

Adadelata

下面我们实现一下 Adadelata

```
def adadelata(parameters, sqrs, deltas, rho):
    eps = 1e-6
    for param, sqr, delta in zip(parameters, sqrs, deltas):
        sqr[:] = rho * sqr + (1 - rho) * param.grad.data ** 2
        cur_delta = torch.sqrt(delta + eps) / torch.sqrt(sqr + eps) * param.grad.data
        delta[:] = rho * delta + (1 - rho) * cur_delta ** 2
        param.data = param.data - cur_delta
```

```
import numpy as np
import torch
from torchvision.datasets import MNIST # 导入 pytorch 内置的 mnist 数据
from torch.utils.data import DataLoader
from torch import nn
from torch.autograd import Variable
import time
import matplotlib.pyplot as plt
%matplotlib inline

def data_tf(x):
    x = np.array(x, dtype='float32') / 255
    x = (x - 0.5) / 0.5 # 标准化, 这个技巧之后会讲到
    x = x.reshape((-1,)) # 拉平
    x = torch.from_numpy(x)
    return x

train_set = MNIST('./data', train=True, transform=data_tf, download=True) # 载入数据集, 申明定义的数据变换
test_set = MNIST('./data', train=False, transform=data_tf, download=True)

# 定义 loss 函数
criterion = nn.CrossEntropyLoss()
```

```
train_data = DataLoader(train_set, batch_size=64, shuffle=True)
# 使用 Sequential 定义 3 层神经网络
net = nn.Sequential(
    nn.Linear(784, 200),
    nn.ReLU(),
```

```

        nn.Linear(200, 10),
    )

    # 初始化梯度平方项和 delta 项
    sqrs = []
    deltas = []
    for param in net.parameters():
        sqrs.append(torch.zeros_like(param.data))
        deltas.append(torch.zeros_like(param.data))

    # 开始训练
    losses = []
    idx = 0
    start = time.time() # 计时开始
    for e in range(5):
        train_loss = 0
        for im, label in train_data:
            im = Variable(im)
            label = Variable(label)
            # 前向传播
            out = net(im)
            loss = criterion(out, label)
            # 反向传播
            net.zero_grad()
            loss.backward()
            adadelta(net.parameters(), sqrs, deltas, 0.9) # rho 设置为 0.9
            # 记录误差
            train_loss += loss.data[0]
            if idx % 30 == 0:
                losses.append(loss.data[0])
            idx += 1
        print('epoch: {}, Train Loss: {:.6f}'
              .format(e, train_loss / len(train_data)))
    end = time.time() # 计时结束
    print('使用时间: {:.5f} s'.format(end - start))

```

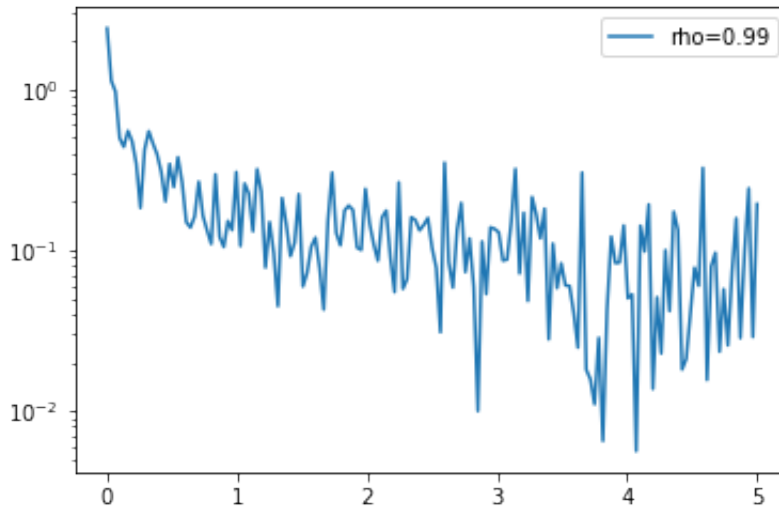
```

epoch: 0, Train Loss: 0.365601
epoch: 1, Train Loss: 0.159966
epoch: 2, Train Loss: 0.123347
epoch: 3, Train Loss: 0.102201
epoch: 4, Train Loss: 0.087986
使用时间: 59.26491 s

```

```
x_axis = np.linspace(0, 5, len(losses), endpoint=True)
plt.semilogy(x_axis, losses, label='rho=0.99')
plt.legend(loc='best')
```

```
<matplotlib.legend.Legend at 0x103f3a5f8>
```



可以看到使用 adadelta 跑 5 次能够得到更小的 loss

小练习：思考一下为什么 **Adadelta** 没有学习率这个参数，它是被什么代替了

当然 pytorch 也内置了 adadelta 的方法，非常简单，只需要调用 `torch.optim.Adadelta()` 就可以了，下面是例子

```
train_data = DataLoader(train_set, batch_size=64, shuffle=True)
# 使用 Sequential 定义 3 层神经网络
net = nn.Sequential(
    nn.Linear(784, 200),
    nn.ReLU(),
    nn.Linear(200, 10),
)

optimizer = torch.optim.Adadelta(net.parameters(), rho=0.9)

# 开始训练
start = time.time() # 记时开始
for e in range(5):
    train_loss = 0
```

```
for im, label in train_data:
    im = Variable(im)
    label = Variable(label)
    # 前向传播
    out = net(im)
    loss = criterion(out, label)
    # 反向传播
    optimizer.zero_grad()
    loss.backward()
    optimizer.step()
    # 记录误差
    train_loss += loss.data[0]
print('epoch: {}, Train Loss: {:.6f}'
      .format(e, train_loss / len(train_data)))
end = time.time() # 计时结束
print('使用时间: {:.5f} s'.format(end - start))
```

```
epoch: 0, Train Loss: 0.356505
epoch: 1, Train Loss: 0.158333
epoch: 2, Train Loss: 0.120510
epoch: 3, Train Loss: 0.100807
epoch: 4, Train Loss: 0.084741
使用时间: 47.90947 s
```

小练习：看看 pytorch 中的 `adadelta`，里面是有学习率这个参数，但是前面我们讲过 `adadelta` 不用设置学习率，看看这个学习率到底是干嘛的