

Pytorch (mini) tutorial

Timon Deschamps

timon.deschamps@univ-lyon1.fr

Original slides by Mathieu Lefort

September 2025

Deep learning framework

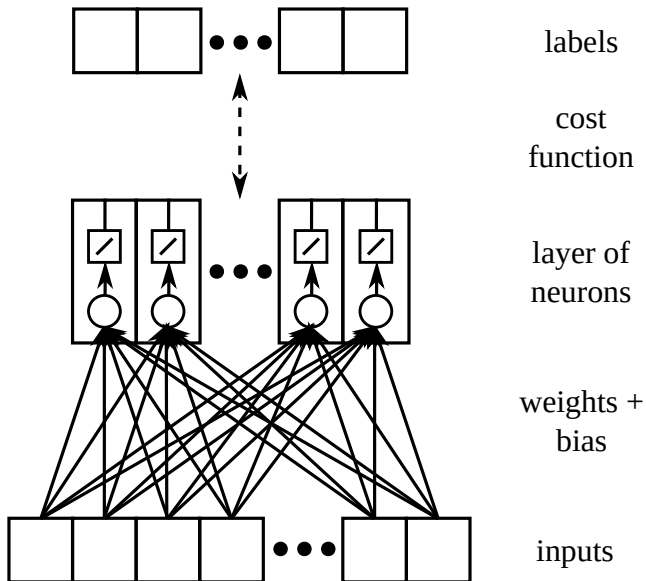
Objectives

- automatic differentiation (on the dependence tree of variables)
- easy to develop
- code sharing

Available

- Theano (Montreal university)
- Pytorch (Facebook)
- Tensorflow (Google)
- Keras

Perceptron architecture



Tensors

Hyperparameters

```
batch_size = 5, nb_epochs = 10, eta = 0.00001, w_min = -0.001, w_max = 0.001
```

Model

```
w = torch.empty((data.shape[1],label.shape[1]))  
b = torch.empty((1,label.shape[1]))  
torch.nn.init.uniform_(w,w_min,w_max)  
torch.nn.init.uniform_(b,w_min,w_max)
```

Data

```
(data,label) = torch.load('mnist.pkl')  
indices = numpy.arange(data.shape[0],step=batch_size)  
for n in range(nb_epochs):  
    numpy.random.shuffle(indices)  
    for i in indices:  
        x = data[i:i+batch_size]  
        y = label[i:i+batch_size]
```

Output

```
 $\hat{y}$  = torch.mm(x,w)+b
```

Learning

```
grad = (y- $\hat{y}$ )  
w += eta * torch.mm(x.T,grad)  
b += eta * grad.sum(axis=0)
```

Loaders

Hyperparameters

```
batch_size = 5, nb_epochs = 10, eta = 0.00001, w_min = -0.001, w_max = 0.001
```

Model

```
w = torch.empty((data.shape[1],label.shape[1]))  
b = torch.empty((1,label.shape[1]))  
torch.nn.init.uniform_(w,w_min,w_max)  
torch.nn.init.uniform_(b,w_min,w_max)
```

Data

```
(data,label) = torch.load('mnist.pkl')  
dataset = torch.utils.data.TensorDataset(data,label)  
loader = torch.utils.data.DataLoader(dataset, batch_size, shuffle=True)  
for n in range(nb_epochs):  
    for x,y in train_loader:
```

Output

```
 $\hat{y} = \text{torch.mm}(x,w)+b$ 
```

Learning

```
grad = ( $\hat{y}$ -y)  
w += eta * torch.mm(x.T,grad)  
b += eta * grad.sum(axis=0)
```

Automatic differentiation

Model

```
w = torch.empty((data.shape[1],label.shape[1]),requires_grad=True)
b = torch.empty((1,label.shape[1]),requires_grad=True)
torch.nn.init.uniform_(w,w_min,w_max)
torch.nn.init.uniform_(b,w_min,w_max)
```

Data

```
(data,label) = torch.load('mnist.pkl')
dataset = torch.utils.data.TensorDataset(data,label)
loader = torch.utils.data.DataLoader(dataset, batch_size, shuffle=True)
for n in range(nb_epochs):
    for x,y in train_loader:
```

Output

```
 $\hat{y}$  = torch.mm(x,w)+b
```

Learning

```
loss = ((y- $\hat{y}$ ).pow(2)).sum()
loss.backward()
with torch.no_grad():
    w -= eta*w.grad
    b -= eta*b.grad
    w.grad.zero_()
    b.grad.zero_()
```

Layers

Hyperparameters

```
batch_size = 5, nb_epochs = 10, eta = 0.00001, w_min = -0.001, w_max = 0.001
```

Model

```
model = torch.nn.Linear(data_train.shape[1], label_train.shape[1])  
torch.nn.init.uniform_(model.weight, w_min, w_max)
```

Data

```
(data, label) = torch.load('mnist.pkl')  
dataset = torch.utils.data.TensorDataset(data, label)  
loader = torch.utils.data.DataLoader(dataset, batch_size, shuffle=True)  
for n in range(nb_epochs):  
    for x, y in train_loader:
```

Output

```
 $\hat{y} = \text{model}(x)$ 
```

Learning

```
loss = ((y -  $\hat{y}$ ).pow(2)).sum()  
loss.backward()  
with torch.no_grad():  
    w -= eta * w.grad  
    b -= eta * b.grad  
    w.grad.zero_()  
    b.grad.zero_()
```

Optimizers

Hyperparameters

```
batch_size = 5, nb_epochs = 10, eta = 0.00001, w_min = -0.001, w_max = 0.001
```

Model

```
model = torch.nn.Linear(data_train.shape[1], label_train.shape[1])  
torch.nn.init.uniform_(model.weight, w_min, w_max)  
loss_func = torch.nn.MSELoss()  
optim = torch.optim.SGD(model.parameters(), lr=eta)
```

Data

```
(data, label) = torch.load('mnist.pkl')  
dataset = torch.utils.data.TensorDataset(data, label)  
loader = torch.utils.data.DataLoader(dataset, batch_size, shuffle=True)  
for n in range(nb_epochs):  
    for x, y in train_loader:
```

Output

```
 $\hat{y}$  = model(x)
```

Learning

```
loss = loss_func(y,  $\hat{y}$ )  
loss.backward()  
optim.step()  
optim.zero_grad()
```


For CNNs

Convolution layer

```
torch.nn.Conv2d(in_channels, out_channels, kernel_size, stride,  
padding, bias=True)
```

Max pooling layer

```
torch.nn.MaxPool2d(kernel_size, stride, padding)
```

Or *max pooling* function

```
torch.nn.functional.max_pool2d(input, kernel_size, stride,  
padding)
```

Modules

```
import torch
import torch.nn as nn
import torch.nn.functional as F
```

```
class MyModel(nn.Module):
```

Constructeur

```
    def __init__(self):
        super(MyModel, self).__init__()
        self.conv1 = nn.Conv2d(1, 6, 5)
        self.conv2 = nn.Conv2d(6, 16, 5)
        self.fc1 = nn.Linear(16 * 5 * 5, 120)
        self.fc2 = nn.Linear(120, 84)
        self.fc3 = nn.Linear(84, 10)
```

Constructeur

```
    def forward(self, x):
        x = F.max_pool2d(F.relu(self.conv1(x)), (2, 2))
        x = F.max_pool2d(F.relu(self.conv2(x)), 2)
        x = torch.flatten(x, 1)
        x = F.relu(self.fc1(x))
        x = F.relu(self.fc2(x))
        x = self.fc3(x)
        return x
```

```
model = MyModel()
```