

Introduction to PyTorch

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What is PyTorch?

- `Ndarray` library with GPU support
- Automatic differentiation engine
 - Gradient based optimisation package
- Neural network primitives (cuDNN integration)
- Utilities: Data loading, model zoo, distributed computation, etc.

Tensor library

`torch.Tensor`

- `torch.Tensor` is similar to `ndarray` in NumPy
- `np.ndarray` \leftrightarrow `torch.Tensor` can be replaced in quite a few places
- A lot of operations defined for Tensors
<http://pytorch.org/docs/master/tensors.html>
- Acceleration through CUDA

```
[In [26]: %timeit a + b
100 loops, best of 3: 2.64 ms per loop

[In [27]: %timeit a_thc + b_thc
10000 loops, best of 3: 125 us per loop]
```

Automatic Differentiation through autograd

- Compute gradients for Tensor operations
- `X = Variable(torch.randn(10))`
- `W = Variable(torch.randn(10))`
- `B = Variable(torch.randn(1))`
- `T = Variable(torch.randn(1))`
- `Y = W.X + B`
- `L = Y - T`
- Gradient: dL / dW

Automatic Differentiation through autograd

- Wrap Tensor objects in `Variable`
- `autograd` tracks the sequence of operations applied to `Variables`
- Call `.backward()` to compute gradients
 - It is called `.backward()` because of the back propagation algorithm
- `Variable` has two important attributes:
 - `grad_fn`: A `Function` which created the `Variable`
 - `grad`: `Variable` holding the gradient

Automatic Differentiation through autograd

- autograd can be extended to include operations which are not defined
- Look at implementations of existing functions:
https://github.com/pytorch/pytorch/tree/master/torch/autograd/_functions
- Check correctness against numerical gradients:
<https://github.com/pytorch/pytorch/blob/master/torch/autograd/gradcheck.py>
- Documentation: <http://pytorch.org/docs/master/autograd.html>
- The API will change soon. Wrapping in `Variable` will not be required.

Neural Networks

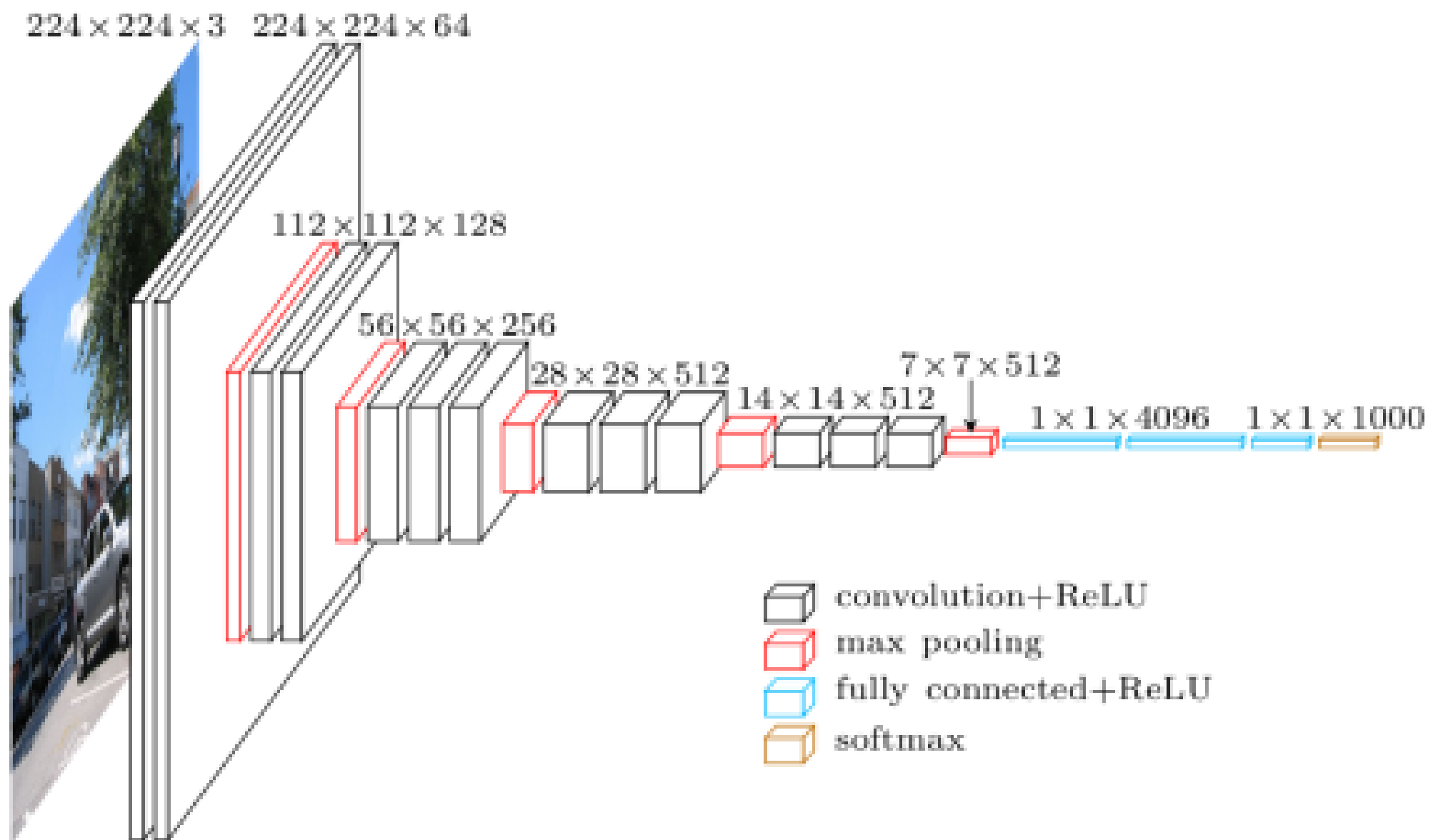
`torch.nn`

- Neural network primitives present in `torch.nn` package
- `nn.Module` is used to encapsulate trainable parameters and define their operations
- `nn.Parameter` is a kind of `Variable` which is registered for *training*.

Neural Networks

torch.nn

```
class Net(nn.Module):  
    def __init__(self):  
        super(Net, self).__init__()  
        self.conv1 = nn.Conv2d(3, 6, 5)  
        self.pool = nn.MaxPool2d(2, 2)  
        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)  
  
    def forward(self, x):  
        x = self.pool(F.relu(self.conv1(x)))  
        x = self.pool(F.relu(self.conv2(x)))  
        x = x.view(-1, 16 * 5 * 5)  
        x = F.relu(self.fc1(x))  
        x = F.relu(self.fc2(x))  
        x = self.fc3(x)  
        return x
```

Recap

- `torch.Tensor` - *A multi-dimensional array.*
- `autograd.Variable` - *Wraps a Tensor and records the history of operations applied to it. Has the same API as a `Tensor`, with some additions like `backward()`. Also holds the gradient w.r.t. the tensor.*
- `nn.Module` - *Neural network module. Convenient way of encapsulating parameters, with helpers for moving them to GPU, exporting, loading, etc.*
- `nn.Parameter` - *A kind of Variable, that is automatically registered as a parameter when assigned as an attribute to a `Module`.*
- `autograd.Function` - *Implements forward and backward definitions of an autograd operation. Every `Variable` operation, creates at least a single `Function` node, that connects to functions that created a `Variable` and encodes its history.*

Next Steps

- Advanced examples: <https://github.com/pytorch/examples>
- Explore models:
<https://github.com/pytorch/vision/tree/master/torchvision/models>
- Introductory DL course with coding assignments:
<http://cs231n.github.io>
- Theory of DL: <https://stats385.github.io>

PyTorch Alternatives

- TensorFlow Eager
- CuPy Chainer (~99% NumPy replacement)
- MxNet