

Regression from Scratch in Numpy vs. PyTorch

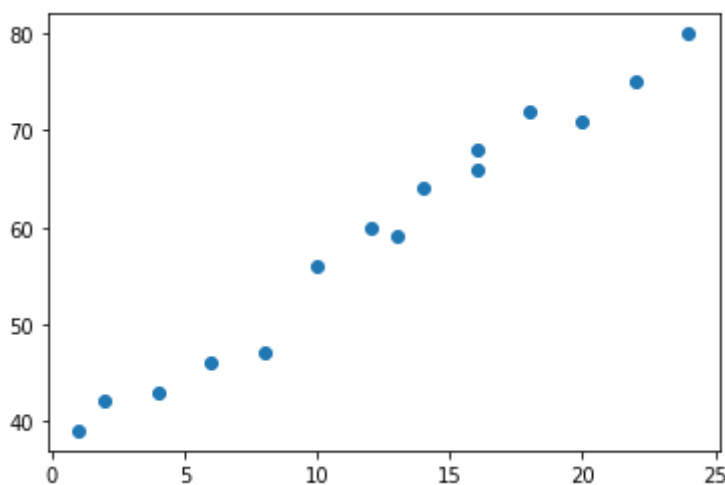
Regression in Numpy

```
import numpy as np
import matplotlib.pyplot as plt
```

```
X = np.array([1,2,4,6,8,10,12,13,14,16,16,18,20,22,24])
Y = np.array([39,42,43,46,47,56,60,59,64,66,68,72,71,75,80])
```

```
plt.scatter(X,Y)
```

↳ <matplotlib.collections.PathCollection at 0x7fc3f3b97c18>



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ip between X and Y.(We'll discuss more about correlation in

''' $Y = aX + b$ is the equation of line/linear regression model.

Goal here is to find the values of a and b .

There are multiple techniques to achieve this:

- 1.Matrix calculations: Put all data into matrices to perform optimization.Used for small d
- 2.Gradient Descent : Try to minimize error/difference between actual and predicted values
- 3.Regularization: While minimizing error,also try to reduce impact of unnecessary features
- 4.Simple linear regression:If there are single input variable and single output variable,u

More detailed explanation of above techniques is not in the scope here.

We'll implement method 2 i.e Gradient Descent here-more specific-Batch Gradient Descent.

Weights(a, b) are updated at end of complete batch/all rows as follow:

new a = old a - (learning_rate*gradient_a)

new b = old b - (learning_rate*gradient_b)

...

↳ " $Y = aX + b$ is the equation of line/linear regression model.\nGoal here is to find the

```
np.random.seed(2)
epochs=15
learning_rate = 0.001
w = np.random.randn()
b = np.random.randn()
y_pred = np.empty(len(Y))

for i in range(epochs):
    print("-----epoch:{}-----".format(i))
    #prediction
    y_pred = w*X +b

    #Error/loss calculation is Mean Squared Error
    error = np.mean((Y - y_pred)**2)
    print('Total Error:{}'.format(error))

    #Gradient calculation
    gradient_a = np.mean(-2*X*(Y-y_pred))
    gradient_b = np.mean(-2*(Y-y_pred))

    #Update weights
    w -= learning_rate*gradient_a
    b -= learning_rate*gradient_b
```



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```
print(w,b)
```

```
↳ 4.042799282999869 0.4771951521774575
```

```
epoch:1
```

```
'''Error is reducing with increment in epochs. Number of epochs and learning rate are hype
Let's not play around with it and jump to PyTorch'''
```

```
↳ "Error is reducing with increment in epochs. Number of epochs and learning rate are h
```

```
-----epoch:1-----
```

```
-----epoch:1-----
```

Regression in PyTorch

```
Total Error:315.4/66/816/40153
```

```
import torch
```

```
-----epoch:10-----
```

```
#initialise data/features and target
```

```
X_tensor = torch.from_numpy(X)
```

```
Y_tensor = torch.from_numpy(Y)
```

```
Total Error:315.4/66/816/40153
```

```
#Initialise weights
```

```
'''Here unlike numpy we have to mention that these variables are trainable(need to calcula
This can be done using requires_grad:'''
```

```
↳ 'Here unlike numpy we have to mention that these variables are trainable(need to calc
```

```
torch.random.seed = 2
```

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```
requires_grad=True,dtype=torch.float)
```

```
requires_grad=True,dtype=torch.float)
```

```
learning_rate = 0.001
```

```
w_tensor
```

```
↳ tensor([-0.6845], requires_grad=True)
```

```
#Model without PyTorch in-built methods
```

```
for i in range(epochs):
```

```
    print("-----epoch:{}-----".format(i))
```

```
    #prediction
```

```
    y_pred = w_tensor*X_tensor +b_tensor
```

```
    #Error/loss calculation is Mean Squared Error
```

```
    error = ((Y_tensor - y_pred)**2).mean()
```

```
    print('Total Error:{}'.format(error))
```

```
'''Now no need to calculate gradients,PyTorch will do it if we tell which function/var
error.backward()
```

```
'''Actual values of gradients can be seen using grad attribute'''
```



```
'''Now no need to calculate gradients,PyTorch will do it if we tell which function/var
error.backward()
```

```
#Update weights using Optimizer
optimizer.step()
```

```
#After each step,Reinitilaise gradients because PyTorch holds on to gradients
#Reinitilaise gradients using Optimizer
optimizer.zero_grad()
```

```

-----epoch:0-----
Total Error:297.0393981933594
-----epoch:1-----
Total Error:296.7607727050781
-----epoch:2-----
Total Error:296.48272705078125
-----epoch:3-----
Total Error:296.2050476074219
-----epoch:4-----
Total Error:295.927734375
-----epoch:5-----
Total Error:295.6506652832031
-----epoch:6-----
Total Error:295.3738098144531
-----epoch:7-----
Total Error:295.0972595214844
-----epoch:8-----
Total Error:294.8209533691406
-----epoch:9-----
Total Error:294.5448913574219
-----epoch:10-----
Total Error:294.2691345214844
-----epoch:11-----
Total Error:293.99365234375

```

Saved successfully!



```

-----epoch:12-----
Total Error:293.4433898925781
-----epoch:13-----
Total Error:293.16864013671875

```

'''Till now,we've explored loss calculation and Optimizers.
The only manual step remaining is prediction step. Let's remove that also'''

```

"-----epoch:14-----
Total Error:293.16864013671875
'''Till now,we've explored loss calculation and Optimizers.\n
The only manual step remai

```

```
#Create Network by extending parent nn.Module.
```

```
'''We have to implement __init__ and forward methods '''
```

```
class Network(torch.nn.Module):
```

```
    def __init__(self):
```

```
        super().__init__()
```

```
        #Intialise parameters whcih should be trained. Note that parameters need to be wrapped
```

```
        self.w_tensor = torch.nn.Parameter(torch.randn(1,requires_grad=True,dtype=torch.float))
```

```
        self.b_tensor = torch.nn.Parameter(torch.randn(1,requires_grad=True,dtype=torch.float))
```

```
def forward(self,x):
    #Output prediction calculation
    return w_tensor*x +b_tensor

#Model with PyTorch in-built methods
model = Network()
optimizer = torch.optim.SGD(model.parameters(), lr=learning_rate)
loss = torch.nn.MSELoss(reduction='mean')
for i in range(epochs):
    print("-----epoch:{}-----".format(i))
    #This will not do actual training but will set model in training mode.
    model.train()

    #prediction
    y_pred = model(X_tensor)

    #Error/loss calculation is Mean Squared Error
    error = loss(Y_tensor, y_pred)
    print('Total Error:{}'.format(error))

    '''Now no need to calculate gradients,PyTorch will do it if we tell which function/var
    error.backward()

    #Update weights using Optimizer
    optimizer.step()

    #After each step,Reinitilaise gradients because PyTorch holds on to gradients
    #Reinitilaise gradients using Optimizer
    optimizer.zero_grad()
```



Saved successfully!



```

-----epoch:0-----
Total Error:288.8082275390625
-----epoch:1-----
Total Error:288.8082275390625
-----epoch:2-----
Total Error:288.8082275390625
-----epoch:3-----
Total Error:288.8082275390625
-----epoch:4-----
Total Error:288.8082275390625
-----epoch:5-----
Total Error:288.8082275390625
-----epoch:6-----
Total Error:288.8082275390625
-----epoch:7-----
Total Error:288.8082275390625
-----epoch:8-----
Total Error:288.8082275390625
-----epoch:9-----
Total Error:288.8082275390625
-----epoch:10-----
Total Error:288.8082275390625
-----epoch:11-----
Total Error:288.8082275390625
-----epoch:12-----
Total Error:288.8082275390625
-----epoch:13-----
Total Error:288.8082275390625
-----epoch:14-----
Total Error:288.8082275390625

```

'''To summarize, following are steps for model creation PyTorch:

1. Create Model class in which `__init__()` method contains trainable parameter and forward method

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```

model.train()--- Set model in training mode
pred = model(X)-- Prediction
loss = LossFunction(pred,actual)-- Loss calculation
loss.backward()-- Gradient calculation
optimizer.step()-- Update weights/parameters
optimizer.zero_grad()-- Reset gradients'''

```

☞ 'To summarize, following are steps for model creation PyTorch:\n1. Create Model class i

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