O PyTorch Basics

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What is Tensor?

A tensor is a mathematical object that generalizes the concepts of scalars, vectors, and matrices to higher dimensions. It can be thought of as a multi-dimensional array of numbers that can represent data or physical quantities.

Order	Example	Shape	Interpretation
0 (Scalar)	s = 5	No shape	Single number
1 (Vector)	[3, 4, 5]	3 imes 1	List of numbers
2 (Matrix)	$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$	2 imes 2	Grid of numbers
3 (Tensor)	RGB image	32 imes 32 imes 3	Cube of numbers
4+ (Tensor)	Batch of images	$100\times32\times32\times3$	Multi-dimensional array

- 1. Tensor Intitalization
- 2. Tensor Maths
- 3. Tensor Indexing
- 4. Tensor Reshaping

Why Pytorch?

Dynamic Computation Graphs – Build networks on the fly.

Pythonic – Easy to learn and implement.

Extensive Community Support – Strong community contributions and resources.

Research to Production – PyTorch serves both research purposes and production-

level deployment.

Seamless GPU Support – Simplifies GPU-based computation.

1. Tensor Intialization

```
device = "cuda" if torch.cuda.is_available() else "cpu"
```

```
torch_tensor = torch.tensor([[1,2,3],[4,5,6]], dtype = torch.float32, device = device, requires_grad = True)
```

Other methods for torch tensor declaration

```
X = torch.empty((2,2)),
X = torch.rand((2,2))= torch.empty((2,2)).uniform_(0,1),
X = torch.ones((2,2)),
X = torch.zeros((2,2)),
X = torch.eye((2,2)) = torch.diag(torch.ones((2,2))),
X = torch.arange(start=0,end=5, step=1),
X = torch.linspace(start=0.1, end=1, steps=10)
```

1. Tensor Intialization

Datatype conversion in torch

```
X = torch.empty((5,5)), # dtype = int-32
X = X.bool()
                  # Convert into binary form
X = X.short()
                  # Convert into dtype int-16
X = X.long()
                  # Convert into dtype int-64
X = X.half()
                  # Convert into dtype float-16
X = X.float()
                  # Convert into dtype float-32
X = X.double()
                    # Convert into dtype float-64
```

Torch to array conversion and vise-versa

```
X = np.zeros((6,6))
                   # Declare numpy array
X = torch.from numpy(X) # Convert to torch tensor
X = X.numpy()
                   # Convert to numpy array
```

2. Tensor Maths

```
X1 = torch.tensor([1,2,3]), X2 = torch.tensor([4,5,6])
```

Addition: X1 + X2 = torch.add(X1, X2)

Inplace operation --> X2 += X1 or X2._add(X1)

Subraction: X1 - X2

Division: torch.true_divide(X1, X2)

Exponential: z = X1.pow(2) = X1 ** 2

Comparision: Z = X1>0, X2>X1, X2<X1

2. Tensor Maths

```
X1 = torch.tensor([1,2,3]), X2 = torch.tensor([4,5,6])
```

Metrix Multiplication: torch.mm(X1, X2) = X1.mm(X2)

Element-wise Multiplication: X1 * X2

Dot product: torch.dot(X1, X2)

Batch Metrix Multiplication:

batch =
$$2$$
, a = 3 , b = 5 , c = 4

$$Z = torch.bmm(X1,X2)$$

2. Tensor Maths

```
X1 = torch.tensor([1,2,3]), X2 = torch.tensor([4,5,6])
indices, values = torch.max(X1, dim=0), indices = torch.argmax(X1, dim=0)
Absolute tensor: torch.abs(X1)
```

Tensor Clamping: torch.clamp(X1, min = 0)

3. Tensor Indexing

```
batch_size = 2, features = 5
x = torch.rand((batch_size), features)
z = x[0].shape # Get the shape of first tensor
z = x[:,0].shape # Get the shape of all tensors at the first dimension
Conditional Formatting: Z = torch.arange(10)
Z[(Z>2) | (Z<8)] # Get the elements greater than 2 and less than 8
Z[Z.remainder(2)==0] # Get the even numbered elements
Z = torch.where(X>5, X, X*2)
```

4. Tensor Reshaping

```
X = torch.tensor([1,2,3,4])
Z = X.view(2,2) # When X is contiguous
Z = X.reshape(2,2) # Independent of contiguity of X
Concatenation: X1 = torch.rand([2,4]), X2 = torch.rand([2,3])
Z1 = torch.cat((X1, X2), dim=0), Z2 = torch.cat(X1, X2, dim=1)
T = torch.tensor([1,2,3])
T = T.permute(0,2,1)
T = T.unsqueeze(0)
```

THANK YOU



https://www.iitg.ac.in/stud/suklav/



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