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In [1]: #This is a supplementary material to the lecture "2D Lists and Numpy" to quickly revise, whenever need
In [2]: #2D list
        #creating a 2D list
        #if we were to create an n sized 1d list with all zeros, then
        n = 5
        1 1d = [0]*n
        #or alternatively
        1_1d = [0 for i in range(n)]
        print(l 1d)
        #2D list can be thought of as collection of multiple 1D lists
        #for example, if we want to create a 2D list with say 3 rows and 5 columns with all zeros
        #it can be thought of as collection of 3 nos. 1D Lists, each containing 5 items
        n rows = 3
        n cols = 5
        1_2d = [0]*n_cols]*n_rows
        print(1_2d)
        [0, 0, 0, 0, 0]
        [[0, 0, 0, 0, 0], [0, 0, 0, 0, 0], [0, 0, 0, 0, 0]]
In [3]: #alternatively, same 2D list can be created as
        1_2d = [[0 for j in range(n_cols)] for i in range(n_rows)]
        1 2d
Out[3]: [[0, 0, 0, 0, 0], [0, 0, 0, 0], [0, 0, 0, 0]]
In [4]: #taking 2d list as input from user
        #say user gives two lines of input, first line contains #rows and #cols separated by space
        #next line contains all the elements of 2D list separated by space
        n_rows, n_cols = [int(item) for item in input().strip().split(' ')]
        elements = [int(item) for item in input().strip().split(' ')]
        list_2d = [[elements[i * n_cols + j] for j in range(n_cols)] for i in range(n_rows)]
        list 2d
        3 4
        1 2 3 4 5 6 7 8 9 10 11 12
Out[4]: [[1, 2, 3, 4], [5, 6, 7, 8], [9, 10, 11, 12]]
In [5]: ## accessing an element in 2d list
        list_2d[2][1]
Out[5]: 10
In [6]: #NumPy
        #NumPy is the fundamental package for scientific computing with Python.
        #to use numpy functions, we need to import the package
        import numpy as np
In [7]: #creating numpy array
        1 = [1, 2, 3, 4]
        np_arr = np.array(1)
        np_arr
Out[7]: array([1, 2, 3, 4])
In [8]: np_arr.dtype
Out[8]: dtype('int32')
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In [9]: 1 = [1, 2, '3']
                                    #if some of the items are strings in addition to int, it converts all item
         s to string
         np\_arr = np.array(1)
         np_arr
Out[9]: array(['1', '2', '3'], dtype='<U11')</pre>
In [10]: #to create numpy array of all zeros or ones
                                  #will create 1d numpy array of size 5 with all zeros
         arr zeros = np.zeros(5)
                                       #will create 1d numpy array of size 5 with all ones
         arr_ones = np.ones(5)
         print(arr_zeros)
         print(arr_ones)
         [0. 0. 0. 0. 0.]
         [1. 1. 1. 1. 1.]
In [11]: arr_zeros.dtype
Out[11]: dtype('float64')
In [12]: #if you want to create array of int type
         arr zeros = np.zeros(5, int)
         print(arr_zeros)
         arr_zeros.dtype
         [0 0 0 0 0]
Out[12]: dtype('int32')
In [13]: #creating multidimensional numpy arrays
         arr 2d = np.zeros((3,5)) #will create 2d numpy arrays of 3x5 conatining all zeros
         arr_2d
Out[13]: array([[0., 0., 0., 0., 0.],
                [0., 0., 0., 0., 0.]
                [0., 0., 0., 0., 0.]])
In [14]: arr_3d = np.ones((3, 3, 5))
                                           #think of it as 3 nos. of 2d numpy arrays of size 3x5 each containin
         g all ones
         arr_3d
Out[14]: array([[[1., 1., 1., 1., 1.],
                 [1., 1., 1., 1., 1.],
                 [1., 1., 1., 1., 1.]],
                [[1., 1., 1., 1., 1.],
                 [1., 1., 1., 1., 1.],
                 [1., 1., 1., 1., 1.]],
                [[1., 1., 1., 1., 1.],
                 [1., 1., 1., 1., 1.],
                 [1., 1., 1., 1., 1.]])
In [15]: arr_3d.shape
                        #to check the shape or dimensions of numpy array
Out[15]: (3, 3, 5)
In [16]: #Returns an array containing the same data with a new shape.
         arr_2d = arr_3d.reshape(9,5)
In [17]: #accessing the items and slicing work same as we have studied in lists
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In [18]: #one nice feature is that we can extarct sub-matrix/sub-array of the original matrix/array using slici
         arr = np.array([[1,2,3,4], [5,6,7,8], [9, 10, 11, 12]])
                                       #it will give a 2d array containing items from arr present in rows and
         arr_sub = arr[1:3, 1:3]
         cols with index 1 and 2
         arr_sub
Out[18]: array([[ 6, 7],
                [10, 11]])
In [19]: | #numpy operations
         arr_new = arr + 2
                                 #will add 2 to each item in arr
         print(arr new)
         print(arr new - arr)
                                 #will do element-wise subtraction of elements
         [[ 3 4 5 6]
          [7 8 9 10]
          [11 12 13 14]]
         [[2 2 2 2]
          [2 2 2 2]
          [2 2 2 2]]
                               #return sum of all elements in arr
In [20]: print(arr.sum())
                             #return mean of all elements in arr
         print(arr.mean())
         78
         6.5
In [21]: arr * arr_new
                             #will do element-wise multiplication of these matrices
Out[21]: array([[ 3, 8, 15, 24],
                [ 35, 48, 63, 80],
                [ 99, 120, 143, 168]])
In [22]: #if, you want to compute the dot product of two numpy arrays, provided dimensions are compatible
         a = np.array([[1,2], [3,4]])
         b = np.array([[5, 6], [7,8]])
         c = a.dot(b)
                            #will compute the dot product
         C
Out[22]: array([[19, 22],
                [43, 50]])
In [23]: #say, you want to calculate sum of first row of array c
         np.sum(c[0,:])
Out[23]: 41
In [24]: #similarly, for sum of elements in second column of c
         np.sum(c[:, 1])
Out[24]: 72
                                   #column-wise sum
In [25]: np.sum(c, axis = 0)
Out[25]: array([62, 72])
In [26]: np.sum(c, axis = 1)
                                  #row-wise sum
Out[26]: array([41, 93])
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In [27]:
         #Numpy also has lots of ways to create random number arrays:
          #Create an array of the given shape and populate it with random samples from a uniform distribution ov
          er [0, 1).
          np.random.rand(2)
Out[27]: array([0.46616062, 0.90643978])
In [28]: np.random.rand(5,5)
Out[28]: array([[0.46139565, 0.61125049, 0.58092469, 0.96934139, 0.64813732],
                 [0.42130718, 0.26109802, 0.30697161, 0.63601977, 0.57425207],
                 [0.96214112, 0.57649256, 0.73205837, 0.00952067, 0.80429736],
                 [0.42951367, 0.39888115, 0.72890438, 0.4917824, 0.51476326],
                 [0.8908243 , 0.38508019, 0.61038847, 0.40601544, 0.14917933]])
In [29]: #Return a sample (or samples) from the "standard normal" distribution. Unlike rand which is uniform:
          np.random.randn(2)
Out[29]: array([-1.39718995, 1.77372121])
In [30]: np.random.randn(5,5)
Out[30]: array([[-0.37923369, -0.56998801, 0.3712431, 0.78905378, 0.10938529],
                 \hbox{\tt [-0.12640059, -0.15656451, 0.05231141, 0.83405347, 0.68454559],}
                 [-0.15418843, 0.04013789, 0.61515418, 0.24499148, -0.34176554],
                 [-2.4499439 , -1.98456664 , -2.26153452 , 0.8622766 , 0.76724156], [-0.97318391 , -1.33416268 , 0.89568369 , 0.20433311 , 1.00688174]])
In [31]: #Return random integers from low (inclusive) to high (exclusive).
          np.random.randint(1,100)
Out[31]: 59
In [32]: #Following will randomly sample 10 values in the interval [1, 100)
          np.random.randint(1,100,10)
Out[32]: array([74, 3, 27, 54, 32, 87, 50, 93, 8, 48])
In [33]: #Let's discuss some useful attributes and methods or an array:
          arr = np.arange(25)
          ranarr = np.random.randint(0,50,10)
In [34]: arr
Out[34]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16,
                 17, 18, 19, 20, 21, 22, 23, 24])
In [35]: ranarr
Out[35]: array([38, 43, 39, 6, 36, 37, 18, 17, 6, 31])
In [36]: #Thanks, Happy Coding!
 In [ ]: #To download .ipynb notebook, right click the following link and click save as
          https://ninjasfiles.s3.amazonaws.com/0000000000003219.ipynb
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