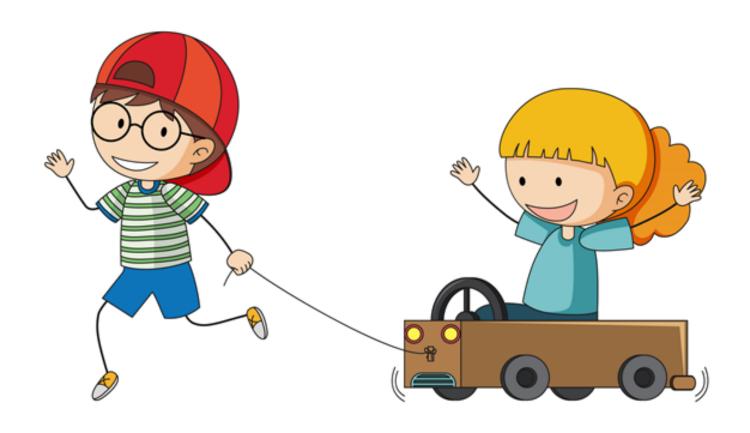
https://graphicmama.com/blog/free-school-clipart-for-your-education-projects/



- October 2019
- Dr. Marco A Arocha

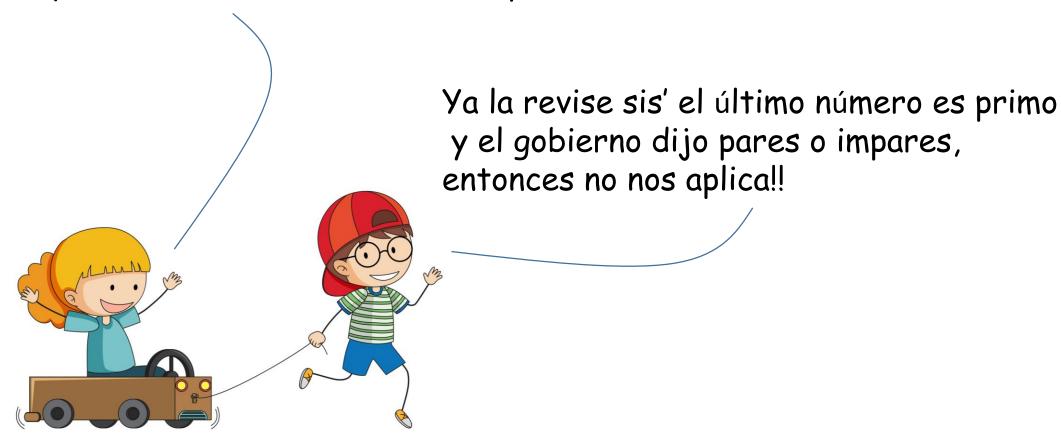
L#12 Arrays [2D]

https://docs.python.org/2.5/lib/typesseq.html



¿ Tu crees que podremos salir el domingo a jugar ? ¿ Nos dejara la orden ejecutiva del gobierno ?

Bro' chequéate el número de la tablilla, por favor



Creating 2D Arrays

Statement:

lista =[[5,3,9,7],[7,9,3,5]]

datos = np.array(lista)

easier:

datos=np.array([[5,3,9,7],[7,9,3,5]])

tupla=(2,4) # (row,col)

unos=np.ones(tupla)

easier: unos = np.ones((2,4))

ceros =np.zeros((3,3))







5 3 9 7

7 9 3 5

datos

1 1 1

1 1 1 1

0 0 0

0 0 0

0 0 0

unos

ceros



Indexing: Lists versus Arrays

```
# L is a list; x is an array
L= [ [1,2,3], [4,5,6], [7,8,9], [10,11,12] ]
x=np.array([[1,2,3], [4,5,6], [7,8,9], [10,11,12]])
```

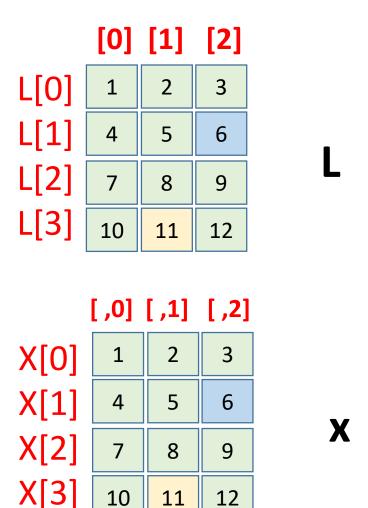
```
# Reference element valued-11:
```

```
L[3][1]
         # L[row][element within a row]
x[3,1]
         # x[row, col]
```

```
# Reference element valued-6:
```

```
L[1][2]
```

x[1,2]



10

11

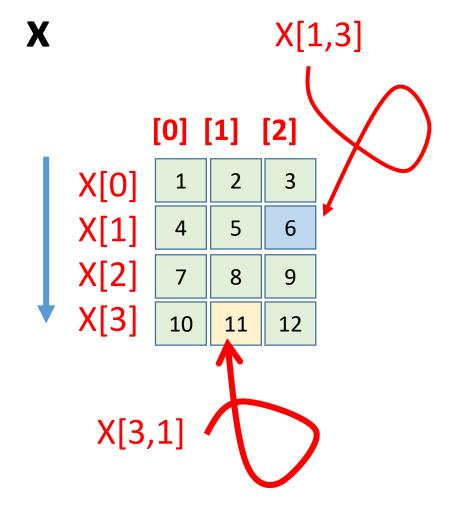
12

Array construction and storage is row-by-row

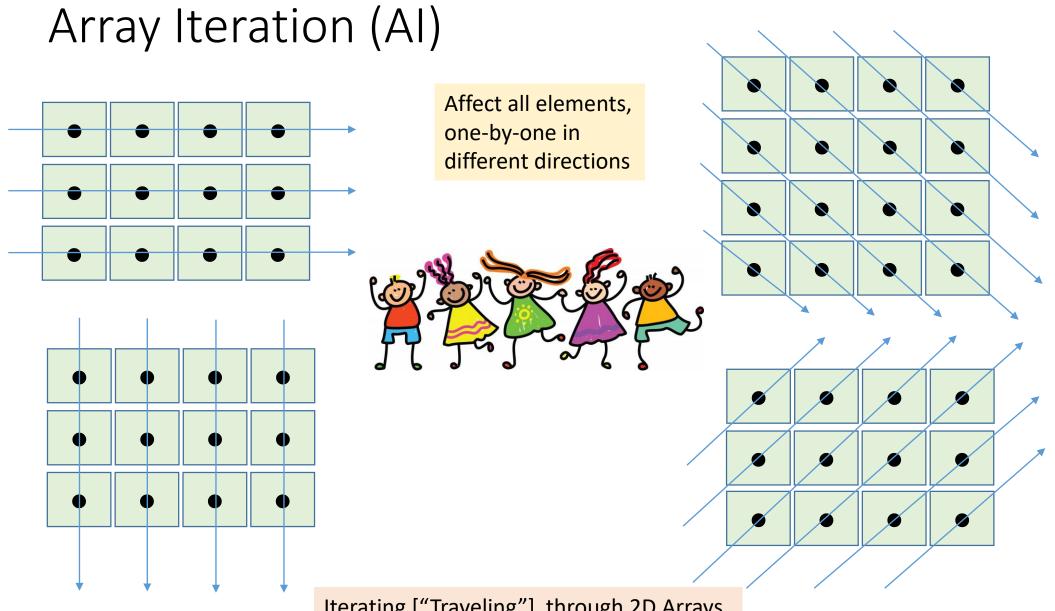


Arrays simple operations with single elements (scalar)

```
# x is an array
x=np.array([[1,2,3], [4,5,6], [7,8,9], [10,11,12]])
# Print element valued-11:
print(x[3,1])
                                      # 11
# Multiply 6 and 11:
print(x[1,2]*x[3,1])
                                      # 6*11=66
# Compute sine of element valued-6:
print(np.sin(x[1,2]))
                              # sin(6)=0.104528
```



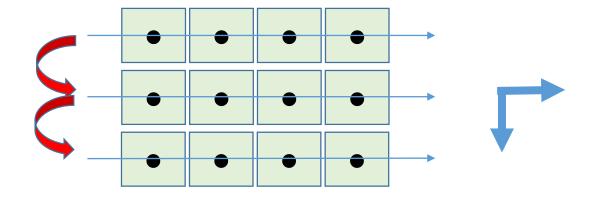


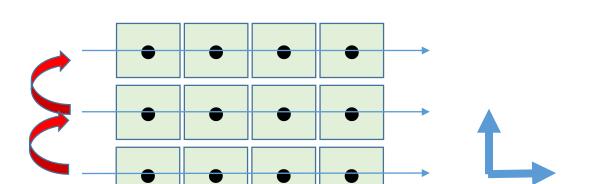


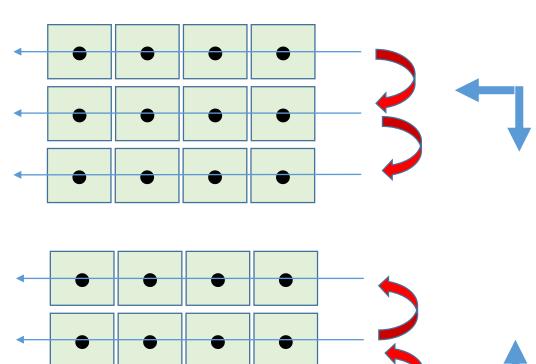
Array Iteration (AI)

Traveling row-wise





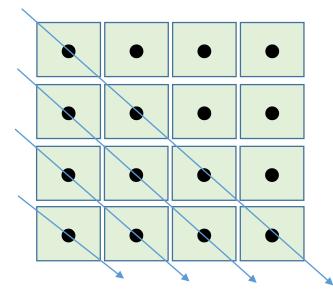


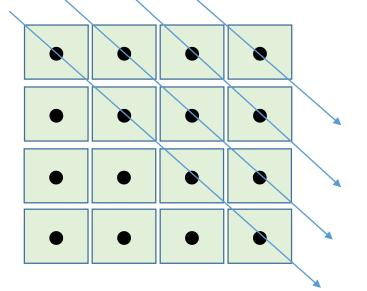


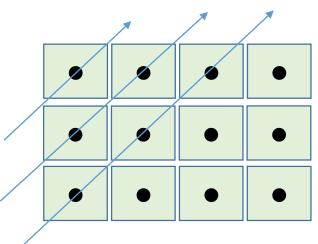
Array Iteration

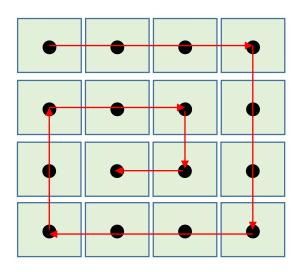
Affect partial elements

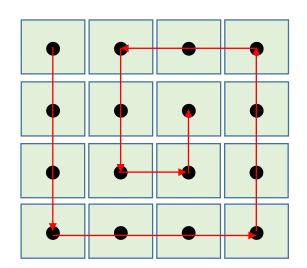


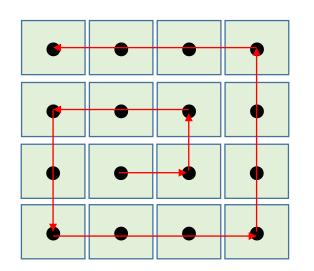












Al: row-wise



1	2	3	4
5	6	7	8
9	10	11	12

X

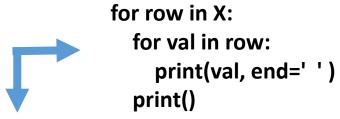
]]	1	2	3	4]
[5	6	7	8]
[9	10	11	12]]

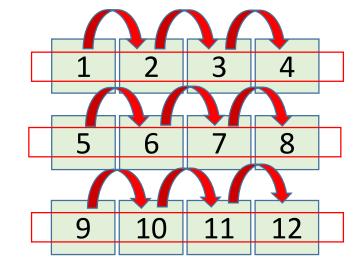
for row in X: print(row)

1	2	3	4	
5	6	7	8	
9	10	11	12	

[1 2 3 4] [5 6 7 8] [9 10 11 12]

FILE: iterationArrays01.py





1 2 3 4 5 6 7 8 9 10 11 12

Al: row-wise processing

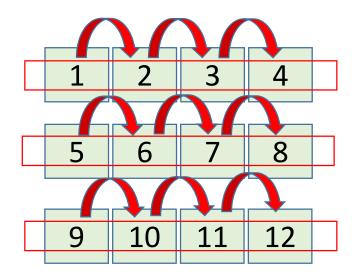


```
for row in X:

for val in row:

print(val, end=' ')

print()
```



```
1 2 3 4
5 6 7 8
9 10 11 12
```

```
# Sum over each element:
S=0
for row in X:
    for val in row:
        S=S+val
print("Sum is",S) # Sum is 78
```

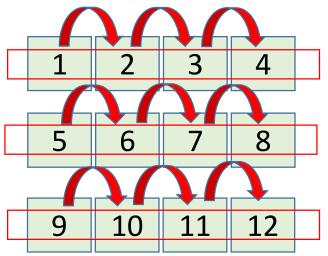
```
# Even elements:
even=np.array([])
for row in X:
    for val in row:
        if val%2==0:
             even=np.append(even,val)
print("Even are",even)
# Even are [ 2.  4.  6.  8.  10.  12.]
```

FILE: iterationArrays01.py

Al, row-wise: values vs indexes

FILE: iterationArrays01.py

```
# By values:
for row in X:
    for val in row:
        print(val, end=' ')
    print()
```



```
# By indexes:
m,n = np.shape(X)
for i in range(m):
    for j in range(n):
        print(X[i,j],end=',')
    print()
```

```
# traveling row-by-row by indexes and List Comprehension
m,n=np.shape(X)

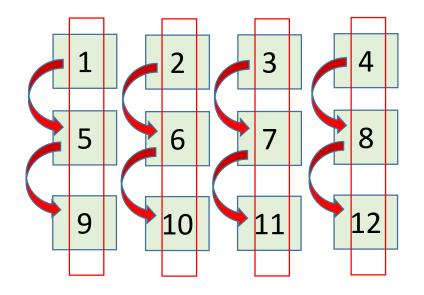
XXX=np.array([[X[i,j] for j in range(n)] for i in range(m)])
print(XXX)
```

Al, column-wise: values vs indexes



```
# By values:
for col in X:
    for val in col:
        print(val, end=',')
    print()
```

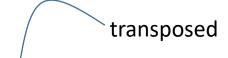
From previous solution row-wise solution, we switch the nested loops

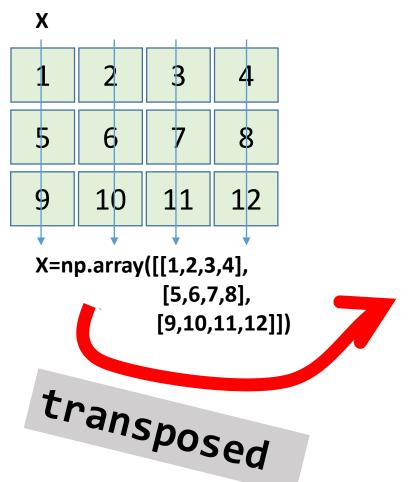


```
# By indexes:
m,n=np.shape(X)
for j in range(n):
    for i in range(m):
        print(X[i,j],end=',')
    print()
```

```
1,5,9,
2,6,10,
3,7,11,
4,8,12,
```

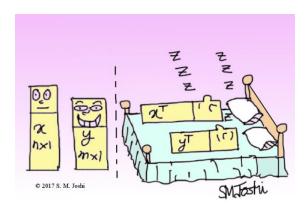
Al: column-wise





for col in X.T:
 print(col)

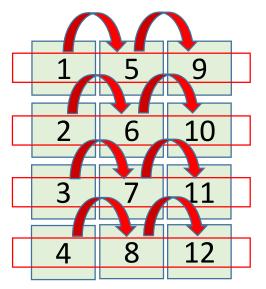
1	5	9	
2	6	10	
3	7	11	
4	8	12	



x and y transpose themselves to get some z's



for col in X.T:
 for val in col:
 print(val, end=',')
 print()



Al: By List Comprehension



FILE: Arrays08.py

```
import numpy as np
     X=np.array([[1,2,3,4],[5,6,7,8],[9,10,11,12]])
     print(X)
10
11
     # traveling row-by-row by values
12
     XX=np.array([[val for val in x_row] for x_row in X])
13
     print(XX)
14
15
     # traveling row-by-row by indexes
16
     m,n=np.shape(X)
     XXX=np.array([[X[i,j] for j in range(n)] for i in range(m)])
18
     print(XXX)
19
20
     # traveling column by column by indexes
     XXXX=np.array([[X[i,j] for i in range(m)] for j in range(n)])
22
     print(XXXX)
     print(XXXX.T)
23
24
25
     # traveling column by column by indexes
     XXXXX=np.array([[X[i,j] for i in range(m)] for j in range(n)]).T
26
     print(XXXXX)
```

```
[[1234]
[5678]
[ 9 10 11 12]]
```

```
[[1234]
13
    [5678]
     [ 9 10 11 12]]
```

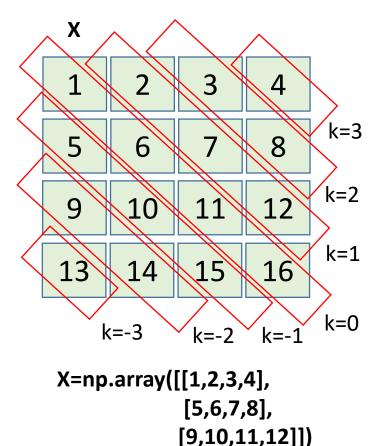
```
[[1234]
18
    [5678]
    [ 9 10 11 12]]
```

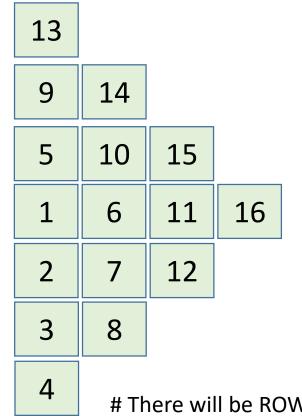
```
[[159]
[2610]
[3711]
[4812]]
```

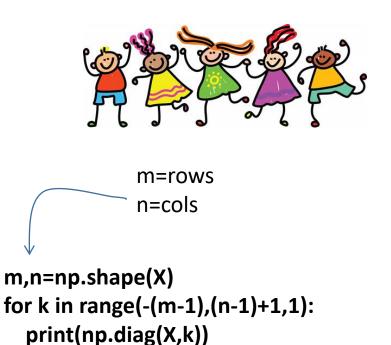
```
[[1234]
23
    [5678]
    [ 9 10 11 12]]
```

```
[[1234]
27
    [5678]
    [9 10 11 12]
```

Al: diagonal wise







There will be ROW+COL-1 lines in the output. i.e., k values 4 rows + 4 columns-1 = 7 (k values)

https://www.geeksforgeeks.org/zigzag-or-diagonal-traversal-of-matrix/

FILE: iterationArrays04.py

- # Data for next slide:
- x=np.array([[1,2,3], [4,5,6], [7,8,9], [10,11,12]])



1	2	3
4	5	6
7	8	9
10	11	12

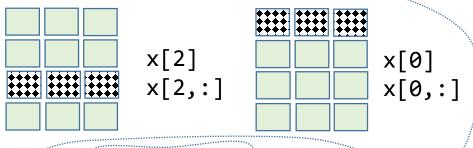
X

Cartoon from: https://es.vecteezy.com/arte-vectorial/360345-conjunto-de-ninos-del-doodle

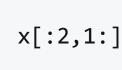
Subarrays

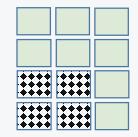




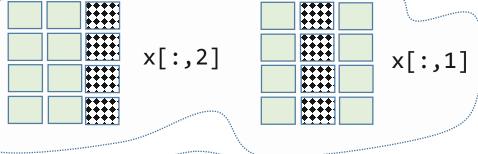


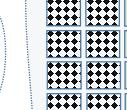


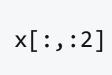


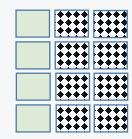


One column

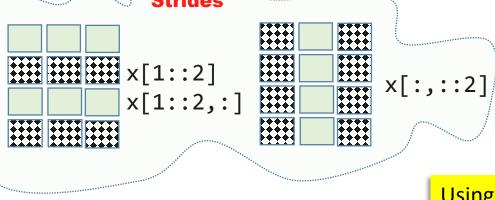




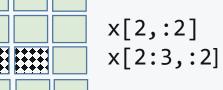


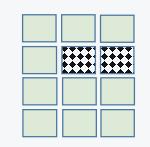


Strides







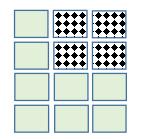


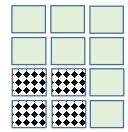
Subarrays





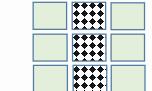




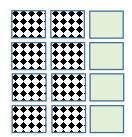


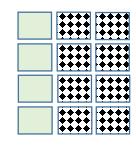
One column



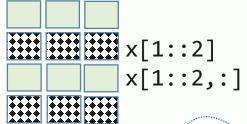


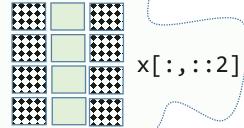


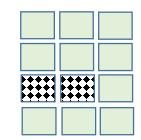


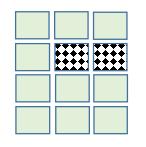


Strides



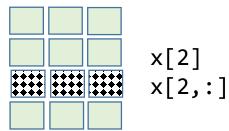


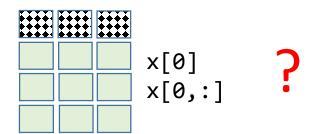


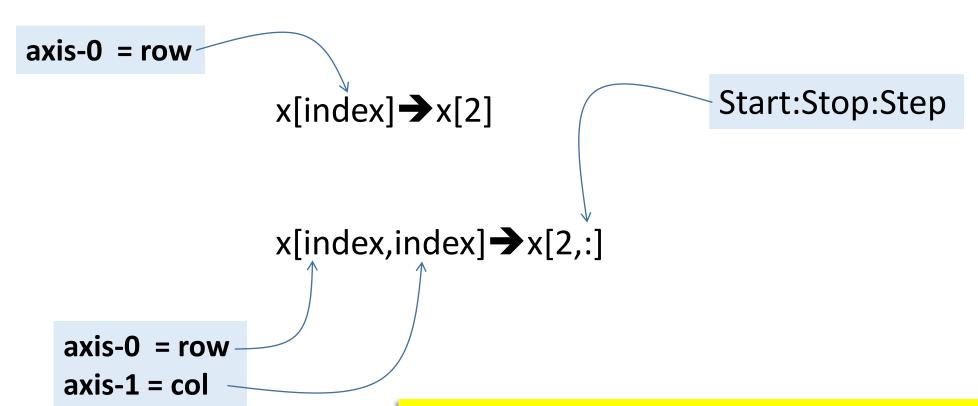






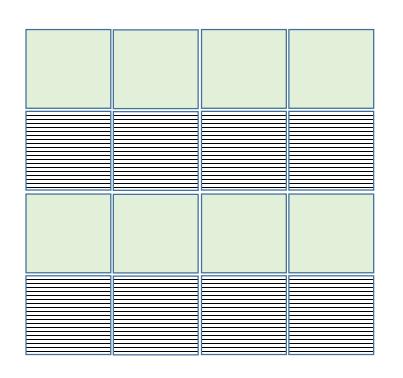


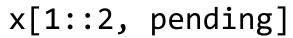


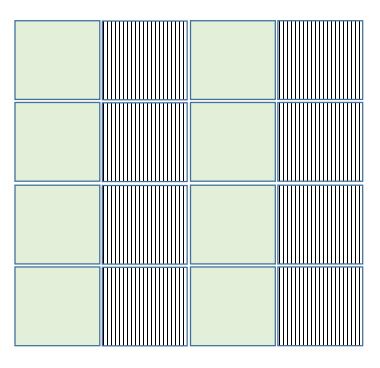


Indexing & Slicing: Demystified

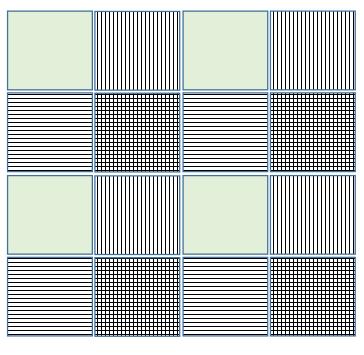




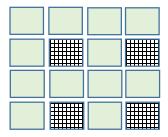




x[pending, 1::2]



x[1::2,1::2]



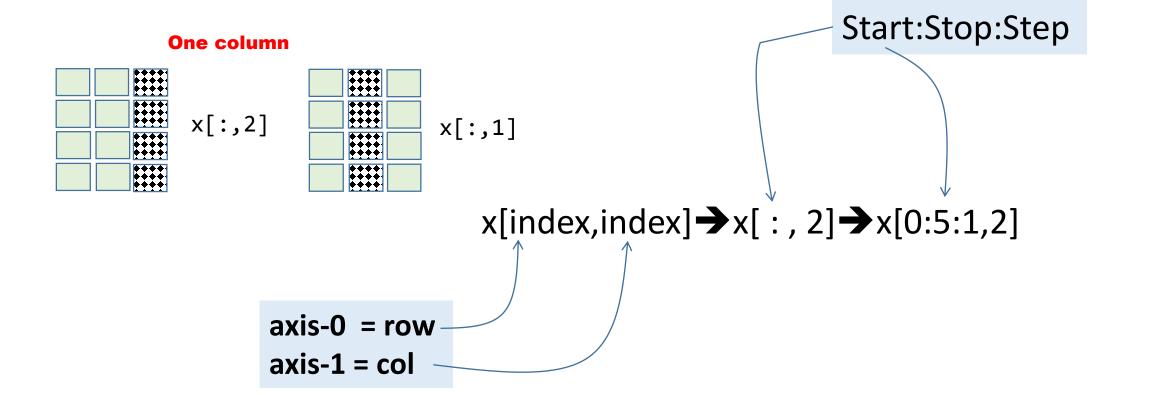
Using the data in previous slide, what is the output?

Processing a Slice

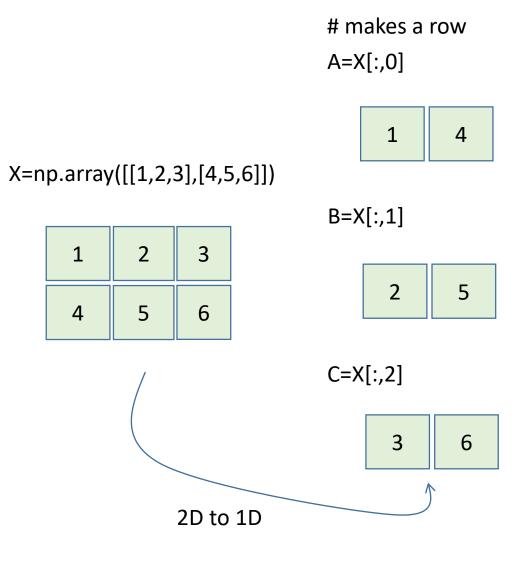
1	3	13	7	9	11	5
2	4	14	8	10	12	6
7	9	19	13	15	17	11
4	6	16	10	12	14	8
5	7	17	11	13	15	9
6	8	18	12	14	16	10
3	5	15	9	11	13	7
8	10	20	14	16	18	12

Given X with all elements, sum all orange elements:





Array Slicing and Gluing pieces back



Slicing columnwise makes the original slice loose its shape



Array Slicing and Gluing pieces back

Slicing columnwise makes the original slice loose its shape

makes a row

A=X[:,0]

1 4

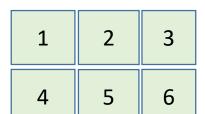
makes a column vector AA=A[:,np.newaxis]

1

4



X=np.array([[1,2,3],[4,5,6]])



B=X[:,1]

2 5

BB=B[:,np.newaxis]

2

5

s]

[[1] [4]] BB=

AA=

BB= [[2] [5]]

CC=

[[3]

3

4 5 6

YY=np.concatenate((AA,BB,CC),axis=1)

C=X[:,2]

3 6

3

6

CC=C[:,np.newaxis]

[6]]

2D to 2D

2D to 1D

1D to 2D

Array Slicing and Gluing pieces back

Slicing columnwise makes the original slice loose its shape FILE: ArraySlicing01.py

```
import numpy as np; X=np.array([[1,2,3],[4,5,6]])
print("X=\n",X)
A=X[:,0] # makes a row
AA=A[:,np.newaxis] # makes a column
B=X[:,1] # makes a row
BB=B[:,np.newaxis] # makes a column
C=X[:,2] # makes a row
CC=C[:,np.newaxis] # makes a column
print("AA=\n",AA); print("BB=\n",BB); print("CC=\n",CC)
print("Concatenate them along axis=1:")
YY=np.concatenate((AA,BB,CC),axis=1); print(YY)
print("Concatenate them along axis=0:")
YY=np.concatenate((AA,BB,CC),axis=0); print(YY)
```

```
X=
 [[1 2 3]
 [4 5 6]]
AA=
 [[1]
BB=
 [[2]
 [5]]
CC=
 [[3]
 [6]]
Concatenate them along axis=1:
Concatenate them along axis=0:
[[1]
 [5]
 [3]
 [6]]
```

Delete parts of an array

0/1/None/blank

FILE: deleteArray01.py

The axis along which to delete the subarray defined by indexes. If axis is None, indexes are applied to the flattened array.

Possibilities:

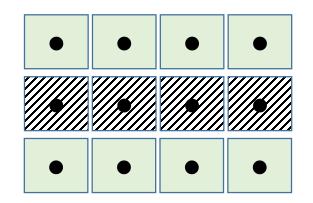
Syntax

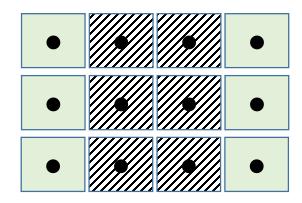
numpy.delete(arr, indexes, axis=None)

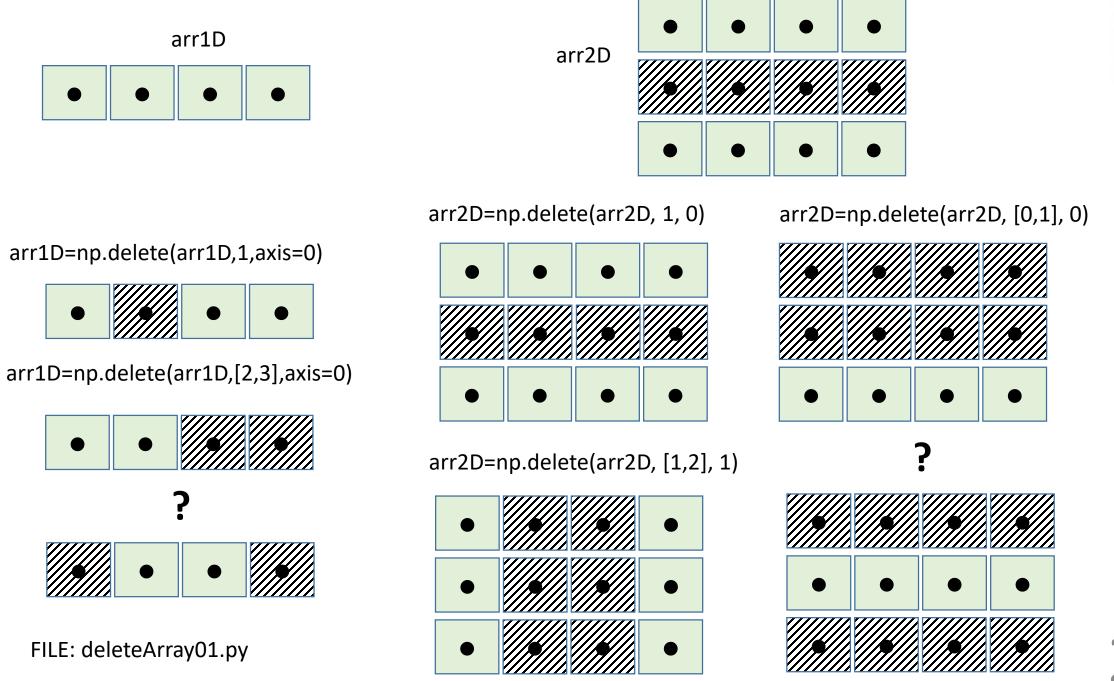
One index or array of indexes along axis to be deleted











More on Delete: Questions?



OUTPUT

```
[[1, 2, 3, 4],
arr = np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]])
                                                                [5, 6, 7, 8],
print(arr)
                                                                [ 9, 10, 11, 12]]
print(np.delete(arr, np.s [::2], 1))
#delete columns with indexes: 0,2 (step-2)
                                                               [[2, 4],
                                                                 [6, 8],
arr = np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]])
                                                                 [10, 12]
print(np.delete(arr, [1,3,5], None))
#delete indexes: 1,3,5 in a flattened array
                                                               [1, 3, 5, 7, 8, 9, 10, 11, 12]
```

https://docs.scipy.org/doc/numpy/reference/generated/numpy.s_.html

numpy.s_[::2] # a way to build index tuples. In this example indexes are: 0,2

Copy and Deep Copy: Scalar example



Scalars	Output
# Example 1	
a=5	
b=a	
b=b+1	
print("a = ", a)	a= 5
print("b = ", b)	b= 6
# Example 2	
a=5	
b=a	
a=a+1	
print("a=", a)	a= 6
print("b=", b)	b= 5

Nothing to worry about with scalars. Copying happens as expected. This is hard copying.

b is modified but a is not

a is modified but b is not

Copy and Deep Copy: Array case



Assuming A and B are arrays, we can have to types of copying:

Shallow copy

B = A

Deep copy

B = np.copy(A)

Copy and Deep Copy. File: copyDeepCopy.py



Arrays	Output	
# Arrays	A = [[1 2]	
A=np.array([[1,2],[3,4]]) print("A={}".format(A))	[3 4]]	
D-A # Copy A	B = [[5 2]	
B=A # Copy A B[0,0]=5	[3 4]]	
print("B={}".format(B))	Λ = [[E 2]	
print("A={}".format(A))	A = [[5 2] [3 4]]	
print("")		
A=np.array([[1,2],[3,4]])	B = [[5 2]	
B=np.copy(A) # Deep copy	[3 4]]	
B[0,0]=5		
print("B={}".format(B))	A = [[1 2]	
print("A={}".format(A))	[3 4]]	2