Logistic Regression - Demo

Input: Breast Cancer Data from sklearn.datasets

Code:

1. Importing Libraries

import torch
import torch.nn as nn
import numpy as np
from sklearn import datasets
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler

#model
#loss and optimizer
#training loop

2. Loading and preparing data

#preparing data - taking data from sklearn datasets
bc=datasets.load_breast_cancer()
X,y=bc.data,bc.target # X contains features, y contains labels 0 or 1

n_samples, n_features=X.shape
print(n_samples, n_features)
#splitting data in sets
80% training and 20% testing
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.2,random_state=1234)

3. Scaling Features

#scaling features sc=StandardScaler() X_train=sc.fit_transform(X_train) X_test=sc.transform(X_test)

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#converting numpy arrays to PyTorch tensors of type float32
X train=torch.from numpy(X train.astype(np.float32))
X test=torch.from numpy(X test.astype(np.float32))
y train=torch.from numpy(y train.astype(np.float32))
y_test=torch.from_numpy(y_test.astype(np.float32))
#reshape y tensor
#Reshape y train and y test to be column vectors
y_train=y_train.view(y_train.shape[0],1)
y_test=y_test.view(y_test.shape[0],1)
   4. Defining the model
#setting up model
class LogisticRegression(nn.Module):
  def init (self,n input features):
    super(LogisticRegression,self). init ()
    self.linear=nn.Linear(n_input_features,1)
  def forward(self,x):
    #sigmoid - logistic function
    y pred=torch.sigmoid(self.linear(x))
    return y_pred #returns 0 or 1
model=LogisticRegression(n features) #30 input features, 1 output feature
   5. Loss and Optimizer
#loss and optimizer
criterion=nn.BCELoss()
                              #binary cross entropy loss
optimizer=torch.optim.SGD(model.parameters(),lr=0.01) #learning rate is 0.01
                                                                              #using
stochastic gradient descent
   Training Loop
#training loop
num epochs=100
for epoch in range(num epochs):
  #forward pass and loss
  y pred=model(X train)
```

```
loss=criterion(y pred,y train)
  #backward pass and optimizer step
  loss.backward()
  if (epoch + 1) \% 10 == 0:
     print(f'Epoch {epoch + 1} - Before updating: Weights:
{model.linear.weight.data.numpy()}')
  #update weight
  optimizer.step()
  # Print the values of theta (weights) after updating
  if (epoch + 1) \% 10 == 0:
     print(f'Epoch {epoch + 1} - After updating: Weights:
{model.linear.weight.data.numpy()}')
  #empty gradient
  optimizer.zero grad()
  if(epoch+1)%10==0: #after every 10th step
     print(f'epoch:{epoch+1}, loss={loss.item():.4f}')
   7. Evaluation
with torch.no grad():
  y predicted=model(X test)
  #if y is greater than 0.5 than 1 otherwise 0
  y predicted cls=y predicted.round()
  #calculating accuracy
  acc=y predicted cls.eq(y test).sum()/float(y test.shape[0])
  print()
  print(f'accuracy={acc.item():.4f}')
```

The code demonstrates a full machine learning pipeline using PyTorch: data loading, preprocessing, model definition, training, and evaluation. It uses logistic regression to classify breast cancer data and achieves this by defining a model class, using loss functions and optimizers, and iterating through training epochs while monitoring the model's performance.

Output:

```
569 30
Epoch 10 - Before updating: Weights: [[-0.17073406 -0.1711105 -0.13568045
-0.02915274 -0.16110763 -0.09371217
  0.0763087 -0.02072659 -0.16975053 -0.0309427 -0.00556845 -0.03620943
  0.05817804 0.1473701 0.16461124 0.1604254 0.10457321 -0.06082732
 -0.11116552 -0.09127568 -0.16230337 -0.06632227 -0.20284383 0.07208515
 -0.12065051 -0.06398452 -0.04981894 0.11550106 0.16690008
-0.0182646611
Epoch 10 - After updating: Weights: [[-0.17301682 -0.17222755 -0.13801351
-0.03139529 -0.16219965 -0.09567025
 0.07396628 -0.0232381 -0.17072085 -0.03096887 -0.00750027 -0.03602823
  0.05626854 0.14543271 0.16457392 0.15931483 0.10358009 -0.06227916
 -0.11111563 -0.09168877 -0.16476777 -0.06757452 -0.20533884 0.06973836
 -0.12209271 -0.06593622 -0.05205654 0.11287782 0.16553918
-0.01941744]]
epoch:10, loss=0.4870
Epoch 20 - Before updating: Weights: [[-0.19135179 -0.18135822 -0.15672535
-0.04937538 -0.1706881 -0.11100165
  0.0553793 \quad -0.04327925 \quad -0.17825574 \quad -0.03072538 \quad -0.02281404 \quad -0.03443651
  0.04118913 0.13009918 0.16432303 0.15089591 0.09603389 -0.0735964
 -0.11044271 -0.0944775 -0.18462431 -0.07787882 -0.2254004 0.05086967
 -0.13376732 -0.08145928 -0.06994706 0.09181805 0.1544983
-0.0285260211
Epoch 20 - After updating: Weights: [[-0.19318125 -0.18228605 -0.15858969
-0.05116667 -0.17150615 -0.11249018
  0.05354844 -0.04526543 -0.1789812 -0.03065387 -0.02432154 -0.03426554
  0.03971044 0.128592 0.16430229 0.15010852 0.09532756 -0.07468774
 -0.11034843 -0.09469938 -0.18661195 -0.0789325 -0.22740448 0.0489844
 -0.13494076 -0.08299085 -0.0717217 0.08971956 0.15338622 -0.0294174
epoch:20, loss=0.4129
Epoch 30 - Before updating: Weights: [[-0.20822072 -0.19003777 -0.17389622
-0.06587987 -0.17802043 -0.12441476
  0.03867775 - 0.06149107 - 0.18475068 - 0.02971203 - 0.03658122 - 0.03279112
  -0.10939056 -0.09608827 -0.20299526 -0.08777841 -0.24389474 0.03346243
 -0.14463726 -0.09543103 -0.08620635 0.07251409 0.14418341
-0.03659105]]
Epoch 30 - After updating: Weights: [[-0.20975463 -0.19084115 -0.17545538
-0.06737968 -0.17866327 -0.12559828
  0.03717995 -0.06313524 -0.18531898 -0.02957915 -0.03781933 -0.0326346
 0.02651779 0.11510852 0.16413592 0.14345896 0.08937427 -0.08420321
 -0.10927507 -0.09618411 -0.2046705 -0.08869925 -0.24557805 0.03187664
 -0.14563069 -0.09668304 -0.08767146 0.0707653 0.14323947
-0.03730547]]
```

epoch:30, loss=0.3643

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Epoch 40 - Before updating: Weights: [[-0.22257762 -0.19765273 -0.18847445
-0.07991655 -0.18387742 -0.1352407
 0.02480417 -0.07679739 -0.18991724 -0.02818853 -0.04809534 -0.03128986
 0.01653086 0.10484661 0.16401388 0.13895679 0.08537385 -0.09105352
 -0.10818789 -0.09663111 -0.21870504 -0.09653374 -0.2596591 0.01859735
 -0.15396243 -0.10700937 -0.09981182 0.05620527 0.13531868
-0.04313461]]
Epoch 40 - After updating: Weights: [[-0.22390564 -0.1983681 -0.18982124
-0.08121511 -0.18440105 -0.13621299
 0.02353767 -0.07820367 -0.1903776 -0.02801538 -0.04915288 -0.03114728
 0.01550583 0.10379014 0.16400056 0.13852447 0.08499211 -0.09173524
 -0.10806326 -0.09664021 -0.22016157 -0.09735918 -0.2611183 0.01721956
 -0.15482788 -0.10806365 -0.10105728 0.05470413 0.13449574
-0.04372285]]
epoch:40, loss=0.3296
Epoch 50 - Before updating: Weights: [[-0.23513842 -0.20449613 -0.20120125
-0.09220279 -0.18870737 -0.144\overline{2326}
  0.01294266 -0.0900318 -0.1941498 -0.02632674 -0.0580579 -0.02992082
  0.00689481 0.09488824 0.16387632 0.13513213 0.0820194 -0.09728307
-0.10692288 -0.09642725 -0.23250403 -0.10444853 -0.27346686 0.00554434
-0.16216666 -0.11685827 -0.11149379 0.04206502 0.12751916
-0.0485742911
Epoch 50 - After updating: Weights: [[-0.23631428 -0.20514576 -0.20239131
-0.09335358 -0.18914555 -0.14505069
  0.01184583 -0.09126305 -0.19453202 -0.02612659 -0.05898652 -0.02979052
 0.00599892 0.0939592 0.16386178 0.13480477 0.08173524 -0.0978408
 -0.1067948 -0.09637449 -0.23379843 -0.10520193 -0.2747602 0.00431978
 -0.16293687 -0.11776582 -0.11257578 0.04074823 0.12678736
-0.0490689911
epoch:50, loss=0.3035
Epoch 60 - Before updating: Weights: [[-0.24634446 -0.21075211 -0.21253346
-0.10317601 -0.19278787 -0.15186132
0.00258506 -0.101712 -0.19769438 -0.02423737 -0.06688683 -0.02866665
 -0.00160703 0.08604818 0.16372128 0.13223036 0.0795249 -0.10241697
 -0.1056373 -0.09568485 -0.24485853 -0.11171757 -0.2857979 -0.00614618
 -0.16952358 -0.1254026 -0.12172064 0.02956699 0.1205348 -0.0531846
]]
Epoch 60 - After updating: Weights: [[-0.24740265 -0.21135052 -0.2136025
-0.10421301 -0.19316229 -0.15256223
0.00161799 -0.10280879 -0.19801779 -0.02401901 -0.0677185 -0.02854691
 -0.00240606 0.08521457 0.16370448 0.1319817 0.07931424 -0.10288067
 -0.10550851 -0.09558687 -0.24602738 -0.11241443 -0.286963 -0.0072526
 -0.17022039 -0.1261972 -0.12267646 0.02839279 0.1198741
-0.05360783]]
epoch:60, loss=0.2830
Epoch 70 - Before updating: Weights: [[-0.25648636 -0.2165438 -0.22277178
-0.11312157 -0.1963013 -0.15843983
```

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-0.00660583 -0.11218073 -0.20071423 -0.02199503 -0.07484785 -0.0275113
 -0.00924263 0.0780616 0.16354136 0.13002916 0.07768508 -0.10671018
 -0.10435173 -0.0945456 -0.2560774 -0.11847305 -0.29696962 -0.01676971
-0.17621946 -0.13292965 -0.13081 0.01835623 0.11419427
-0.05715435]]
Epoch 70 - After updating: Weights: [[-0.25745037 -0.217101 -0.22374408
-0.11406774 -0.19662665 -0.15904893
 -0.00747047 \ -0.11317084 \ -0.20099205 \ -0.02176453 \ -0.07560366 \ -0.02740066
 -0.00996603 0.07730252 0.16352186 0.12984113 0.07753111 -0.10710069
 -0.10422363 -0.09441397 -0.25714573 -0.11912422 -0.29803216 -0.01778184
 -0.17685813 -0.13363478 -0.13166563 0.01729564 0.11359061
-0.0575216411
epoch:70, loss=0.2663
Epoch 80 - Before updating: Weights: [[-0.26576626 -0.2219576 -0.23212524
-0.12223656 -0.19937345 -0.16418639
 -0.01486555 -0.12167708 -0.20332295 -0.01965166 -0.08212034 -0.02644142
 -0.01619235 0.0707513 0.16333318 0.12837262 0.07635409 -0.11034291
 -0.10307679 -0.09310968 -0.2663761 -0.12480877 -0.3072034 -0.0265312
 -0.18238625 -0.13964255 -0.13898712 0.00818213 0.10837533
-0.06061825]]
Epoch 80 - After updating: Weights: [[-0.26665285 -0.22248077 -0.23301813
-0.12310822 -0.19966011 -0.16472177
-0.01564732 -0.12258034 -0.20356457 -0.0194133 -0.08281495 -0.02633869
 -0.01685484 0.07005233 0.16331074 0.1282323 0.07624459 -0.11067526
 -0.10295013 -0.09295277 -0.2673618 -0.12542208 -0.3081818 -0.02746602
 -0.18297778 -0.14027515 -0.13976143 0.00721422 0.10781843
-0.0609408611
epoch:80, loss=0.2524
Epoch 90 - Before updating: Weights: [[-0.27433088 -0.22705656 -0.24074553
-0.1306634 -0.20209451 -0.16925925
-0.02236534 -0.13037467 -0.20560251 -0.01724432 -0.08883145 -0.02544614
 -0.02258367 0.06399258 0.16309524 0.12714766 0.07542478 -0.11344698
 -0.10181822 -0.09145021 -0.2759115 -0.13079362 -0.31666014 -0.03557906
 -0.1881206 -0.14568987 -0.14641738 -0.00113887 0.10298782
-0.06367521]]
Epoch 90 - After updating: Weights: [[-0.2751525 -0.22755107 -0.24157187
-0.13147259 -0.20235002 -0.1697343
-0.02307876 -0.13120581 -0.20581485 -0.01700117 -0.08947551 -0.02535037
 -0.02319592 0.0633433 0.16306978 0.12704545 0.07535065 -0.11373235
 -0.1016934 -0.09127419 -0.27682787 -0.13137493 -0.317568 -0.03644913
 -0.18867321 -0.14626256 -0.14712442 -0.00202969 0.10247002
-0.06396156]]
epoch:90, loss=0.2406
Epoch 100 - Before updating: Weights: [[-0.28229064 -0.23188816
-0.24874668 -0.13850884 -0.20453097 -0.17377664
 -0.02923374 -0.13840383 -0.20761387 -0.01479989 -0.09507484 -0.02451668
 -0.02851036 0.05769395 0.1628271 0.12626974 0.07481632 -0.11612135
```

```
-0.10057913 -0.08962179 -0.28480145 -0.1364797 -0.3254607 -0.04402452 -0.19349517 -0.15118356 -0.15322582 -0.0097453 0.0979634 -0.06639975]]

Epoch 100 - After updating: Weights: [[-0.2830568 -0.2323581 -0.24951631 -0.13926475 -0.20476101 -0.17420152 -0.02988987 -0.13917406 -0.20780215 -0.01455422 -0.09567636 -0.02442706 -0.02908037 0.05708657 0.16279863 0.12619841 0.07477058 -0.11636825 -0.10045636 -0.08943135 -0.28565866 -0.13703352 -0.32630846 -0.04483942 -0.1940151 -0.15170598 -0.1538764 -0.01057099 0.09747878 -0.06665625]]

epoch:100, loss=0.2305
```

accuracy=0.9123