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In [4]: #IMPORTING THE numpy array and make them numpy array
import numpy as np
a = np.array([12,23,45])
b = np.array([11,56,78])
print(a)
print(b)

[12 23 45]
[11 56 78]

In [5]: #call the b array
b

Out[5]: array([11, 56, 78])

In [8]: #changing vlaue of the value of b array at the idex position 1
b[1]=19

In [9]: #then print the b array
print(b)

[11 19 78]

In [12]: #creat the new array with the with help of 1d array
c = np.array([[a],[b]])
print(c)

[[[12 23 45]]
 [[11 19 78]]]

In [14]: #change the index value of a 1-d array to the 23 at position 2
a[2]=23
print(a)

[12 23 23]

In [15]: print(c)

[[[12 23 45]]
 [[11 19 78]]]

In [16]: d = np.array([22,76,45])
e = np.array([59,21,98])
print(d)
print(e)

[22 76 45]
[59 21 98]

In [17]: d[2]=23
print(d)

[22 76 23]

In [35]: f = np.array([[d],[e]])
print(f)

[[[22 76 23]]
 [[59 21 98]]]

In [28]: #calling the 2nd row of the f array
f[1]

Out[28]: array([[59, 21, 98]])

In [29]: #calling the 1st row of the f array
f[0]

Out[29]: array([[22, 76, 23]])

In [31]: #calling the element of a 1-d array from the elemenet 1st position to 2nd and here 2nd excuded
a[1:2]

Out[31]: array([23])

In [32]: #calling the element of a 1-d array from the elemenet 0th position to 2nd and here 2nd excuded
a[0:2]

Out[32]: array([12, 23])

In [40]: #calling the element of a 1-d array from the elemenet 0th position to last position of the a array
a[0:]
a[0:3]

Out[40]: array([12, 23, 23])

In [47]: #this is all about the masking in numpy arrays it will gives us by default arrays.
print('numpy array initialization')

numpy array initialization

In [49]: #here we paassed the tuple in which we have to speeify the row and column(3,4)
j = np.zeros((3,4))
print(j)

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

In [51]: #here we have to use the arange in that we have to pass the (start,end,interval)
k = np.arange(20,25,1)
print(k)

[20 21 22 23 24]

In [55]: #here we have to use the linspace in that paassing argument (start,stop,how much points in that like below)
l = np.linspace(1,5,5)
print(l)

[1.  2.  3.  4.  5.]

In [58]: #if we want the same number in all row and column then we need to use the full with the arument i.e.((row,column),value)
m = np.full((3,3),5)
print(m)

[[5 5 5]
 [5 5 5]
 [5 5 5]]

In [59]: #this will print any random values in the row and column(row,column)
n = np.random.random((3,4))
print(n)

[[0.01895143 0.46790746 0.29178308 0.89767245]
 [0.56564337 0.72539292 0.18306492 0.26201124]
 [0.45594671 0.57263548 0.80163887 0.35893378]]

In [64]: #how can we access the array menas changes be done in the array according to the command
#we got the shape of the array by using the method that is shape which can help to get tuple (row,column) form
o = np.array([[23,76,90],[56,75,78]])
print(o)
print(o.shape)

[[23 76 90]
 [56 75 78]]
(2, 3)

In [65]: #use of the shape function
#we change the tuple that we got by using shape function last time
#we can access the individual element of the tuple.

In [67]: o.shape = (3,2)

In [71]: print(o)
print(o.shape[0])

[[23 76]
 [90 56]
 [75 78]]
3

In [74]: #size
#n.arange(number of points needed)
#will place the numbers zero to till that number
#but the size will count the all the element which will come inbetween 0 to 10 that how it works
p = np.arange(10)
print(p)
print(p.size)

[0 1 2 3 4 5 6 7 8 9]
10

In [75]: #ndim
#ndim returns of the dimnesions of the array it could be 1d,2d or upto nd..
print(o.ndim)

2

In [76]: #it will gives the what type of array it is because array should be homogeous not hetrogeous.
print(p.dtype)

int32

In [89]: #we can do the mathematical calculations with numpy arrays
#like np.subtract,np.mulipty and np.log,np.exp,np.divide,np.sqrt and so on..
q = np.array([5,10])
np.sum(q)
print(np.sum(q))
r = np.array([3,7])
np.subtract(q,r)
print(np.subtract(q,r))
u = np.divide(q,r)
print(u)
r =np.multiply(q,r)
print(r)
#it works like the matrix operations

15
[2 3]
[1.66666667 1.42857143]
[15 70]

In [87]: #sum method with the axis
#in this kind of sum method axis taken zero afterwardds they sum with the corresponding element in the vertical
#for axis =1 voice vesa.
s = np.sum([[5,1],[6,4]], axis = 0)
print(s)
t = np.sum([[9,6],[23,4]], axis = 1)
print(t)

[11  5]
[15 27]

In [106]: # equal function it works to compare the coresponding element of the array
np.equal(q,r)
print(np.equal(q,r))
# equal function it also works to compare the two array with the comparision of an element it will return in true or false
np.array_equal(q,r)
print(np.array_equal(q,r))

[False False]
False

In [121]: #aggregate function it always work on single array
v = np.array([23,56,98])
print(v)
print(np.min(v)) #min value
print(np.max(v)) #max value
print(np.mean(v)) #mean value
print(np.sum(v)) #sum of vaalue
print(np.std(v)) #standard deviation its nothing but how much the value is differ from the mean value
print(np.median(v)) #median
print(np.corrcoef(v)) #ML term

[23 56 98]
23
98
59.0
177
30.692018506445613
56.0
1.0

In [131]: #Concept of barocasting
#it can happen with any operation here we taken addition for understading the concept. of it.
#when the one array is three dimensional is added or subtracted with the one dimensional array that time its()
#to get three dimensional array after that it gets merge
#2d array manage itself with the dimensional of the first array then only the addition or subtraction can done
#manage the first array.

w = np.array([1,2,3],[4,4,5])
wa = np.array([3,4,5])
print(np.sum([w,wa]))

[[ 4  6  8]
 [ 7  8 10]]

In [132]: #indexing & slicing in the numpy array
#here we having the three dimensional array in we want to solice or index it
i=np.array([[23,76,98],[45,91,34],[12,32,93]])
print(i)

[[23 76 98]
 [45 91 34]
 [12 32 93]]

In [139]: #in the 3-d np array here we are calling the 1st row of the array
print(i[0])

[23 76 98]

In [24]: #all_based_on_zero_based_indexing
#when we call for the 2nd dimensional array that time de use the [3,4] But here we having the three dimensional array
#here we want to sollice like [:0, :]
#in that first use for the row and column similarly for the three dimensional array
#here,we having more row and column so three need of indexing themself is necessary so that is we have taken (:)
import numpy as np
i=np.array([[23,76,98],[45,91,34],[12,32,93]])
print(i)
#first part of [ : , : ] work for the row 2nd for the column.
print(i[1,:])
#if we write it down as below for column then we wiling get the 1 column to 0th column and here 1st column is excluded
#so we are getting the zeroth element of zeroth column in the zeroth row that we called first argument.
print(i[:,1])
#similarly in that we are getting the 2nd and first element of column in thhe zeoroth row
print(i[:,1:1])
#it means having the all rows but the column available from then 1 to zeroth here 1st coumn is included.and removed the 1st column
print(i[:, 1 : 1])
#here we wanting the 2nd row of the 3x3 array and we wanted the column upto the 3rd element according to indexing
print(i[2,:])
#we want till the 2nd row that why first two row(0,1) come into picture and we had also mention that we want till the 2nd row
print(i[ : 2,:])
#here we wanted the 2nd row and 2nd column upto the end
print(i[2, : 2 :])
#here we will get the entire array
print(i[0, :0:])
#here we wanted the 1st row and coulumn 1st upto the end
print(i[1:,1:])
#here we wanted the 2nd row and 1st collumn till the end
print(i[2:,1:])
#here we wanted the row from the 1st to zeroth and 2 nd will be excluded and 1st column will be excluded.
print(i[:2,:])

[[23 76 98]
 [45 91 34]
 [12 32 93]]
[[23 76 98]]
[[23]]
[[76 98]]
[[91 34]]
[[76 98]
 [91 34]
 [32 93]]
[[12 32 93]]
[[98]]
[[34]]
[[93]]
[[23 76 98]
 [45 91 34]
 [12 32 93]]
[[91 34]
 [32 93]]
[[91 34]
 [32 93]]
[[12 32 93]]
[[23]]
[[45]]

In [34]: #array manipulation
#this the way we can can make the horizontally concatenation with the two arrays.
aa = np.array([23,98,76])
bb =np.array([45,67,23])
ss =np.concatenate([aa,bb],axis=0)
print(ss)
#if we want the veritically(column wise) concatenation that time we have two multidimensional array at condition
#column wise stacking
aal=np.array([[22,78,56],[28,67,45]])
bb1=np.array([[21,98,90],[29,90,78]])
sbl = np.concatenate([aal,bb1],axis=1)
print(ss1)
#if we want the horizontally concatenation means row wise that time we need to use the axis=0
#row wise stacking
ss2 = np.concatenate([aal,bb1],axis=0)
print(ss2)

[23 98 76 45 67 23]
[22 78 56 21 98 90]
[28 67 45 29 90 78]
[[22 78 56]
 [28 67 45]
 [21 98 90]
 [29 90 78]]

In [41]: #v-stack (vertical stacking) and h-stack (horizontal stacking) in numpy array
#stacking- same thing can be concatenated with the small difference.
#in stacking we required 1-d array
cc = np.array([23,56,78])
dd = np.array([32,87,55])
print(dd)
#in that we are stacking the two rows of the arrays
ee =np.stack((cc,dd),axis =0)
print(ee)
#similarly, here we are stacking the two column of the arrays
ff = np.stack((cc,dd), axis=1)
print(ff)

[23 56 78]
[32 87 55]
[[23 56 78]
 [32 87 55]]
[[23 32]
 [56 87]
 [78 55]]

In [51]: #v-stack (vertical stacking) and h-stack (horizontal stacking) in numpy array
s1 =np.array([[23,87,90],[20,30,56]])
s2 =np.array([[33,67,90],[12,34,89]])
print('stack of a nd b')
print(np.stack((s1,s2)))
print('marix a')
print(s1)
print('matrix b')
print(s2)

stack of a nd b
[[[23 87 90]
 [20 30 56]]

 [[33 67 90]
 [12 34 89]]]
marix a
[[23 87 90]
 [20 30 56]]
matrix b
[[33 67 90]
 [12 34 89]]

In [59]: #horizontal stacking
print('horizontal staking')
print(np.hstack((s1,s2)))
print('horizontal concatenate')
print(np.concatenate((s1,s2),axis=0))
print('vertical stack')
print(np.vstack((s1,s2)))
print('vertical concatenate')
print(np.concatenate((s1,s2),axis=1))
print('column stack') #not used generally in data science
print(np.column_stack((s1,s2)))

horizontal staking
[[23 87 90 33 67 90]
 [20 30 56 12 34 89]]
horizontal concatenate
[[23 87 90]
 [20 30 56]
 [33 67 90]
 [12 34 89]]
vertical stack
[[23 87 90]
 [20 30 56]
 [33 67 90]
 [12 34 89]]
vertical concatenate
[[23 87 90 33 67 90]
 [20 30 56 12 34 89]]
column stack
[[23 87 90 33 67 90]
 [20 30 56 12 34 89]]

In [92]: #split function
#syntax
#np.split(array,index,axis) here array is any numpy array ,index =int or list
print(s1)
print('split appy on it')#we have pass the argument according to requirement
print('this is about the row based')
print(np.split(s1,2,axis=0))
print('this is about the column based')
print(np.split(s1,3,axis=1))

[[23 87 90]
 [20 30 56]]
split appy on it
this is about the row based
array([[23, 87, 90]], array([[20, 30, 56]]))
this is about the column based
[array([[23],
       [87]], array([[87],
       [30]]), array([[90],
       [56]])]

In [87]: print('if we use the list instaed of integer then let's see")
print('this will be for the row wise')
print(np.split(s1,[2,1],axis=0))

print('lets ccheck the the spliting of the each array')

if we use the list instaed of integer then let's see
[array([[23, 87, 90]], array([[20, 30, 56]]), array([], shape=(0, 3), dtype=int32), array([[20, 30, 56]])])
lets ccheck the the spliting of the each array

In [90]: print('part 1')
print(s1[:1])

part 1
[[23 87 90]]

In [114]: print('part 2')
print(s1[1:2])

part 2
[[20 30 56]]

In [88]: print('part 3')
print(s1[2:])

part 3
[]

In [135]: j =np.array([[23,21,76],[98,67,54]])
print(j)
print('for the column wise we need to take the axis= 1')
k =np.split(j,[1,2],axis=1)
print(k)
print('how the spliting works lets see'),
print('here everything done for the column because its column wise spliting')

[[23 21 76]
 [98 67 54]]
for the column wise we need to take the axis= 1
[array([[23],
       [67]]), array([[21],
       [67]]), array([[76],
       [54]])]
how the spliting works lets see
here everything done for the column because its column wise spliting

In [134]: j[:,1:]

Out[134]: array([[23],
 [98]])

In [136]: j[:,1:2]

Out[136]: array([[21],
 [67]])

In [137]: j[:,2:]

Out[137]: array([[76],
 [54]])

In [ ]: 
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