

Deep Learning Toolkits II

PyTorch

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A graph is created on the fly

```
from torch.autograd import Variable

x = Variable(torch.randn(1, 10))
prev_h = Variable(torch.randn(1, 20))
W_h = Variable(torch.randn(20, 20))
W_x = Variable(torch.randn(20, 10))
```



Tensors and Dynamic neural networks in Python with strong GPU acceleration.

PyTorch is a deep learning framework that puts Python first.

We are in an early-release Beta. Expect some adventures.

[Learn More](#)

Tensor computation (like numpy) with GPU

Deep learning with automatic differentiation

Developers

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Outlines

- Examples
 - PyTorch as a fast calculator
 - Train a CNN with PyTorch
- Basic concepts
- Write new models
- HOWTOs

Example: A fast calculator

```
1 import numpy as np
2 import torch
3
4 # Task: compute matrix multiplication C = AB
5 d = 3000
6
7 # using numpy
8 A = np.random.rand(d, d).astype(np.float32)
9 B = np.random.rand(d, d).astype(np.float32)
10 C = A.dot(B)
11
12 # using torch with gpu
13 A = torch.rand(d, d).cuda()
14 B = torch.rand(d, d).cuda()
15 C = torch.mm(A, B)
```



350 ms



0.1 ms

Example: Auto Differentiate

```
1  import torch
2  from torch.autograd import Variable
3
4  # Task: compute  $d(||x||^2)/dx$ 
5  x = Variable(torch.range(1, 5), requires_grad=True)
6  print(x.data)  # x.data = [1, 2, 3, 4, 5]
7
8  f = x.dot(x)
9  print(f.data)  # f.data = 55
10
11 f.backward()
12 print(x.grad)  # x.grad = [2, 4, 6, 8, 10]
```

Example: Train a CNN

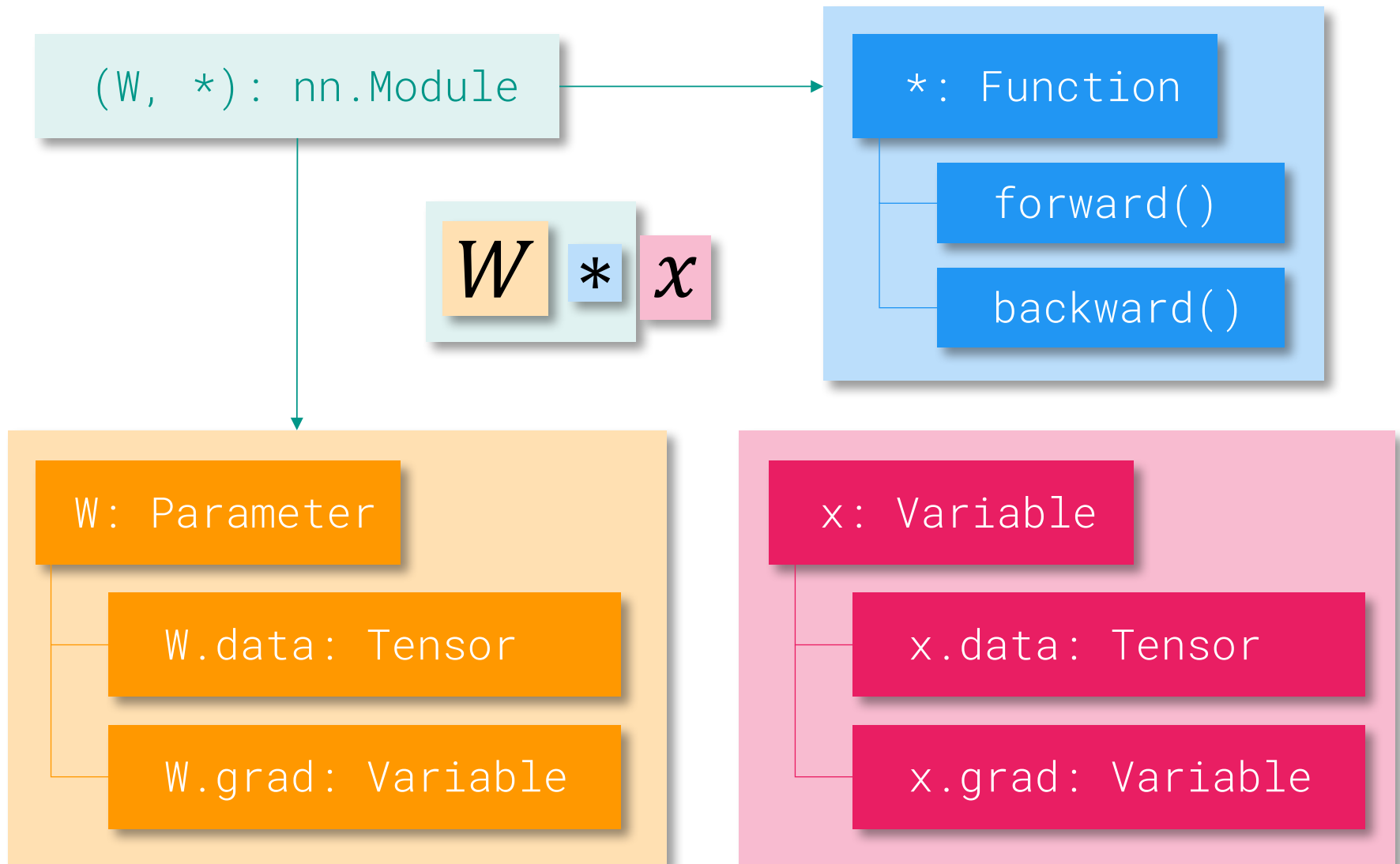
- See our notebook example

<http://nbviewer.jupyter.org/gist/Cysu/320960298b6ccaedb778a2a4de2dd2db>

Basic Concepts

- **torch.Tensor** – similar to `numpy.array`, with GPU
- **autograd.Variable** – wraps a Tensor and enables auto differentiation
- **autograd.Function** – operate on Variables. Implement forward and backward.
- **nn.Parameter** – a special Variable
- **nn.Module** – contain Parameters and define functions on input Variables

Basic Concepts



Write New Models

```
1  from torch import nn
2
3  class SimpleCNN(nn.Module):
4      def __init__(self):
5          super(SimpleCNN, self).__init__()
6
7          self.block = nn.Sequential(
8              nn.Conv2d(3, 16, kernel_size=3, padding=1),
9              nn.BatchNorm2d(16),
10             nn.ReLU(),
11             nn.MaxPool2d(kernel_size=4, stride=4))
12         self.fc = nn.Linear(64*16, 10)
13
14     def forward(self, x):
15         x = self.block(x)
16         x = x.view(x.size(0), -1)
17         x = self.fc(x)
18         return x
```

Group into
nn.Sequential

Write New Models

```
1 from torch import nn
2 import torch.nn.functional as F
3
4 class SimpleCNN(nn.Module):
5     def __init__(self):
6         super(SimpleCNN, self).__init__()
7         self.conv = nn.Conv2d(3, 16, kernel_size=3, padding=1)
8         self.bn = BatchNorm2d(16)
9         self.fc = nn.Linear(64*16, 10)
10
11     def forward(self, x):
12         x = self.conv(x)
13         x = F.relu(x)
14         x = F.max_pool2d(x, 4, stride=4)
15         x = x.view(x.size(0), -1)
16         x = self.fc(x)
17         return x
```

Functional
interface

Not recommended, but
sometimes useful

Write New Models

- Combine high-level `nn.Module` and low-level math operations on `Variable`
- Use functional interfaces smartly, e.g., convolve two feature maps using `F.conv2d`
- See <http://pytorch.org/docs/notes/extending.html> for more details

HOWTOs

- Initialize parameters
- Fine-tuning
- Extract features
- Two-branch modules

Initialize Parameters

```
1 class SimpleCNN(nn.Module):
2     def __init__(self):
3         # Create some modules here
4         ...
5         for m in self.modules():
6             if isinstance(m, nn.Conv2d):
7                 n = m.kernel_size[0] * m.kernel_size[1] * m.out_channels
8                 m.weight.data.normal_(0, math.sqrt(2. / n))
9                 if m.bias is not None:
10                     m.bias.data.zero_()
11             elif isinstance(m, nn.BatchNorm2d):
12                 m.weight.data.fill_(1)
13                 m.bias.data.zero_()
14             elif isinstance(m, nn.Linear):
15                 m.weight.data.normal_(0, 0.001)
16                 if m.bias is not None:
17                     m.bias.data.zero_()
```

Any operation that changes a tensor in-place is post-fixed with an _

Fine-tuning

```
1 model = torchvision.models.resnet18(pretrained=True)
2 model.fc = nn.Linear(512, 100)
3
4 new_param_ids = set(id(p) for p in model.fc.parameters())
5 base_params = [p for p in model.parameters() if id(p) not in new_param_ids]
6
7 optimizer = optim.SGD([
8     {'params': base_params},
9     {'params': model.fc.parameters(), 'lr': 0.1}],
10    lr=0.01, momentum=0.9)
```

Replace the classifier

Different lr for pretrained layers and new layers

Extract Features

- Make a new class FeatureExtractor that holds a pretrained model
- Redefine the forward function, call the layers in the pretrained model explicitly, save the features we want for output
- Use `model._modules`

Extract Features – model._modules

```
1 OrderedDict([
2   ('conv1', Conv2d(3, 64, kernel_size=(7, 7), stride=(2, 2), padding=(3, 3), bias=False)),
3   ('bn1', BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True)),
4   ('relu', ReLU (inplace)),
5   ('maxpool', MaxPool2d (size=(3, 3), stride=(2, 2), padding=(1, 1), dilation=(1, 1))),
6   ('layer1', Sequential (
7     (0): BasicBlock (
8       (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
9       (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True)
10      (relu): ReLU (inplace)
11      (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
12      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True)
13    )
14    (1): BasicBlock (
15      (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
16      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True)
17      (relu): ReLU (inplace)
18      (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
19      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True)
20    )
21  )),
22  ...
```

Extract Features

```
1 class FeatureExtractor(nn.Module):
2     def __init__(self, model, layer_names):
3         super(FeatureExtractor, self).__init__()
4         self._model = model
5         self._layer_names = set(layer_names)
6
7     def forward(self, x):
8         outs = {}
9         for name, module in self._model._modules.items():
10             if isinstance(module, nn.Linear):
11                 x = x.view(x.size(0), -1)
12                 x = module(x)
13                 if name in self._layer_names:
14                     outs[name] = x
15         return outs
16
```

Not included in the `_modules`.
So call it explicitly.

Extract Features

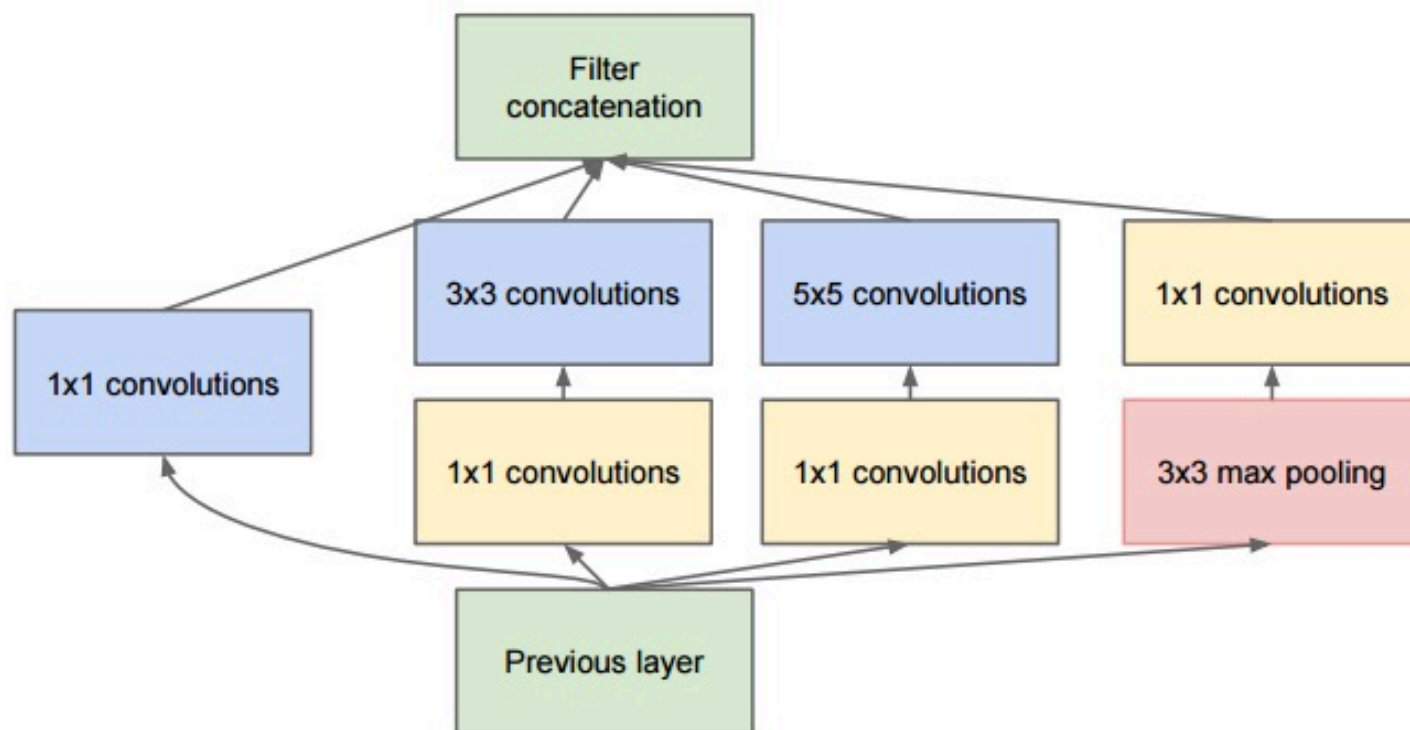
Remember to call `model.eval()`.
Affects BN, Dropout, etc.

```
17 model = torchvision.models.resnet18(pretrained=True).cuda().eval()
18 extractor = FeatureExtractor(model, ['avgpool', 'fc'])
19
20 x = Variable(torch.randn(1, 3, 224, 224).cuda(), volatile=True)
21 outs = extractor.forward(x)
```

Don't call `extractor(x)`
directly

Reduces memory cost significantly!
All non-output Tensors could be corrupted.

Two-branch Modules



Two-branch Modules

```
1 class InceptionBlock(nn.Module):
2     def __init__(self, num_in, num_out):
3         super(InceptionBlock, self).__init__()
4         self.branches = [
5             nn.Sequential(
6                 nn.Conv2d(num_in, num_out, kernel_size=1),
7                 nn.ReLU()),
8             nn.Sequential(
9                 nn.Conv2d(num_in, num_out, kernel_size=1),
10                nn.ReLU(),
11                nn.Conv2d(num_out, num_out, kernel_size=3, padding=1),
12                nn.ReLU()),
13            ...
14        ]
15        for i, branch in enumerate(self.branches):
16            self.add_module(str(i), branch)
17
18    def forward(self, x):
19        # Concatenate branch results along channels
20        return torch.cat([b(x) for b in self.branches], 1)
```

Need to call `add_module` to register the branches!

Summary

- Examples
- Basic concepts
 - torch – Tensor
 - autograd – Variable, Function
 - nn – Module, Parameter
- Write new models
- HOWTOs

More Adventures

- Tutorials
 - official <https://github.com/pytorch/tutorials>
 - jcjohnson <https://github.com/jcjohnson/pytorch-examples>
 - RNN and NLP <https://github.com/spro/practical-pytorch>
- Examples <https://github.com/pytorch/examples>
- Docs <http://pytorch.org/docs/>
- Discussions <https://discuss.pytorch.org/>

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