第十一章 Numpy库

11.1 为什么要用Numpy

11.1.1 低效的Python for循环

【例】 求100万个数的倒数

In [9]:

```
def compute_reciprocals(values):
    res = []
    for value in values: #每遍历到一个元素,就要判断其类型,并查找适用于该数据类型的正确是
    res. append(1/value)
    return res

values = list(range(1, 1000000))

timeit compute_reciprocals(values)
```

145 ms \pm 13.7 ms per loop (mean \pm std. dev. of 7 runs, 10 loops each)

%timeit: ipython中统计运行时间的魔术方法(多次运行取平均值)

In [10]:

```
import numpy as np
values = np. arange(1, 1000000)
timeit 1/values
```

5.99 ms \pm 33.9 μ s per loop (mean \pm std. dev. of 7 runs, 100 loops each)

实现相同计算,Numpy的运行速度是Python循环的25倍,产生了质的飞跃

11.1.2 Numpy为什么如此高效

Numpy 是由C语言编写的

1、编译型语言VS解释型语言

C语言执行时,对代码进行整体编译,速度更快

2、连续单一类型存储VS分散多变类型存储

(1) Numpy数组内的数据类型必须是统一的,如全部是浮点型,而Python列表支持任意类型数据的填充

(2) Numpy数组内的数据连续存储在内存中,而Python列表的数据分散在内存中

这种存储结构,与一些更加高效的底层处理方式更加的契合

3、多线程VS线程锁

Python语言执行时有线程锁,无法实现真正的多线程并行,而C语言可以

11.1.3 **什么时候用Numpy**

在数据处理的过程中,遇到使用"Python for循环"实现一些向量化、矩阵化操作的时候,要优先考虑用Numpy

如: 1、两个向量的点乘

2、矩阵乘法

11.2 Numpy数组的创建

11.2.1 从列表开始创建

In [11]:

```
1  import numpy as np
2  
3  x = np.array([1, 2, 3, 4, 5])
4  print(x)

import numpy as np

x = np.array([1, 2, 3, 4, 5])

print(x)
```

[1 2 3 4 5]

In [12]:

```
1 print(type(x))
2 print(x. shape)
```

<class 'numpy.ndarray'>
(5,)

• 设置数组的数据类型

In [13]:

```
1  x = np.array([1, 2, 3, 4, 5], dtype="float32")
2  print(x)
3  print(type(x[0]))
```

```
[1. 2. 3. 4. 5.] <class 'numpy.float32'>
```

• 二维数组

```
In [15]:
```

```
[[1 2 3]
[4 5 6]
[7 8 9]]
(3, 3)
```

11.2.2 从头创建数组

(1) 创建长度为5的数组,值都为0

```
In [16]:
```

```
1 np. zeros(5, dtype=int)
```

Out[16]:

```
array([0, 0, 0, 0, 0])
```

(2) 创建一个2*4的浮点型数组,值都为1

```
In [17]:
```

```
1 np.ones((2, 4), dtype=float)
```

Out[17]:

```
array([[1., 1., 1., 1.], [1., 1., 1., 1.]])
```

(3) 创建一个3*5的数组,值都为8.8

In [18]:

```
1 np. full((3, 5), 8.8)
```

Out[18]:

```
array([[8.8, 8.8, 8.8, 8.8, 8.8], [8.8, 8.8, 8.8, 8.8, 8.8, 8.8], [8.8, 8.8, 8.8, 8.8, 8.8]])
```

(4) 创建一个3*3的单位矩阵

```
In [19]:
    np. eye (3)
Out[19]:
array([[1., 0., 0.],
      [0., 1., 0.],
      [0., 0., 1.]
 (5) 创建一个线性序列数组,从1开始,到15结束,步长为2
In [20]:
    np. arange (1, 15, 2)
Out[20]:
array([ 1, 3, 5, 7, 9, 11, 13])
 (6) 创建一个4个元素的数组,这四个数均匀的分配到0~1
In [21]:
    np. linspace (0, 1, 4)
Out[21]:
                                                ])
array([0.
               , 0.33333333, 0.66666667, 1.
(7) 创建一个10个元素的数组,形成1~10^9的等比数列
In [22]:
    np. logspace (0, 9, 10)
Out[22]:
array([1. e+00, 1. e+01, 1. e+02, 1. e+03, 1. e+04, 1. e+05, 1. e+06, 1. e+07,
      1. e+08, 1. e+09])
 (8) 创建一个3*3的,在0~1之间均匀分布的随机数构成的数组
In [23]:
   np. random. random((3, 3))
Out[23]:
array([[0.24347952, 0.41715541, 0.41363866],
      [0.44869706, 0.18128167, 0.18568051],
      [0.05705023, 0.0689205, 0.74837661]])
```

(9) 创建一个3*3的,均值为0,标准差为1的正态分布随机数构成的数组

```
In [24]:
```

```
1 np. random. normal(0, 1, (3,3))
```

Out[24]:

(10) 创建一个3*3的,在[0,10)之间随机整数构成的数组

In [25]:

```
1 np. random. randint(0, 10, (3, 3))
```

Out[25]:

```
array([[9, 1, 9], [0, 3, 9], [8, 5, 4]])
```

(11) 随机重排列

In [26]:

```
1 x = np. array([10, 20, 30, 40])
2 np. random. permutation(x) # 生产新列表
```

Out[26]:

array([20, 40, 10, 30])

In [27]:

```
1 print(x)
2 np. random. shuffle(x) # 修改原列表
3 print(x)
```

[10 20 30 40] [20 40 10 30]

(12) 随机采样

• 按指定形状采样

In [28]:

```
1 x = np.arange(10, 25, dtype = float)
2 x
```

Out[28]:

```
array([10., 11., 12., 13., 14., 15., 16., 17., 18., 19., 20., 21., 22., 23., 24.])
```

```
In [29]:
```

```
1 np. random. choice(x, size=(4, 3))
```

Out[29]:

```
array([[19., 23., 22.], [22., 21., 13.], [15., 21., 17.], [14., 23., 19.]])
```

• 按概率采样

In [30]:

```
np.random.choice(x, size=(4, 3), p=x/np.sum(x))
```

Out[30]:

```
array([[15., 21., 20.], [23., 17., 18.], [23., 15., 17.], [19., 24., 22.]])
```

11.3 Numpy数组的性质

11.3.1 数组的属性

In [31]:

```
1 x = np.random.randint(10, size=(3, 4))
2 x
```

Out[31]:

```
array([[5, 5, 2, 7], [2, 3, 0, 8], [3, 8, 1, 7]])
```

1、数组的形状shape

In [32]:

```
1 x. shape
```

Out[32]:

(3, 4)

2、数组的维度ndim

```
In [33]:
1 x. ndim
Out[33]:
In [34]:
1 y = np. arange (10)
Out[34]:
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
In [35]:
1 y.ndim
Out[35]:
1
3、数组的大小size
In [36]:
1 x. size
Out[36]:
12
4、数组的数据类型dtype
In [37]:
1 x. dtype
Out[37]:
dtype('int32')
```

11.3.2 数组索引

1、一维数组的索引

```
In [38]:
1 x1 = np. arange(10)
2 x1
Out[38]:
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
In [39]:
1 \mid x1[0]
Out[39]:
In [40]:
1 | x1[5]
Out[40]:
5
In [41]:
1 | x1[-1]
Out[41]:
9
2、多维数组的索引——以二维为例
In [42]:
1 x2 = np. random. randint(0, 20, (2, 3))
Out[42]:
array([[11, 3, 11],
 [6, 1, 5]
In [43]:
1 \ x2[0, \ 0]
```

Out[43]:

11

```
In [44]:
 1 \ x2[0][0]
Out[44]:
11
注意: numpy数组的数据类型是固定的,向一个整型数组插入一个浮点值,浮点值会向下进行取整
In [46]:
1 \times 2[0, 0] = 1.618
In [47]:
1 \mid x2
Out[47]:
array([[ 1, 3, 11],
    [ 6, 1, 5]])
11.3.3 数组的切片
1、一维数组——跟列表一样
In [48]:
1 \mid x1 = np. arange(10)
 2 x1
Out[48]:
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
In [49]:
1 x1[:3]
Out[49]:
array([0, 1, 2])
In [50]:
1 \mid x1[3:]
```

Out[50]:

array([3, 4, 5, 6, 7, 8, 9])

```
In [51]:
1 x1[::-1]
Out[51]:
array([9, 8, 7, 6, 5, 4, 3, 2, 1, 0])
2、多维数组——以二维为例
In [52]:
1 x^2 = \text{np. random. randint}(20, \text{size}=(3, 4))
2 x2
Out[52]:
array([[14, 9, 15, 8],
      [18, 8, 16, 17],
      [ 0, 8, 2, 17]])
In [53]:
1 | x2[:2, :3]
                # 前两行,前三列
Out[53]:
array([[14, 9, 15],
   [18, 8, 16]])
In [54]:
                 # 前两行 前三列 (每隔一列)
1 \times 2[:2, 0:3:2]
Out[54]:
array([[14, 15],
     [18, 16]])
In [55]:
1 \mid x2[::-1, ::-1]
```

Out[55]:

```
array([[17, 2, 8, 0],
      [17, 16, 8, 18],
      [ 8, 15, 9, 14]])
```

3、获取数组的行和列

```
In [56]:
 1 \mid x3 = \text{np. random. randint}(20, \text{ size}=(3, 4))
 2 x3
Out [56]:
array([[ 8, 13, 15, 7],
     [19, 13, 17, 6],
      [11, 2, 0, 12]])
In [57]:
   x3[1, :] #第一行 从0开始计数
Out[57]:
array([19, 13, 17, 6])
In [58]:
1 x3[1] # 第一行简写
Out[58]:
array([19, 13, 17, 6])
In [59]:
1 x3[:, 2] # 第二列 从0开始计数
Out[59]:
array([15, 17, 0])
4、切片获取的是视图,而非副本
In [60]:
1 x4 = np. random. randint(20, size=(3, 4))
 2 x4
Out[60]:
array([[ 5, 14, 7, 2],
   [ 8, 12, 9, 3],
     [19, 0, 10, 7]]
In [61]:
 1 | x5 = x4[:2, :2]
 2 x5
Out[61]:
```

array([[5, 14],

[8, 12]])

注意: 视图元素发生修改, 则原数组亦发生相应修改

```
In [62]:
```

```
\begin{array}{c|cccc}
1 & x5[0, & 0] &= 0 \\
2 & x5
\end{array}
```

Out[62]:

In [63]:

```
1 \quad x4
```

Out[63]:

修改切片的安全方式: copy

In [64]:

```
1 x4 = np.random.randint(20, size=(3,4))
2 x4
```

Out[64]:

```
array([[18, 14, 10, 12], [10, 16, 7, 19], [3, 16, 3, 12]])
```

In [65]:

```
1 x6 = x4[:2, :2].copy()
2 x6
```

Out[65]:

```
array([[18, 14], [10, 16]])
```

In [66]:

```
\begin{array}{c|cccc}
1 & x6[0, & 0] &= 0 \\
2 & x6
\end{array}
```

Out[66]:

```
array([[ 0, 14], [10, 16]])
```

```
In [67]:
```

```
1 x4
```

Out[67]:

```
array([[18, 14, 10, 12], [10, 16, 7, 19], [3, 16, 3, 12]])
```

11.3.4 数组的变形

In [68]:

```
1 x5 = np. random. randint(0, 10, (12,))
2 x5
```

Out[68]:

```
array([9, 8, 5, 9, 2, 6, 2, 9, 4, 5, 1, 7])
```

In [69]:

```
1 x5. shape
```

Out[69]:

(12,)

In [70]:

```
x6 = x5. \text{ reshape}(3, 4)
x6 = x6
```

Out[70]:

```
array([[9, 8, 5, 9], [2, 6, 2, 9], [4, 5, 1, 7]])
```

注意: reshape返回的是视图, 而非副本

In [71]:

```
\begin{array}{c|cccc}
1 & x6[0, & 0] &= 0 \\
2 & x5
\end{array}
```

Out[71]:

```
array([0, 8, 5, 9, 2, 6, 2, 9, 4, 5, 1, 7])
```

一维向量转行向量

```
In [72]:
```

```
1 x7 = x5. reshape(1, x5. shape[0])
2 x7
```

Out[72]:

```
array([[0, 8, 5, 9, 2, 6, 2, 9, 4, 5, 1, 7]])
```

In [73]:

```
1 x8 = x5[np. newaxis, :]
2 x8
```

Out[73]:

```
array([[0, 8, 5, 9, 2, 6, 2, 9, 4, 5, 1, 7]])
```

一维向量转列向量

In [74]:

```
1 x7 = x5. reshape(x5. shape[0], 1)
2 x7
```

Out[74]:

array([[0],

[8],

[5],

[9],

[2],

[6],

[2],

L4J,

[9],

[4],

[5],

[1], [7]])

```
In [75]:
```

Out[75]:

多维向量转一维向量

In [76]:

```
1 x6 = np. random. randint(0, 10, (3, 4))
2 x6
```

Out[76]:

```
array([[3, 7, 6, 4], [4, 5, 6, 3], [7, 6, 2, 3]])
```

flatten返回的是副本

In [77]:

```
1 x9 = x6. flatten()
2 x9
```

Out[77]:

```
array([3, 7, 6, 4, 4, 5, 6, 3, 7, 6, 2, 3])
```

In [78]:

```
1 x9[0]=0
2 x6
```

Out[78]:

```
array([[3, 7, 6, 4], [4, 5, 6, 3], [7, 6, 2, 3]])
```

ravel返回的是视图

```
In [79]:
```

Out[79]:

```
array([3, 7, 6, 4, 4, 5, 6, 3, 7, 6, 2, 3])
```

In [80]:

```
\begin{array}{c|c}
1 & x10[0]=0 \\
2 & x6
\end{array}
```

Out[80]:

```
array([[0, 7, 6, 4], [4, 5, 6, 3], [7, 6, 2, 3]])
```

reshape返回的是视图

In [81]:

```
1 x11 = x6. reshape(-1)
2 x11
```

Out[81]:

```
array([0, 7, 6, 4, 4, 5, 6, 3, 7, 6, 2, 3])
```

In [82]:

```
1 x11[0]=10 x6
```

Out[82]:

11.3.5 数组的拼接

In [84]:

1、水平拼接——非视图

```
In [85]:
```

```
1 x3 = np. hstack([x1, x2])
2 x3
```

Out[85]:

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

In [86]:

```
1 x3[0][0] = 0
2 x1
```

Out[86]:

```
array([[1, 2, 3], [4, 5, 6]])
```

In [87]:

Out[87]:

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

In [88]:

Out[88]:

```
array([[1, 2, 3], [4, 5, 6]])
```

2、垂直拼接——非视图

In [216]:

```
In [89]:
```

```
1 x5 = np. vstack([x1, x2])
2 x5
```

Out[89]:

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

In [90]:

```
1 x6 = np. r_[x1, x2]
2 x6
```

Out[90]:

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

11.3.6 数组的分裂

1、split的用法

In [91]:

```
1 x6 = np. arange (10)
2 x6
```

Out[91]:

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

In [92]:

```
1 x1, x2, x3 = np. split(x6, [2, 7])
2 print(x1, x2, x3)
```

```
[0 1] [2 3 4 5 6] [7 8 9]
```

2、hsplit的用法

```
In [97]:
```

```
1 x7 = np.arange(1, 26).reshape(5, 5)
2 x7
```

Out [97]:

```
array([[ 1, 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 [99]:

```
1 left, middle, right = np. hsplit(x7, [2,4])
2 print("left:\n", left) #第0~1列
3 print("middle:\n", middle) #第2~3列
4 print("right:\n", right) #第4列
```

```
left:
```

 $[[1 \ 2]$

[6 7]

 $[11 \ 12]$

[16 17]

 $[21 \ 22]]$

middle:

[[3 4]

[8 9]

 $[13 \ 14]$

[18 19]

[23 24]]

right:

[[5]

[10]

[15]

[15]

[20] [25]]

3、vsplit的用法

In [100]:

```
1 x7 = np. arange(1, 26). reshape(5, 5)
2 x7
```

Out[100]:

```
array([[ 1, 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 [101]:
```

```
1 upper, middle, lower = np.vsplit(x7, [2, 4])
2 print("upper:\n", upper) # 第0~1行
3 print("middle:\n", middle) # 第2~3行
4 print("lower:\n", lower) # 第4行
```

```
upper:
[[ 1 2 3 4 5]
[ 6 7 8 9 10]]
middle:
[[11 12 13 14 15]
[16 17 18 19 20]]
lower:
[[21 22 23 24 25]]
```

11.4 Numpy四大运算

11.4.1 向量化运算

1、与数字的加减乘除等

```
In [102]:
```

```
1 x1 = np. arange (1, 6)
2 x1
```

Out[102]:

array([1, 2, 3, 4, 5])

In [103]:

```
1 print("x1+5", x1+5)
2 print("x1-5", x1-5)
3 print("x1*5", x1*5)
4 print("x1/5", x1/5)
```

```
x1+5 [ 6 7 8 9 10]
x1-5 [-4 -3 -2 -1 0]
x1*5 [ 5 10 15 20 25]
x1/5 [0.2 0.4 0.6 0.8 1. ]
```

In [104]:

```
print("-x1", -x1)
print("x1**2", x1**2)
print("x1//2", x1//2)
print("x1%2", x1%2)
```

```
-x1 [-1 -2 -3 -4 -5]
x1**2 [ 1 4 9 16 25]
x1//2 [0 1 1 2 2]
x1%2 [1 0 1 0 1]
```

2、绝对值、三角函数、指数、对数

(1) 绝对值

```
In [105]:
```

```
1 x2 = np. array([1, -1, 2, -2, 0])
2 x2
```

Out[105]:

```
array([ 1, -1, 2, -2, 0])
```

In [106]:

```
1 \quad abs(x2)
```

Out[106]:

```
array([1, 1, 2, 2, 0])
```

In [107]:

```
1 np. abs (x2)
```

Out[107]:

```
array([1, 1, 2, 2, 0])
```

(2) 三角函数

In [108]:

```
theta = np.linspace(0, np.pi, 3)
theta
```

Out[108]:

```
array([0. , 1.57079633, 3.14159265])
```

In [109]:

```
print("sin(theta)", np. sin(theta))
print("con(theta)", np. cos(theta))
print("tan(theta)", np. tan(theta))
```

```
sin(theta) [0.0000000e+00 1.0000000e+00 1.2246468e-16]
con(theta) [ 1.000000e+00 6.123234e-17 -1.000000e+00]
tan(theta) [ 0.00000000e+00 1.63312394e+16 -1.22464680e-16]
```

```
In [110]:
```

```
1  x = [1, 0, -1]
2  print("arcsin(x)", np. arcsin(x))
3  print("arccon(x)", np. arccos(x))
4  print("arctan(x)", np. arctan(x))
```

(3) 指数运算

In [111]:

```
1 x = np. arange (3)
2 x
```

Out[111]:

```
array([0, 1, 2])
```

In [112]:

```
1 np. exp(x)
```

Out[112]:

array([1. , 2.71828183, 7.3890561])

(4) 对数运算

In [113]:

```
1  x = np.array([1, 2, 4, 8, 10])
2  print("ln(x)", np.log(x))
3  print("log2(x)", np.log2(x))
4  print("log10(x)", np.log10(x))
```

3、两个数组的运算

In [114]:

Out[114]:

```
array([1, 2, 3, 4, 5])
```

```
In [115]:
```

```
1 x2 = np. arange (6, 11)
2 x2
```

Out[115]:

```
array([ 6, 7, 8, 9, 10])
```

In [116]:

```
print("x1+x2:", x1+x2)
print("x1-x2:", x1-x2)
print("x1*x2:", x1*x2)
print("x1/x2:", x1/x2)
```

11.4.2 矩阵运算

In [132]:

```
1 x = np. arange(9). reshape(3, 3)
2 x
```

Out[132]:

```
array([[0, 1, 2], [3, 4, 5], [6, 7, 8]])
```

• 矩阵的转置

In [138]:

```
\begin{bmatrix} 1 & y = x. T \\ 2 & y \end{bmatrix}
```

Out[138]:

```
array([[0, 3, 6],
[1, 4, 7],
[2, 5, 8]])
```

• 矩阵乘法

```
In [139]:
 1 \mid x = \text{np. array}([[1, 0],
 2
                  [1, 1]]
 3 \mid y = \text{np.array}([[0, 1],
                   [1, 1]])
 4
In [140]:
1 | x. dot (y)
Out[140]:
array([[0, 1],
 [1, 2]]
In [141]:
1 \operatorname{np.dot}(x, y)
Out[141]:
array([[0, 1],
     [1, 2]]
In [142]:
1 y. dot(x)
Out[142]:
array([[1, 1],
      [2, 1]
In [143]:
1 | \text{np. dot}(y, x) |
Out[143]:
array([[1, 1],
      [2, 1]
注意跟x*y的区别
In [144]:
1 | x∗y
Out[144]:
array([[0, 0],
   [1, 1]]
```

11.4.3 广播运算

```
In [145]:
```

```
1  x = np. arange(3). reshape(1, 3)
2  x
```

Out[145]:

```
array([[0, 1, 2]])
```

In [146]:

```
1 x+5
```

Out[146]:

```
array([[5, 6, 7]])
```

规则

如果两个数组的形状在维度上不匹配

那么数组的形式会沿着维度为1的维度进行扩展以匹配另一个数组的形状。

In [147]:

```
    \begin{array}{c|c}
      1 & x1 = np. ones ((3, 3)) \\
      2 & x1
    \end{array}
```

Out[147]:

```
array([[1., 1., 1.], [1., 1., 1.], [1., 1., 1.]])
```

In [148]:

```
1 x2 = np. arange(3). reshape(1, 3)
2 x2
```

Out[148]:

```
array([[0, 1, 2]])
```

In [149]:

```
1 x1+x2
```

Out[149]:

```
In [150]:
```

```
1 x3 = np. logspace(1, 10, 10, base=2).reshape(2, 5)
2 x3
```

Out[150]:

```
array([[ 2., 4., 8., 16., 32.], [ 64., 128., 256., 512., 1024.]])
```

In [151]:

```
1 x4 = np. array([[1, 2, 4, 8, 16]])
2 x4
```

Out[151]:

```
array([[ 1, 2, 4, 8, 16]])
```

In [152]:

```
1 \quad x3/x4
```

Out[152]:

```
array([[ 2., 2., 2., 2., 2.], [64., 64., 64., 64., 64.])
```

In [153]:

```
1 x5 = np. arange(3). reshape(3, 1)
2 x5
```

Out[153]:

```
array([[0],
[1],
[2]])
```

In [154]:

```
1 x6 = np.arange(3).reshape(1, 3)
2 x6
```

Out[154]:

```
array([[0, 1, 2]])
```

In [155]:

```
1 x5+x6
```

Out[155]:

```
array([[0, 1, 2], [1, 2, 3], [2, 3, 4]])
```

.

11.4.4 比较运算和掩码

1、比较运算

In [156]:

```
1 x1 = np.random.randint(100, size=(10, 10))
2 x1
```

Out[156]:

```
array([[37, 44, 58, 79, 1, 24, 85, 90, 27, 56],
       [74, 68, 88, 27, 46, 34, 92,
                                    1, 35, 45],
       [84, 80, 83, 72, 98, 15,
                                4, 77, 14,
                                            98],
       [19, 85, 98, 32, 47, 50, 73, 3, 24,
       [ 5, 28, 26, 31, 48, 43,
                                72, 73, 53, 64],
       [81, 87, 56, 59, 24, 42,
                               84, 34, 97,
       [74, 9, 41, 54, 78, 62,
                                53, 49,
                                         8,
                                            70],
       [63, 44, 33, 35, 26, 83,
                                7, 14, 65, 84],
       [57, 10, 62, 8, 74, 47, 90, 25, 78, 48],
       [36, 31, 45, 39, 66, 82, 42, 25, 33, 84]])
```

In [157]:

Out[157]:

```
array([[False, False,
                       True,
                             True, False, False,
                                                   True, True, False,
         True],
                       True, False, False, False,
       [ True, True,
                                                   True, False, False,
       False],
       [True,
                       True,
                              True,
                                    True, False, False,
                                                           True, False,
                True,
         True],
       [False, True,
                       True, False, False, False,
                                                   True, False, False,
       False],
       [False, False, False, False, False, False,
                                                   True,
                                                           True,
                                                                  True,
         True],
       [ True, True, True,
                              True, False, False,
                                                    True, False,
                                                                  True,
         True],
       [ True, False, False,
                              True,
                                    True,
                                            True,
                                                    True, False, False,
         True],
       [ True, False, False, False, False,
                                            True, False, False,
                                                                  True,
         True],
       [ True, False, True, False,
                                    True, False,
                                                   True, False,
                                                                  True,
       [False, False, False, False, True, True, False, False, False,
         True]])
```

2、操作布尔数组

```
In [158]:
 1 x2 = np. random. randint(10, size=(3, 4))
 2 x2
Out[158]:
array([[1, 4, 2, 9],
     [8, 8, 2, 4],
      [9, 5, 3, 6]])
In [159]:
 1 | print(x2 > 5) |
 2 np. sum (x2 > 5)
[[False False False True]
[ True True False False]
[ True False False True]]
Out[159]:
5
In [160]:
1 | np. all (x2 > 0)
Out[160]:
True
In [161]:
1 | np. any (x2 == 6)
Out[161]:
True
In [163]:
1 np. al1(x2 < 9, axis=1) # 按行进行判断
Out[163]:
array([False, True, False])
In [164]:
1 \mid x2
Out[164]:
array([[1, 4, 2, 9],
      [8, 8, 2, 4],
       [9, 5, 3, 6]])
```

```
In [165]:
   (x2 < 9) & (x2 > 5)
Out[165]:
array([[False, False, False, False],
       [ True, True, False, False],
       [False, False, False, True]])
In [166]:
 1 | np. sum((x2 < 9) & (x2 > 5))
Out[166]:
3
3、将布尔数组作为掩码
In [167]:
1 \mid x2
Out[167]:
array([[1, 4, 2, 9],
       [8, 8, 2, 4],
       [9, 5, 3, 6]])
In [168]:
 1 \mid x2 > 5
Out[168]:
array([[False, False, False, True],
       [ True, True, False, False],
       [ True, False, False, True]])
In [169]:
1 | x2[x2 > 5]
Out[169]:
array([9, 8, 8, 9, 6])
```

11.4.5 花哨的索引

1、一维数组

```
In [170]:
```

```
x = np. random. randint (100, size=10)
x
```

Out[170]:

```
array([43, 69, 67, 9, 11, 27, 55, 93, 23, 82])
```

注意: 结果的形状与索引数组ind一致

In [171]:

```
1 ind = [2, 6, 9]
2 x[ind]
```

Out[171]:

```
array([67, 55, 82])
```

In [172]:

Out[172]:

```
array([[69, 43], [67, 9]])
```

2、多维数组

In [173]:

```
1  x = np. arange(12). reshape(3, 4)
2  x
```

Out[173]:

In [174]:

```
1 row = np.array([0, 1, 2])
2 col = np.array([1, 3, 0])
3 x[row, col] # x(0, 1) x(1, 3) x(2, 0)
```

Out[174]:

```
array([1, 7, 8])
```

```
In [175]:
   row[:, np.newaxis]
                       # 列向量
Out[175]:
array([[0],
      [1],
      [2]
In [176]:
   x[row[:, np.newaxis], col] # 广播机制
Out[176]:
array([[ 1, 3, 0],
      [5, 7, 4],
      [ 9, 11, 8]])
11.5 其他Numpy通用函数
11.5.1 数值排序
In [402]:
   x = \text{np. random. randint (20, 50, size=10)}
 2
Out[402]:
array([48, 27, 44, 24, 34, 21, 24, 30, 34, 46])
 • 产生新的排序数组
In [403]:
   np. sort(x)
Out[403]:
array([21, 24, 24, 27, 30, 34, 34, 44, 46, 48])
In [404]:
 1
Out [404]:
array([48, 27, 44, 24, 34, 21, 24, 30, 34, 46])
```

替换原数组

```
In [406]:
```

```
1 x. sort()
2 x
```

Out [406]:

array([21, 24, 24, 27, 30, 34, 34, 44, 46, 48])

• 获得排序索引

In [407]:

```
1 x = np. random. randint (20, 50, size=10)
2 x
```

Out[407]:

array([27, 36, 35, 28, 34, 20, 21, 49, 48, 30])

In [409]:

Out[409]:

array([5, 6, 0, 3, 9, 4, 2, 1, 8, 7], dtype=int64)

11.5.2 最大最小值

In [411]:

```
1 x = np.random.randint(20, 50, size=10)
2 x
```

Out[411]:

array([48, 31, 30, 44, 48, 33, 44, 48, 39, 35])

In [413]:

```
print("max:", np. max(x))
print("min:", np. min(x))
```

max: 48 min: 30

In [414]:

```
print("max_index:", np.argmax(x))
print("min_index:", np.argmin(x))
```

max_index: 0
min_index: 2

11.5.3 数值求和、求积

array([3, 5, 7])

```
In [183]:
1 \mid x = np. arange(1, 6)
Out[183]:
array([1, 2, 3, 4, 5])
In [417]:
1 x. sum()
Out[417]:
15
In [419]:
1 np. sum(x)
Out[419]:
15
In [177]:
1 \mid x1 = \text{np. arange } (6). \text{ reshape } (2, 3)
 2 x1
Out[177]:
array([[0, 1, 2],
      [3, 4, 5]])
 • 按行求和
In [178]:
 1 np. sum(x1, axis=1)
Out[178]:
array([ 3, 12])
 • 按列求和
In [179]:
1 | np. sum(x1, axis=0)
Out[179]:
```

• 全体求和

```
In [181]:
    np. sum(x1)
Out[181]:
15
 求积
In [184]:
1 \mid \mathbf{x}
Out[184]:
array([1, 2, 3, 4, 5])
In [420]:
1 x. prod()
Out[420]:
120
In [421]:
    np. prod(x)
Out[421]:
120
```

11.5.4 中位数、均值、方差、标准差

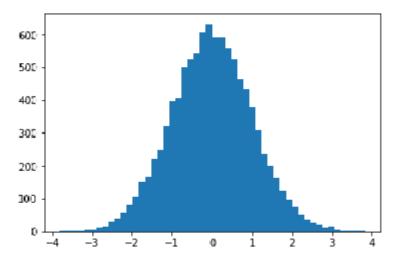
```
In [424]:
```

```
1 x = \text{np. random. normal}(0, 1, \text{size}=10000)
```

```
In [425]:
```

```
import matplotlib.pyplot as plt

plt.hist(x, bins=50)
plt.show()
```



• 中位数

In [427]:

1 np. median(x)

Out[427]:

-0.01024418366119727

均值

In [428]:

```
1 x. mean()
```

Out[428]:

-0. 004164442327293362

In [430]:

```
1 np.mean(x)
```

Out[430]:

-0.004164442327293362

方差

```
In [431]:
 1 x. var()
Out[431]:
1. 0221853234535774
In [432]:
1 \operatorname{np. var}(x)
Out[432]:
1.\ 0221853234535774
 • 标准差
In [433]:
 1 x. std()
Out[433]:
1. 0110318112965473
In [434]:
1 np. std(x)
Out[434]:
```

1. 0110318112965473