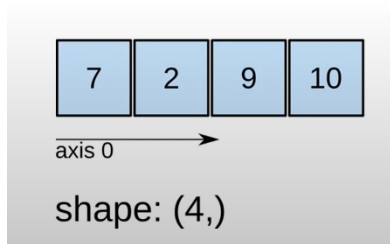


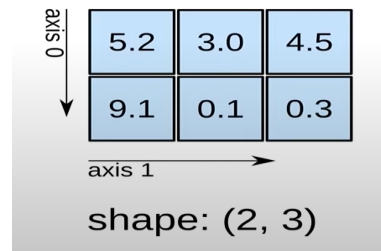
Numpy Libraries

- It is for large, **multi-dimensional array** and **matrix processing**.
- It is very useful for fundamental scientific computations in Machine Learning.

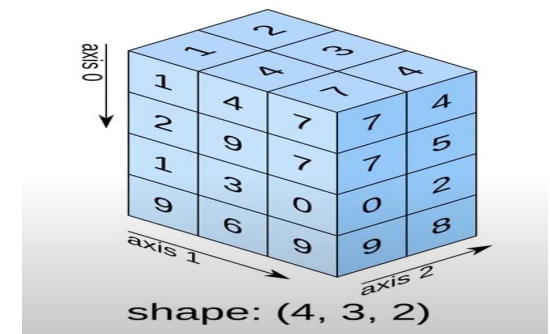
1D array



2D array



3D array

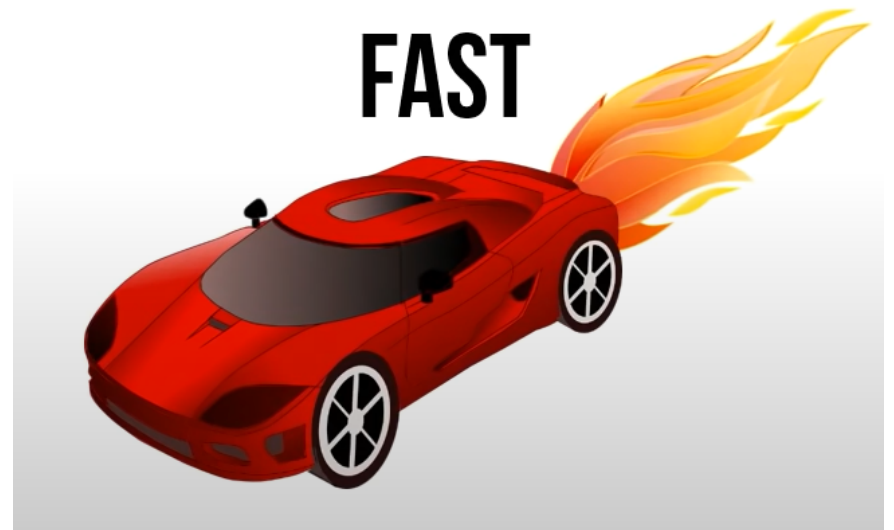


How are Lists different from Numpy

Lists



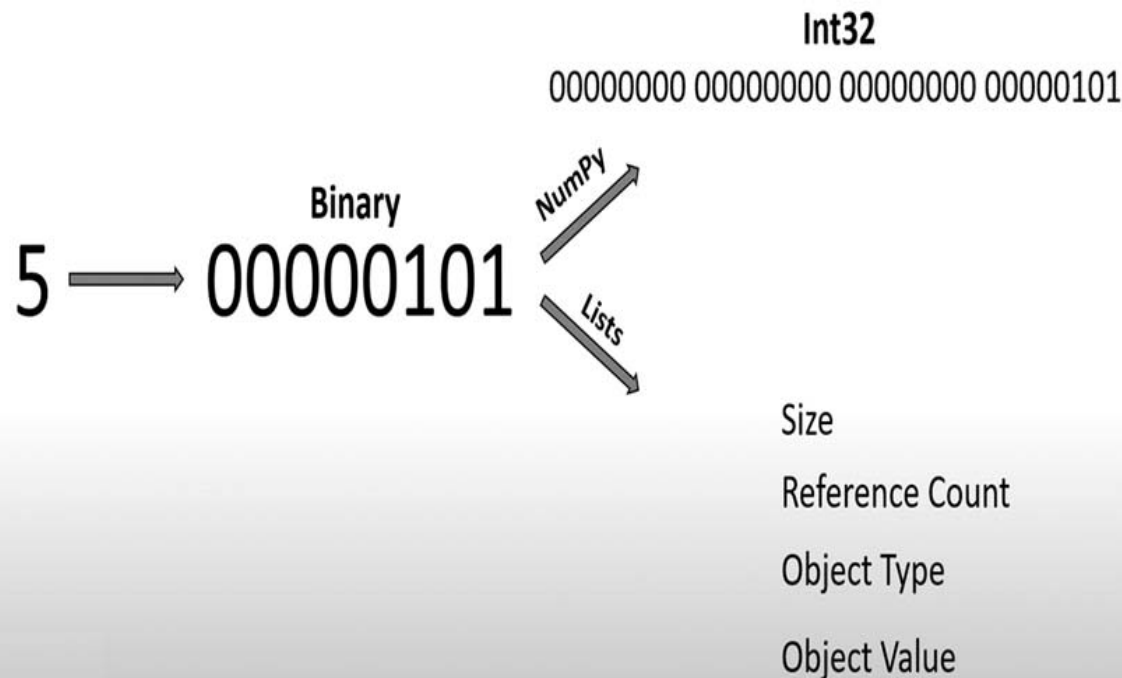
NumPy



Speed

How are Lists different from Numpy

Why is NumPy Faster? - Fixed Type



NumPy:

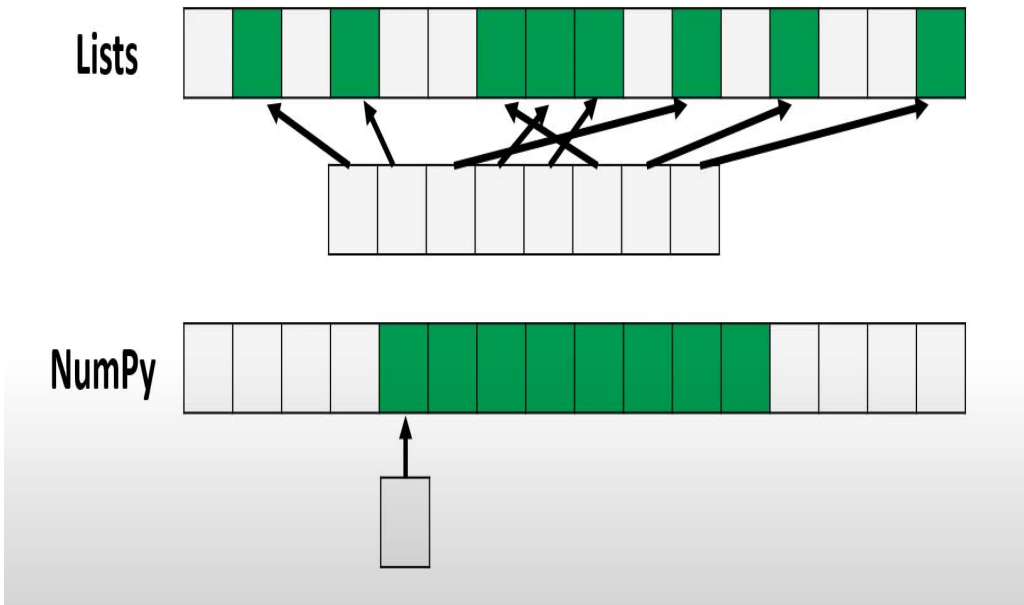
- NumPy uses a fixed type, e.g., int32.
- That means the number is stored in exactly 32 bits (4 bytes).
- So 5 becomes:
00000000 00000000 00000000 0000101
- Stored directly in memory as raw binary.

List:

- Python lists don't store raw numbers, they store references to Python objects.
- Each integer in Python is a full object with: Size, Reference count, Object type, Object value.
- So instead of just 00000101, Python needs to keep extra metadata for every element.

How are Lists different from Numpy

Why is NumPy Faster? - Contiguous Memory



List:

- A Python list is essentially an array of pointers.
- Each list element doesn't store the raw data directly; instead, it stores a reference (pointer) to an object somewhere else in memory.
- This means:
 - Data is not contiguous (scattered across memory).
 - Accessing each element requires following a pointer.

NumPy:

- A NumPy array stores all values as raw binary data in contiguous memory.
- In this diagram, you can see that only one pointer is needed.
- This means:
 - Data is contiguous (tightly packed).
 - The CPU can load and process data much faster.

How are Lists different from Numpy

How are Lists different from Numpy?

Lists

Insertion, deletion,
appending, concatenation,
etc.

NumPy

Insertion, deletion,
appending, concatenation,
etc.

Lots More 😊

How are Lists different from Numpy

How are Lists different from Numpy?

Lists

a = [1,3,5]

b = [1,2,3]

*a*b = ERROR*

NumPy

a = np.array([1,3,5])

b = np.array([1,2,3])

*a*b = np.array([1,6,15])*

How are Lists different from Numpy

Applications of NumPy?

Mathematics (MATLAB Replacement)

Plotting (Matplotlib)

Backend (Pandas, Connect 4, Digital Photography)

Machine Learning

Numpy

Install, import, and upgrade

```
: # pip install
```

```
!pip install numpy
```

Requirement already satisfied: numpy in c:\users\administrator\anaconda3\lib\site-packages (2.3.3)

```
: import numpy #Import
```

```
import numpy as np #standard in python community
```

```
: numpy.__version__
```

```
: '2.3.3'
```

```
: pip install --upgrade numpy
```

Requirement already satisfied: numpy in c:\users\administrator\anaconda3\lib\site-packages (2.3.3)Note: you may need to restart

Numpy

Initializing Array

One-dimensional Array

```
a=np.array([1,2,3])  
a
```

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

Two-dimensional Array

```
a=np.array([[1,2,3],[4,5,6],[7,8,9]])  
a
```

```
array([[1, 2, 3],  
       [4, 5, 6],  
       [7, 8, 9]])
```

Three-dimensional Array

```
# 3-D example  
a=np.array([[[0,1],[2,3]],[[4,5],[6,7]],[[8,9],[10,11]]])  
a
```

```
array([[[ 0,  1],  
        [ 2,  3]],  
       [[ 4,  5],  
        [ 6,  7]],  
       [[ 8,  9],  
        [10, 11]]])
```

Numpy

Why NumPy is Better for Mathematical Computations

```
a = [[1, 3, 5],  
      [4, 5, 6]]  
  
b = [[2, 4, 6],  
      [7, 8, 9]]  
  
c = a * b  # ❌ This does NOT do element-wise multiplication
```

```
-----  
TypeError                                Traceback (most recent call last)  
Cell In[8], line 7  
      1 a = [[1, 3, 5],  
      2       [4, 5, 6]]  
      4 b = [[2, 4, 6],  
      5       [7, 8, 9]]  
----> 7 c = a * b  
  
TypeError: can't multiply sequence by non-int of type 'list'
```

```
: import numpy as np  
  
a = np.array([[1, 3, 5],  
              [4, 5, 6]])  
  
b = np.array([[2, 4, 6],  
              [7, 8, 9]])  
  
c = a * b  
print(c)  
  
[[ 2 12 30]  
 [28 40 54]]
```

Numpy

```
a=np.array([[1,2,3],[4,5,6],[7,8,9]])  
a
```

```
array([[1, 2, 3],  
       [4, 5, 6],  
       [7, 8, 9]])
```

Dimension

```
#Get the dimension (1D array/2D array.....)  
a.ndim
```

2

Shape

```
#Get the shape  
a.shape
```

(3, 3)

Datatype

type of elements stored in a NumPy array

```
#Get type  
a.dtype
```

dtype('int64')

Itemsize

(number of bytes used to store each element in the array)

```
a.itemsize # information of byte
```

8

```
a= np.array([1,2,2.3])  
a
```

```
array([1. , 2. , 2.3])
```

```
a.dtype
```

dtype('float64')

```
a.itemsize # information of byte
```

8

```
a=np.array([1,"Ankit",2.3])  
a.dtype # Unicode
```

dtype('<U32')

```
a.itemsize
```

#In Unicode, every character is stored in exactly 4 bytes

128

string type -not exact length- always bigger length

128

Numpy

Assessing the elements, rows, columns, etc

```
a=np.array([[1,2,3,11,21,23,25,36],[4,5,6,7,8,9,10,0]])  
a
```

```
array([[ 1,  2,  3, 11, 21, 23, 25, 36],  
       [ 4,  5,  6,  7,  8,  9, 10,  0]])
```

Get a specific element

```
#Get a specific element [r,c]  
print(a[0,1]) # a[r,c]
```

```
2
```

Get a specific row

```
#Get a specific row  
print(a[0])
```

```
[1 2 3]
```

Get a specific Column

```
print(a[:,1:2]) # Get a specific Column
```

```
[[2]  
 [5]]
```

- **print(a[0,0:2])**

```
print(a[0,0:2])
```

```
[1 2]
```

- **print(a[:,1:6:2])** # start:stop:step

```
print(a[:,1:6:2])
```

```
[[ 2 11 23]  
 [ 5  7  9]]
```

- **print(a[:,6:1:-1])**

```
print(a[:,6:1:-1])
```

```
[[25 23 21 11  3]  
 [10  9  8  7  6]]
```

Numpy

3-dimensional array

```
# 3-D example
a=np.array([[[0,1],[2,3]],[[4,5],[6,7]],[[8,9],[10,11]]])
a
```

```
array([[[ 0,  1],
        [ 2,  3]],

       [[ 4,  5],
        [ 6,  7]],

       [[ 8,  9],
        [10, 11]]])
```

- **print(a[1,0,1])** # get a specific element

```
#Get a specific element
print(a[1,0,1])
```

5

- **print(a[0])**

```
print(a[0])
```

```
[[0 1]
 [2 3]]
```

- **print(a[0,1,:])**

```
print(a[0,1,:])
```

```
[2 3]
```

- **print(a[:,1,:])**

```
print(a[:,1,:])
```

```
[[ 2  3]
 [ 6  7]
 [10 11]]
```

Numpy

Null Matrix

```
# null Matrix
a=np.zeros(5)
print(a)
```

```
[0. 0. 0. 0. 0.]
```

```
np.zeros((2,5))
```

```
array([[0., 0., 0., 0., 0.],
       [0., 0., 0., 0., 0.]])
```

```
np.zeros(((3,2,5)))
```

```
array([[[0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0.]],
       [[0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0.]],
       [[0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0.]])
```

Full Matrix

```
#Any other number
a=np.full((2,2),1)
a
```

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

```
np.full_like(a,3)
```

```
array([[3, 3],
       [3, 3]])
```

```
np.full(a.shape,5)
```

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

```
np.ones((2,2))
```

```
array([[1., 1.],
       [1., 1.]])
```

Identity Matrix

```
#identity matrix
np.identity(2)
```

```
array([[1., 0.],
       [0., 1.]])
```

```
#identity matrix
matrix = np.eye(2)
print(matrix)
```

```
[[1. 0.]
 [0. 1.]
```

Numpy

Create an array with random numbers (between 0 and 1)

```
#create a array with random number (between 0 and 1)  
np.random.rand(4)  
#np.random.rand(4,2)  
#np.random.rand(4,2,3)
```

```
array([0.80068123, 0.12780885, 0.97459564, 0.5483972 ])
```

```
#create a array with random number (between 0 and 1)  
#np.random.rand(4)  
#np.random.rand(4,2)  
np.random.rand(4,2,3)
```

```
array([[0.42567349, 0.5836056 , 0.74776345],  
       [0.99171107, 0.00296539, 0.98263174]],  
      [[0.38910825, 0.44418267, 0.8932749 ],  
       [0.04631678, 0.29927347, 0.36632794]],  
      [[0.9390877 , 0.75264851, 0.50588025],  
       [0.68145279, 0.02544418, 0.060096  ]],  
      [[0.23270351, 0.47428333, 0.98592488],  
       [0.4426028 , 0.8688685 , 0.86829031]]])
```

Numpy

Create a random number in a given range (eg, between 0 and 50)

```
# create random number between 0 to 50  
np.random.rand(4) * 50
```

```
array([37.05448718, 46.92889767, 27.54244767, 37.11042244])
```

```
#Random Integer value  
np.random.randint(10,size=(3,3)) #between 0 to 9
```

```
array([[4, 1, 2],  
       [5, 6, 5],  
       [2, 4, 9]], dtype=int32)
```

```
np.random.randint(5,10,size=(3,3)) #between 5 to 9
```

```
array([[9, 6, 8],  
       [7, 7, 5],  
       [5, 5, 5]], dtype=int32)
```


Numpy

Concatenate: Joining the arrays (existing axis)

```
arr_1=np.array([[1,2,3],[4,5,6]])  
arr_2=np.array([[7,8,9],[10,11,12]])
```

arr_1

```
array([[1, 2, 3],  
       [4, 5, 6]])
```

arr_2

```
array([[ 7,  8,  9],  
       [10, 11, 12]])
```

Joint vertically rows

```
arr= np.concatenate([arr_1,arr_2]) # axis =0
```

arr

```
array([[ 1,  2,  3],  
       [ 4,  5,  6],  
       [ 7,  8,  9],  
       [10, 11, 12]])
```

Joint the horizontal columns

```
arr= np.concatenate((arr_1,arr_2),axis=1) #axis=1  
arr
```

```
array([[ 1,  2,  3,  7,  8,  9],  
       [ 4,  5,  6, 10, 11, 12]])
```

Numpy

Stack: Joining the arrays along a new axis (Concatenate, merge along an existing axis)

```
arr_1=np.array([[1,2,3],[4,5,6]])  
arr_2=np.array([[7,8,9],[10,11,12]])
```

arr_1

```
array([[1, 2, 3],  
       [4, 5, 6]])
```

arr_2

```
array([[ 7,  8,  9],  
       [10, 11, 12]])
```

```
arr=np.stack((arr_1,arr_2),axis=0)  
print(arr)
```

```
[[[ 1  2  3]  
  [ 4  5  6]]
```

```
[[ 7  8  9]  
 [10 11 12]]]
```

Explore: hstack, vstack, dstack

Numpy

Splitting

<code>np.split()</code>	:	Split array into equal parts (requires size divisible)
<code>np.array_split()</code>	:	Split array into nearly equal parts (works for unequal sizes)
<code>np.hsplit()</code>	:	Split horizontally (columns for 2D arrays)
<code>np.vsplit()</code>	:	Split vertically (rows for 2D arrays)
<code>np.dsplit()</code>	:	Split along the third axis (depth for 3D arrays)

Numpy

np.split()

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

```
new_arr=np.split(arr,2)
print(new_arr)
```

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

np.array_split()

```
arr=np.array([1,2,3,4,5,6,7,8,9])
new_arr=np.array_split(arr,2)
print(new_arr)
```

```
[array([1, 2, 3, 4, 5]), array([6, 7, 8, 9])]
```

np.hsplit()

```
arr=np.array([[1,2,3,4],[5,6,7,8]])
new_arr=np.hsplit(arr,2)
print(new_arr)
```

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

DA-IICT

np.vsplit()

```
arr=np.array([[1,2,3,4],[5,6,7,8]])
new_arr=np.vsplit(arr,2)
print(new_arr)
```

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

np.dsplit()

```
arr=np.array([[[0,1],[2,3]],[[4,5],[6,7]],[[8,9],[10,11]])]
print(new_arr)
new_arr=np.dsplit(arr,2)
print(new_arr)
```

```
[[[ 0  1]
  [ 2  3]]
```

```
[[ 4  5]
 [ 6  7]]
```

```
[[ 8  9]
 [10 11]]]
[array([[[ 0],
         [ 2]],
```

```
[[ 4],
 [ 6]],
```

```
[[ 8],
 [10]]]), array([[[ 1],
 [ 3]],
```

```
[[ 5],
 [ 7]],
```

```
[[ 9],
 [11]])]
```

Numpy

Searching in Numpy Arrays

```
arr=np.array([[1,2,3,4],[5,6,7,8]])  
x=np.where(arr== 4)  
print(x)
```

(array([0]), array([3]))

```
arr=np.array([[1,2,3,4],[5,4,7,8]])  
x=np.where(arr==4)  
print(x)
```

(array([0, 1]), array([3, 1]))

```
arr=np.array([[1,2,3,4],[5,6,7,8]])  
x=np.where((arr==4) | (arr==7))  
print(x)
```

(array([0, 1]), array([3, 2]))

Numpy

Sorting

```
arr=[3,6,4,1,7,9,8,3]
print(np.sort(arr))    #ascending order

[1 3 3 4 6 7 8 9]
```

```
arr_sort=np.sort(arr)
print(arr_sort[::-1]) # reverse order

[9 8 7 6 4 3 3 1]
```

```
print(np.flip(arr)) # flip

[3 8 9 7 1 4 6 3]
```

```
arr=np.array(["cat", "bananan", "dog", "apple"])
print(np.sort(arr))

['apple' 'bananan' 'cat' 'dog']
```

```
: arr=np.array(["cat", "bananan", "Dog", "apple"])
print(np.sort(arr))          # Case senitive

['Dog' 'apple' 'bananan' 'cat']
```

Numpy

Arithmetic Operation with Numpy

```
arr_1=np.array([1,2,3,4,5,6,7])  
arr_2= np.array([8,9,10,11,12,13,14])
```

```
arr_1
```

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

```
arr_2
```

```
array([ 8,  9, 10, 11, 12, 13, 14])
```

```
new_arr=np.add(arr_1,arr_2)  
print(new_arr)
```

```
[ 9 11 13 15 17 19 21]
```

```
new_arr=np.subtract(arr_2,arr_1)  
print(new_arr)
```

```
[7 7 7 7 7 7 7]
```

```
new_arr=np.multiply(arr_2,arr_1)  
print(new_arr)
```

```
[ 8 18 30 44 60 78 98]
```

```
new_arr=np.divide(arr_2,arr_1)  
print(new_arr)
```

```
[8.      4.5      3.3333333 2.75      2.4      2.1666667  
 2.      ]
```

```
new_arr=np.power(arr_2,arr_1)  
print(new_arr)
```

```
[      8      81     1000    14641    248832    4826809   105413504]
```

```
new_arr=np.power(arr_2,2)  
print(new_arr)
```

```
[ 64  81 100 121 144 169 196]
```

Numpy

Arithmetic Operation with Numpy

```
new_arr=np.mod(arr_2,arr_1)
print(new_arr)
```

```
[0 1 1 3 2 1 0]
```

```
new_arr=np.reminder(arr_2,arr_1)
print(new_arr)
```

```
[0 1 1 3 2 1 0]
```

```
arr=np.array([-1,4,-5,7,-9])
print(np.absolute(arr))
```

```
[1 4 5 7 9]
```

```
new_arr=np.divide(arr_2,arr_1)           #Rounding
new_arr_1=np.around(new_arr,2)
print(new_arr,new_arr_1)
```

```
[8.          4.5          3.33333333  2.75          2.4          2.16666667
 2.          ] [8.          4.5          3.33  2.75  2.4          2.17  2.          ]
```

```
arr_1=np.array([1,2,3,4,5,6,7])
arr_2= np.array([8,9,10,11,12,13,14])

print(np.sum([arr_1]))
```

```
28
```

```
print(np.sum([arr_1,arr_2]))    #Sum
```

```
105
```

```
print(np.prod([arr_1]))    # product
```

```
5040
```

and many more ...