

lektion6

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Einführung in NumPy Teil 2

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1 Einführung in NumPy

1.1 Broadcasting

$+$, $-$, $*$, $\%$, $/$ $**$ wirken eintragsweise, **wobei die *singleton* Dimension expandiert wird.**

singleton : fehlende Dimension oder Dimension der Länge 1.

```
[2]: import numpy as np
```

```
[3]: A = np.array([[1, 2], [3, 4]])  
v = np.array([-5, 7])  
display(A, v)
```

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

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

```
[4]: # Das erste Element von v wird zu jeder zu jedem Element der ersten Spalte  
      ↪ addiert  
A+v
```

```
[4]: array([[ -4,  9],  
          [-2, 11]])
```

```
[5]: #' Das erste Element von v wird ...  
A*v
```

```
[5]: array([[ -5,  14],
           [-15,  28]])
```

```
[6]: A**v.astype('float')
```

```
[6]: array([[1.00000000e+00, 1.28000000e+02],
           [4.11522634e-03, 1.63840000e+04]])
```

```
[7]: # zu jedem Eintrag wird 1 addiert
A + 1
```

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

```
[8]: # jeder Eintrag wird durch 3 dividiert
A / 3
```

```
[8]: array([[0.33333333, 0.66666667],
           [1.          , 1.33333333]])
```

```
[9]: # 3 wird durch jeden Eintrag dividiert
3 / A
```

```
[9]: array([[3.  , 1.5 ],
           [1.  , 0.75]])
```

Broadcastingregel (aus <https://numpy.org/doc/stable/user/basics.broadcasting.html>)

When operating on two arrays, NumPy compares their shapes element-wise. It starts with the trailing (i.e. rightmost) dimensions and works its way left. Two dimensions are compatible when

1. they are equal, or
2. one of them is 1

If these conditions are not met, a `ValueError: operands could not be broadcast together` exception is thrown, indicating that the arrays have incompatible shapes. The size of the resulting array is the size that is not 1 along each axis of the inputs.

Arrays do not need to have the same number of dimensions.

```
[10]: DreiArray = np.array(np.arange(3*4*5)).reshape(3, 4, 5)
DreiArray.shape
```

```
[10]: (3, 4, 5)
```

```
[11]: ZweiArray = np.array(np.arange(3*4)).reshape(3, 4)
```

```
[12]: DreiArray + ZweiArray
```

```
-----  
ValueError                                Traceback (most recent call last)  
Cell In[12], line 1  
----> 1 DreiArray + ZweiArray  
  
ValueError: operands could not be broadcast together with shapes (3,4,5) (3,4)
```

```
[13]: DreiArray_neu = np.array(np.arange(3*4*5)).reshape(5, 3, 4)  
DreiArray_neu.shape
```

```
[13]: (5, 3, 4)
```

```
[14]: EinsArray = np.array(np.arange(5)).reshape(5, 1, 1)
```

```
[15]: #DreiArray_neu + ZweiArray + EinsArray
```

```
[34]: EinsArray
```

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

```
[36]: np.squeeze(EinsArray)
```

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

1.2 Beispiel zur Verwendung von Index-Arrays.

Erzeuge einen Vektor, der die Gegendiagonale einer Matrix enthält

```
[16]: A = np.arange(16).reshape(4, 4)  
A
```

```
[16]: array([[ 0,  1,  2,  3],  
          [ 4,  5,  6,  7],  
          [ 8,  9, 10, 11],  
          [12, 13, 14, 15]])
```

```
[17]: i = np.arange(4)  
j = i[::-1]
```

```
j, i
```

```
[17]: (array([3, 2, 1, 0]), array([0, 1, 2, 3]))
```

```
[18]: b = A[i, j]
      b
```

```
[18]: array([ 3,  6,  9, 12])
```

1.3 Verwende boolesche Ausdrücke bei der Definition von Matrizen

```
[19]: # Bsp: Setze alle negativen Matrixeinträge auf Null
      A
      B = A - 2*A.T
      B
```

```
[19]: array([[ 0, -7, -14, -21],
           [ 2, -5, -12, -19],
           [ 4, -3, -10, -17],
           [ 6, -1, -8, -15]])
```

```
[20]: B[B < 0] = 0
      B
```

```
[20]: array([[0, 0, 0, 0],
           [2, 0, 0, 0],
           [4, 0, 0, 0],
           [6, 0, 0, 0]])
```

```
[21]: # elementweises logisches "oder" und "und"
      B = A - A.T
      B
```

```
[21]: array([[ 0, -3, -6, -9],
           [ 3,  0, -3, -6],
           [ 6,  3,  0, -3],
           [ 9,  6,  3,  0]])
```

```
[22]: # Welche Elemente von B sind kleiner als -5 ODER größer als 6?
      index_array = np.logical_or(B < -5, B > 6)
      index_array
```

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

```
[23]: # Welche Elemente von B sind kleiner als 2 UND größer als 0?  
np.logical_and(B < 7, B > 0)
```

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

```
[24]: B[index_array] = 100  
B
```

```
[24]: array([[ 0, -3, 100, 100],  
        [ 3,  0, -3, 100],  
        [ 6,  3,  0, -3],  
        [100,  6,  3,  0]])
```

1.4 Achtung

NumPy macht kein Ducktyping.

... duck typing is an application of the duck test—“If it walks like a duck and it quacks like a duck, then it must be a duck” https://en.wikipedia.org/wiki/Duck_typing

```
[25]: a = np.array([1, -2.25])
```

```
[26]: a.dtype
```

```
[26]: dtype('float64')
```

```
[27]: a[0] = 1.+1j
```

```
-----  
TypeError                                Traceback (most recent call last)  
Cell In[27], line 1  
----> 1 a[0] = 1.+1j  
  
TypeError: float() argument must be a string or a real number, not 'complex'
```

```
[28]: ac = np.array([1, -2.25], dtype='complex')  
ac.dtype
```

```
[28]: dtype('complex128')
```

```
[29]: ac[0] = 10.+10j
```

```
[30]: a + ac
```

```
[30]: array([11. +10.j, -4.5 +0.j])
```

```
[91]: np.sqrt(a)
```

```
/tmp/ipykernel_426820/2835837605.py:1: RuntimeWarning: invalid value encountered  
in sqrt  
  np.sqrt(a)
```

```
[91]: array([ 1., nan])
```

```
[32]: np.sqrt(a.astype('complex'))
```

```
[32]: array([1.+0.j , 0.+1.5j])
```

Die NumPy-Funktionen in `emath` (power, exp, log, und inverse trigonometrische Funktionen) passen den Ausgabedatentyp an, wenn der Ausgabedatentyp vom Eingabedatentyp abweicht.

```
[93]: np.emath.sqrt(a)
```

```
[93]: array([1.+0.j , 0.+1.5j])
```

```
[95]: np.info(np.emath)
```

Wrapper functions to more user-friendly calling of certain math functions whose output data-type is different than the input data-type in certain domains of the input.

For example, for functions like `log` with branch cuts, the versions in this module provide the mathematically valid answers in the complex plane::

```
>>> import math  
>>> np.emath.log(-math.exp(1)) == (1+1j*math.pi)  
True
```

Similarly, `sqrt`, other base logarithms, `power` and trig functions are correctly handled. See their respective docstrings for specific examples.

Functions

```
.. autosummary::  
   :toctree: generated/  
  
   sqrt  
   log  
   log2  
   logn  
   log10
```

```
power
arccos
arcsin
arctanh
```

1.5 Universal functions (ufunc)

Die universellen Funktionen aus NumPy (sin, cos, tan, arcsin, arccos, sinh, cosh, arcsinh, ... exp, log, log2, log10, sqrt, ...) wirken elementweise, wenn sie auf NumPy Arrays angewandt werden. <https://numpy.org/doc/stable/reference/ufuncs.html>

```
[2]: x = np.linspace(0, 5*np.pi, 10) # 10 äquidistante Zahlen zwischen 0 und 5 pi
x, np.sin(x)
```

```
[2]: (array([ 0.          ,  1.74532925,  3.4906585 ,  5.23598776,  6.98131701,
            8.72664626, 10.47197551, 12.21730476, 13.96263402, 15.70796327]),
      array([ 0.00000000e+00,  9.84807753e-01, -3.42020143e-01, -8.66025404e-01,
            6.42787610e-01,  6.42787610e-01, -8.66025404e-01, -3.42020143e-01,
            9.84807753e-01,  6.12323400e-16]))
```

```
[33]: np.info(np.expm1)
```

```
expm1(x, /, out=None, *, where=True, casting='same_kind', order='K', dtype=None,
subok=True[, signature])
```

Calculate `exp(x) - 1` for all elements in the array.

Parameters

`x` : array_like

Input values.

`out` : ndarray, None, or tuple of ndarray and None, optional

A location into which the result is stored. If provided, it must have a shape that the inputs broadcast to. If not provided or None, a freshly-allocated array is returned. A tuple (possible only as a keyword argument) must have length equal to the number of outputs.

`where` : array_like, optional

This condition is broadcast over the input. At locations where the condition is True, the `out` array will be set to the ufunc result. Elsewhere, the `out` array will retain its original value.

Note that if an uninitialized `out` array is created via the default `out=None`, locations within it where the condition is False will remain uninitialized.

****kwargs**

For other keyword-only arguments, see the `:ref:`ufunc docs <ufuncs.kwargs>`.

Returns

out : ndarray or scalar
Element-wise exponential minus one: `out = exp(x) - 1`.
This is a scalar if `x` is a scalar.

See Also

log1p : `log(1 + x)`, the inverse of `expm1`.

Notes

This function provides greater precision than `exp(x) - 1` for small values of `x`.

Examples

The true value of `exp(1e-10) - 1` is `1.00000000005e-10` to about 32 significant digits. This example shows the superiority of `expm1` in this case.

```
>>> import numpy as np
```

```
>>> np.expm1(1e-10)
1.000000000005e-10
>>> np.exp(1e-10) - 1
1.000000082740371e-10
```

```
[4]: %%timeit
      y = np.sin(x)
```

576 ns ± 26.1 ns per loop (mean ± std. dev. of 7 runs, 1000000 loops each)

```
[6]: %%timeit
      l = []
      for xx in x:
          l.append(np.sin(xx))
      y = np.array(l)
```

8.61 µs ± 106 ns per loop (mean ± std. dev. of 7 runs, 100000 loops each)