实验一 矩阵代数

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一、基于1.2.1的R中的一些基本函数及运算

```
In [1]:
import numpy as np
In [2]:
x = np. array([1, 3, 4, 1, 2, 5]) # 创建一个向量
Out[2]:
array([1, 3, 4, 1, 2, 5])
In [3]:
len(x) # 向量的长度
Out[3]:
6
In [4]:
A = x. reshape((2,3), order='F') # 利用 x 数值按列填充创建一个 2 X 3 矩阵
Out[4]:
array([[1, 4, 2],
      [3, 1, 5]])
In [5]:
A. dtype # 数据的模式
Out[5]:
dtype('int32')
```

```
In [6]:
          # 数据的类或类型
type(A)
Out[6]:
numpy.ndarray
In [7]:
         # 矩阵的维度
A. shape
Out[7]:
(2, 3)
In [8]:
    # 矩阵转置
A. T
Out[8]:
array([[1, 3],
      [4, 1],
      [2, 5]]
In [9]:
          # 矩阵求和
A. sum()
Out[9]:
16
In [10]:
A. sum(axis=1)
               # 矩阵按行求和
Out[10]:
array([7, 9])
In [11]:
A. sum(axis=0)
               # 矩阵按列求和
Out[11]:
array([4, 5, 7])
In [12]:
A. mean()
           # 矩阵求平均值
Out[12]:
2.666666666666665
```

```
In [13]:
A. mean (axis=1) # 矩阵按行求平均值
Out[13]:
array([2.33333333, 3.
                      ])
In [14]:
A. mean (axis=0) # 矩阵按列求平均值
Out[14]:
array([2., 2.5, 3.5])
In [15]:
B = \text{np. array}([6, 0, 2, 3, 1, 4]). \text{ reshape}((2, 3))
Out[15]:
array([[6, 0, 2],
      [3, 1, 4]]
In [16]:
A + B # 矩阵相加
Out[16]:
array([[7, 4, 4],
      [6, 2, 9]])
In [17]:
A - B # 矩阵相减
Out[17]:
array([[-5, 4, 0],
      [0, 0, 1]
In [18]:
C = \text{np. array}([1, 2, 1, 3]). \text{ reshape}((2, 2), \text{ order='F'})
С
Out[18]:
array([[1, 1],
       [2, 3]])
```

```
In [19]:
```

```
      Out[19]:

      array([[ 4, 5, 7], [11, 11, 19]])

      In [20]:

      np. multiply(A, B)

      Out[20]:

      array([[ 6, 0, 4], [ 9, 1, 20]])
```

二、计算方阵的一些函数值

In [21]:

```
from numpy.linalg import inv, det, eig, svd
```

In [22]:

Out[22]:

```
array([[ 1, 2, 3, 4, 5], [ 2, 4, 7, 8, 9], [ 3, 7, 10, 15, 20], [ 4, 8, 15, 30, 20], [ 5, 9, 20, 20, 40]])
```

In [23]:

```
np. diag(A) # 由矩阵的对角线元素构成的向量
```

Out[23]:

```
array([ 1, 4, 10, 30, 40])
```

```
In [24]:
```

```
# 由向量diag(A)的元素创建对角矩阵
np. diag(np. diag(A))
Out [24]:
array([[ 1,
            0, 0,
                    0,
                       0],
       [ 0,
            4, 0,
                    0,
                        0],
       [ 0,
            0, 10, 0,
                        07,
       Γ0,
            0,
               0, 30,
                        07,
      [ 0,
                0, 0, 40]
           0,
In [25]:
            # 创建一个5阶单位矩阵
np. eye (5)
Out [25]:
array([[1., 0., 0., 0., 0.],
      [0., 1., 0., 0., 0.]
      [0., 0., 1., 0., 0.]
       [0., 0., 0., 1., 0.],
       [0., 0., 0., 0., 1.]
In [26]:
inv(A)
         # 矩阵的逆
Out [26]:
array([[ 9.78873239e+00, -2.18309859e+00, -1.85915493e+00,
         1. 12676056e-01, 1. 40845070e-01],
                                          7.88732394e-01,
       [-2.18309859e+00, 7.74647887e-01,
       -1.69014085e-01, -2.11267606e-01],
       [-1.85915493e+00, 7.88732394e-01, 3.94366197e-02,
       -8. 45070423e-03, 3. 94366197e-02],
       [ 1.12676056e-01, -1.69014085e-01, -8.45070423e-03,
        7. 32394366e-02, -8. 45070423e-03],
       [ 1.40845070e-01, -2.11267606e-01, 3.94366197e-02,
       -8. 45070423e-03, 3. 94366197e-02]])
In [27]:
det(A)
         # 矩阵的行列式
Out [27]:
-355.000000000000006
In [28]:
eig(A)
         # 矩阵的特征值与特征向量
Out [28]:
(array([70.33488803, 14.44024095, 1.997606 , 0.09374538, -1.86648037]),
array([[ 0.10513926, 0.00733125, 0.26673691, 0.95627367, -0.05730686],
                      0.05549834, 0.82858975, -0.22629386, 0.46554035],
        L 0. 20596656,
        [0.39707684, -0.02585507, 0.32382661, -0.18402887, -0.83840992],
       [0.5462168, 0.78569385, -0.25851683, 0.01391472, 0.13155919],
        \begin{bmatrix} 0.70035756, -0.61553463, -0.26569873, 0.0164776 \end{bmatrix}
                                                             0. 24443626]]))
```

```
In [29]:
```

```
values, vectors = eig(A)[0], eig(A)[1] # 分别获取特征值与特征向量
```

In [30]:

```
value = values[0] # 取第一个特征值
```

In [31]:

```
vector = vectors[:,0] # 取第一个特征向量
```

In [32]:

```
left= A. dot (vector)
```

In [33]:

```
right = value * vector
```

In [34]:

```
np. allclose(left, right) # 比较是否相等(可忽略小数点后五位)
```

Out[34]:

True

In [35]:

```
np. sum(np. diag(A)) # 矩阵的迹,即对向量diag(A)中的元素求和
```

Out[35]:

85

三、例1.6.6中的奇异值分解

In [36]:

```
A = np.array([1, 1, 2, -2, 2, 2]).reshape((2,3), order="F")
A
```

Out[36]:

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

In [37]:

```
svd(A) # 奇异值分解
```

Out[37]: