

# TensorFlow线性代数

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## 1. 标量运算

```
In [1]: import tensorflow as tf
```

```
In [2]: x = tf.constant([3.0])
```

```
In [3]: y = tf.constant([2.0])
```

```
In [4]: sess = tf.Session()
```

```
In [5]: sess.run([x + y, x * y, x / y, x**y])
```

```
Out[5]: [array([5.], dtype=float32),  
         array([6.], dtype=float32),  
         array([1.5], dtype=float32),  
         array([9.], dtype=float32)]
```

## 2. 向量运算

```
In [6]: x = tf.range(12)
```

```
In [7]: sess.run(x)
```

```
Out[7]: array([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11])
```

```
In [8]: sess.run(x[3])
```

```
Out[8]: 3
```

```
In [9]: x.shape
```

```
Out[9]: TensorShape([Dimension(12)])
```

```
In [10]: y = tf.ones(12, dtype=tf.float32)
```

```
In [11]: sess.run(y)
```

```
Out[11]: array([1., 1., 1., 1., 1., 1., 1., 1., 1., 1., 1., 1.], dtype=float32)
```

## 转换数据类型

```
In [12]: x = tf.cast(x, "float32")
```

点积：给定两个向量 $\mathbf{x}, \mathbf{y} \in \mathbb{R}^d$ ，它们的点积 (dotproduct)  $\mathbf{x}^\top \mathbf{y}$  (或 $\langle \mathbf{x}, \mathbf{y} \rangle$ ) 是相同位置的按元素乘积的和： $\mathbf{x}^\top \mathbf{y} = \sum_{i=1}^d x_i y_i$ 。

```
In [13]: sess.run(tf.tensordot(x, y, axes=1))
```

```
Out[13]: 66.0
```

```
In [14]: sess.run(tf.reduce_sum(x * y))
```

```
Out[14]: 66.0
```

### 向量所有元素相乘

```
In [15]: sess.run(tf.reduce_prod(x))
```

```
Out[15]: 0.0
```

```
In [16]: sess.run(tf.reduce_prod(y))
```

```
Out[16]: 1.0
```

```
In [17]: sess.run(x + y)
```

```
Out[17]: array([ 1.,  2.,  3.,  4.,  5.,  6.,  7.,  8.,  9., 10., 11., 12.],  
              dtype=float32)
```

### 向量所有元素相加

```
In [18]: sess.run(tf.reduce_sum(x))
```

```
Out[18]: 66.0
```

```
In [19]: sess.run(tf.reduce_sum(y))
```

```
Out[19]: 12.0
```

### 求均值

```
In [20]: sess.run(tf.reduce_mean(x))
```

```
Out[20]: 5.5
```

```
In [21]: sess.run(tf.size(x))
```

```
Out[21]: 12
```

```
In [22]: x_size = tf.cast(tf.size(x), "float32")
```

```
In [23]: sess.run(tf.reduce_sum(x) / x_size)
```

```
Out[23]: 5.5
```

将向量转化为矩阵

```
In [24]: sess.run(tf.reshape(x, (3, 4)))
```

```
Out[24]: array([[ 0.,  1.,  2.,  3.],
                [ 4.,  5.,  6.,  7.],
                [ 8.,  9., 10., 11.]], dtype=float32)
```

### 3. 矩阵/张量运算

```
In [25]: X = tf.reshape(tf.range(12, dtype=tf.float32), (3, 4))
```

```
In [26]: sess.run(X)
```

```
Out[26]: array([[ 0.,  1.,  2.,  3.],
                [ 4.,  5.,  6.,  7.],
                [ 8.,  9., 10., 11.]], dtype=float32)
```

axis=0时, 返回矩阵X每一列最大元素所在下标

```
In [27]: sess.run(tf.argmax(X, 0))
```

```
Out[27]: array([2, 2, 2, 2], dtype=int64)
```

axis=1时, 返回矩阵X每一行最大元素所在下标

```
In [28]: sess.run(tf.argmax(X, 1))
```

```
Out[28]: array([3, 3, 3], dtype=int64)
```

axis=0时, 返回矩阵X每一列求和结果

```
In [29]: sess.run(tf.reduce_sum(X, axis=0))
```

```
Out[29]: array([12., 15., 18., 21.], dtype=float32)
```

axis=1时, 返回矩阵X每一行求和结果

```
In [30]: sess.run(tf.reduce_sum(X, axis=1))
```

```
Out[30]: array([ 6., 22., 38.], dtype=float32)
```

axis=[0, 1], 先对列求和, 再对行求和, 即矩阵所有元素相加的结果

```
In [31]: sess.run(tf.reduce_sum(X, axis=[0, 1]))
```

```
Out[31]: 66.0
```

```
In [32]: sess.run(tf.reduce_sum(X))
```

```
Out[32]: 66.0
```

```
In [33]: Y = tf.constant([[2.0, 1, 4, 3], [1, 2, 3, 4], [4, 3, 2, 1]])
```

```
In [34]: sess.run(Y)
```

```
Out[34]: array([[2., 1., 4., 3.],  
               [1., 2., 3., 4.],  
               [4., 3., 2., 1.]], dtype=float32)
```

```
In [35]: sess.run(tf.argmax(Y, 0))
```

```
Out[35]: array([2, 2, 0, 1], dtype=int64)
```

```
In [36]: sess.run(tf.argmax(Y, 1))
```

```
Out[36]: array([2, 3, 0], dtype=int64)
```

**axis=0时, X与Y按行连接**

```
In [37]: sess.run(tf.concat([X, Y], axis=0))
```

```
Out[37]: array([[ 0.,  1.,  2.,  3.],  
               [ 4.,  5.,  6.,  7.],  
               [ 8.,  9., 10., 11.],  
               [ 2.,  1.,  4.,  3.],  
               [ 1.,  2.,  3.,  4.],  
               [ 4.,  3.,  2.,  1.]], dtype=float32)
```

**axis=1时, X与Y按列连接**

```
In [38]: sess.run(tf.concat([X, Y], axis=1))
```

```
Out[38]: array([[ 0.,  1.,  2.,  3.,  2.,  1.,  4.,  3.],  
               [ 4.,  5.,  6.,  7.,  1.,  2.,  3.,  4.],  
               [ 8.,  9., 10., 11.,  4.,  3.,  2.,  1.]], dtype=float32)
```

## 矩阵对应元素相加

```
In [39]: sess.run(X + Y)
```

```
Out[39]: array([[ 2.,  2.,  6.,  6.],  
               [ 5.,  7.,  9., 11.],  
               [12., 12., 12., 12.]], dtype=float32)
```

## 矩阵的转置

```
In [40]: Z = tf.transpose(X)
```

```
In [41]: sess.run(Z)
```

```
Out[41]: array([[ 0.,  4.,  8.],  
               [ 1.,  5.,  9.],  
               [ 2.,  6., 10.],  
               [ 3.,  7., 11.]], dtype=float32)
```

## 矩阵对应元素相乘

```
In [42]: sess.run(X * Y)
```

```
Out[42]: array([[ 0.,  1.,  8.,  9.],  
               [ 4., 10., 18., 28.],  
               [32., 27., 20., 11.]], dtype=float32)
```

矩阵相乘  $A=Z*Z'$

```
In [43]: A = tf.matmul(Z, tf.transpose(Z))
```

```
In [44]: sess.run(A)
```

```
Out[44]: array([[ 80.,  92., 104., 116.],
                [ 92., 107., 122., 137.],
                [104., 122., 140., 158.],
                [116., 137., 158., 179.]], dtype=float32)
```

构建对称矩阵,  $A\_symm=(A+A')/2$

```
In [45]: A_symm = (A + tf.transpose(A)) / 2.0
```

```
In [46]: sess.run(A_symm)
```

```
Out[46]: array([[ 80.,  92., 104., 116.],
                [ 92., 107., 122., 137.],
                [104., 122., 140., 158.],
                [116., 137., 158., 179.]], dtype=float32)
```

判断 $A\_symm$ 是否为对称阵, 即 $A\_symm=A\_symm'$

```
In [47]: sess.run(tf.equal(A_symm, tf.transpose(A_symm)))
```

```
Out[47]: array([[ True,  True,  True,  True],
                [ True,  True,  True,  True],
                [ True,  True,  True,  True],
                [ True,  True,  True,  True]])
```



计算总和或均值时保持轴数不变

```
In [48]: sum_X = tf.reduce_sum(X, axis=1, keepdims=True)
```

```
In [49]: sess.run(sum_X)
```

```
Out[49]: array([[ 6.],
               [22.],
               [38.]], dtype=float32)
```

由于sum\_X在对每行进行求和后仍保持两个轴，我们可以通过广播将X除以sum\_X。

```
In [50]: sess.run(X / sum_X)
```

```
Out[50]: array([[0.          , 0.16666667, 0.33333334, 0.5          ],
               [0.18181819, 0.22727273, 0.27272728, 0.3181818  ],
               [0.21052632, 0.23684211, 0.2631579 , 0.28947368]], dtype=float32)
```

沿某个轴计算X元素的累积总和，比如axis=0（按行计算），我们可以调用cumsum函数。此函数不会沿任何轴降低输入张量的维度。

```
In [51]: sess.run(tf.cumsum(X, axis=0))
```

```
Out[51]: array([[ 0.,  1.,  2.,  3.],
               [ 4.,  6.,  8., 10.],
               [12., 15., 18., 21.]], dtype=float32)
```

```
In [52]: sess.run(tf.cumsum(X, axis=1))
```

```
Out[52]: array([[ 0.,  1.,  3.,  6.],
               [ 4.,  9., 15., 22.],
               [ 8., 17., 27., 38.]], dtype=float32)
```

```
In [53]: Z = tf.reshape(tf.range(24), (2, 3, 4))
```

```
In [54]: sess.run(Z)
```

```
Out[54]: array([[ 0,  1,  2,  3],
                [ 4,  5,  6,  7],
                [ 8,  9, 10, 11]],

              [[12, 13, 14, 15],
               [16, 17, 18, 19],
               [20, 21, 22, 23]])
```

### 标量乘以矩阵

```
In [55]: a = 2
```

```
In [56]: sess.run(a + Z)
```

```
Out[56]: array([[ 2,  3,  4,  5],
                [ 6,  7,  8,  9],
                [10, 11, 12, 13]],

              [[14, 15, 16, 17],
               [18, 19, 20, 21],
               [22, 23, 24, 25]])
```

```
In [57]: sess.run(a * Z)
```

```
Out[57]: array([[ 0,  2,  4,  6],
                [ 8, 10, 12, 14],
                [16, 18, 20, 22]],

              [[24, 26, 28, 30],
               [32, 34, 36, 38],
               [40, 42, 44, 46]])
```

矩阵乘以向量  $\mathbf{X}\mathbf{b} = \begin{bmatrix} \mathbf{x}^{\text{top}_1} \\ \mathbf{x}^{\text{top}_2} \\ \vdots \\ \mathbf{x}^{\text{top}_m} \end{bmatrix} \mathbf{b} = \begin{bmatrix} \mathbf{x}^{\text{top}_1} \mathbf{b} \\ \mathbf{x}^{\text{top}_2} \mathbf{b} \\ \vdots \\ \mathbf{x}^{\text{top}_m} \mathbf{b} \end{bmatrix}$

```
In [58]: b = tf.constant([2.0, 1, 4, 3])
```

```
In [59]: sess.run(b)
```

```
Out[59]: array([2., 1., 4., 3.], dtype=float32)
```

```
In [60]: sess.run(X)
```

```
Out[60]: array([[ 0.,  1.,  2.,  3.],
                 [ 4.,  5.,  6.,  7.],
                 [ 8.,  9., 10., 11.]], dtype=float32)
```

把向量b扩展成与矩阵X大小一致

```
In [61]: b = tf.expand_dims(b, 1)
```

```
In [62]: sess.run(b)
```

```
Out[62]: array([[2.],
                 [1.],
                 [4.],
                 [3.]], dtype=float32)
```

```
In [63]: sess.run(tf.matmul(X, b))
```

```
Out[63]: array([[18.],
                 [58.],
                 [98.]], dtype=float32)
```

## 4. 范数

2范数

$$\|\mathbf{x}\|_2 = \sqrt{\sum_{i=1}^n x_i^2}$$

```
In [64]: u = tf.constant([3.0, -4.0])
```

```
In [65]: sess.run(tf.norm(u, ord=2))
```

```
Out[65]: 5.0
```

1范数

$$\|\mathbf{x}\|_1 = \sum_{i=1}^n |x_i|$$

```
In [66]: sess.run(tf.reduce_sum(tf.abs(u)))
```

```
Out[66]: 7.0
```

```
In [67]: sess.run(tf.norm(u, ord=1))
```

```
Out[67]: 7.0
```

$\infty$  范数

$$\|\mathbf{x}\|_{\infty} = \max(|x_i|)$$

```
In [68]: import numpy as np
```

```
In [69]: sess.run(tf.norm(u, ord=np.inf))
```

```
Out[69]: 4.0
```