

```
import torch
import torch.nn as nn # Neural network module
import numpy as np
import matplotlib.pyplot as plt # For plotting
%matplotlib inline
```

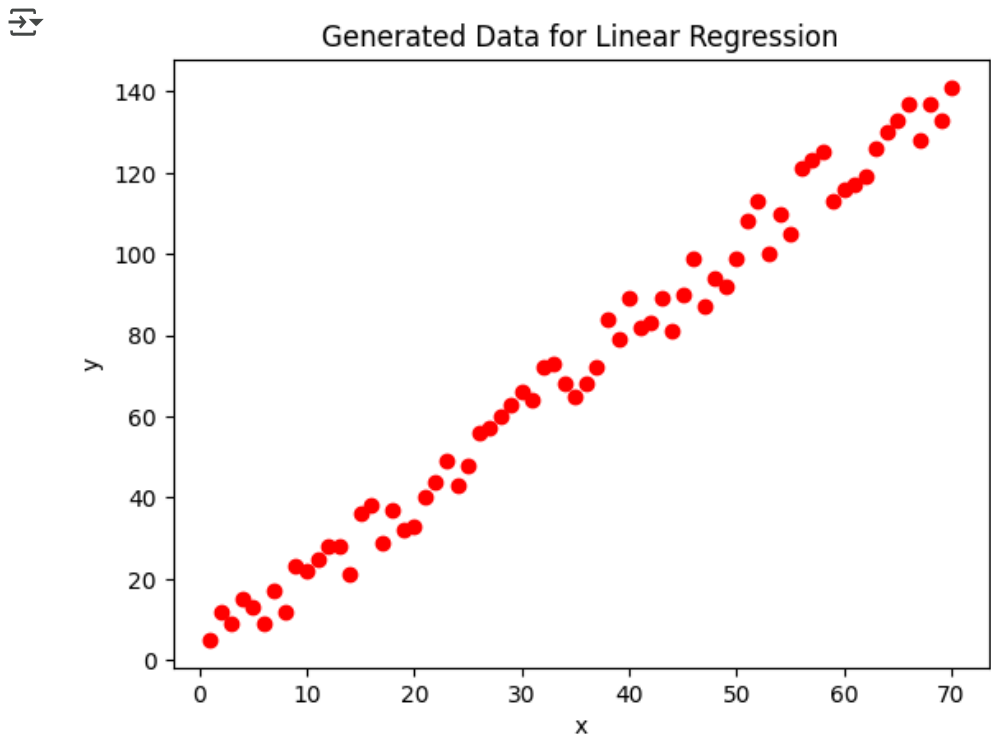
```
X = torch.linspace(1,70,70).reshape(-1,1)
```

```
torch.manual_seed(71) # to obtain reproducible results
e = torch.randint(-8,9,(70,1),dtype=torch.float)
#print(e.sum())
```

```
y = 2*X + 1 + e
print(y.shape)
```

```
↗ ↘ torch.Size([70, 1])
```

```
plt.scatter(X.numpy(), y.numpy(),color='red') # Scatter plot of data points
plt.xlabel('x')
plt.ylabel('y')
plt.title('Generated Data for Linear Regression')
plt.show()
```



```
torch.manual_seed(59)

# Defining the model class
class Model(nn.Module):
    def __init__(self, in_features, out_features):
        super().__init__()
        self.linear = nn.Linear(in_features, out_features)

    def forward(self, x):
        y_pred = self.linear(x)
        return y_pred
```

```
torch.manual_seed(59)
model = Model(1, 1)
print('Weight:', model.linear.weight.item())
print('Bias: ', model.linear.bias.item())
```

```
↗ ↘ Weight: 0.10597813129425049
      Bias:  0.9637961387634277
```

```
loss_function = nn.MSELoss() # Mean Squared Error (MSE) loss

optimizer = torch.optim.SGD(model.parameters(), lr=0.0001)
```

```
epochs = 50 # Number of training iterations
losses = [] # List to store loss values

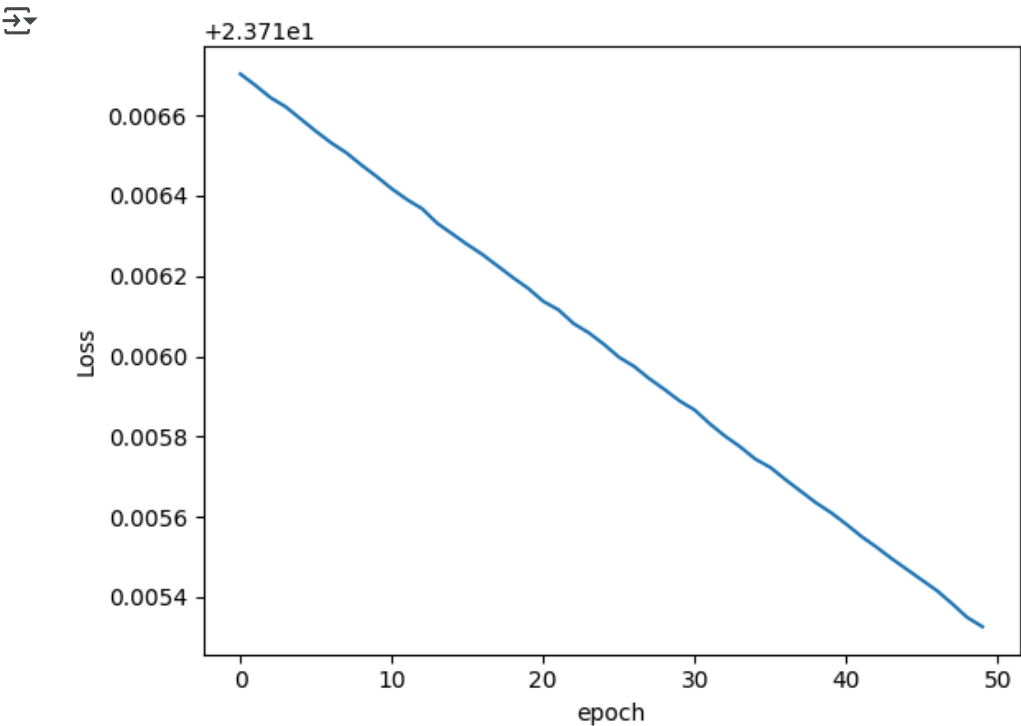
for epoch in range(1, epochs + 1): # Start from 1 to 50
    optimizer.zero_grad() # Clear previous gradients
    y_pred = model(X) # Forward pass
    loss = loss_function(y_pred, y) # Compute loss
    losses.append(loss.item()) # Store loss value

    loss.backward() # Compute gradients
    optimizer.step() # Update weights

# Print loss, weight, and bias for EVERY epoch (1 to 50)
print(f'epoch: {epoch:2}  loss: {loss.item():10.8f}  '
      f'weight: {model.linear.weight.item():10.8f}  '
      f'bias: {model.linear.bias.item():10.8f}')
```

```
↗ epoch: 1 loss: 23.71670341 weight: 1.99034202 bias: 1.00660098
epoch: 2 loss: 23.71667480 weight: 1.99034083 bias: 1.00665390
epoch: 3 loss: 23.71664429 weight: 1.99033976 bias: 1.00670683
epoch: 4 loss: 23.71662140 weight: 1.99033856 bias: 1.00675976
epoch: 5 loss: 23.71659088 weight: 1.99033749 bias: 1.00681269
epoch: 6 loss: 23.71656036 weight: 1.99033642 bias: 1.00686562
epoch: 7 loss: 23.71653175 weight: 1.99033523 bias: 1.00691855
epoch: 8 loss: 23.71650696 weight: 1.99033415 bias: 1.00697148
epoch: 9 loss: 23.71647644 weight: 1.99033296 bias: 1.00702441
epoch: 10 loss: 23.71644783 weight: 1.99033189 bias: 1.00707734
epoch: 11 loss: 23.71641731 weight: 1.99033070 bias: 1.00713027
epoch: 12 loss: 23.71639061 weight: 1.99032962 bias: 1.00718319
epoch: 13 loss: 23.71636772 weight: 1.99032843 bias: 1.00723612
epoch: 14 loss: 23.71633148 weight: 1.99032736 bias: 1.00728905
epoch: 15 loss: 23.71630478 weight: 1.99032629 bias: 1.00734198
epoch: 16 loss: 23.71627808 weight: 1.99032509 bias: 1.00739491
epoch: 17 loss: 23.71625328 weight: 1.99032402 bias: 1.00744784
epoch: 18 loss: 23.71622467 weight: 1.99032283 bias: 1.00750077
epoch: 19 loss: 23.71619606 weight: 1.99032176 bias: 1.00755370
epoch: 20 loss: 23.71616936 weight: 1.99032056 bias: 1.00760663
epoch: 21 loss: 23.71613693 weight: 1.99031949 bias: 1.00765955
epoch: 22 loss: 23.71611595 weight: 1.99031830 bias: 1.00771248
epoch: 23 loss: 23.71608162 weight: 1.99031723 bias: 1.00776541
epoch: 24 loss: 23.71605873 weight: 1.99031603 bias: 1.00781834
epoch: 25 loss: 23.71603012 weight: 1.99031496 bias: 1.00787127
epoch: 26 loss: 23.71599770 weight: 1.99031377 bias: 1.00792420
epoch: 27 loss: 23.71597481 weight: 1.99031270 bias: 1.00797713
epoch: 28 loss: 23.71594429 weight: 1.99031150 bias: 1.00803006
epoch: 29 loss: 23.71591759 weight: 1.99031043 bias: 1.00808299
epoch: 30 loss: 23.71588898 weight: 1.99030936 bias: 1.00813591
epoch: 31 loss: 23.71586609 weight: 1.99030828 bias: 1.00818884
epoch: 32 loss: 23.71583176 weight: 1.99030709 bias: 1.00824177
epoch: 33 loss: 23.71580124 weight: 1.99030602 bias: 1.00829470
epoch: 34 loss: 23.71577454 weight: 1.99030483 bias: 1.00834763
epoch: 35 loss: 23.71574402 weight: 1.99030375 bias: 1.00840056
epoch: 36 loss: 23.71572304 weight: 1.99030256 bias: 1.00845349
epoch: 37 loss: 23.71569252 weight: 1.99030149 bias: 1.00850642
epoch: 38 loss: 23.71566391 weight: 1.99030030 bias: 1.00855935
epoch: 39 loss: 23.71563530 weight: 1.99029922 bias: 1.00861228
epoch: 40 loss: 23.71561050 weight: 1.99029803 bias: 1.00866508
epoch: 41 loss: 23.71558189 weight: 1.99029696 bias: 1.00871789
epoch: 42 loss: 23.71555138 weight: 1.99029577 bias: 1.00877070
epoch: 43 loss: 23.71552467 weight: 1.99029458 bias: 1.00882351
epoch: 44 loss: 23.71549606 weight: 1.99029350 bias: 1.00887632
epoch: 45 loss: 23.71546936 weight: 1.99029243 bias: 1.00892913
epoch: 46 loss: 23.71544266 weight: 1.99029136 bias: 1.00898194
epoch: 47 loss: 23.71541595 weight: 1.99029016 bias: 1.00903475
epoch: 48 loss: 23.71538353 weight: 1.99028909 bias: 1.00908756
epoch: 49 loss: 23.71534920 weight: 1.99028790 bias: 1.00914037
epoch: 50 loss: 23.71532631 weight: 1.99028683 bias: 1.00919318
```

```
plt.plot(range(epochs), losses)
plt.ylabel('Loss')
plt.xlabel('epoch');
plt.show()
```



```
# Automatically determine x-range
x1 = torch.tensor([X.min().item(), X.max().item()])

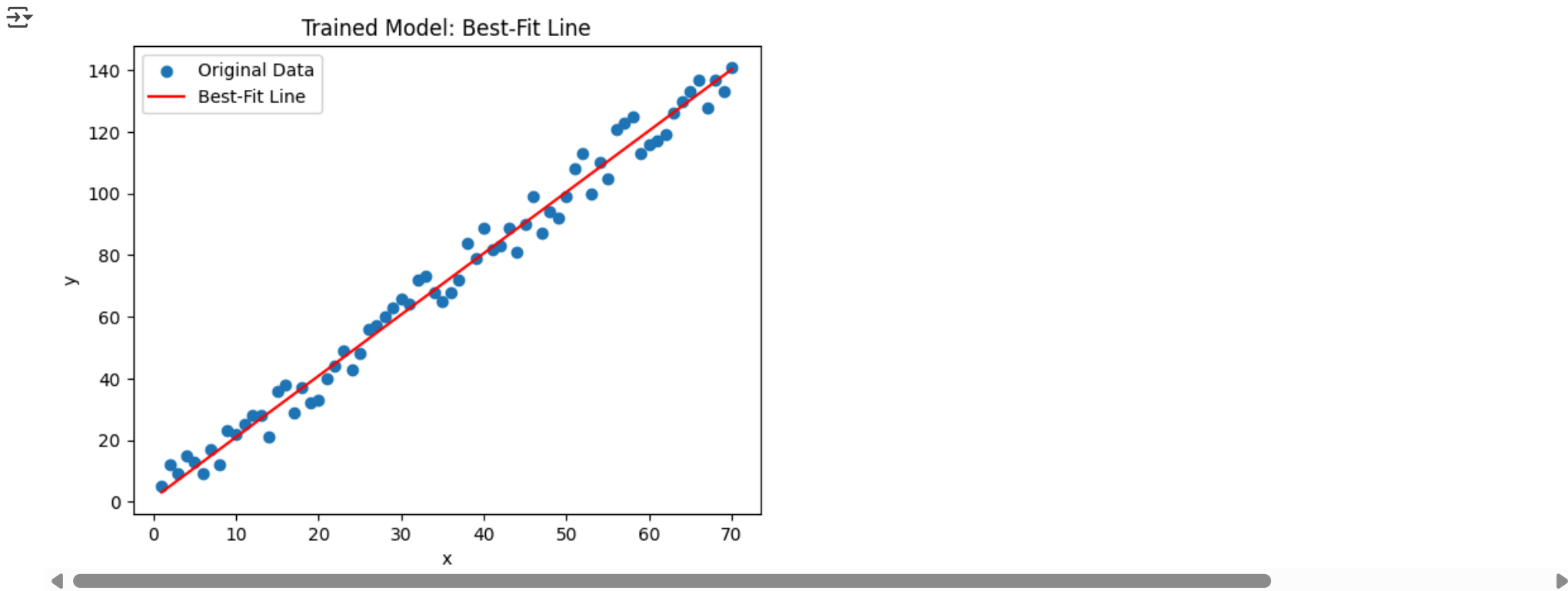
# Extract model parameters
w1, b1 = model.linear.weight.item(), model.linear.bias.item()

# Compute y1 (predicted values)
y1 = x1 * w1 + b1
```

```
# Print weight, bias, and x/y values
print(f'Final Weight: {w1:.8f}, Final Bias: {b1:.8f}')
print(f'X range: {x1.numpy()}')
print(f'Predicted Y values: {y1.numpy()}')
```

↗ Final Weight: 1.99028683, Final Bias: 1.00919318
X range: [1. 70.]
Predicted Y values: [2.99948 140.32927]

```
# Plot original data and best-fit line
plt.scatter(X.numpy(), y.numpy(), label="Original Data")
plt.plot(x1.numpy(), y1.numpy(), 'r', label="Best-Fit Line")
plt.xlabel('x')
plt.ylabel('y')
plt.title('Trained Model: Best-Fit Line')
plt.legend()
plt.show()
```



```
torch.save(model.state_dict(), 'Sana Fathima H regression.pt')
```