

Numpy

Dimensions

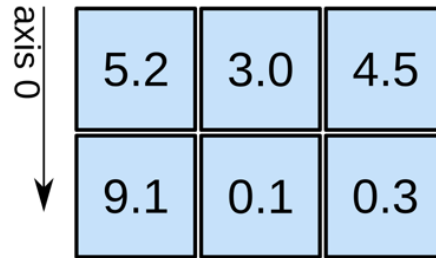
1D array



axis 0 →

shape: (4,)

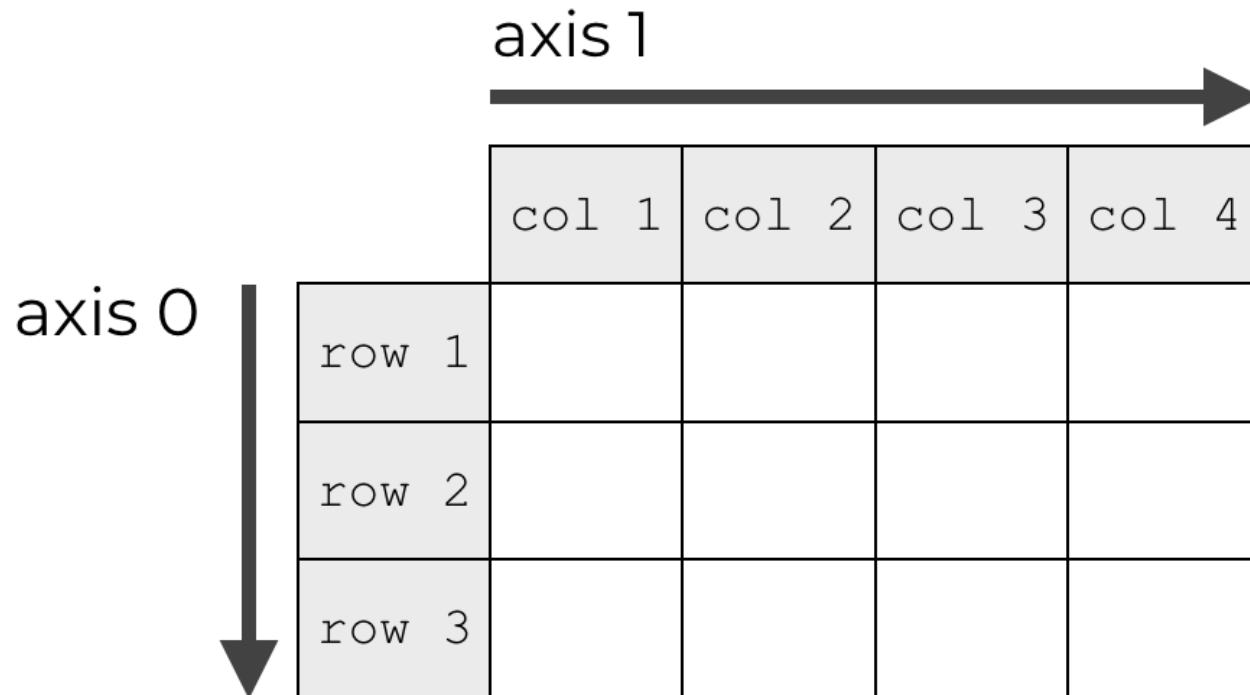
2D array



axis 1 →

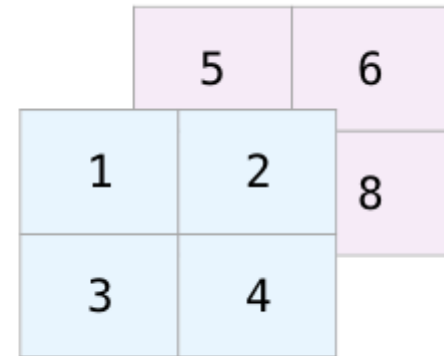
shape: (2, 3)

axis



3-Dimensions

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



```
[[[1 2]  
  [3 4]]
```

```
[[5 6]  
 [7 8]]]
```

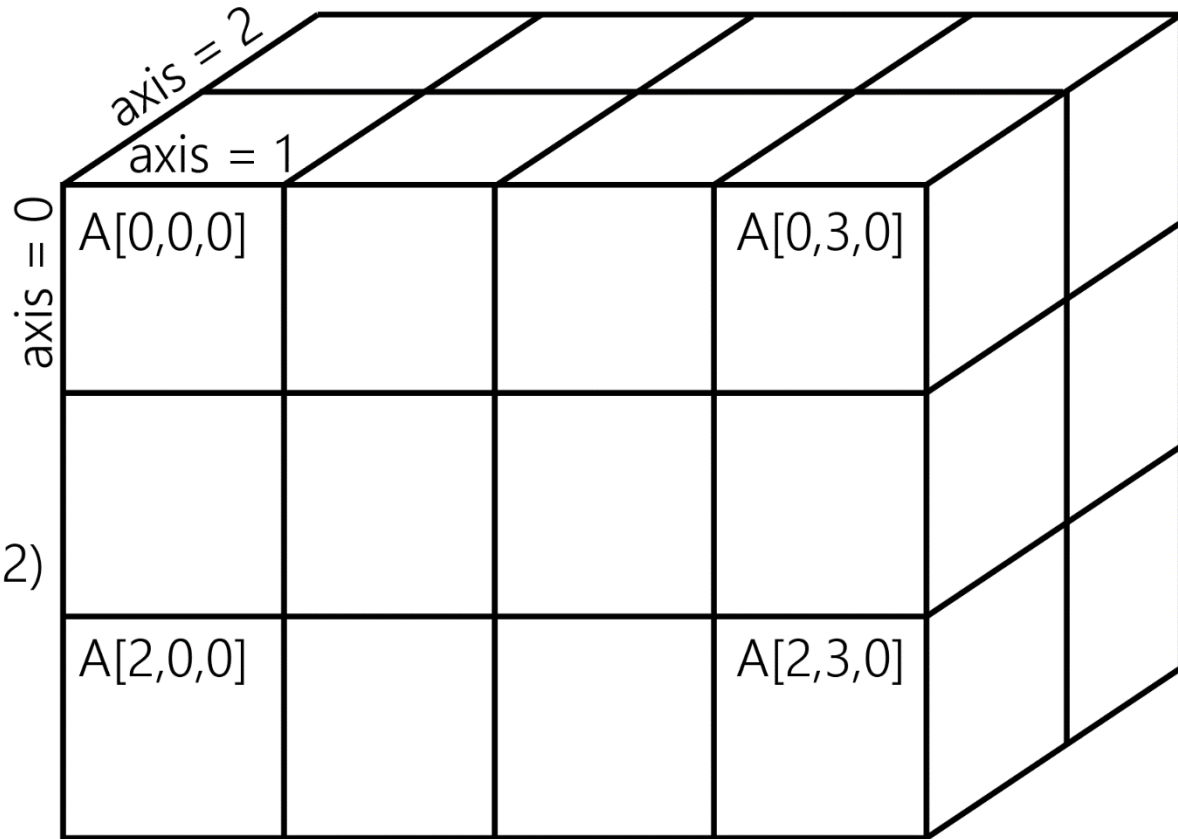
3

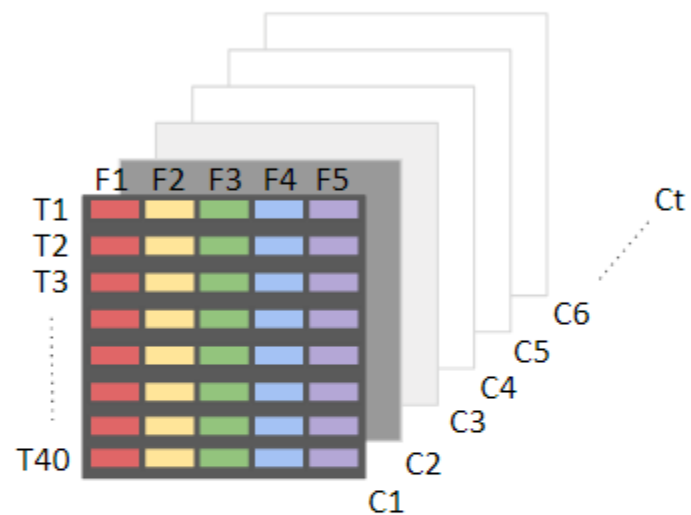
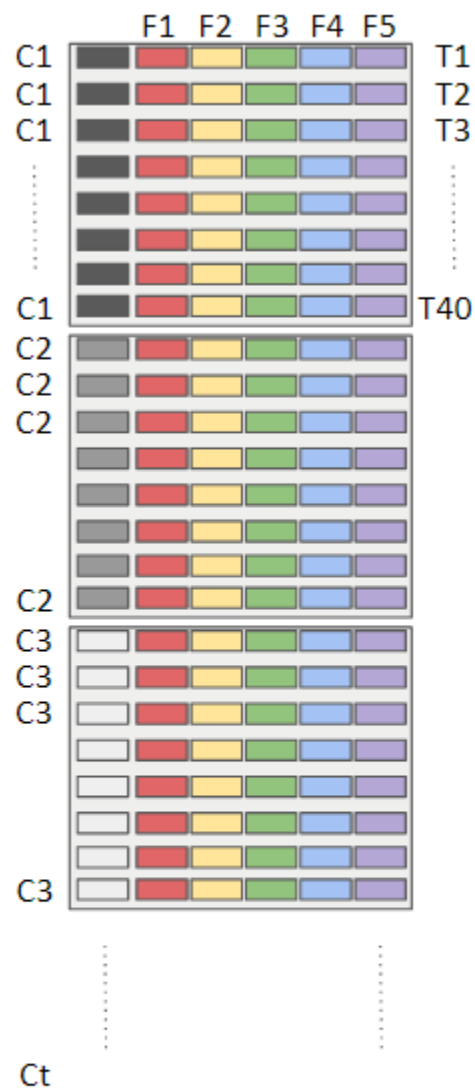
(2, 2, 2)

(Row,Col,Depth/layers)

Numpy -3d

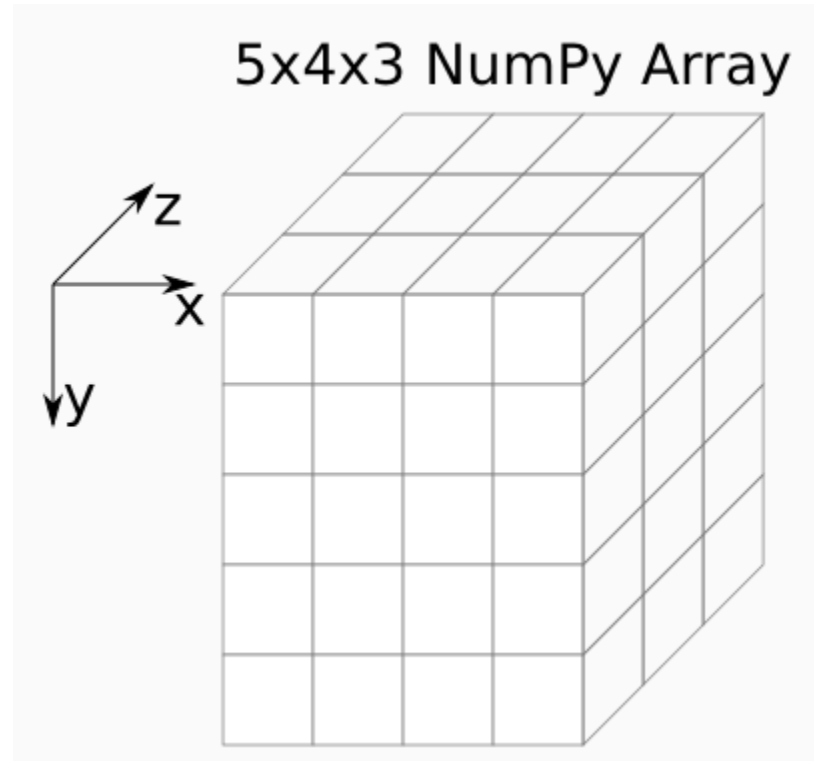
ndarray
ndim = 3
shape = (3, 4, 2)





$\text{shape} = (t, 40, 5)$

Numpy 3d

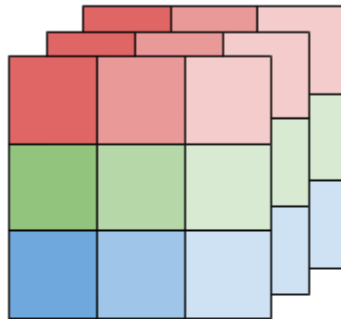


row,col

0,0	0,1	0,2
1,0	1,1	1,2
2,0	2,1	2,2

			0,0	0,1	0,2	1,0	1,1	1,2	2,0	2,1	2,2			
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row,col,depth



0,0,0	0,1,0	0,2,0
1,0,0	1,1,0	1,2,0
2,0,0	2,1,0	2,2,0

0,0,1	0,1,1	0,2,1
1,0,1	1,1,1	1,2,1
2,0,1	2,1,1	2,2,1

0,0,2	0,1,2	0,2,2
1,0,2	1,1,2	1,2,2
2,0,2	2,1,2	2,2,2



3-D Data

- Panel data can be represented in three dimensions. Data that tracks attributes of a cohort (group) of individuals over time could be structured as (**respondents, dates, attributes**). The 1979 National Longitudinal Survey of Youth follows **12,686 respondents over 27 years**.
- Assuming that you have **~500 directly** asked or derived data points per individual, per year, this data would have shape (12686, 27, 500) for a total of 177,604,000 data points.

4-D data

- Color-image data for multiple images is typically stored in four dimensions. Each image is a three-dimensional array of (height, width, channels), where the channels are usually red, green, and blue (RGB) values. A collection of images is then just (image_number, height, width, channels). One thousand 256x256 RGB images would have shape (1000, 256, 256, 3). (An extended representation is RGBA, where the A—alpha—denotes the level of opacity.)