

```
In [2]: import torch
```

```
In [3]: x = torch.tensor(3.0)
y = torch.tensor(2.0)

x+y, x*y, x/y, x**y
```

```
Out[3]: (tensor(5.), tensor(6.), tensor(1.5000), tensor(9.))
```

```
In [4]: x = torch.arange(3)
x
```

```
Out[4]: tensor([0, 1, 2])
```

```
In [5]: x[2]
```

```
Out[5]: tensor(2)
```

```
In [6]: len(x)
```

```
Out[6]: 3
```

```
In [7]: x.shape
```

```
Out[7]: torch.Size([3])
```

```
In [8]: A = torch.arange(6).reshape(3,2)
A
```

```
Out[8]: tensor([[0, 1],
               [2, 3],
               [4, 5]])
```

```
In [9]: A.T
```

```
Out[9]: tensor([[0, 2, 4],
               [1, 3, 5]])
```

```
In [10]: A = torch.tensor([[1, 2, 3], [2, 0, 4], [3, 4, 5]])
A == A.T
```

```
Out[10]: tensor([[True, True, True],
                [True, True, True],
                [True, True, True]])
```

```
In [11]: torch.arange(24).reshape(2,3,4)
```

```
Out[11]: tensor([[[ 0,  1,  2,  3],
                  [ 4,  5,  6,  7],
                  [ 8,  9, 10, 11]],
                 [[12, 13, 14, 15],
                  [16, 17, 18, 19],
                  [20, 21, 22, 23]]])
```

```
In [12]: A = torch.arange(6, dtype=torch.float32).reshape(2,3)
B = A.clone()
A, A+B
```

```
Out[12]: (tensor([[0., 1., 2.],
                  [3., 4., 5.]]),
          tensor([[0., 2., 4.],
                  [6., 8., 10.]])
```

```
In [13]: A*B
```

```
Out[13]: tensor([[0., 1., 4.],
                  [9., 16., 25.]])
```

```
In [14]: a = 2
X = torch.arange(24).reshape(2,3,4)
a*X, (a*x).shape
```

```
Out[14]: (tensor([[[ 2,  3,  4,  5],
                   [ 6,  7,  8,  9],
                   [10, 11, 12, 13]],

                  [[14, 15, 16, 17],
                   [18, 19, 20, 21],
                   [22, 23, 24, 25]]]),
          torch.Size([3]))
```

```
In [15]: x = torch.arange(3, dtype=torch.float32)
x, x.sum()
```

```
Out[15]: (tensor([0., 1., 2.]), tensor(3.))
```

```
In [16]: A.shape, A.sum()
```

```
Out[16]: (torch.Size([2, 3]), tensor(15.))
```

```
In [17]: A.shape, A.sum(axis=0).shape
```

```
Out[17]: (torch.Size([2, 3]), torch.Size([3]))
```

```
In [18]: A.shape, A.sum(axis=1).shape
```

```
Out[18]: (torch.Size([2, 3]), torch.Size([2]))
```

```
In [20]: A.sum(axis=[0,1]) == A.sum()
```

```
Out[20]: tensor(True)
```

```
In [21]: A.mean(), A.sum() / A.numel()
```

```
Out[21]: (tensor(2.5000), tensor(2.5000))
```

```
In [22]: A.mean(axis=0), A.sum(axis=0) / A.shape[0]
```

```
Out[22]: (tensor([1.5000, 2.5000, 3.5000]), tensor([1.5000, 2.5000, 3.5000]))
```

```
In [23]: sum_A = A.sum(axis=1, keepdims=True)
sum_A, sum_A.shape
```

```
Out[23]: (tensor([[ 3.],
                  [12.]]),
          torch.Size([2, 1]))
```

```
In [24]: A/sum_A
```

```
Out[24]: tensor([[0.0000, 0.3333, 0.6667],
                 [0.2500, 0.3333, 0.4167]])
```

```
In [25]: A.cumsum(axis=0)
```

```
Out[25]: tensor([[0., 1., 2.],
                 [3., 5., 7.]])
```

```
In [26]: y = torch.ones(3, dtype=torch.float32)
x,y,torch.dot(x,y)
```

```
Out[26]: (tensor([0., 1., 2.]), tensor([1., 1., 1.]), tensor(3.))
```

```
In [27]: torch.sum(x*y)
```

```
Out[27]: tensor(3.)
```

```
In [29]: A.shape, x.shape, torch.mv(A,x), A@x
```

```
Out[29]: (torch.Size([2, 3]), torch.Size([3]), tensor([ 5., 14.]), tensor([ 5., 14.]))
```

```
In [30]: B = torch.ones(3,4)
torch.mm(A,B), A@B
```

```
Out[30]: (tensor([[ 3.,  3.,  3.,  3.],
                  [12., 12., 12., 12.]]),
          tensor([[ 3.,  3.,  3.,  3.],
                  [12., 12., 12., 12.]])
```

```
In [31]: u = torch.tensor([3.0,-4.0])
torch.norm(u)
```

```
Out[31]: tensor(5.)
```

```
In [32]: torch.abs(u).sum()
```

```
Out[32]: tensor(7.)
```

```
In [34]: torch.norm(torch.ones((4,9)))
```

```
Out[34]: tensor(6.)
```

Discussion: Linear algebra was one of the first subjects I studied in my first semester, but unfortunately, I wasn't very focused on my studies back then, so there are many gaps in my understanding of the concepts. Nevertheless, while working on this unit's exercises, I was able to revisit parts that I had forgotten, such as matrix transposition. The most useful part of this unit was realizing how easily matrix-vector and matrix-matrix multiplication can be done using the "@" operator.

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