0.00031734377262182534 Training accuracy is 1.0 Epoch 750 Loss 0.0003117427113465965 Training accuracy is 1.0 Epoch 760 Loss 0.0003063501790165901 Training accuracy is 1.0 Epoch 770 Loss 0.00030113637330941856 Training accuracy is 1.0 Epoch 780 Loss 0.00029604171868413687 Training accuracy is 1.0 Epoch 790 Loss 0.00029112579068169 Training accuracy is 1.0 Epoch 800 Loss 0.00028635881608352065 Training accuracy is 1.0 Epoch 810 Loss 0.0002817109925672412 Training accuracy is 1.0 Epoch 820 Loss 0.0002771823201328516 Training accuracy is 1.0 Epoch 830 Loss 0.00027280260110273957 Training accuracy is 1.0 Epoch 840 Loss 0.0002685718354769051 Training accuracy is 1.0 Epoch 850 Loss 0.0002644304186105728 Training accuracy is 1.0 Epoch 860 Loss 0.0002604081528261304 Training accuracy is 1.0 Epoch 870 Loss 0.00025653489865362644 Training accuracy is 1.0 Epoch 880 Loss 0.00025272119091823697 Training accuracy is 1.0 Epoch 890 Loss 0.00024899683194234967 Training accuracy is 1.0 Epoch 900 Loss 0.00024533207761123776 Training accuracy is 1.0 Epoch 910 Loss 0.0002418460207991302 Training accuracy is 1.0 Epoch 920 Loss 0.00023844940005801618 Training accuracy is 1.0 Epoch 930 Loss 0.00023511234030593187 Training accuracy is 1.0 Epoch 940 Loss 0.00023183487064670771 Training accuracy is 1.0 Epoch 950 Loss 0.00022867656662128866 Training accuracy is 1.0 Epoch 960 Loss 0.0002255480212625116 Training accuracy is 1.0 Epoch 970 Loss 0.00022250888287089765 Training accuracy is 1.0 Epoch 980 Loss 0.0002195591077907011 Training accuracy is 1.0 Epoch 990 Loss 0.0002166987251257524 Training accuracy is 1.0 Epoch 1000 Loss 0.00021389791800174862 Training accuracy is 1.0 Epoch 1010 Loss 0.0002111566864186898 Training accuracy is 1.0

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Epoch 1020 Loss	0.00020847504492849112
Training accuracy	
Epoch 1030 Loss	0.0002058529935311526
Training accuracy	
Epoch 1040 Loss	0.00020326071535237134
Training accuracy	
Epoch 1050 Loss	0.00020075784414075315
Training accuracy	
Epoch 1060 Loss	0.0001982847461476922
Training accuracy	
Epoch 1070 Loss	0.00019590104056987911
Training accuracy	
Epoch 1080 Loss	0.0001935471227625385
Training accuracy	
Epoch 1090 Loss	0.00019125279504805803
Training accuracy	
Epoch 1100 Loss	0.0001890180428745225
Training accuracy	
Epoch 1110 Loss	0.00018678329070098698
Training accuracy	
Epoch 1120 Loss	0.000184667733265087
Training accuracy	
Epoch 1130 Loss	0.0001825521612772718
Training accuracy	
Epoch 1140 Loss	0.00018046637705992907
Training accuracy	
Epoch 1150 Loss	0.0001784401829354465
Training accuracy	
Epoch 1160 Loss	0.0001764735789038241
Training accuracy	
Epoch 1170 Loss	0.00017453677719458938
Training accuracy	
Epoch 1180 Loss	0.00017262976325582713
Training accuracy	
Epoch 1190 Loss	
Training accuracy	
	0.0001689051277935505
Training accuracy	
Epoch 1210 Loss	
Training accuracy	
Epoch 1220 Loss	
Training accuracy	
Epoch 1230 Loss	
Training accuracy	
Epoch 1240 Loss	
Training accuracy	
Epoch 1250 Loss	
Training accuracy	
Epoch 1260 Loss	
Training accuracy	
Epoch 1270 Loss	
Training accuracy	
Epoch 1280 Loss	
Training accuracy	
Epoch 1290 Loss	
Training accuracy Epoch 1300 Loss	
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Training accuracy Epoch 1310 Loss	
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Training accuracy Epoch 1320 Loss	
Epoch 1320 Loss	0.00014941/53/613//186

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m	i- 1 0
Training accuracy Epoch 1330 Loss	
Training accuracy	
Epoch 1340 Loss	0.00014655693667009473
Training accuracy	
Epoch 1350 Loss	0.00014515644579660147
Training accuracy	
Epoch 1360 Loss	0.0001437857426935807
Training accuracy	
Epoch 1370 Loss	0.00014244482736103237
Training accuracy	
Epoch 1380 Loss	0.00014110391202848405
Training accuracy	
Epoch 1390 Loss	0.00013982260134071112
Training accuracy	
Epoch 1400 Loss	0.00013851147377863526
Training accuracy	
Epoch 1410 Loss	0.0001372897531837225
Training accuracy	
Epoch 1420 Loss	0.00013603823026642203
Training accuracy	
Epoch 1430 Loss	0.00013481650967150927
Training accuracy	is 1.0
Epoch 1440 Loss	0.00013359477452468127
Training accuracy	is 1.0
Epoch 1450 Loss	0.00013243264402262866
Training accuracy	is 1.0
Epoch 1460 Loss	0.00013130030129104853
Training accuracy	
Epoch 1470 Loss	0.00013013817078899592
Training accuracy	
Epoch 1480 Loss	0.00012903561582788825
Training accuracy	
Epoch 1490 Loss	0.0001279330754186958
Training accuracy	
Epoch 1500 Loss	0.00012680073268711567
Training accuracy	0.00012575779692269862
Training accuracy	
	0.0001247446343768388
Training accuracy	
	0.00012367189629003406
Training accuracy	
	0.00012262894597370178
Training accuracy	
	0.00012161578342784196
Training accuracy	is 1.0
Epoch 1560 Loss	0.00012066222552675754
Training accuracy	is 1.0
Epoch 1570 Loss	0.00011964907025685534
Training accuracy	
Epoch 1580 Loss	
Training accuracy	
Epoch 1590 Loss	
Training accuracy	
Epoch 1600 Loss	
Training accuracy	
_	0.0001158944214694202
Training accuracy	
Epoch 1620 Loss Training accuracy	0.00011494086356833577
Training accuracy	το 1.0

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T 1 1620 T	0.00011404600202606010
Epoch 1630 Loss	0.00011404690303606912
Training accuracy	is 1.0
Epoch 1640 Loss	0.00011318273027427495
Training accuracy	is 1.0
Epoch 1650 Loss	0.00011231856478843838
Training accuracy	is 1.0
Epoch 1660 Loss	0.0001114245897042565
Training accuracy	is 1.0
Epoch 1670 Loss	0.00011056041694246233
Training accuracy	is 1.0
Epoch 1680 Loss	0.00010978565114783123
Training accuracy	is 1.0 0.00010892147838603705
Epoch 1690 Loss	
Training accuracy	is 1.0 0.00010814670531544834
Epoch 1700 Loss	
Training accuracy Epoch 1710 Loss	is 1.0 0.00010734212992247194
-	is 1.0
Training accuracy Epoch 1720 Loss	0.00010650775220710784
Training accuracy	is 1.0
Epoch 1730 Loss	0.00010576278145890683
Training accuracy	is 1.0
Epoch 1740 Loss	0.0001049581915140152
Training accuracy	is 1.0
Epoch 1750 Loss	0.00010421322076581419
Training accuracy	is 1.0
Epoch 1760 Loss	0.00010343844041926786
Training accuracy	is 1.0
Epoch 1770 Loss	0.00010269346239510924
Training accuracy	is 1.0
Epoch 1780 Loss	0.00010200807446381077
Training accuracy	is 1.0
Epoch 1790 Loss	0.00010126309643965214
Training accuracy	is 1.0
Epoch 1800 Loss	0.0001005777157843113
Training accuracy	
Epoch 1810 Loss	9.983273048419505e-05
Training accuracy	is 1.0
Epoch 1820 Loss	
Training accuracy	is 1.0
Epoch 1830 Loss	9.846196189755574e-05
Training accuracy	is 1.0
Epoch 1840 Loss	
Training accuracy	
Epoch 1850 Loss	
Training accuracy	
Epoch 1860 Loss	
Training accuracy	
Epoch 1870 Loss	
Training accuracy	
Epoch 1880 Loss	
Training accuracy	
Epoch 1890 Loss	
Training accuracy	
Epoch 1900 Loss	
Training accuracy	
Epoch 1910 Loss	
Training accuracy	
Epoch 1920 Loss	
Training accuracy	
Epoch 1930 Loss	9.202527871821076e-05

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Training accuracy	
Epoch 1940 Loss	
Training accuracy	
Epoch 1950 Loss	9.083329496206716e-05
Training accuracy	
Epoch 1960 Loss	9.023729944601655e-05
Training accuracy	
Epoch 1970 Loss	8.967111352831125e-05
Training accuracy	
Epoch 1980 Loss	8.910492761060596e-05
Training accuracy	
Epoch 1990 Loss	8.853872714098543e-05
Training accuracy	
Epoch 2000 Loss	8.797253394732252e-05
Training accuracy	
Epoch 2010 Loss	8.74063407536596e-05
Training accuracy	8.686994260642678e-05
Epoch 2020 Loss	
Training accuracy Epoch 2030 Loss	8.633354445919394e-05
Training accuracy	
Epoch 2040 Loss	8.582695591030642e-05
-	
Training accuracy Epoch 2050 Loss	8.529056503903121e-05
Training accuracy	
Epoch 2060 Loss	8.475415961584076e-05
Training accuracy	
Epoch 2070 Loss	8.424757106695324e-05
Training accuracy	
Epoch 2080 Loss	8.37111656437628e-05
Training accuracy	
Epoch 2090 Loss	8.323437941726297e-05
Training accuracy	
Epoch 2100 Loss	
Training accuracy	
Epoch 2110 Loss	
Training accuracy	
	8.171458466676995e-05
Training accuracy	
	8.12377838883549e-05
Training accuracy	is 1.0
	8.079079270828515e-05
Training accuracy	is 1.0
Epoch 2150 Loss	8.028418233152479e-05
Training accuracy	is 1.0
Epoch 2160 Loss	7.983719115145504e-05
Training accuracy	
	7.936039037303999e-05
Training accuracy	
	7.891339191701263e-05
Training accuracy	
Epoch 2190 Loss	
Training accuracy	
Epoch 2200 Loss	
Training accuracy	
	7.754259422654286e-05
Training accuracy	
	7.70955957705155e-05
Training accuracy	
Epoch 2230 Loss Training accuracy	7.667839236091822e-05

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Epoch 2240 Loss	
Training accuracy	
Epoch 2250 Loss	
Training accuracy	
Epoch 2260 Loss	
Training accuracy	
Epoch 2270 Loss	
Training accuracy	
Epoch 2280 Loss	
Training accuracy	
Epoch 2290 Loss	
Training accuracy	
Epoch 2300 Loss	
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Epoch 2310 Loss	
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Epoch 2320 Loss	
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Epoch 2330 Loss	
Training accuracy	
Epoch 2340 Loss	
Training accuracy	
Epoch 2350 Loss	
Training accuracy	is 1.0
Epoch 2360 Loss	
Training accuracy	
Epoch 2370 Loss	7.101637311279774e-05
Training accuracy	is 1.0
Epoch 2380 Loss	7.068857667036355e-05
Training accuracy	
Epoch 2390 Loss	
Training accuracy	
Epoch 2400 Loss	
Training accuracy	
Epoch 2410 Loss	
Training accuracy	
Epoch 2420 Loss	
Training accuracy	
	6.88111613271758e-05
Training accuracy	
	6.848335760878399e-05
Training accuracy	
	6.812575156800449e-05
Training accuracy	
-	6.77681528031826e-05
Training accuracy	
	6.744035636074841e-05
Training accuracy	
	6.70827430440113e-05
Training accuracy	
	6.675493932561949e-05
Training accuracy	
	6.642713560722768e-05
Training accuracy	
	6.60695368424058e-05
Training accuracy	
	6.574173312401399e-05
Training accuracy	
_	6.544372445205227e-05
Training accuracy	is 1.0 6.511592800961807e-05

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Training accuracy	
-	6.475832196883857e-05
Training accuracy	
Epoch 2560 Loss	
Training accuracy	
Epoch 2570 Loss	
Training accuracy	
Epoch 2580 Loss	
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Training accuracy Epoch 2600 Loss	
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Epoch 2660 Loss	
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Epoch 2670 Loss	
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Epoch 2680 Loss	
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Epoch 2690 Loss	
Training accuracy	
Epoch 2700 Loss	
Training accuracy	
Epoch 2710 Loss	
Training accuracy	
Epoch 2720 Loss	
Training accuracy	
Epoch 2730 Loss	
Training accuracy	
Epoch 2740 Loss	
Training accuracy	is 1.0
Epoch 2750 Loss	5.8917430578731e-05
Training accuracy	is 1.0
Epoch 2760 Loss	5.867902291356586e-05
Training accuracy	is 1.0
Epoch 2770 Loss	5.841082020197064e-05
Training accuracy	is 1.0
Epoch 2780 Loss	5.8142617490375414e-05
Training accuracy	
Epoch 2790 Loss	5.790421346318908e-05
Training accuracy	
Epoch 2800 Loss	
Training accuracy	
Epoch 2810 Loss	
Training accuracy	
Epoch 2820 Loss	
Training accuracy	
Epoch 2830 Loss	
Training accuracy	
-	5.6622786360094324e-05
Training accuracy	1S 1.0

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Epoch 2850 Loss 5.638437869492918e-05 Training accuracy is 1.0 Epoch 2860 Loss 5.611617598333396e-05 Training accuracy is 1.0 Epoch 2870 Loss 5.5907570640556514e-05 Training accuracy is 1.0 Epoch 2880 Loss 5.563936792896129e-05 Training accuracy is 1.0 Epoch 2890 Loss 5.5371157941408455e-05 Training accuracy is 1.0 Epoch 2900 Loss 5.5132750276243314e-05 Training accuracy is 1.0 Epoch 2910 Loss 5.489434624905698e-05 Training accuracy is 1.0 Epoch 2920 Loss 5.4655942221870646e-05 Training accuracy is 1.0 Epoch 2930 Loss 5.44473368790932e-05 Training accuracy is 1.0 Epoch 2940 Loss 5.4208932851906866e-05 Training accuracy is 1.0 Epoch 2950 Loss 5.403012619353831e-05 Training accuracy is 1.0 Epoch 2960 Loss 5.3791722166351974e-05 Training accuracy is 1.0 Epoch 2970 Loss 5.358311682357453e-05 Training accuracy is 1.0 Epoch 2980 Loss 5.328511178959161e-05 Training accuracy is 1.0 Epoch 2990 Loss 5.3076506446814165e-05 Training accuracy is 1.0

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