

lab1-lab-2-Numpy

February 23, 2023

```
[41]: def fun():  
      L = []  
      for n in range(10000000):  
          L.append(n**2)  
      return L
```

```
[42]: %timeit fun()
```

2.41 s \pm 12.3 ms per loop (mean \pm std. dev. of 7 runs, 1 loop each)

```
[39]: # can we optimise the above code?  
def fun2():  
    return [n**2 for n in range(10000000)]
```

```
[40]: %timeit fun2()
```

2.18 s \pm 16.8 ms per loop (mean \pm std. dev. of 7 runs, 1 loop each)

```
[14]: # how to find out documentation of any inbuilt function  
      # np.append?
```

```
[43]: # find the error in the below code.  
def func1(a, b):  
    return a / b  
  
def func2(x):  
    a = x  
    b = x - 1  
    return func1(a, b)
```

```
[45]: # func2(1)
```

0.1 Numpy

```
[46]: import numpy as np
```

```
[47]: # import matplotlib.pyplot as plt
```

```
[48]: np.__version__
```

```
[48]: '1.21.6'
```

```
[51]: x = [1,2,4]
      print(type(x))
      x = np.array(x)
      print(type(x))
```

```
<class 'list'>
<class 'numpy.ndarray'>
```

```
[52]: # integer array:
      np.array([1, 4, 2, 5, 3])
```

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

```
[61]: x = np.random.rand(3,4)
      print(type(x))
      x = np.array(x)
      print("\n\n")
      print(type(x))
```

```
<class 'numpy.ndarray'>
```

```
<class 'numpy.ndarray'>
```

```
[68]: # nested lists result in multi-dimensional arrays
      np.array([range(i, i + 3) for i in [2, 4, 6]])
```

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

```
[77]: # Create a length-10 integer array filled with zeros
      np.zeros((3,5), dtype=int)
```

```
[77]: array([[0, 0, 0, 0, 0],
           [0, 0, 0, 0, 0],
           [0, 0, 0, 0, 0]])
```

```
[76]: # Create a 3x5 floating-point array filled with ones
      np.ones((3, 5), dtype=float)
```

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

```
[80]: # Create a 3x5 array filled with 3.14  
np.full((3, 5), 3.14)
```

```
[80]: array([[3.14, 3.14, 3.14, 3.14, 3.14],  
           [3.14, 3.14, 3.14, 3.14, 3.14],  
           [3.14, 3.14, 3.14, 3.14, 3.14]])
```

```
[88]: # Create an array filled with a linear sequence  
# Starting at 0, ending at 20, stepping by 2  
# (this is similar to the built-in range() function)  
np.arange(0, 20, 2)
```

```
[88]: array([ 0,  2,  4,  6,  8, 10, 12, 14, 16, 18])
```

```
[92]: # Create an array of five values evenly spaced between 0 and 1  
np.linspace(0, 1, 5)
```

```
[92]: array([0.   , 0.25, 0.5  , 0.75, 1.   ])
```

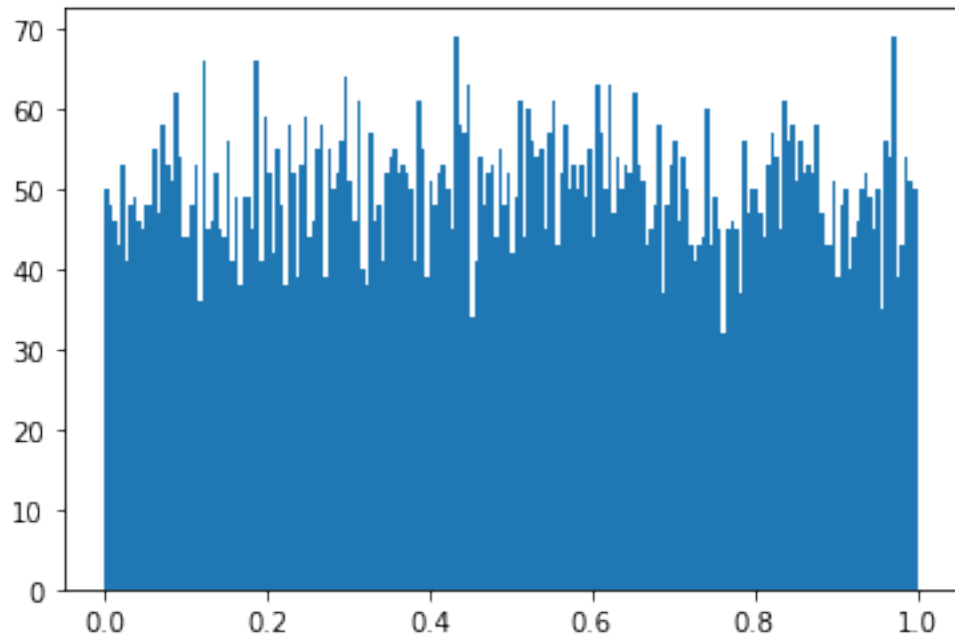
```
[31]: # Create a 3x3 array of uniformly distributed  
# random values between 0 and 1  
x = np.random.random((10000))
```

```
[99]: import matplotlib.pyplot as plt
```

```
[100]: x
```

```
[100]: array([4.87137687e-01, 4.57356103e-01, 3.83035179e-01, 8.13774995e-01,  
           1.79689897e-01, 4.45249867e-01, 4.28580573e-01, 3.66515480e-01,  
           6.94895119e-01, 3.64226033e-04])
```

```
[121]: plt.hist(x, bins=200);
```

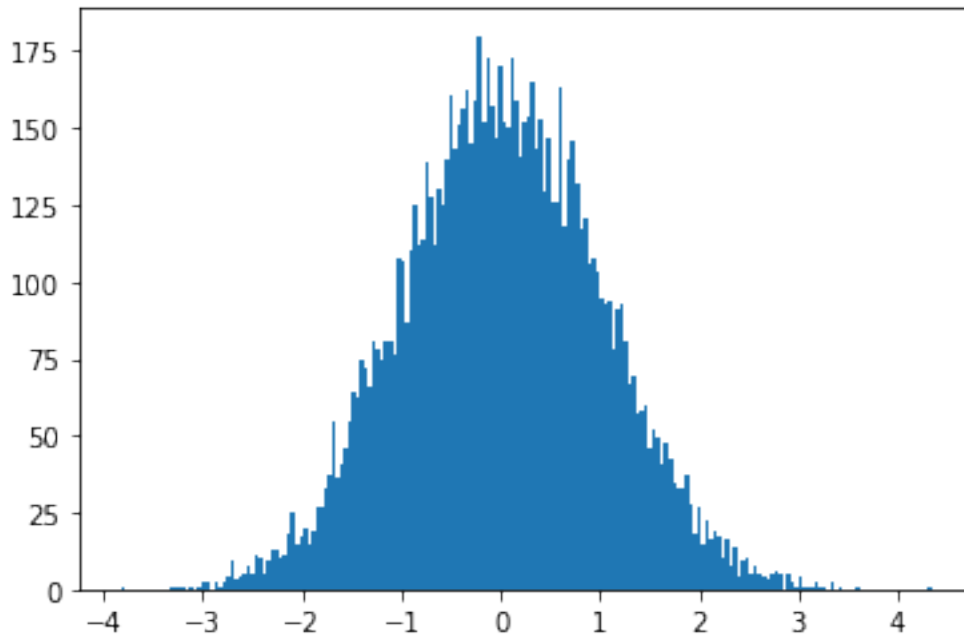


```
[30]: # Create a 3x3 array of normally distributed random values  
# with mean 0 and standard deviation 1  
np.random.normal(0, 1, (3, 3))
```

```
[30]: array([[ 0.26010113,  0.40018921,  0.35291973],  
          [-0.39158533, -0.34958723,  0.06219367],  
          [-1.56564613,  1.23545203,  0.50353038]])
```

```
[123]: x = np.random.normal(0, 1, (10000))
```

```
[124]: plt.hist(x, bins=200);
```



```
[132]: # Create a 3x3 array of random integers in the interval [0, 10)
np.random.randint(0, 100, 10)
```

```
[132]: array([39, 51, 11, 44, 66, 95, 90, 16, 77, 89])
```

```
[134]: # Create a 3x3 identity matrix
np.eye(3)
```

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

```
[133]: # Data Types
np.zeros(10, dtype='int16')
```

```
[133]: array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0], dtype=int16)
```

```
[ ]: # Data Types
np.zeros(10, dtype=np.int16)
```

0.1.1 Numpy Arrays

```
[44]: np.random.seed(0) # seed for reproducibility
x1 = np.random.randint(10, size=6) # One-dimensional array
x2 = np.random.randint(10, size=(3, 4)) # Two-dimensional array
```

```
x3 = np.random.randint(10, size=(3, 4, 5)) # Three-dimensional array
```

```
[47]: x2
```

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

```
[48]: print("x3 ndim: ", x3.ndim)
      print("x3 shape:", x3.shape)
      print("x3 size: ", x3.size)
```

```
x3 ndim: 3
x3 shape: (3, 4, 5)
x3 size: 60
```

```
[139]: # convert list to numpy array
      x = [[1,2,3], [4,5,6], [7,8,9], [1,1,1]] # 4 x 3
      y = [[1,2,3,1], [4,5,6,1], [7,8,9,4]] # 3 x 4
      type(x)
      # convert to numpy array
```

```
[139]: list
```

0.1.2 Numpy Array Indexing and Slicing

```
[49]: x2
```

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

```
[153]: # with 2D array
      x2[0:2,1]
```

```
[153]: array([5, 6])
```

```
[ ]: # with 2D array
```

```
[100]: x2
```

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

```
[101]: x2[:,2, :3]
```

```
[101]: array([[3, 5, 2],
             [7, 6, 8]])
```

```
[50]: x2[::-1, ::-1]
```

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

0.1.3 Be careful while creating copies

```
[58]: x2_sub_copy = x2[:2, :2]
      print(x2_sub_copy)
```

```
[[3 5]
 [7 6]]
```

```
[59]: x2_sub_copy[0, 0] = 42
      print(x2_sub_copy)
```

```
[[42  5]
 [ 7  6]]
```

```
[60]: print(x2)
```

```
[[42  5  2  4]
 [ 7  6  8  8]
 [ 1  6  7  7]]
```

0.1.4 Reshaping

```
[72]: grid = np.arange(1, 10).reshape((3,3))
      print(grid)
```

```
[[1 2 3]
 [4 5 6]
 [7 8 9]]
```

```
[70]: x = np.array([1, 2, 3])
      print(x.shape)
      # row vector via reshape
      x = x.reshape((1, 3))
      print(x.shape)
```

```
(3,)
(1, 3)
```

0.1.5 Concatenation

```
[71]: x = np.array([1, 2, 3])
      y = np.array([3, 2, 1])
      np.concatenate([x, y])
```

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

```
[74]: grid, grid
```

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

```
[80]: # concatenate along the first axis
      np.concatenate([grid, grid], axis=0) # rows will change, columns are expanded
```

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

```
[83]: # concatenate along the second axis (zero-indexed)
      np.concatenate([grid, grid], axis=1)
```

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

```
[85]: x = np.array([1, 2, 3]) # 1,3

      grid = np.array([[9, 8, 7],
                       [6, 5, 4]]) # (2,3)

      # vertically stack the arrays
      np.vstack([grid, x])
```

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

```
[87]: grid.shape
```



```
[87]: (2, 3)
```

```
[88]: y = np.array([[99],  
                  [99]])
```

```
[89]: y.shape
```

```
[89]: (2, 1)
```

```
[86]: # horizontally stack the arrays  
#y = np.array([[99],  
#             [99]])  
  
np.hstack([grid, y])
```

```
[86]: array([[ 9,  8,  7, 99],  
          [ 6,  5,  4, 99]])
```

0.1.6 Example of matrix multiplication

```
[91]: # vanilla python code to multiply two matrix  
  
def matmul_python(x, y):  
    result = [[0,0,0,0], [0,0,0,0], [0,0,0,0], [0,0,0,0]] # 4 x 4  
    colY = len(y[0])  
    rowX = len(x)  
    for row in range(rowX):  
        for col in range(colY):  
            for i,itemX in enumerate(x[row]):  
                result[row][col]+=itemX*y[i][col]  
    return result
```

```
[94]: x = [[1,2,3], [4,5,6], [7,8,9], [1,1,1]] # 4 x 3  
y = [[1,2,3,1], [4,5,6,1], [7,8,9,4]] # 3 x 4  
%timeit matmul_python(x,y)
```

9.8 μ s \pm 29.1 ns per loop (mean \pm std. dev. of 7 runs, 100,000 loops each)

```
[95]: x = np.array(x)  
y = np.array(y)  
%timeit np.matmul(x,y)
```

941 ns \pm 1.8 ns per loop (mean \pm std. dev. of 7 runs, 1,000,000 loops each)

```
[ ]: # Hint: 1 us = 1000ns
```

0.1.7 Summary Stats of Data

```
[1]: import numpy as np
```

```
[91]: # np.random.seed(0) # <- what happens with this line?
      L = np.random.random(100)
      sum(L)
```

```
[91]: 52.13902574466732
```

```
[92]: np.sum(L)
```

```
[92]: 52.139025744667336
```

```
[93]: big_array = np.random.rand(1000000)
      %timeit sum(big_array)
```

71.1 ms \pm 187 μ s per loop (mean \pm std. dev. of 7 runs, 10 loops each)

```
[94]: %timeit np.sum(big_array)
```

201 μ s \pm 2.36 μ s per loop (mean \pm std. dev. of 7 runs, 1,000 loops each)

```
[95]: # maximum and minimum
      min(big_array), max(big_array)
```

```
[95]: (1.4057692298008462e-06, 0.9999994392723005)
```

```
[96]: np.min(big_array), np.max(big_array)
```

```
[96]: (1.4057692298008462e-06, 0.9999994392723005)
```

```
[97]: # sum of multidimensional data
      M = np.random.random((3, 4))
      print(M)
```

```
[[0.35747479 0.16525562 0.22801649 0.01116173]
 [0.96484772 0.93042983 0.28571123 0.18633432]
 [0.84657553 0.11751108 0.43646768 0.48335078]]
```

```
[99]: M.shape, #(3,1)
```

```
[99]: (3, 4)
```

```
[98]: M.sum()
```

```
[98]: 5.013136800754203
```

```
[109]: M.sum(axis=1)
```

```
[109]: array([0.76190863, 2.3673231 , 1.88390506])
```

```
[106]: M.min(axis=0) # find sum of each column
```

```
[106]: array([0.35747479, 0.11751108, 0.22801649, 0.01116173])
```

```
[107]: M.min(axis=1)
```

```
[107]: array([0.01116173, 0.18633432, 0.11751108])
```

- `axis=0` is column
- `axis=1` is row
- The `axis` keyword specifies the dimension of the array that will be collapsed, rather than the dimension that will be returned

0.1.8 Broadcasting

```
[110]: a = np.array([0, 1, 2])  
b = np.array([5, 5, 5])  
a + b
```

```
[110]: array([5, 6, 7])
```

```
[113]: a + 5  
# [0,1,2] + [5,5,5]
```

```
[113]: array([5, 6, 7])
```

```
[115]: M, a
```

```
[115]: (array([[1., 1., 1.],  
             [1., 1., 1.],  
             [1., 1., 1.]]),  
       array([0, 1, 2]))
```

```
[116]: M = np.ones((3, 3))  
M+a # (3,3) (1,3)
```

```
[116]: array([[1., 2., 3.],  
             [1., 2., 3.],  
             [1., 2., 3.]])
```

Here the one-dimensional array `a` is stretched, or broadcast across the second dimension in order to match the shape of `M`

`numpy.newaxis` is used to increase the dimension of the existing array by one more dimension, when used once.

```
[117]: a = np.arange(3)
      b = np.arange(3)[: , np.newaxis] # same as saying, "I want every element spread
      ↪ across row but make it a 2d array"
```

```
[124]: a.reshape(3,1)
```

```
[124]: array([[0],
             [1],
             [2]])
```

```
[123]: b
```

```
[123]: array([[0],
             [1],
             [2]])
```

```
[120]: print(a.shape)
      print(b.shape)
```

```
(3,)
(3, 1)
```

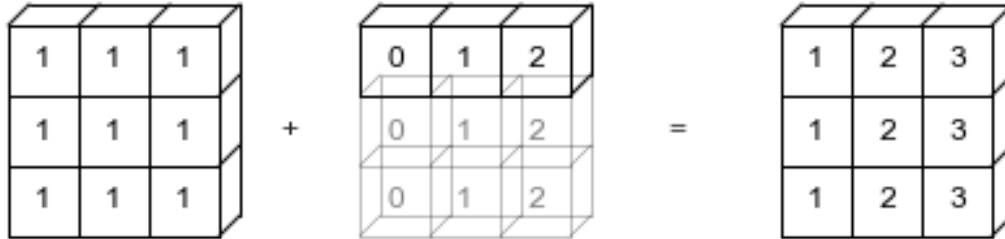
```
[121]: a + b
```

```
[121]: array([[0, 1, 2],
             [1, 2, 3],
             [2, 3, 4]])
```

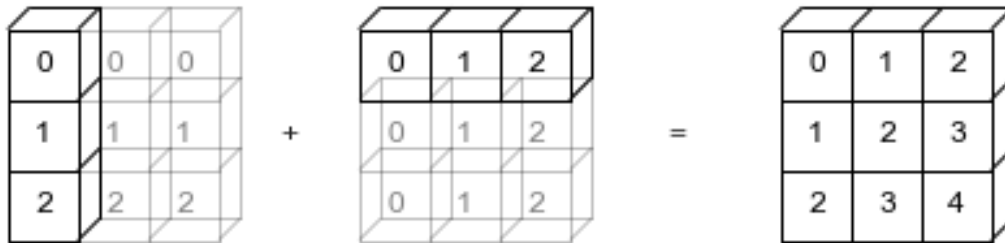
`np.arange(3)+5`



`np.ones((3, 3))+np.arange(3)`



`np.arange(3).reshape((3, 1))+np.arange(3)`



0.1.9 Rules of Broadcasting

Broadcasting in NumPy follows a strict set of rules to determine the interaction between the two arrays:

- Rule 1: If the two arrays differ in their number of dimensions, the shape of the one with fewer dimensions is padded with ones on its leading (left) side.
- Rule 2: If the shape of the two arrays does not match in any dimension, the array with shape equal to 1 in that dimension is stretched to match the other shape.
- Rule 3: If in any dimension the sizes disagree and neither is equal to 1, an error is raised.

```
[135]: # example of Rule 1
M = np.ones((2, 3)) # (2,3)
a = np.arange(3) # (1,3) -> (2,3)
```

```
[129]: M, a
```

```
[129]: (array([[1., 1., 1.],
               [1., 1., 1.]]),
       array([0, 1, 2]))
```

```
[136]: M+a
```

```
[136]: array([[1., 2., 3.],
            [1., 2., 3.]])
```

```
[140]: # example of Rule 2
a = np.arange(3).reshape((3, 1))
b = np.arange(3)
```

```
[141]: a+b
```

```
[141]: array([[0, 1, 2],
            [1, 2, 3],
            [2, 3, 4]])
```

```
[ ]: # example of Rule 3
```

```
[142]: M = np.ones((3, 2)) # (3,2)
a = np.arange(3)          #(3,3)
```

```
[143]: M+a
```

```
-----
ValueError                                Traceback (most recent call last)
Input In [143], in <cell line: 1>()
----> 1 M+a

ValueError: operands could not be broadcast together with shapes (3,2) (3,)
```

0.1.10 Usefulness of Broadcasting

```
[144]: # mean centering
X = np.random.random((10, 3))
```

```
[147]: X
```

```
[147]: array([[0.34284584, 0.00846426, 0.85343236],
            [0.38841378, 0.3388289 , 0.93265682],
            [0.46910315, 0.74570254, 0.9263706 ],
            [0.17372444, 0.02774391, 0.05559845],
            [0.67021314, 0.43505254, 0.90258482],
            [0.16114474, 0.56536336, 0.56557665],
            [0.28697744, 0.68061159, 0.54282346],
            [0.87071569, 0.97289904, 0.49027435],
            [0.80780529, 0.20026995, 0.33832142],
            [0.61004664, 0.87544312, 0.45195328]])
```

```
[146]: Xmean = X.mean(0)
Xmean
```

```
[146]: array([0.47809902, 0.48503792, 0.60595922])
```

```
[148]: X_centered = X - Xmean # (10,3) (10,3) ->
```

```
[149]: X_centered
```

```
[149]: array([[ -0.13525318, -0.47657366,  0.24747314],
             [-0.08968523, -0.14620902,  0.32669759],
             [-0.00899587,  0.26066462,  0.32041138],
             [-0.30437457, -0.45729401, -0.55036077],
             [ 0.19211412, -0.04998538,  0.2966256 ],
             [-0.31695427,  0.08032544, -0.04038257],
             [-0.19112157,  0.19557367, -0.06313576],
             [ 0.39261667,  0.48786112, -0.11568487],
             [ 0.32970628, -0.28476797, -0.2676378 ],
             [ 0.13194762,  0.3904052 , -0.15400594]])
```

0.1.11 Easy Sorting

```
[150]: x = np.array([2, 1, 4, 3, 5])
np.sort(x)
```

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

```
[151]: x = np.array([2, 1, 4, 3, 5])
i = np.argsort(x)
print(i)
```

```
[1 0 3 2 4]
```

```
[155]: np.argmin(x)
```

```
[155]: 1
```

```
[156]: # with 2d array
rand = np.random.RandomState(42)
X = rand.randint(0, 10, (4, 6))
print(X)
```

```
[[6 3 7 4 6 9]
 [2 6 7 4 3 7]
 [7 2 5 4 1 7]
 [5 1 4 0 9 5]]
```

```
[ ]: # sort each column of X
np.sort(X, axis=0)
```

```
[ ]: # sort each row of X
np.sort(X, axis=1)
```

0.1.12 Partial Sorts: Partitioning

```
[158]: x = np.array([7, 2, 3, 1, 6, 5, 4])
np.partition(x, 2)
```

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

0.1.13 2D plots

```
[165]: %matplotlib inline
import matplotlib.pyplot as plt
```

```
[159]: # x and y have 50 steps from 0 to 5
x = np.linspace(0, 5, 50)
y = np.linspace(0, 5, 50)[: , np.newaxis]
```

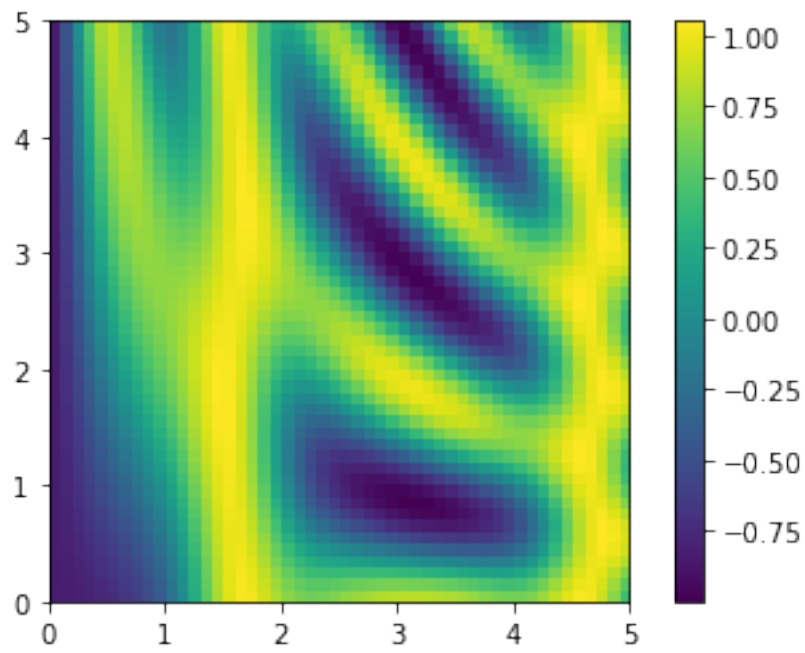
```
[168]: # can you explain whether the input to trigonometric functions are in degrees or
↳ radians?

# z = f(x,y)
z = np.sin(x) ** 10 + np.cos(10 + y * x) * np.cos(x)
```

```
[169]: z.shape
```

```
[169]: (50, 50)
```

```
[170]: plt.imshow(z, origin='lower', extent=[0, 5, 0, 5],
               cmap='viridis')
plt.colorbar();
```

[]: