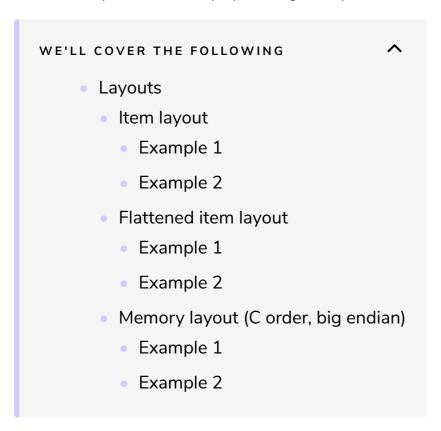
Memory layout

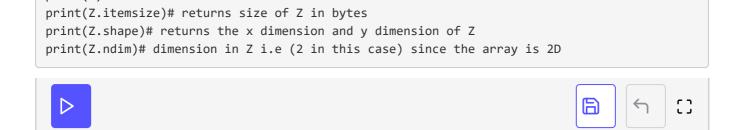
This lesson explains the memory layout using NumPy.



The NumPy documentation defines the ndarray class very clearly:

An instance of class **ndarray** consists of a contiguous one-dimensional segment of computer memory (owned by the array, or by some other object), combined with an indexing scheme that maps N integers into the location of an item in the block.

Said differently, an array is mostly a contiguous block of memory whose parts can be accessed using an indexing scheme. Such indexing scheme is in turn defined by a shape and a data type and this is precisely what is needed when you define a new array:



Here, we know that itemsize is 2 bytes (int16), the shape is (3,3) and the number of dimensions is 2.

To calculate the dimension we can also use len(Z.shape).

Furthermore, we can deduce the strides of the array that define the number of bytes to step in each dimension when traversing the array.

```
import numpy as np
Z = np.arange(9).reshape(3,3).astype(np.int16)
stride = Z.shape[1]*Z.itemsize, Z.itemsize # store stride of Z
print("Stride(as np.int16):",stride)
print("Z.stride(np.16):",Z.strides)
Z = np.arange(9).reshape(3,3).astype(np.int32)
stride= Z.shape[1]*Z.itemsize, Z.itemsize #stores stride of Z
print("Stride(as np.int32):",stride)
print("Z.stride(np.32):",Z.strides)
```

Here in this example, we have to skip 2 bytes (1 value) to move to the next column, but 6 bytes (3 values) to get to the same position in the next row. As such, the **strides** for the array **Z** will be **(6, 2)**.

With all this information, we know how to access a specific item (designed by an index tuple) and more precisely, how to compute the start and end offsets:

```
import numpy as np
Z = np.arange(9).reshape(3,3).astype(np.int16)
offset_start = 0
for i in range(Z.ndim):
   offset_start += Z.strides[i] * i #compute the start offset of Z
   offset_end = offset_start + Z.itemsize #compute the end offset of Z

print("Starting offset:", offset_start)
print("Ending offset:", offset_end)
```

Let's see if this is correct using the tobytes conversion method that construct

Python bytes containing the raw data bytes in the array:

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```
import numpy as np
Z = np.arange(9).reshape(3, 3).astype(np.int16)
index = 1, 1
print(Z[index].tobytes())
    #b'\x04\x00'
offset_start = 0
for i in range(Z.ndim):
    offset_start += Z.strides[i] * index[i]
    offset_end = offset_start + Z.itemsize

print(Z.tobytes()[offset_start:offset_end])#b'\x04\x00'
```

Layouts

This array can be actually considered from different perspectives (i.e. layouts):

Item layout

Consider item layout as a 2-Dimensional Matrix with x rows and y columns.

Example 1

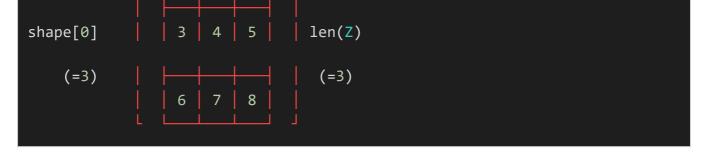
Consider the following example:

```
Z = np.arange(9).reshape(3, 3)
```

Here takes 9 values from 0-8 and is reshaped in 2-D matrix format having dimensions (3 * 3).

Z has the following item layout:

```
shape[1]
(=3)
```



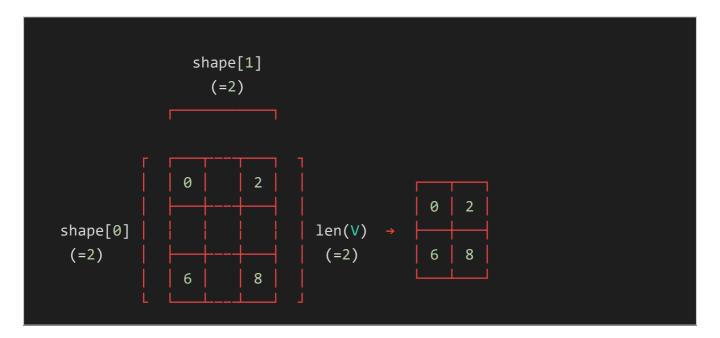
Example 2 #

Consider the following example

```
Z = np.arange(9).reshape(3,3)
V= Z[::2,::2])
```

Here Z takes up 9 values from 0-8 and arranges them in a 3 * 3 matrix V takes values from corners from the grid. i.e V has 4 values.

V has the following item layout:



Flattened item layout

Consider flattened item layout as 1-Dimensional Matrix with 1 row and n columns.

Example 1#

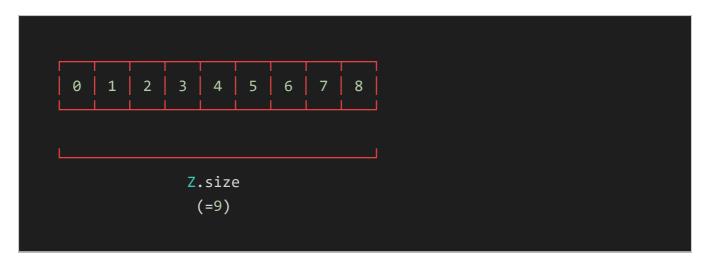
Consider the following example:

```
Z = np.arange(9)
```

It makes 9 indices in the computer's memory and places values from 0 to 8.

7 has the following flattened item levels.

L has the following hatteried ftern layout.

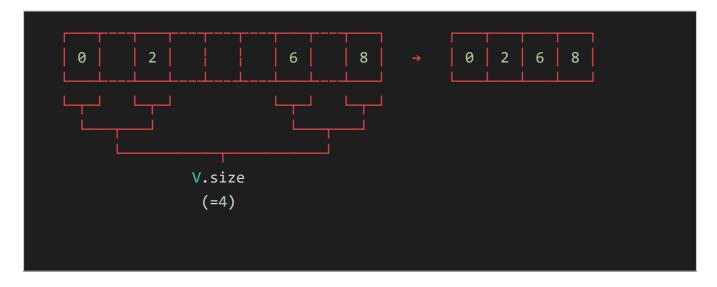


Example 2

Consider the following example

```
Z = np.arange(9).reshape(3,3).astype(np.int16)
V = Z[::2,::2]
V=V.reshape(1,4)
```

Here Z takes up 9 values from 0-8 and arranges them in a 3 * 3 matrix V takes values from corners from the grid. i.e V has 4 values. V has the following flattened item layout:



Memory layout (C order, big endian)

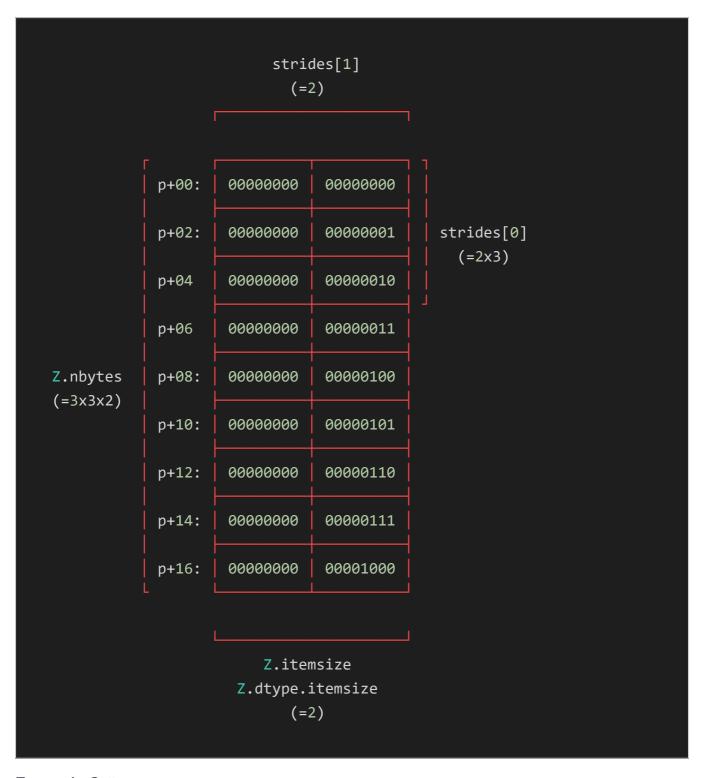
Consider memory layout with rows equal to the number of bytes and columns equal to the number of bytes divided by 8 (i.e Z.itemsize).

Example 1#

Consider the following example:

```
Z = np.arange(9).reshape(3, 3).astype(np.int16)
```

Here, the number of rows is 16 and the number of columns is 2. The total number of bytes is 3 * 3 * 2 where 3 * 3 is the size of the grid and each cell takes 2 bytes, so total 18 bytes. Z has the following memory layout:



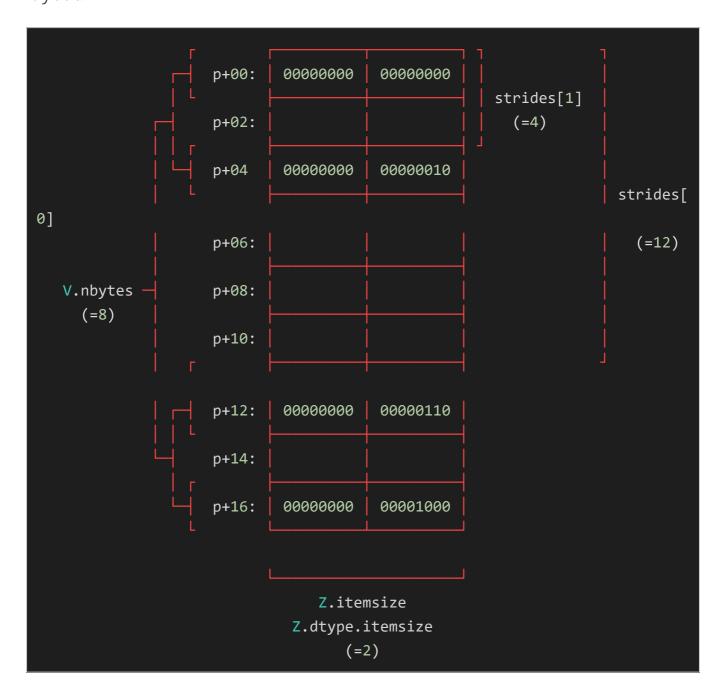
Example 2 #
Consider the following example

```
Z = np.arange(9).reshape(3,3).astype(np.int16)
V = Z[::2,::2]
```

Here we take a slice of \mathbf{z} , the result is a view of the base array \mathbf{z} . In the

memory layout below, since the array takes up only 4 value. and each value is

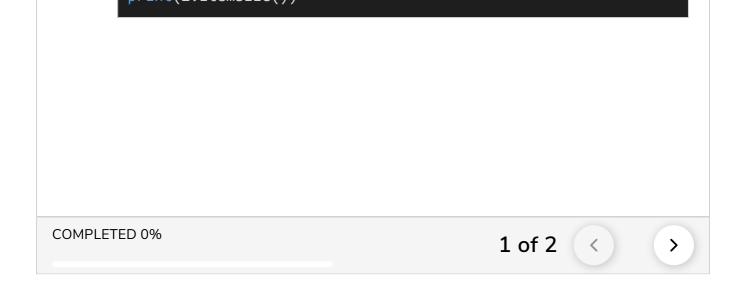
2 bytes so the total bytes are 2 * 4 = 8 bytes. V has the following memory layout:



Such a view is specified using a shape, a dtype and strides because strides cannot be deduced anymore from the dtype and shape only.

```
What is the output of the following code?

Z = np.arange(9).reshape(3,3).astype(np.int32)
print(Z.itemsize())
```



Now, that we have viewed memory layouts, we'll look at views and copies in the next lesson.