

# Python NumPy



Linear Algebra and  
Scientific Computing

By: Mouafak Dakhlaoui

```
def filterStudies(studies, filterByOrg):  
    filteredStudies = []  
    for study in studies:  
        if study.lead_organization == filterByOrg:  
            filteredStudies.append(study)  
    return filteredStudies
```

```
def filterStudies(studies, filterByOrg):  
    filteredStudies = studies.filter(study  
    return filteredStudies
```



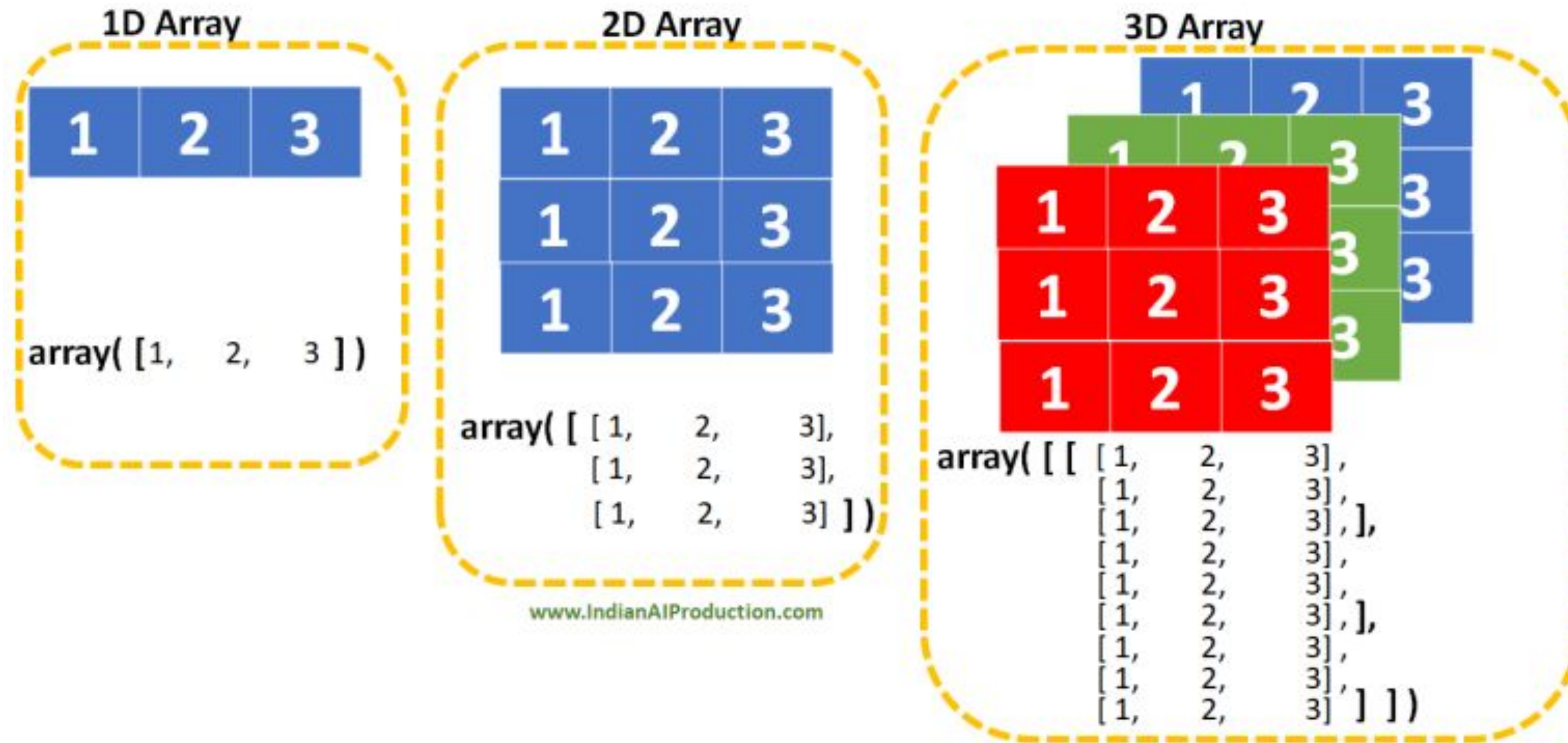
# What is NumPy?

NumPy is the fundamental package for scientific computing in Python. It is a Python library that provides a multidimensional array object, various derived objects (such as masked arrays and matrices), and an assortment of routines for fast operations on arrays, including mathematical, logical, shape manipulation, sorting, selecting, I/O, discrete Fourier transforms, basic linear algebra, basic statistical operations, random simulation and much more.






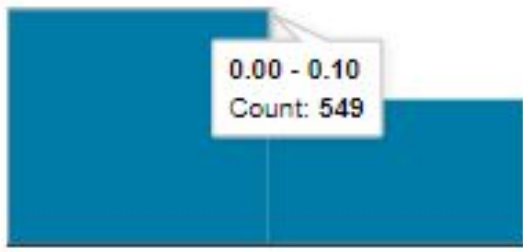

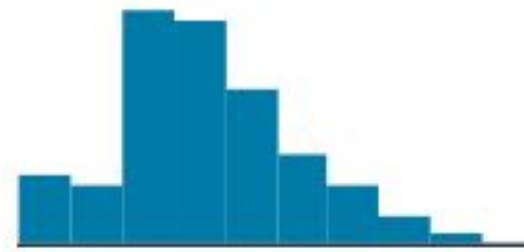
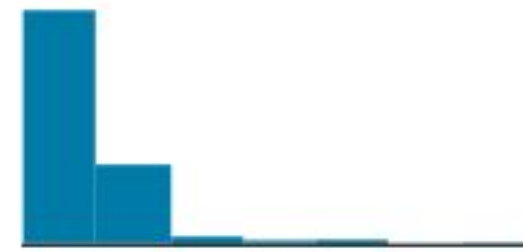
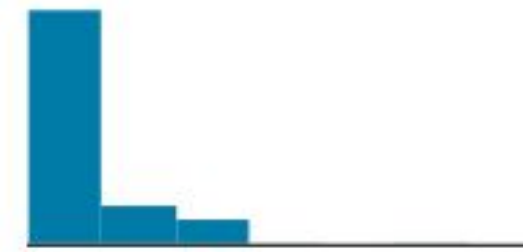




# NumPy - ndarray

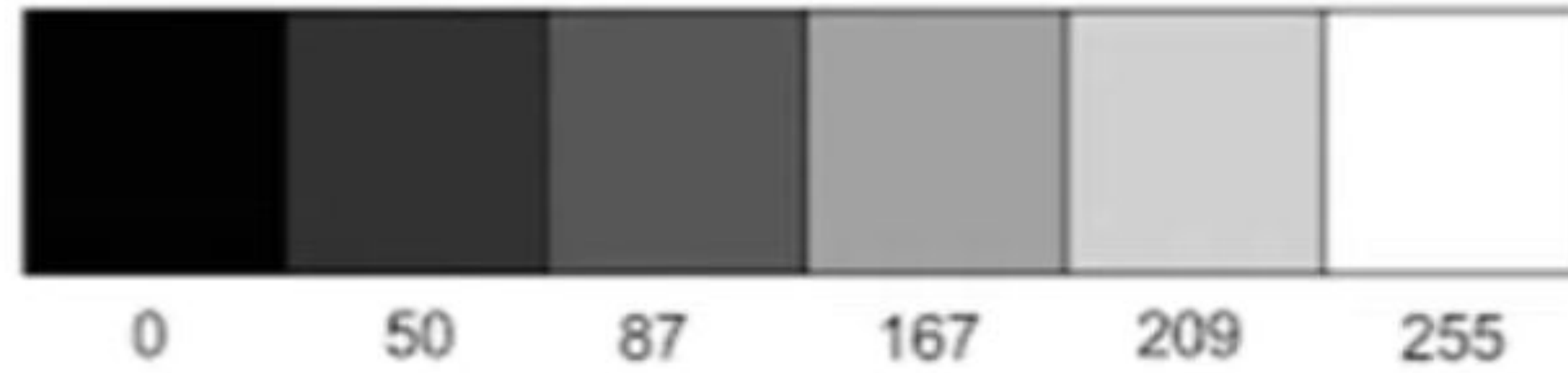


# NumPy - 2D Array (1)

 PassengerId	# Survived	# Pclass	 Name	 Sex	# Age	# SibSp	# Parch	 Ticket
 1891	 01	 13	891 unique values	male female 65% 35%	 0.4280	 08	 06	6 unique values
1	0	3	Braund, Mr. Owen Harris	male	22	1	0	A/5 21171
2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Thayer)	female	38	1	0	PC 17599
3	1	3	Heikkinen, Miss. Laina	female	26	0	0	STON/O2. 3
4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35	1	0	113803
5	0	3	Allen, Mr. William Henry	male	35	0	0	373450
6	0	3	Moran, Mr. James	male		0	0	330877
7	0	1	McCarthy, Mr. Timothy J	male	54	0	0	17463
8	0	3	Palsson, Master. Gosta Leonard	male	2	3	1	349909
9	1	3	Johnson, Mrs. Oscar	female	27	0	2	347742



# NumPy - 2D Array (2)



157	153	174	168	150	152	129	151	172	161	155	156
155	182	163	74	75	62	33	77	110	170	180	154
188	180	50	14	34	6	10	33	48	186	159	181
206	109	5	134	131	111	120	204	166	15	56	180
194	68	137	151	237	239	239	228	227	87	71	201
172	105	207	233	233	214	220	229	228	98	74	206
188	88	179	209	185	215	211	158	138	75	20	169
188	97	165	84	10	168	134	11	31	62	22	148
199	168	191	193	158	227	178	143	182	186	36	190
205	174	155	152	236	231	149	178	228	43	95	234
198	216	118	149	236	187	85	150	79	38	218	241
198	234	147	108	227	210	127	102	36	181	255	224
198	214	173	66	103	143	96	50	2	189	249	215
187	196	235	75	1	81	47	0	6	217	258	211
183	202	237	145	0	0	12	108	200	138	243	226
195	206	123	207	177	121	123	200	175	13	96	218

157	153	174	168	150	152	129	151	172	161	155	156
155	182	163	74	75	62	33	77	110	170	180	154
188	180	50	14	34	6	10	33	48	186	159	181
206	109	5	134	131	111	120	204	166	15	56	180
194	68	137	251	237	239	238	228	227	87	71	201
172	105	207	233	233	214	220	229	228	98	74	206
188	88	179	209	185	215	211	158	138	75	20	169
188	97	165	84	10	168	134	11	31	62	22	148
199	168	191	193	158	227	178	143	182	186	36	190
205	174	155	252	236	231	149	178	228	43	95	234
198	216	118	149	236	187	85	150	79	38	218	241
198	234	147	108	227	210	127	102	36	181	255	224
198	214	173	66	103	143	96	50	2	189	249	215
187	196	235	75	1	81	47	0	6	217	258	211
183	202	237	145	0	0	12	108	200	138	243	226
195	206	123	207	177	121	123	200	175	13	96	218



# NumPy - 3D Array





# NumPy - shape

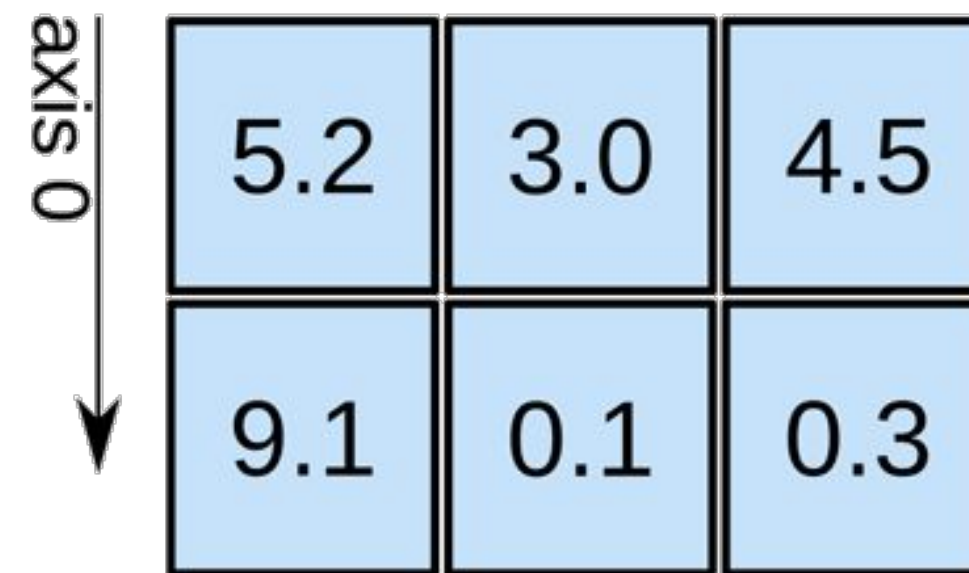
1D array



axis 0 →

shape: (4,)

2D array

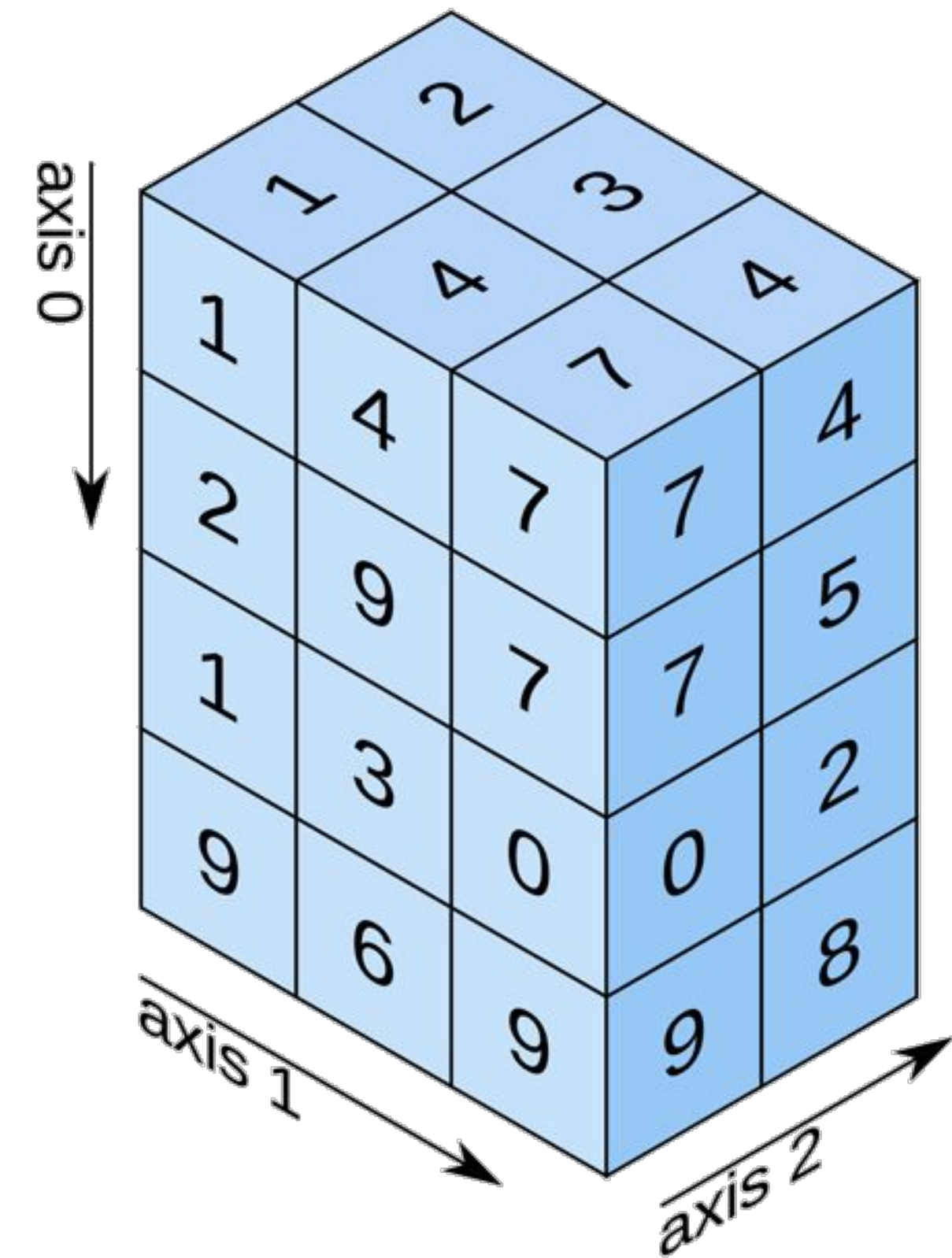


axis 0 ↓

axis 1 →

shape: (2, 3)

3D array



shape: (4, 3, 2)

# numpy.array

Create an array

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

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





# NumPy - other ndarray Constructors

numpy.**full**(**shape**, **fill\_value**)

*Return a new array of given **shape**, filled with fill value.*

numpy.**zeros**(**shape**)

*Return a new array of given **shape**, filled with zeros.*

numpy.**ones**(**shape**)

*Return a new array of given **shape**, filled with ones.*

numpy.**eye**()

*Return a 2-D array with ones on the diagonal and zeros elsewhere.*

numpy.**random.randn**(**shape**)

*Return a sample (or samples) from the “standard normal” distribution.*





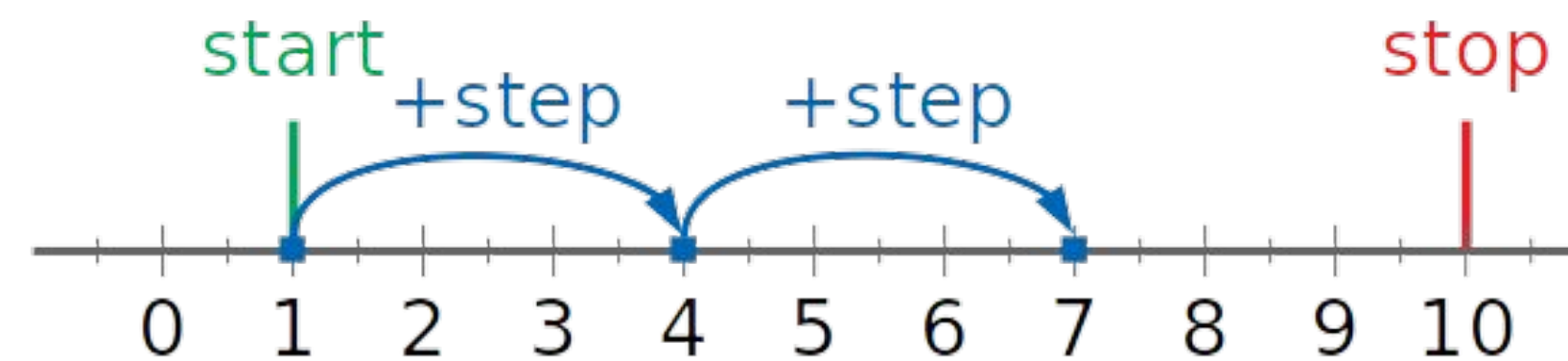
# NumPy - 1D array Constructors

numpy.linspace(start, stop, num)

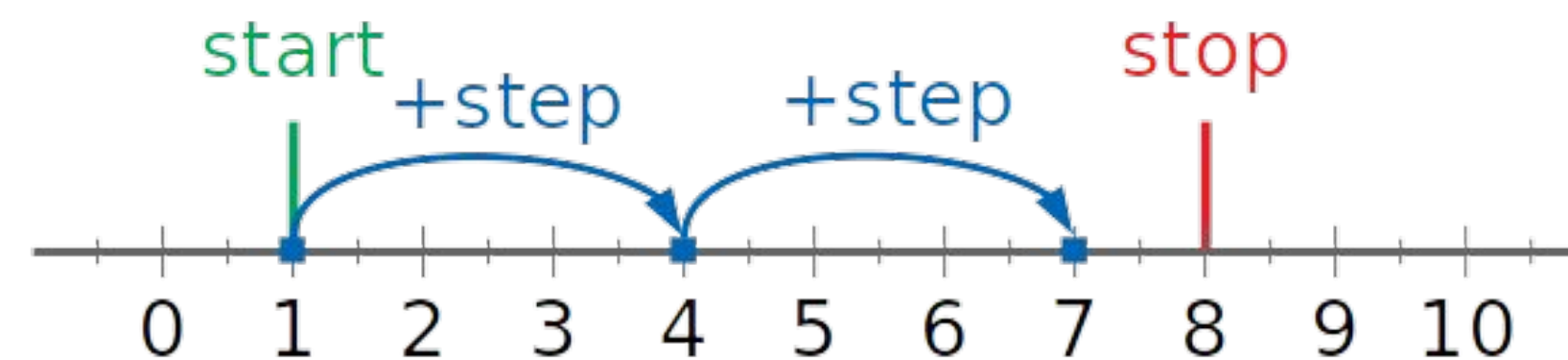
numpy.arange(start, stop, step)

Both **linspace** and **arange** return evenly spaced numbers over a specified interval.

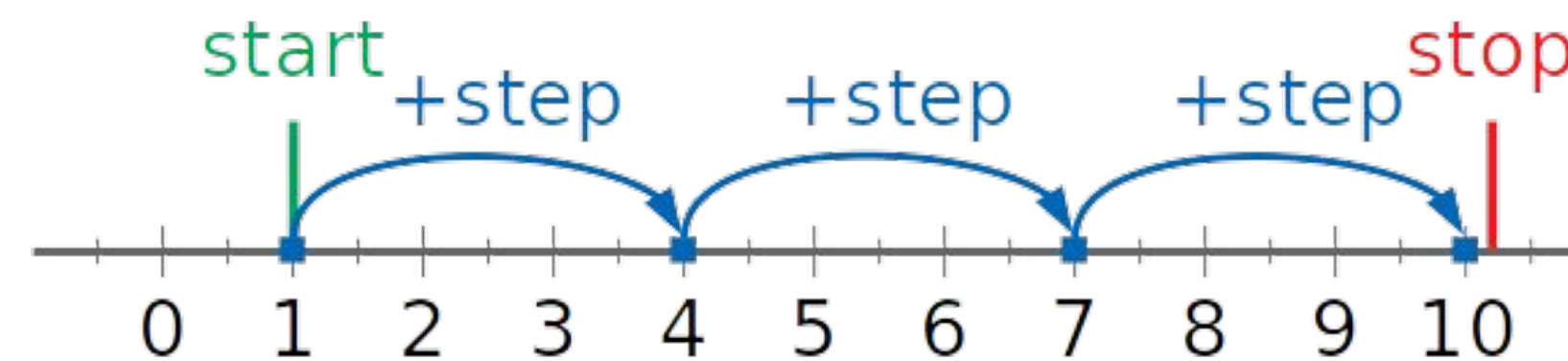
```
>>> np.arange(1, 10, 3)  
array([1, 4, 7])
```



```
>>> np.arange(1, 8, 3)  
array([1, 4, 7])
```



```
>>> np.arange(1, 10.1, 3)  
array([1., 4., 7., 10.])
```





# NumPy - dtypes

## Array types and conversions between types

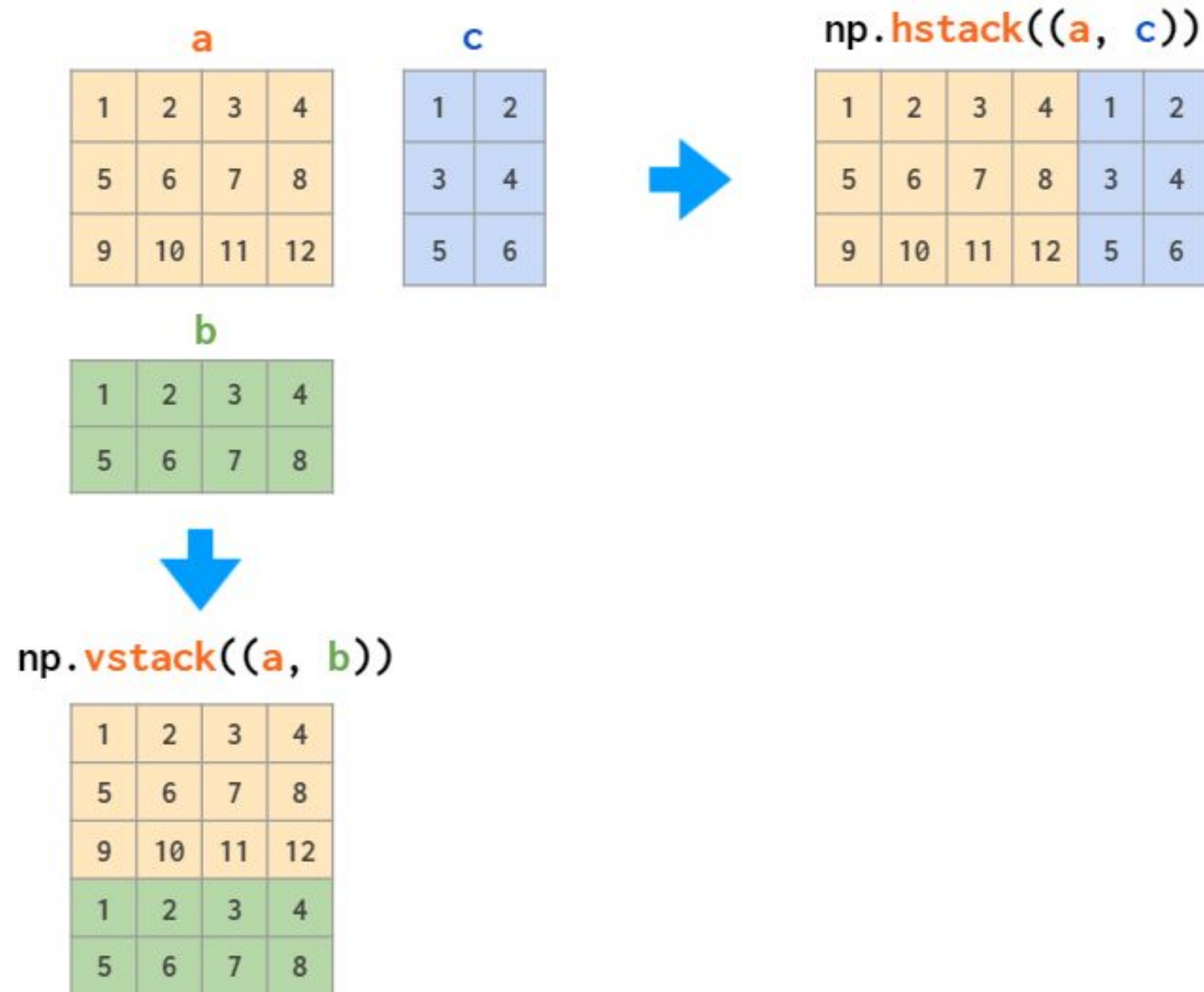
NumPy supports a much greater variety of numerical types than Python does. This section shows which are available, and how to modify an array's data-type.

The primitive types supported are tied closely to those in C:

Numpy type	C type	Description
<code>numpy.bool_</code>	<code>bool</code>	Boolean (True or False) stored as a byte
<code>numpy.byte</code>	<code>signed char</code>	Platform-defined
<code>numpy.ubyte</code>	<code>unsigned char</code>	Platform-defined
<code>numpy.short</code>	<code>short</code>	Platform-defined
<code>numpy.ushort</code>	<code>unsigned short</code>	Platform-defined
<code>numpy.intc</code>	<code>int</code>	Platform-defined
<code>numpy.uintc</code>	<code>unsigned int</code>	Platform-defined
<code>numpy.int</code>	<code>long</code>	Platform-defined



# NumPy - np.hstack, np.vstack

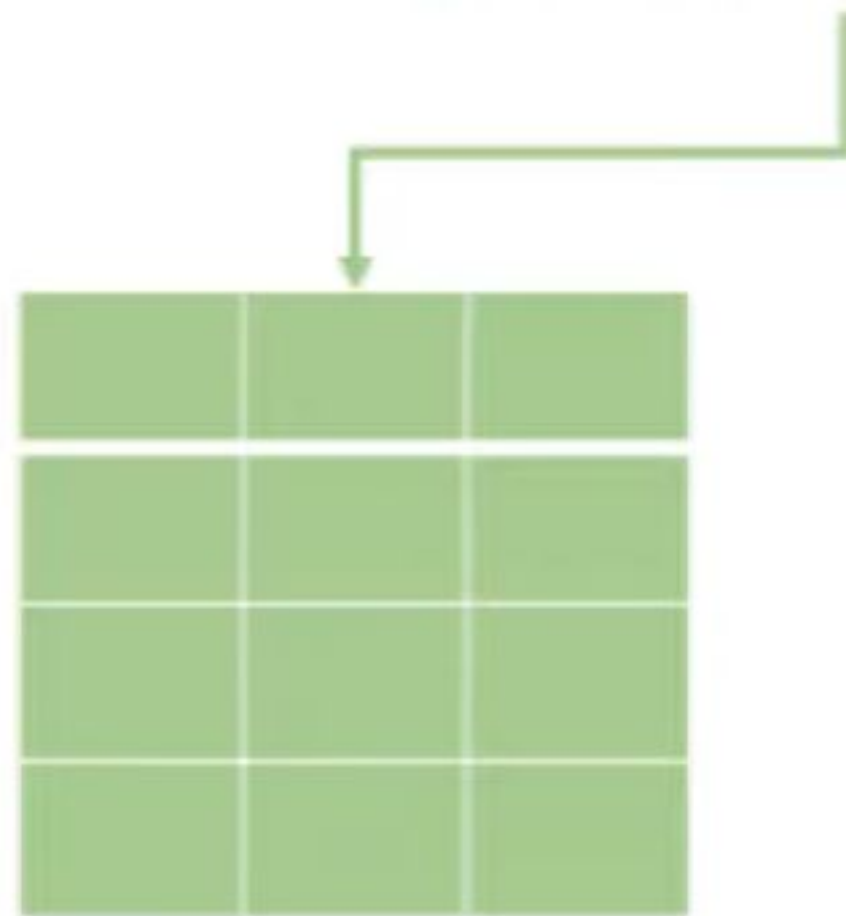




# NumPy - ndarray.reshape



```
C = C.reshape((3, 4))
```

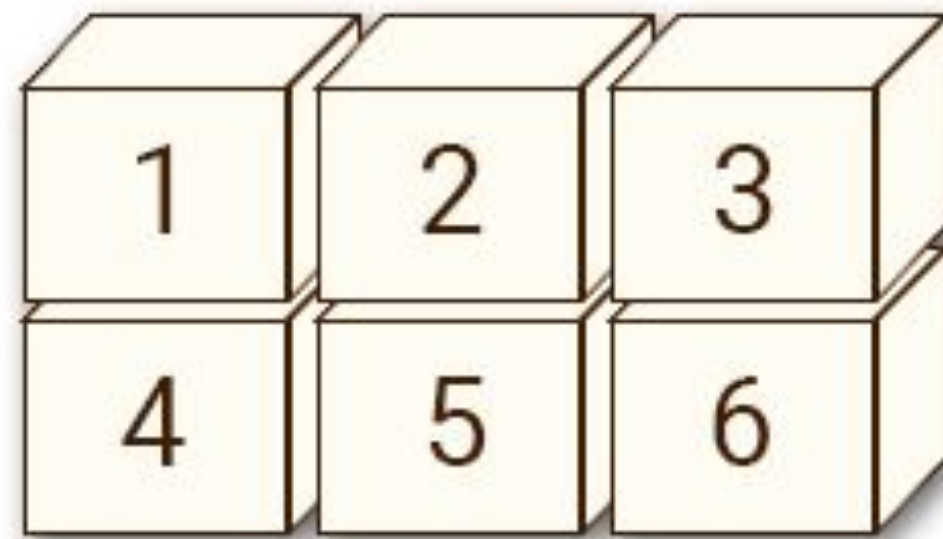


ndarray.**reshape**(**shape**)

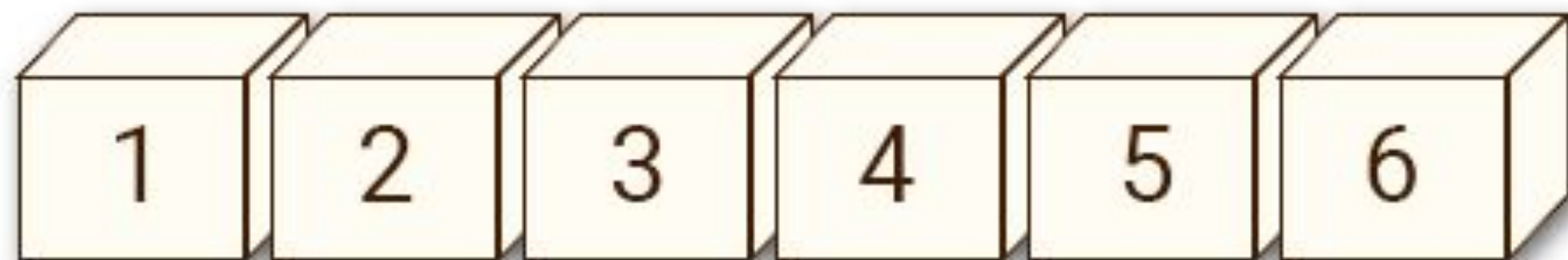
*Returns an array containing the same data with a new **shape**.*



# NumPy - ndarray.ravel



np.ravel()



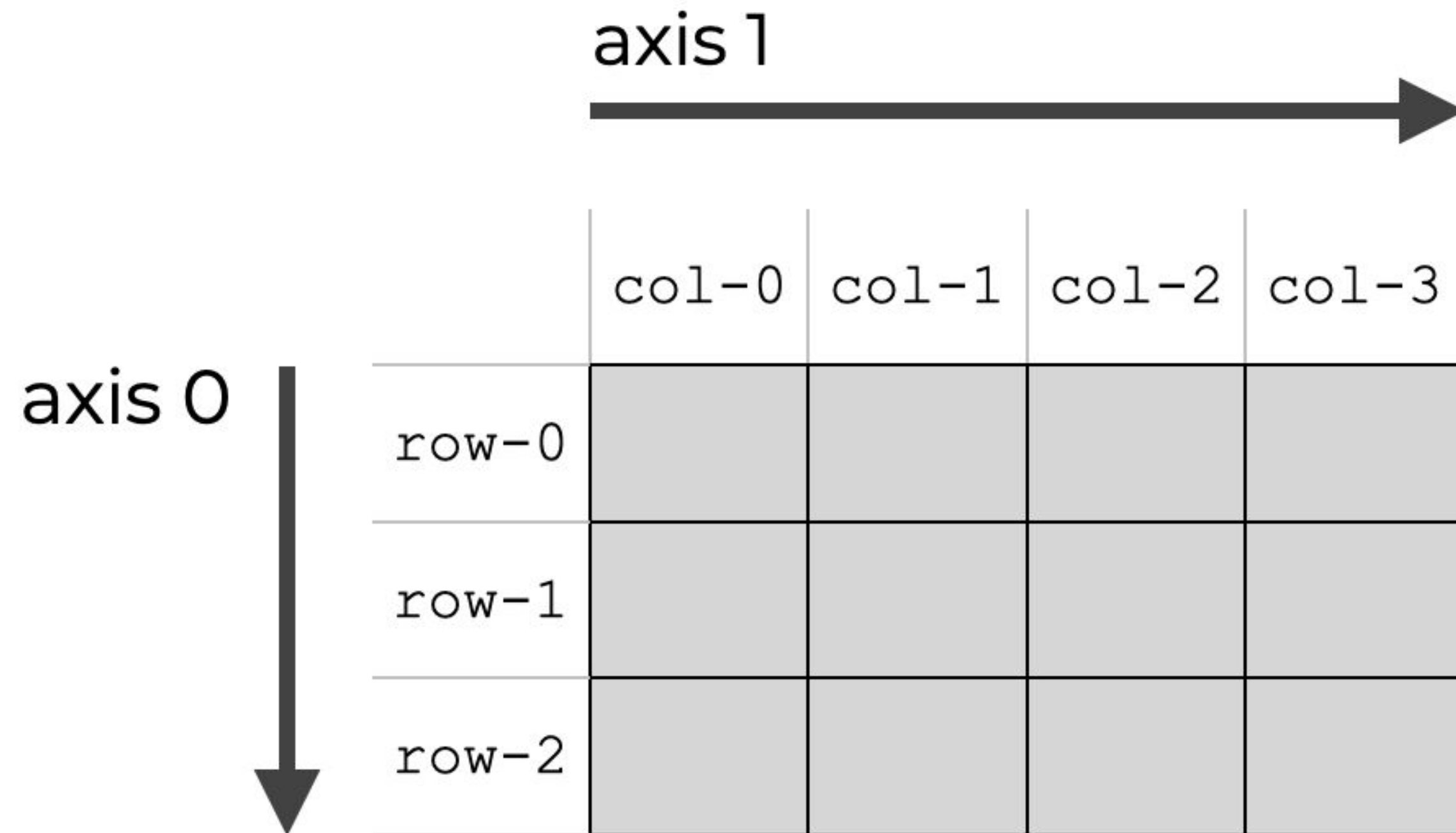
ndarray.**ravel**()  
*Returns a flattened array.*

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# NumPy - Axes

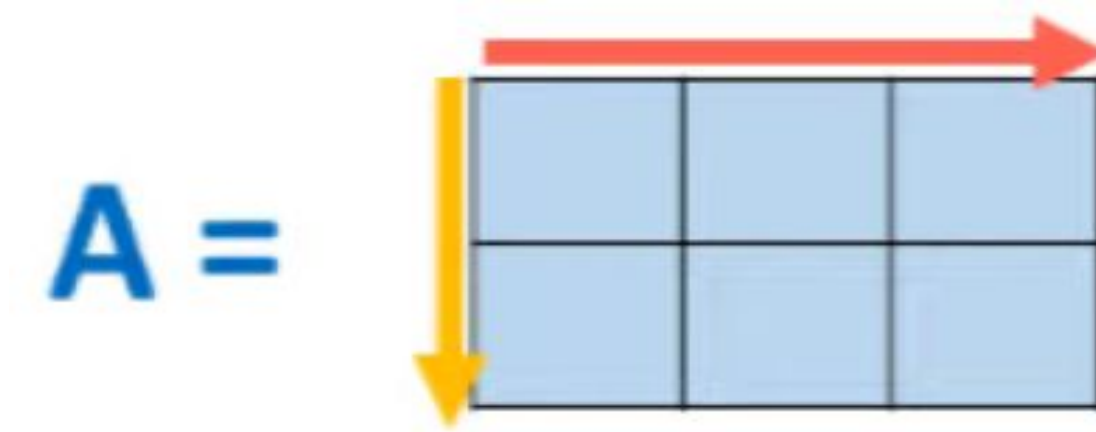


NumPy axes are **the directions along the rows and columns**. Just like coordinate systems, NumPy arrays also have axes. In a 2-dimensional NumPy array, the axes are the directions along the rows and columns.



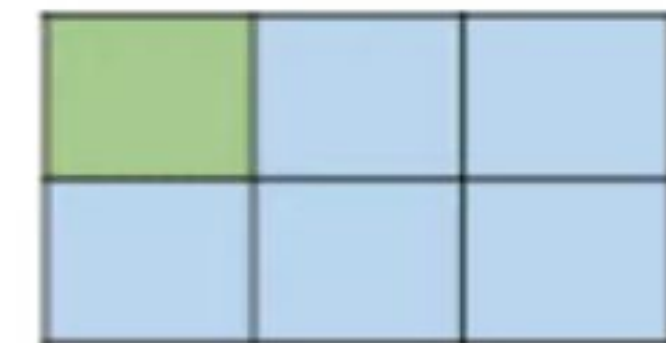
# NumPy - Indexing

**Array indexing is the same as accessing an array element.** You can access an array element by referring to its index number. The indexes in NumPy arrays start with 0, meaning that the first element has index 0, and the second has index 1 etc.

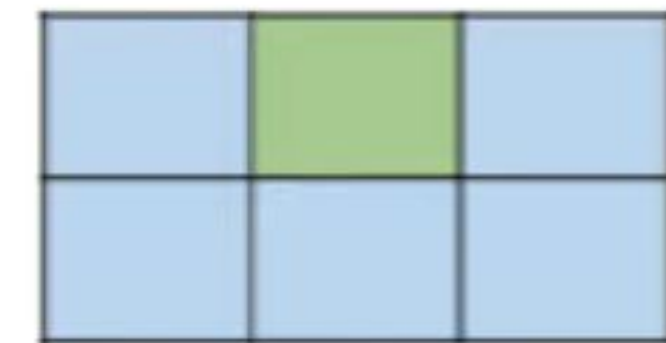


$A[\textit{ligne}, \textit{colonne}]$

$A[0, 0]$



$A[0, 1]$





# NumPy - Slicing

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	16

Row one, columns two to four

```
>>> arr[1, 2:4]
array([7, 8])
```

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	16

All rows in column one

```
>>> arr[:, 1]
array([2, 6, 10, 14])
```

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	16

All rows after row two,  
all columns after column two

```
>>> arr[2:, 2:]
array([[11, 12],
       [15, 16]])
```

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	16

Every other row after row one,  
every other column

```
>>> arr[1::2, ::2]
array([[5, 7],
       [13, 15]])
```



# NumPy - Boolean Indexing

**A =**

5	4	5
4	4	4
5	4	5

**A < 5** →

F	T	F
T	T	T
F	T	F

**A[A < 5] = 10** →

5	10	5
10	10	10
5	10	5

You can index specific values from a NumPy array using another NumPy array of Boolean values on one axis to specify the indices you want to access.





# NumPy - ndarray.sum

*axis 1*

*axis 0*

5	0	3
3	7	9

5	0	3
3	7	9

= = =

8	7	12
---	---	----

**A.sum(*axis=0*) =**

5	0	3
3	7	9

=

8
19

**A.sum(*axis=1*) =**

# NumPy - ndarray.min, ndarray.max

*axis 1*

*axis 0*

5	0	3
3	7	9

5	0	3
3	7	9

= = =

`np.min(axis=0)` = 

3	0	3
---	---	---

`np.min(axis=1)` = 

5	0	3
3	7	9

 = 

0
3





# NumPy - ndarray.argmax, ndarray.argmin

*axis 0* ↓

*axis 1* →

5	0	3
3	7	9

Return indices of the minimum values along the given **axis**.

*0* ↓

*1* ↓

5	0	3
3	7	9

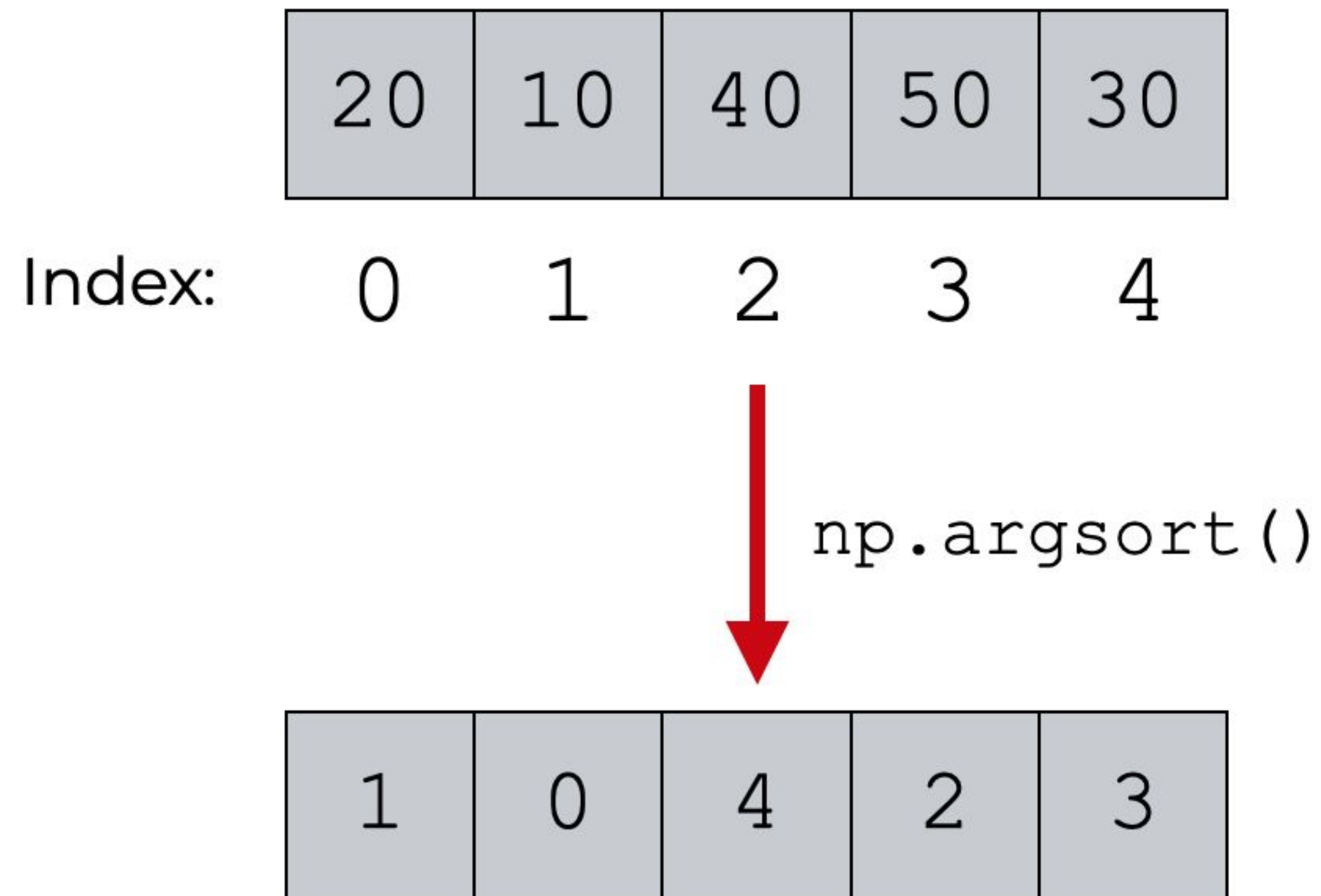
**argmin(*axis=0*)** = 

1	0	0
---	---	---



# NumPy - argsort

`np.argsort()` RETURNS THE INDEX VALUES  
IN AN ORDER THAT WOULD SORT THE INPUT ARRAY





# NumPy - Mathematical Functions

## Trigonometric functions

<code>sin</code> (x, /[, out, where, casting, order, ...])	Trigonometric sine, element-wise.
<code>cos</code> (x, /[, out, where, casting, order, ...])	Cosine element-wise.
<code>tan</code> (x, /[, out, where, casting, order, ...])	Compute tangent element-wise.
<code>arcsin</code> (x, /[, out, where, casting, order, ...])	Inverse sine, element-wise.
<code>arccos</code> (x, /[, out, where, casting, order, ...])	Trigonometric inverse cosine, element-wise.
<code>arctan</code> (x, /[, out, where, casting, order, ...])	Trigonometric inverse tangent, element-wise.
<code>hypot</code> (x1. x2. /[, out, where, casting, ...])	Given the "legs" of a right triangle. return its hypotenuse.

## Exponents and logarithms #

<code>exp</code> (x, /[, out, where, casting, order, ...])	Calculate the exponential of all elements in the input array.
<code>expm1</code> (x, /[, out, where, casting, order, ...])	Calculate $\exp(x) - 1$ for all elements in the array.
<code>exp2</code> (x, /[, out, where, casting, order, ...])	Calculate $2^{*p}$ for all $p$ in the input array.
<code>log</code> (x, /[, out, where, casting, order, ...])	Natural logarithm, element-wise.
<code>log10</code> (x, /[, out, where, casting, order, ...])	Return the base 10 logarithm of the input array, element-wise.
<code>log2</code> (x, /[, out, where, casting, order, ...])	Base-2 logarithm of x.
<code>log1p</code> (x, /[, out, where, casting, order, ...])	Return the natural logarithm of one plus the input array, element-wise.

## Hyperbolic functions

<code>sinh</code> (x, /[, out, where, casting, order, ...])	Hyperbolic sine, element-wise.
<code>cosh</code> (x, /[, out, where, casting, order, ...])	Hyperbolic cosine, element-wise.
<code>tanh</code> (x, /[, out, where, casting, order, ...])	Compute hyperbolic tangent element-wise.
<code>arcsinh</code> (x, /[, out, where, casting, order, ...])	Inverse hyperbolic sine element-wise.
<code>arccosh</code> (x, /[, out, where, casting, order, ...])	Inverse hyperbolic cosine, element-wise.
<code>arctanh</code> (x, /[, out, where, casting, order, ...])	Inverse hyperbolic tangent element-wise.

## Sums, products, differences

<code>prod</code> (a[, axis, dtype, out, keepdims, ...])	Return the product of array elements over a given axis.
<code>sum</code> (a[, axis, dtype, out, keepdims, ...])	Sum of array elements over a given axis.
<code>nanprod</code> (a[, axis, dtype, out, keepdims, ...])	Return the product of array elements over a given axis treating Not a Numbers (NaNs) as ones.
<code>nansum</code> (a[, axis, dtype, out, keepdims, ...])	Return the sum of array elements over a given axis treating Not a Numbers (NaNs) as zero.
<code>cumprod</code> (a[, axis, dtype, out])	Return the cumulative product of elements along a given axis.
<code>cumsum</code> (a[, axis, dtype, out])	Return the cumulative sum of the elements along a given axis.



# NumPy - Statistics

## Order statistics

<code>ptp(a[, axis, out, keepdims])</code>	Range of values (maximum - minimum) along an axis.
<code>percentile(a, q[, axis, out, ...])</code>	Compute the q-th percentile of the data along the specified axis.
<code>nanpercentile(a, q[, axis, out, ...])</code>	Compute the qth percentile of the data along the specified axis, while ignoring nan values.
<code>quantile(a, q[, axis, out, overwrite_input, ...])</code>	Compute the q-th quantile of the data along the specified axis.
<code>nanquantile(a, q[, axis, out, ...])</code>	Compute the qth quantile of the data along the specified axis, while ignoring nan values.

## Correlating

<code>corrcoef(x[, y, rowvar, bias, ddof, dtype])</code>	Return Pearson product-moment correlation coefficients.
<code>correlate(a, v[, mode])</code>	Cross-correlation of two 1-dimensional sequences.
<code>cov(m[, y, rowvar, bias, ddof, fweights, ...])</code>	Estimate a covariance matrix, given data and weights.

## Averages and variances

<code>median(a[, axis, out, overwrite_input, keepdims])</code>	Compute the median along the specified axis.
<code>average(a[, axis, weights, returned, keepdims])</code>	Compute the weighted average along the specified axis.
<code>mean(a[, axis, dtype, out, keepdims, where])</code>	Compute the arithmetic mean along the specified axis.
<code>std(a[, axis, dtype, out, ddof, keepdims, where])</code>	Compute the standard deviation along the specified axis.
<code>var(a[, axis, dtype, out, ddof, keepdims, where])</code>	Compute the variance along the specified axis.
<code>nanmedian(a[, axis, out, overwrite_input, ...])</code>	Compute the median along the specified axis, while ignoring NaNs.
<code>nanmean(a[, axis, dtype, out, keepdims, where])</code>	Compute the arithmetic mean along the specified axis, ignoring NaNs.

## Histograms #

<code>histogram(a[, bins, range, normed, weights, ...])</code>	Compute the histogram of a dataset.
<code>histogram2d(x, y[, bins, range, normed, ...])</code>	Compute the bi-dimensional histogram of two data samples.
<code>histogramdd(sample[, bins, range, normed, ...])</code>	Compute the multidimensional histogram of some data.
<code>bincount(x, /[, weights, minlength])</code>	Count number of occurrences of each value in array of non-negative ints.
<code>histogram_bin_edges(a[, bins, range, weights])</code>	Function to calculate only the edges of the bins used by the <code>histogram</code> function.
<code>digitize(x, bins[, right])</code>	Return the indices of the bins to which each value in input array belongs.



# NumPy - Linear Algebra

## Matrix and vector products

<code>dot(a, b[, out])</code>	Dot product of two arrays.
<code>linalg.multi_dot(arrays, *[, out])</code>	Compute the dot product of two or more arrays in a single function call, while automatically selecting the fastest evaluation order.
<code>vdot(a, b, /)</code>	Return the dot product of two vectors.
<code>inner(a, b, /)</code>	Inner product of two arrays.
<code>outer(a, b[, out])</code>	Compute the outer product of two vectors.
<code>matmul(x1, x2, /[, out, casting, order, ...])</code>	Matrix product of two arrays.
<code>tensordot(a, b[, axes])</code>	Compute tensor dot product along specified axes.
<code>einsum(subscripts, *operands[, out, dtype, ...])</code>	Evaluates the Einstein summation convention on the operands.

## Solving equations and inverting matrices

<code>linalg.solve(a, b)</code>	Solve a linear matrix equation, or system of linear scalar equations.
<code>linalg.tensorsolve(a, b[, axes])</code>	Solve the tensor equation $\mathbf{a} \cdot \mathbf{x} = \mathbf{b}$ for $\mathbf{x}$ .
<code>linalg.lstsq(a, b[, rcond])</code>	Return the least-squares solution to a linear matrix equation.
<code>linalg.inv(a)</code>	Compute the (multiplicative) inverse of a matrix.
<code>linalg.pinv(a[, rcond, hermitian])</code>	Compute the (Moore-Penrose) pseudo-inverse of a matrix.
<code>linalg.tensorinv(a[, ind])</code>	Compute the 'inverse' of an N-dimensional array.

## Decompositions

<code>linalg.cholesky(a)</code>	Cholesky decomposition.
<code>linalg.qr(a[, mode])</code>	Compute the qr factorization of a matrix.
<code>linalg.svd(a[, full_matrices, compute_uv, ...])</code>	Singular Value Decomposition.

## Matrix eigenvalues

<code>linalg.eig(a)</code>	Compute the eigenvalues and right eigenvectors of a square array.
<code>linalg.eigh(a[, UPLO])</code>	Return the eigenvalues and eigenvectors of a complex Hermitian (conjugate symmetric) or a real symmetric matrix.
<code>linalg.eigvals(a)</code>	Compute the eigenvalues of a general matrix.
<code>linalg.eigvalsh(a[, UPLO])</code>	Compute the eigenvalues of a complex Hermitian or real symmetric matrix.

## Norms and other numbers

<code>linalg.norm(x[, ord, axis, keepdims])</code>	Matrix or vector norm.
<code>linalg.cond(x[, p])</code>	Compute the condition number of a matrix.
<code>linalg.det(a)</code>	Compute the determinant of an array.
<code>linalg.matrix_rank(A[, tol, hermitian])</code>	Return matrix rank of array using SVD method
<code>linalg.slogdet(a)</code>	Compute the sign and (natural) logarithm of the determinant of an array.
<code>trace(a[, offset, axis1, axis2, dtype, out])</code>	Return the sum along diagonals of the array.

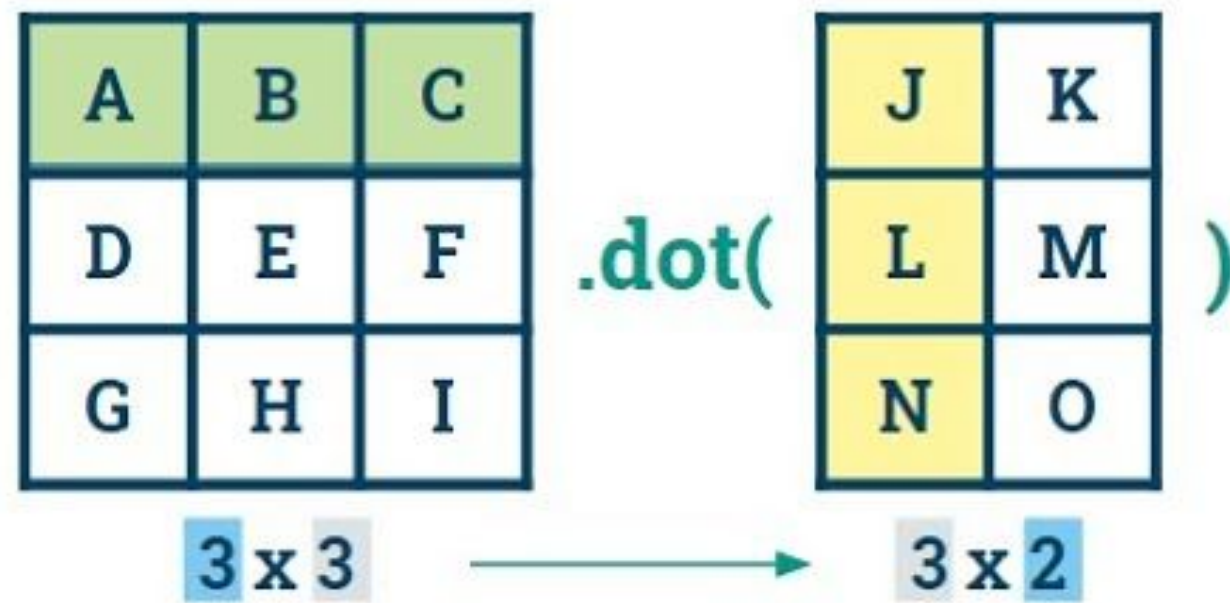


# NumPy - Dot Product

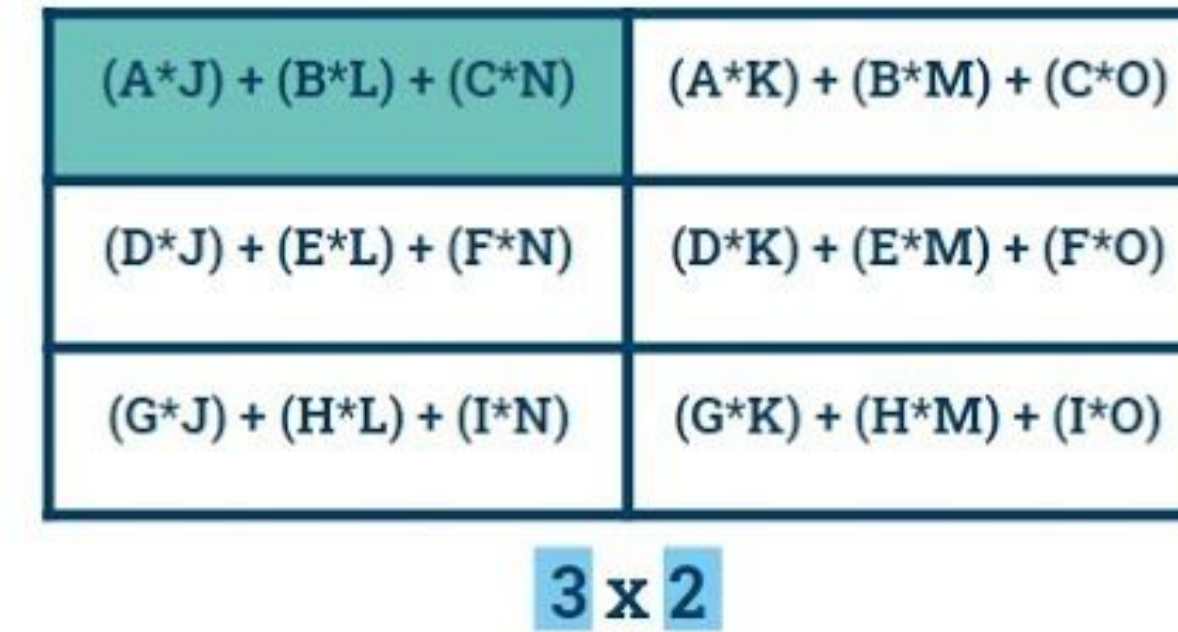
Video 16

Machine Learning

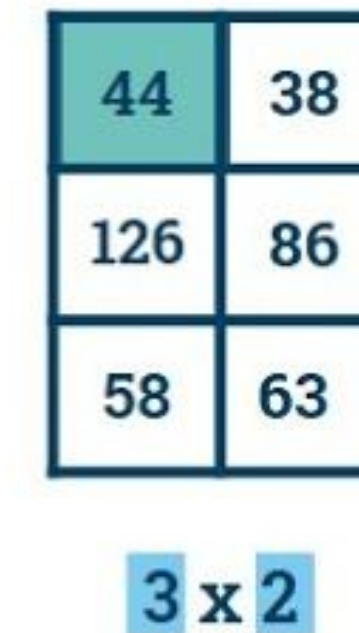
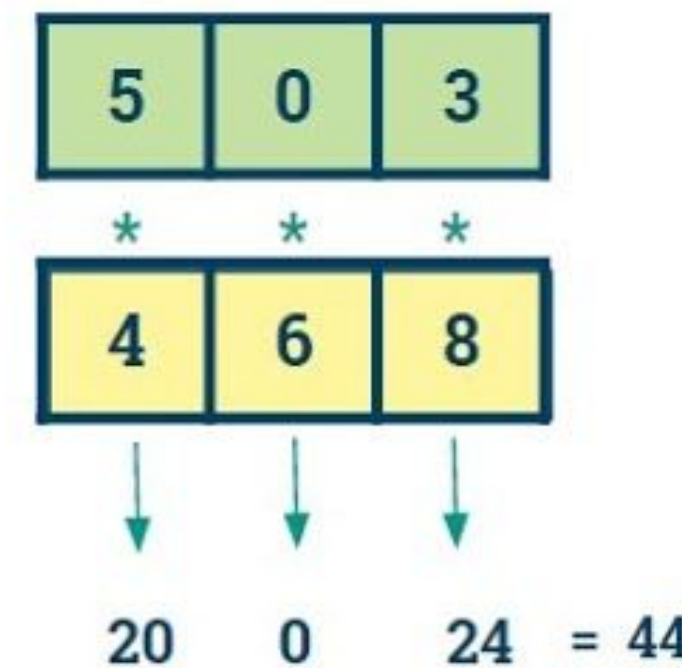
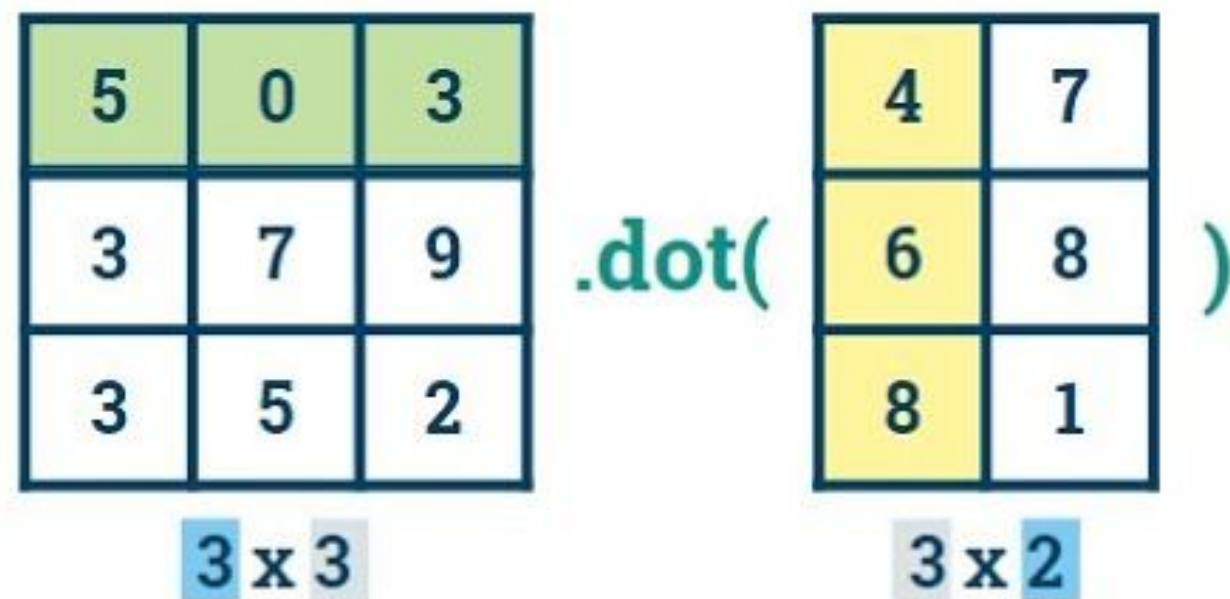
## Dot Product in NumPy



Numbers on the inside must match



New size is same as outside numbers





# NumPy - Broadcasting

(3,3)      (3,) or (1,3)      (3,3)

1	2	3
4	5	6
7	8	9

\*

-1	0	1
-1	0	1
-1	0	1

=

-1	0	3
-4	0	6
-7	0	9

(3,3)      (3,1)      (3,3)

1	2	3
4	5	6
7	8	9

/

3	3	3
6	6	6
9	9	9

=

.3	.7	1.
.6	.8	1.
.8	.9	1.

(3,) or (1,3)      (3,1)      (3,3)

1	2	3
1	2	3
1	2	3

\*

1	1	1
2	2	2
3	3	3

=

1	2	3
2	4	6
3	6	9

