```
import torch
 from torch.autograd import Variable
x_data = Variable(torch.Tensor([[1.0], [2.0], [3.0]]))
y_data = Variable(torch.Tensor([[2.0], [4.0], [6.0]]))
 class LinearRegressionModel(torch.nn.Module):
                           init (self):
                           super(LinearRegressionModel, self).__init__()
self.linear = torch.nn.Linear(1, 1) # One in and one out
def forward(self, x):
    y_pred = self.linear(x)
    return y_pred
our_model = LinearRegressionModel()
criterion = torch.nn.MSELoss(size_average = False)
optimizer = torch.optim.SGD(our_model.parameters(), lr
                 /usr/local/lib/python3.7/dist-packages/torch/nn/_reduction.py:42: UserWarning: size_average and reduce args will be deprecated, please use reduction='sum' instead. warnings.warn(warning.format(ret))

    Perform a forward pass bypassing our data and finding out the predicted value of y

            · Compute the loss using MSE
            · Reset all the gradients to 0, perform a backpropagation and then, update the weights
               # Forward pass: Compute predicted y by passing
                # x to the model
              pred_y = our_model(x_data)
               # Compute and print loss
              loss = criterion(pred_y, y_data)
              # Zero gradients, perform a backward pass,
               # and undate the weights.
              optimizer.zero_grad()
loss.backward()
              print('epoch {}, loss {}'.format(epoch, loss.item()))
                print( epoth (), 1055 (). Tommat(epoth, epoth 442, 1055 0.00044(275035043514845 epoth 443, 1055 0.00041618343675500005 epoth 444, 1055 0.000440430085973216593 epoth 445, 1055 0.000384967479482293 epoth 446, 1055 0.00038716910019628704 epoth 447, 1055 0.00038711910019628704
              epoch 446, loss 0.00033849674794822931
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epoch 448, loss 0.00038711910019628704
epoch 449, loss 0.00038716764666367322
epoch 450, loss 0.00037607646666367322
epoch 451, loss 0.00037607646666367322
epoch 451, loss 0.0003760761472553909
epoch 452, loss 0.00036534370156006025
epoch 453, loss 0.0003549152461346239
epoch 455, loss 0.0003549152461346239
epoch 455, loss 0.000334915214613972091
epoch 456, loss 0.0003349157355673015
epoch 457, loss 0.00033938539112907201
epoch 458, loss 0.0003393853911256454
epoch 457, loss 0.00033164161612564504
epoch 460, loss 0.00033164161612564504
epoch 461, loss 0.000331041616115264504
epoch 462, loss 0.000331041616115264504
epoch 464, loss 0.000331041616115264504
epoch 465, loss 0.0003317356673015
epoch 464, loss 0.000331749208215252
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epoch 466, loss 0.00031608635161097
epoch 469, loss 0.00031608635161097
epoch 469, loss 0.000238209560625555
epoch 464, loss 0.0002983209560625555
epoch 467, loss 0.0002283209560625555
epoch 469, loss 0.0002283209560625555
epoch 470, loss 0.000227489881943342
epoch 475, loss 0.000227489881943342
epoch 476, loss 0.000225877514881376
epoch 473, loss 0.00022587571489434342
epoch 476, loss 0.00022587571489433432
epoch 476, loss 0.0002258757148943742
epoch 473, loss 0.0002258757148943742
epoch 473, loss 0.0002258757148943742
epoch 474, loss 0.0002258757148943743
epoch 475, loss 0.0002258757148953953604
epoch 480, loss 0.0002258757148963439332
epoch 481, loss 0.000223866458677429847192938644
epoch 476, loss 0.00022382652777187258
epoch 481, loss 0.0002238366658658674
epoch 482, loss 0.0002238366658684977
epoch 483, loss 0.00022383666586684977
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                epoch 491, loss 0.002297739327919377514
epoch 492, loss 0.00029475551279614
epoch 493, loss 0.0002018150844378397
epoch 494, loss 0.0001981905221582576
epoch 495, loss 0.0001932340528583154
epoch 497, loss 0.0001932340528583154
epoch 497, loss 0.0001946114175580442
epoch 499, loss 0.00018772498879116029
epoch 499, loss 0.00018572604507375509
 new_var = Variable(torch.Tensor([[4.0]]))
 pred_y = our_model(new_var)
 print("predict (after training)", 4, our_model(new_var).item())
                 predict (after training) 4 7.984364032745361
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