

# Deep Learning with PyTorch

이동민

삼성전자 서울대 공동연구소 Jul 15, 2019

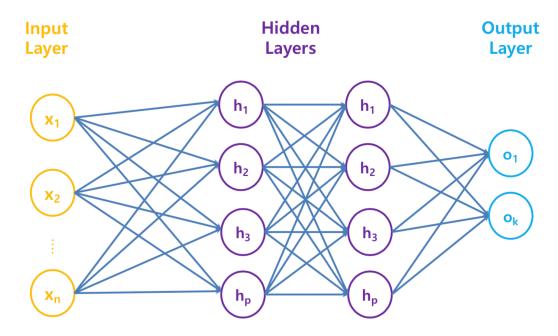
#### Outline

- Deep Learning with PyTorch
  - Tensor Manipulation
  - Deep Neural Network
  - Make Model with PyTorch



## **PyTorch**

- Deep Neural Network도 numpy로 표현할 수 있을까?
  - Deep Neural Network (DNN)



• 신경망이 깊어질수록 numpy로 표현하는 것이 점점 더 어려워짐
→ 따라서 딥러닝 프레임워크를 통해 손쉽게 표현하고자 함



# **PyTorch**











- Numpy array → Torch tensor
  - torch.Tensor Constructs a tensor with data

```
### torch.Tensor
     array = np.array([[1,2,3,4], [5,6,7,8]])
11
12
     tensor = torch.Tensor(array)
     print(array)
13
     print(tensor)
15
     [[1 2 3 4]
16
     [5 6 7 8]]
17
     tensor([[1., 2., 3., 4.],
18
            [5., 6., 7., 8.]])
19
20
21
     array[0,0] = 10
22
     print(array)
23
     print(tensor)
24
25
26
     [[10 2 3 4]
27
     [5 6 7 8]]
     tensor([[1., 2., 3., 4.],
28
29
           [5., 6., 7., 8.]])
30
```



- Numpy array → Torch tensor
  - torch.from\_numpy Creates a Tensor from a numpy.ndarray

```
### torch.from_numpy
     array = np.array([[1,3,5,7], [9,11,13,15]])
33
     tensor = torch.from_numpy(array)
     print(array)
     print(tensor)
37
     [[ 1 3 5 7]
    [ 9 11 13 15]]
     tensor([[ 1, 3, 5, 7],
41
             [ 9, 11, 13, 15]])
42
43
     array[0][0] = 10
     print(array)
45
     print(tensor)
47
     [[10 3 5 7]
     [ 9 11 13 15]]
     tensor([[10, 3, 5, 7],
50
             [ 9, 11, 13, 15]])
51
52
```



- Torch tensor → Numpy array
  - numpy() Returns self tensor as a numpy ndarray



- Creating functions
  - Zeros & Ones

```
### Zeros & Ones
     zeros = torch.zeros((2, 5))
     print(zeros)
76
     tensor([[0., 0., 0., 0., 0.],
              [0., 0., 0., 0., 0.]
78
79
80
     ones = torch.ones((5, 2))
81
82
     print(ones)
83
     tensor([[1., 1.],
84
              [1., 1.],
85
              [1., 1.],
              [1., 1.],
87
              [1., 1.]]
89
```



- Creating functions
  - Something like



- Creating functions
  - Rand Uniform distribution over [0, 1)

Randn - Standard normal(gaussian) distribution of mean 0 and variance 1



- Operation functions
  - o Sum

```
120
      ### Sum
      tensor = torch.Tensor([[1,2,3,4], [5,6,7,8]])
121
122
      sum_ = torch.sum(tensor)
      sum_0 = torch.sum(tensor, dim=0)
123
      sum_1 = torch.sum(tensor, dim=1)
124
125
      print(sum_)
      print(sum 0)
126
      print(sum_1)
127
128
129
      tensor(36.)
130
      tensor([ 6., 8., 10., 12.])
      tensor([10., 26.])
131
132
```



- Operation functions
  - Max

```
134
       tensor = torch.Tensor([[1,2], [3,4], [5,6], [7,8]])
135
136
       max_ = torch.max(tensor)
137
       print(max_)
138
139
       tensor(8.)
       max_0 = torch.max(tensor, dim=0)
                                                                 max_1 = torch.max(tensor, dim=1)
       value, index = torch.max(tensor, dim=0)
                                                                 value, index = torch.max(tensor, dim=1)
                                                                 \max 1 0 = \text{torch.max}(\text{tensor, dim}=1)[0]
       \max 0 0 = \operatorname{torch.max}(\operatorname{tensor}, \operatorname{dim}=0)[0]
       max 0 1 = torch.max(tensor, dim=0)[1]
                                                                 max_1_1 = torch.max(tensor, dim=1)[1]
       print(max 0)
                                                                 print(max_1)
                                                                 print(value)
       print(value)
                                                                 print(index)
149
       print(index)
                                                                 print(max 1 0)
       print(max_0_0)
150
                                                                 print(max_1_1)
       print(max_0_1)
                                                          170
       (tensor([7., 8.]), tensor([3, 3]))
                                                          171
                                                                 tensor([2., 4., 6., 8.])
       tensor([7., 8.])
                                                                 tensor([1, 1, 1, 1])
       tensor([3, 3])
                                                                 tensor([2., 4., 6., 8.])
       tensor([7., 8.])
                                                                 tensor([1, 1, 1, 1])
       tensor([3, 3])
```



- Operation functions
  - Dot product
    - torch.dot Computes the dot product (inner product) of two tensors (1-Dimension)

$$\langle [x_1, ..., x_n], [y_1, ..., y_n] \rangle = x^T y = \sum_{i=1}^n x_i y_i = x_1 y_1 + \dots + x_n y_n$$

```
### Dot product
tensor = torch.Tensor([1,2,3,4,5])

dot = torch.dot(tensor, tensor)
print(dot)

tensor(55.)

183
```



- Operation functions
  - Mathematical functions
    - torch.sqrt:  $\sqrt{x}$
    - torch.exp:  $e^x$
    - torch.log:  $\log_e x$

```
### Mathematical functions
185
      tensor = torch. Tensor([[1,2,3,4], [5,6,7,8]])
186
187
      sqrt = torch.sqrt(tensor)
188
      exp = torch.exp(tensor)
189
      log = torch.log(tensor)
190
191
      print(sqrt)
192
      print(exp)
      print(log)
193
194
195
      tensor([[1.0000, 1.4142, 1.7321, 2.0000],
196
              [2.2361, 2.4495, 2.6458, 2.8284]])
197
      tensor([[ 2.7183, 7.3891, 20.0855,
                                                   54.5982],
198
              [ 148.4132, 403.4288, 1096.6332, 2980.9580]])
      tensor([[0.0000, 0.6931, 1.0986, 1.3863],
199
200
              [1.6094, 1.7918, 1.9459, 2.0794]])
201
```



- Operation functions
  - Concatenate

```
### Concatenate
204
      tensor_a = torch.Tensor([[1,2,3,4], [5,6,7,8]])
      tensor_b = torch.Tensor([[1,3,5,7], [2,4,6,8]])
      cat = torch.cat([tensor_a, tensor_b]) # vstack
208
      print(cat)
209
210
      tensor([[1., 2., 3., 4.],
              [2., 4., 6., 8.]])
      cat_0 = torch.cat([tensor_a, tensor_b], dim=0) # vstack
216
      print(cat_0)
218
219
              [5., 6., 7., 8.],
              [2., 4., 6., 8.]])
      cat_1 = torch.cat([tensor_a, tensor_b], dim=1) # hstack
226
      print(cat_1)
      tensor([[1., 2., 3., 4., 1., 3., 5., 7.],
              [5., 6., 7., 8., 2., 4., 6., 8.]])
```



- Operation functions
  - View
    - torch.view Returns a new tensor with the same data as the self tensor but of a different shape

```
232
      ### View
      tensor_a = torch.Tensor([[1,3,5,7], [2,4,6,8]])
233
234
235
      tensor_b = tensor_a.view(8)
236
      print(tensor b.shape)
237
238
      torch.Size([8])
239
240
241
      tensor c = tensor a.view(-1, 2)
      print(tensor_c.shape)
242
243
244
      torch.Size([4, 2])
245
246
247
      tensor_d = tensor_a.view(-1)
248
      print(tensor_d.shape)
249
250
      torch.Size([8])
251
```



- Operation functions
  - Squeeze torch.squeeze(input, dim=None)
    - Returns a tensor with all the dimensions of input of size 1 removed

```
253
       ### Squeeze & Unsqueeze
       tensor = torch.zeros(2, 1, 1, 5)
254
255
256
       squ 0 = torch.squeeze(tensor)
257
       print(squ 0.shape)
258
       torch.Size([2, 5])
259
260
261
       squ_1 = torch.squeeze(tensor, 1)
262
       print(squ_1.shape)
263
       print(tensor.squeeze(1).shape)
264
265
       torch.Size([2, 1, 5])
266
       torch.Size([2, 1, 5])
267
        \mathbf{I} \cdot \mathbf{I} \cdot \mathbf{I}
268
```



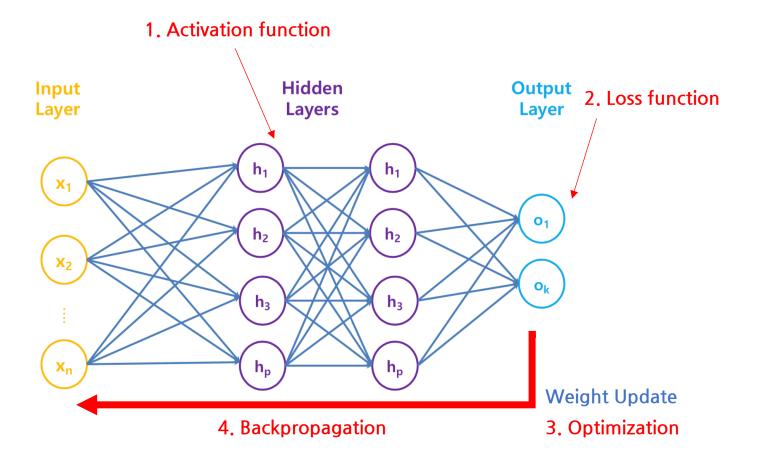
- Operation functions
  - Unsqueeze torch.unsqueeze(input, dim)
    - Returns a new tensor with a dimension of size one inserted at the specified position

```
unsqu 0 = torch.unsqueeze(tensor, 0)
270
271
       print(unsqu_0.shape)
272
        torch.Size([1, 2, 1, 1, 5])
273
274
275
276
       unsqu_1 = torch.unsqueeze(tensor, 1)
       print(unsqu_1.shape)
277
        print(tensor.unsqueeze(1).shape)
278
279
280
       torch.Size([2, 1, 1, 1, 5])
        torch.Size([2, 1, 1, 1, 5])
281
282
        \mathbf{I} \cdot \mathbf{I} \cdot \mathbf{I}
```



# **Deep Neural Network**

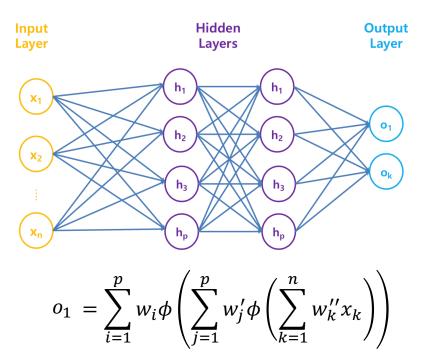
Deep Neural Network (DNN) process





PyTorch example for Deep Neural Network (DNN)

```
294
      import torch
295
      import torch.nn as nn
      import torch.optim as optim
296
297
298
      class Net(nn.Module):
           def __init__(self):
299
               super(Net, self).__init__()
301
               self.fc1 = nn.Linear(4, 64)
               self.fc2 = nn.Linear(64, 64)
302
303
               self.fc3 = nn.Linear(64, 2)
304
305
          def forward(self, x):
               x = torch.tanh(self.fc1(x))
306
               x = torch.tanh(self.fc2(x))
               x = self.fc3(x)
309
               return x
```





PyTorch example for Deep Neural Network (DNN)

```
294
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296
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      class Net(nn.Module):
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           def __init__(self):
               super(Net, self).__init__()
301
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               self.fc2 = nn.Linear(64, 64)
302
303
               self.fc3 = nn.Linear(64, 2)
304
305
           def forward(self, x):
               x = torch.tanh(self.fc1(x))
306
               x = torch.tanh(self.fc2(x))
               x = self.fc3(x)
309
               return x
```

```
311
      net = Net()
312
313
      # Define loss and optimizer
      criterion = torch.nn.MSELoss()
314
315
      optimizer = optim.Adam(net.parameters(), lr=0.001)
316
317
      net.train()
318
319
320
      hypothesis = net(inputs)
       loss = criterion(hypothesis, labels)
321
322
323
      optimizer.zero_grad() # initialize gradient
       loss.backward()
                             # compute gradient
324
325
      optimizer.step()
                             # improve step
```



Save & Load

```
### save
torch.save(net.state_dict(), './save_model/model.pth')
### load
net.load_state_dict(torch.load('./save_model/model.pth'))
```



MNIST example



# Thank you

