```
import torch # Core PyTorch library
import torch.nn as nn # Neural network modules (layers, loss functions, etc.)
import torch.nn.functional as F # Functional interface for activation functions, loss functions, etc.
from torch.utils.data import Dataset, DataLoader # Dataset and DataLoader classes for handling data efficiently
from sklearn.model_selection import train_test_split # Splits dataset into training and test sets
import pandas as pd \, # Data manipulation and loading
import matplotlib.pyplot as plt # Visualization library
%matplotlib inline
### Name : Sana Fathima H
### reg no : 212223240145
class Model(nn.Module):
    def __init__(self, in_features=4, h1=8, h2=9, out_features=3):
        super().__init__()
        self.fc1 = nn.Linear(in_features, h1)
                                                 # Input layer
        self.fc2 = nn.Linear(h1, h2)
                                                 # Hidden layer
        self.out = nn.Linear(h2, out_features) # Output layer
    def forward(self, x):
       x = F.relu(self.fc1(x)) # Activation function for first hidden layer
       x = F.relu(self.fc2(x)) # Activation function for second hidden layer
       x = self.out(x)
                                 # Output layer (no activation, raw scores)
        return x
torch.manual_seed(32) # For reproducibility
model = Model()
df = pd.read_csv('iris.csv')
df.head()
\overline{\mathbf{T}}
        sepal length (cm) sepal width (cm) petal length (cm) petal width (cm) target
     0
                       5.1
                                         3.5
                                                            1.4
                                                                              0.2
                                                                                      0.0
     1
                       4.9
                                         3.0
                                                            1.4
                                                                              0.2
                                                                                      0.0
                                                            1.3
                       4.7
                                                                              0.2
                                                                                      0.0
     3
                                         3.1
                                                            1.5
                                                                                      0.0
                       4.6
                                                                              0.2
                                                                                      0.0
 Next steps: ( Generate code with df
                                   View recommended plots
                                                                New interactive sheet
X = df.drop('target', axis=1).values
y = df['target'].values
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=33)
X_train = torch.FloatTensor(X_train)
X_test = torch.FloatTensor(X_test)
# y_train = F.one_hot(torch.LongTensor(y_train)) # Not needed with CrossEntropyLoss
# y_test = F.one_hot(torch.LongTensor(y_test))
y_train = torch.LongTensor(y_train)
y_test = torch.LongTensor(y_test)
trainloader = DataLoader(X_train, batch_size=60, shuffle=True)
testloader = DataLoader(X_test, batch_size=60, shuffle=False)
torch.manual_seed(4)
model = Model()
criterion = nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.parameters(), lr=0.01)
epochs = 100
losses = []
for i in range(epochs):
   i+=1
    y_pred = model.forward(X_train)
    loss = criterion(y_pred, y_train)
    losses.append(loss)
    # a neat trick to save screen space:
    if i%10 == 1:
        print(f'epoch: {i:2} loss: {loss.item():10.8f}')
    optimizer.zero_grad()
    loss.backward()
    optimizer.step()
→ epoch: 1 loss: 1.09568226
     epoch: 11 loss: 0.98190624
     epoch: 21 loss: 0.75652373
     epoch: 31 loss: 0.49447367
     epoch: 41 loss: 0.34981722
     epoch: 51 loss: 0.22807978
     epoch: 61 loss: 0.13547550
```

```
epoch: 71 loss: 0.09162237
epoch: 81 loss: 0.07378434
epoch: 91 loss: 0.06546316

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```
plt.plot(range(epochs), [loss.detach().numpy() for loss in losses])
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.title('Training Loss Over Time')
plt.show()
```




```
with torch.no_grad():
    y_val = model.forward(X_test)
    loss = criterion(y_val, y_test)
print(f'{loss:.8f}')
```

→ 0.06247779

```
1. tensor([-0.3360, 7.3629, 1.3780])
 2. tensor([0.2770, 8.1552, 0.4267])
 3. tensor([ 11.9968, 6.1842, -19.1980])
 4. tensor([-2.0192, 7.9662, 4.2445])
 5. tensor([-6.1353, 7.9516, 11.0908])
 6. tensor([-10.2640, 8.3102, 17.9992])
 7. tensor([ 12.0541, 6.4316, -19.2913])
 8. tensor([ 12.9496, 6.4815, -20.7530])
                                             0
 9. tensor([-5.7727, 8.2435, 10.5079])
10. tensor([-7.8872, 8.6126, 14.0726])
11. tensor([-8.7060, 8.6074, 15.4331])
12. tensor([ 11.6348, 5.8164, -18.6210])
13. tensor([-8.1013, 8.2331, 14.3883])
14. tensor([-2.0796, 7.7751, 4.3185])
15. tensor([-6.0833, 8.3916, 11.0582])
16. tensor([0.1354, 7.8658, 0.6407])
                                             1
17. tensor([-4.0880, 7.7216, 7.6638])
18. tensor([ 13.1511, 6.5907, -21.0787])
19. tensor([-1.5649, 8.0220, 3.4751])
20. tensor([-6.2865, 8.9727, 11.4244])
21. tensor([ 12.3848, 6.2568, -19.8265]) 0
22. tensor([ 13.8199, 7.0854, -22.1532])
23. tensor([-8.8475, 8.3181, 15.6471])
24. tensor([ 12.1968, 6.1261, -19.5250]) 0
25. tensor([-5.8089, 7.5468, 10.5336]) 2
26. tensor([-4.4530, 7.7875, 8.2861]) 2
27. tensor([-1.4289, 7.7785, 3.2325])
                                             1
28. tensor([ 0.5351, 7.5358, -0.0494])
                                             1
29. tensor([-5.8235, 8.1573, 10.5971])
                                             2
30. tensor([-5.2573, 7.7475, 9.6101])
29 out of 30 = 96.67% correct
```

```
torch.save(model.state_dict(), 'IrisDatasetModel.pt')
```

```
new_model = Model()
new_model.load_state_dict(torch.load('IrisDatasetModel.pt'))
new_model.eval()
```

```
Model(
(fc1): Linear(in_features=4, out_features=8, bias=True)
(fc2): Linear(in_features=8, out_features=9, bias=True)
```

```
29/03/2025, 05:15 (out): Linear(in_features=9, out_features=3, bias=True)
```

```
with torch.no_grad():
    y_val = new_model.forward(X_test)
    loss = criterion(y_val, y_test)
print(f'{loss:.8f}')
```

→ 0.06247779

```
mystery_iris = torch.tensor([5.6,3.7,2.2,0.5])
```

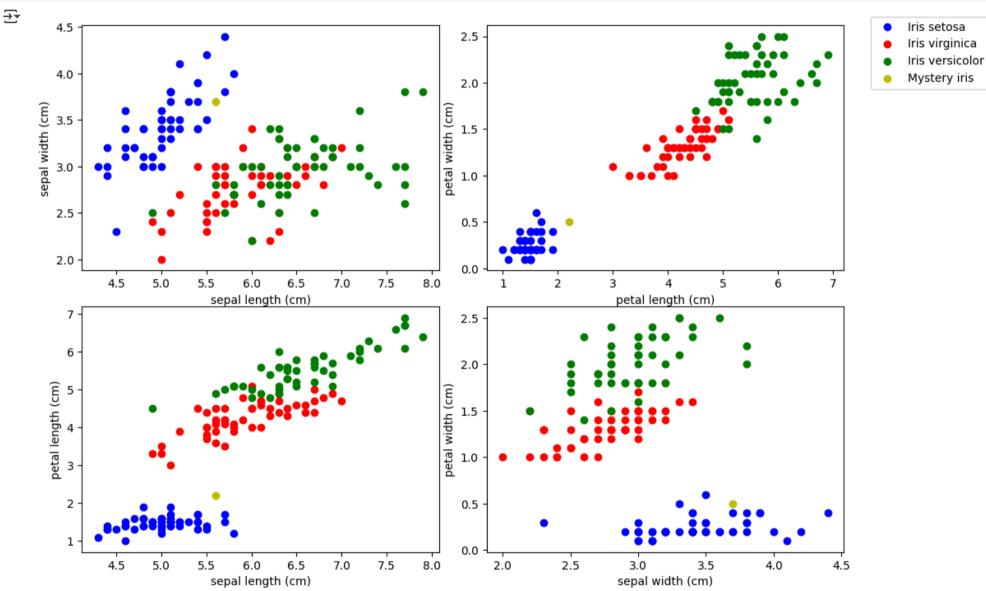
```
fig, axes = plt.subplots(nrows=2, ncols=2, figsize=(10,7))
fig.tight_layout()

plots = [(0,1),(2,3),(0,2),(1,3)]
colors = ['b', 'r', 'g']
labels = ['Iris setosa','Iris virginica','Iris versicolor','Mystery iris']

for i, ax in enumerate(axes.flat):
    for j in range(3):
        x = df.columns[plots[i][0]]
        y = df.columns[plots[i][1]]
        ax.scatter(df[df['target']==j][x], df[df['target']==j][y], color=colors[j])
        ax.set(xlabel=x, ylabel=y)

# Add a plot for our mystery iris:
        ax.scatter(mystery_iris[plots[i][0]],mystery_iris[plots[i][1]], color='y')

fig.legend(labels=labels, loc=3, bbox_to_anchor=(1.0,0.85))
plt.show()
```



```
with torch.no_grad():
    print(new_model(mystery_iris))
    print()
    print(labels[new_model(mystery_iris).argmax()])
```

tensor([12.2112, 7.1279, -19.5248])

 Iris setosa

```
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```