Numpy_2

Data analysis with Numpy

Kunal Khurana

2024-02-16

Table of contents

NumPy 2
data types for ndarrays
arithmetic with numpy arrays
basic indexing and slicing
indexing with slices
Boolean indexing
Fancy indexing
Transposing arrays and swapping axes
Pseudorandom number generation
Universal Functions: Fast Element-Wise Array Functions
Array oriented programming with Arrays
Expressing Conditional Logic as Array Operations
numpy.where
mathematical and statistical methods
methods for boolean arrays
Sorting
unique and other set logic
array set operations
file input and output
Linear Alzerba
Random walks

NumPy

- numerical python function description
- ndarray- multidimensional array providing fast arithmetic operations
- mathematical functions
- tools for reading/writing
- linear algebra, random number generation, fourier transformation

import numpy as np

```
my_arr = np.arange(1000000)
  my_list = list(range(100000))
  %timeit my_arr2 = my_arr * 2
3.91~\text{ms}~\pm~154~\mu\text{s} per loop (mean \pm~\text{std}. dev. of 7 runs, 100 loops each)
  %timeit my_list2 = [x * 2 for x in my_list]
12.3 ms \pm 4.77 ms per loop (mean \pm std. dev. of 7 runs, 10 loops each)
  data = np.array([[1.5, -0.1, 3], [0, -3, 6.5]])
  data
array([[ 1.5, -0.1, 3.],
       [ 0. , -3. , 6.5]])
  data * 10
array([[ 15., -1., 30.],
       [ 0., -30., 65.]])
  data + data
array([[ 3. , -0.2, 6. ],
       [ 0. , -6. , 13. ]])
  data.shape
(2, 3)
  data.dtype
dtype('float64')
```

```
# creating arrays
  data1 = [6, 7.5, 8, 0, 1]
  arr1 = np.array(data1)
  arr1
array([6. , 7.5, 8. , 0. , 1. ])
  data2 = [[1, 2, 3, 4], [5, 6, 7, 8]]
  arr2 = np.array(data2)
  arr2
array([[1, 2, 3, 4],
       [5, 6, 7, 8]])
  arr2.ndim
2
  arr2.shape
(2, 4)
  arr1.dtype
dtype('float64')
  arr2.dtype
dtype('int32')
```

```
np.zeros(10)
array([0., 0., 0., 0., 0., 0., 0., 0., 0., 0.])
  np.ones((3, 6))
array([[1., 1., 1., 1., 1., 1.],
       [1., 1., 1., 1., 1., 1.],
       [1., 1., 1., 1., 1., 1.]])
np.empty((2, 3, 2))
  np.arange(15)
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14])
data types for ndarrays
  arr = np.array([1, 2, 3, 4, 5])
  arr.dtype
dtype('int32')
  float_arr = arr.astype(np.float64)
  float_arr.dtype
dtype('float64')
  int_array = np.arange(10)
```

```
calibers = np.array([.22, .27, .357, .380, .44, .50], dtype = np.float64)
  int_array.astype(calibers.dtype)
array([0., 1., 2., 3., 4., 5., 6., 7., 8., 9.])
arithmetic with numpy arrays
  arr = np.array([[1., 2., 3.], [4., 5., 6.]])
  arr
array([[1., 2., 3.],
       [4., 5., 6.]])
  arr * arr
array([[ 1., 4., 9.],
       [16., 25., 36.]])
  arr + arr
array([[ 2., 4., 6.],
       [8., 10., 12.]])
  1 / arr
                        , 0.33333333],
array([[1.
                , 0.5
                 , 0.2
                            , 0.16666667]])
       [0.25]
  arr ** 2
array([[ 1., 4., 9.],
```

[16., 25., 36.]])

```
# comparions between arrays yield boolean arrays
  arr2 = np.array([[0, 4, 1], [7, 2., 12]])
  arr2
array([[ 0., 4., 1.],
       [7., 2., 12.]])
  arr2 > arr
array([[False, True, False],
       [ True, False, True]])
basic indexing and slicing
  arr = np.arange(10)
  arr
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
  arr[5]
5
  arr[5:8]
array([5, 6, 7])
  arr[5:8]
array([5, 6, 7])
```

```
arr
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
  arr_slice= arr[5:8]
  arr_slice
array([5, 6, 7])
  # changing values
  arr_slice[1] = 12345
  arr
array([
          Ο,
                1, 2, 3, 4, 5, 12345, 7, 8,
          9])
  arr_slice
array([
          5, 12345,
                       7])
  # bare slice
  arr_slice[:] = 64  #assigns all values to the array
  arr
array([ 0, 1, 2, 3, 4, 64, 64, 64, 8, 9])
  arr2d = np.array([[1,2,3], [4,5,6], [7,8,9]])
  arr2d
array([[[1, 2, 3],
       [4, 5, 6],
       [7, 8, 9]]])
```

```
arr2d[1]
array([4, 5, 6])
  arr2d[0][2] #first array third element
3
  arr2d[0,2] # same result
3
  arr3d = np.array([[[1,2,3], [4,5,6]], [[7,8,9], [10, 11, 12]]])
  arr3d
array([[[ 1, 2, 3],
        [4, 5, 6]],
       [[7, 8, 9],
        [10, 11, 12]]])
  arr3d[<mark>0</mark>]
array([[1, 2, 3],
       [4, 5, 6]])
  \# scalar and vector arays can be assigned to arr3d[0]
  old_values = arr3d[0].copy()
  arr3d[0] = 42
  arr3d
```

```
array([[[42, 42, 42],
       [42, 42, 42]],
       [[7, 8, 9],
       [10, 11, 12]]])
  arr3d[0] = old_values
  arr3d
array([[[ 1, 2, 3],
       [4, 5, 6]],
       [[7, 8, 9],
       [10, 11, 12]]])
  arr3d[1,0]
array([7, 8, 9])
  x = arr3d[1]
  X
array([[ 7, 8, 9],
      [10, 11, 12]])
  x[0]
array([7, 8, 9])
indexing with slices
  arr
array([ 0, 1, 2, 3, 4, 64, 64, 64, 8, 9])
```

```
arr[1:6]
array([ 1, 2, 3, 4, 64])
  # slicing a 2d array
  arr2d
array([[1, 2, 3],
       [4, 5, 6],
       [7, 8, 9]])
  arr2d[:2] #selects the first two rows
array([[1, 2, 3],
       [4, 5, 6]])
  arr2d[:2, 1:] #selects first two rows and last two columns
array([[2, 3],
       [5, 6]])
  lower_dim_slice = arr2d[1, :2]
  lower_dim_slice
array([4, 5])
  lower_dim_slice.shape
(2,)
  arr2d[:2, 2] #dots before selects rows before
array([3, 6])
```

```
arr2d[:, :1] #all rows, first column
array([[1],
       [4],
       [7]])
  # assigning value to the section
  arr2d[:2, 1:] = 0
  arr2d
array([[1, 0, 0],
       [4, 0, 0],
       [7, 8, 9]])
Boolean indexing
  names = np.array(['bob', 'joe', 'will', 'zhou', 'lu', 'wei'])
  names
array(['bob', 'joe', 'will', 'zhou', 'lu', 'wei'], dtype='<U4')
  data = np.array([[4,7], [0,2], [-5, 6], [0, 0], [1, 2], [-12, -4], [3, 4]])
  data
array([[ 4,
               7],
       [ 0,
               2],
       [ -5,
               6],
       [ 0,
               0],
       [ 1,
               2],
       [-12,
              -4],
```

4]])

[3,

```
data.shape
(7, 2)
  names.shape
(6,)
  # let's check how many times wei's name come
  names == 'wei'
                   #once
array([False, False, False, False, False, True])
  data[names == "wei"]
IndexError: boolean index did not match indexed array along dimension 0; dimension is 7 but
  # adding a name so that the dimension becomes 7
  names = np.append(names, 'rajwinder')
  names
array(['bob', 'joe', 'will', 'zhou', 'lu', 'wei', 'rajwinder',
       'rajwinder'], dtype='<U9')
  # deleting extra
  names = np.delete(names, 7)
  names
array(['bob', 'joe', 'will', 'zhou', 'lu', 'wei', 'rajwinder'],
      dtype='<U9')
```

```
data[names == 'rajwinder']
array([[3, 4]])
  data[names == 'zhou']
array([[0, 0]])
Fancy indexing
  • indexing using integers
  • indexing gets modified
  arr = np.zeros((8, 4))
  for i in range(8):
       arr[i] = i
  arr
array([[0., 0., 0., 0.],
       [1., 1., 1., 1.],
       [2., 2., 2., 2.],
       [3., 3., 3., 3.],
       [4., 4., 4., 4.],
       [5., 5., 5., 5.],
       [6., 6., 6., 6.],
       [7., 7., 7., 7.]])
  # selecting rows in particular order
  arr[[4,3, 0, 6]]
array([[4., 4., 4., 4.],
       [3., 3., 3., 3.],
```

[0., 0., 0., 0.], [6., 6., 6., 6.]])

```
# using negative indices
  arr[[-2, -4, -7]]
array([[6., 6., 6., 6.],
       [4., 4., 4., 4.],
       [1., 1., 1., 1.]])
  # multiple array indexing
  arr3 = np.arange(32).reshape((8, 4))
  arr3
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, 25, 26, 27],
       [28, 29, 30, 31]])
  # selecting elements based on rows and columns
  arr3[[1, 4, 7, 2], [0, 3, 2, 1]]
array([ 4, 19, 30, 9])
  # selecting complete rows and decding sequence of elements
  arr3[[1, 4, 7, 2]][:, [0, 3, 2, 1]]
array([[ 4, 7, 6, 5],
       [16, 19, 18, 17],
       [28, 31, 30, 29],
       [8, 11, 10, 9]])
```

Transposing arrays and swapping axes

```
arr = np.arange(15). reshape(3, 5)
  arr
array([[ 0, 1, 2, 3, 4],
       [5, 6, 7, 8, 9],
      [10, 11, 12, 13, 14]])
  arr.T
array([[ 0, 5, 10],
      [1, 6, 11],
      [2, 7, 12],
      [3, 8, 13],
       [4, 9, 14]])
  # used often for matrix computation
  arr
array([[ 0, 1, 2, 3, 4],
      [5, 6, 7, 8, 9],
      [10, 11, 12, 13, 14]])
  # multiplied two arrays
  np.dot(arr.T, arr)
array([[125, 140, 155, 170, 185],
       [140, 158, 176, 194, 212],
       [155, 176, 197, 218, 239],
       [170, 194, 218, 242, 266],
       [185, 212, 239, 266, 293]])
  # another way to do it
  arr.T @ arr
```

```
array([[125, 140, 155, 170, 185],
        [140, 158, 176, 194, 212],
        [155, 176, 197, 218, 239],
        [170, 194, 218, 242, 266],
        [185, 212, 239, 266, 293]])

arr

array([[ 0,  1,  2,  3,  4],
        [ 5,  6,  7,  8,  9],
        [10, 11, 12, 13, 14]])

arr.swapaxes(0, 1)  # returns the view without making a copy

array([[ 0,  5, 10],
        [ 1,  6, 11],
        [ 2,  7, 12],
        [ 3,  8, 13],
        [ 4,  9, 14]])
```

Pseudorandom number generation

Universal Functions: Fast Element-Wise Array Functions

```
import numpy as np
  from random import normalvariate
  arr = np.arange(10)
  np.sqrt(arr)
array([0.
                , 1. , 1.41421356, 1.73205081, 2.
      2.23606798, 2.44948974, 2.64575131, 2.82842712, 3.
                                                                ])
  np.exp(arr)
array([1.00000000e+00, 2.71828183e+00, 7.38905610e+00, 2.00855369e+01,
      5.45981500e+01, 1.48413159e+02, 4.03428793e+02, 1.09663316e+03,
      2.98095799e+03, 8.10308393e+03])
  x = rng.standard normal(8)
  y = rng.standard_normal(8)
  X
array([-0.32357072, -1.8494368, -1.89739205, 0.04315429, 1.01046514,
      -0.73625393, 0.46616191, -0.09290374])
  у
array([-0.12705798, -0.64476954, -0.62430977, 0.87432098, 1.55273649,
      -1.53739177, -0.73752509, 0.41995739])
  np.maximum(x, y) #based on element wise comparison
array([-0.12705798, -0.64476954, -0.62430977, 0.87432098, 1.55273649,
      -0.73625393, 0.46616191, 0.41995739])
```

```
arr = rng.standard_normal(7) * 5
  arr
array([-1.61785359, -9.24718402, -9.48696026, 0.21577147, 5.05232568,
       -3.68126964, 2.33080955])
  remainder, whole_part = np.modf(arr)
  remainder
array([-0.61785359, -0.24718402, -0.48696026, 0.21577147, 0.05232568,
       -0.68126964, 0.33080955])
  whole_part
array([-1., -9., -9., 0., 5., -3., 2.])
  arr
array([-1.61785359, -9.24718402, -9.48696026, 0.21577147, 5.05232568,
       -3.68126964, 2.33080955])
  out = np.zeros_like(arr)
  np.add(arr, 1)
array([-0.61785359, -8.24718402, -8.48696026, 1.21577147, 6.05232568,
      -2.68126964, 3.33080955])
  np.add(arr, 1, out= out)
array([-0.61785359, -8.24718402, -8.48696026, 1.21577147, 6.05232568,
      -2.68126964, 3.33080955])
```

```
array([-0.61785359, -8.24718402, -8.48696026, 1.21577147, 6.05232568, -2.68126964, 3.33080955])
```

Array oriented programming with Arrays

• vectorization (faster) than pure Python equivalents

```
points = np.arange(-5, 5, 0.01) #100 equally spaced points
  xs, ys = np.meshgrid(points, points)
  # numpy.meshgrid function takes two one-dimensional arrays and produces two two-dimensional
  уs
array([[-5., -5., -5., ..., -5., -5., -5.],
       [-4.99, -4.99, -4.99, ..., -4.99, -4.99, -4.99],
       [-4.98, -4.98, -4.98, \ldots, -4.98, -4.98, -4.98],
       [4.97, 4.97, 4.97, \ldots, 4.97, 4.97, 4.97],
       [4.98, 4.98, 4.98, \ldots, 4.98, 4.98, 4.98],
       [4.99, 4.99, 4.99, \ldots, 4.99, 4.99, 4.99]])
  z = np.sqrt (xs ** 2 + ys ** 2)
  Z
array([[7.07106781, 7.06400028, 7.05693985, ..., 7.04988652, 7.05693985,
       7.06400028],
       [7.06400028, 7.05692568, 7.04985815, ..., 7.04279774, 7.04985815,
       7.05692568],
       [7.05693985, 7.04985815, 7.04278354, ..., 7.03571603, 7.04278354,
       7.04985815],
       [7.04988652, 7.04279774, 7.03571603, ..., 7.0286414, 7.03571603,
       7.04279774],
```

```
[7.05693985, 7.04985815, 7.04278354, ..., 7.03571603, 7.04278354, 7.04985815],
[7.06400028, 7.05692568, 7.04985815, ..., 7.04279774, 7.04985815, 7.05692568]])
```

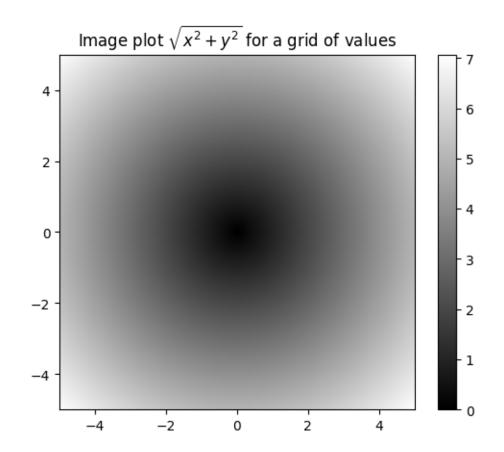
```
# visualizations with 2-d arrays
import matplotlib.pyplot as plt

plt.imshow(z, cmap = plt.cm.gray, extent = [-5, 5, -5, 5])

plt.colorbar()

plt.title("Image plot $\sqrt{x^2 + y^2}$ for a grid of values")
```

Text(0.5, 1.0, 'Image plot $\x^2 + y^2$ ' for a grid of values')



```
plt.close('all')
```

Expressing Conditional Logic as Array Operations

numpy.where

```
# numpy.where (replace all positive values with 2 and negative with -2)
arr = rng.standard_normal((4, 4))
arr

array([[-0.09290374, -0.12705798, -0.64476954, -0.62430977],
        [ 0.87432098,  1.55273649, -1.53739177, -0.73752509],
        [ 0.41995739, -0.93658739,  0.62072248,  0.81057914],
        [-0.21398203,  0.67748945, -1.54002066, -0.9638457 ]])

arr > 0

array([[False, False, False, False],
        [ True, True, False, False],
```

```
[ True, False, True, True],
       [False, True, False, False]])
  np.where (arr > 0, 2, -2)
array([[-2, -2, -2, -2],
      [2, 2, -2, -2],
      [2,-2,2,2],
      [-2, 2, -2, -2]
  # or set only the positive values to 2
  np.where (arr > 0, 2, arr)
array([[-0.09290374, -0.12705798, -0.64476954, -0.62430977],
       [ 2.
                             , -1.53739177, -0.73752509],
                  , 2.
                  , -0.93658739, 2.
       [ 2.
                        , -1.54002066, -0.9638457 ]])
       [-0.21398203, 2.
mathematical and statistical methods
  arr = rng.standard_normal ((5, 4))
  arr
array([[-0.64316368, -0.48860061, -1.41271857, -0.10120962],
       [-0.70385422, 2.41319157, -0.54405393, -0.90339244],
       [0.82712685, -0.62647321, -0.13480887, 0.03956079],
       [0.56044129, 0.34237924, -0.6576538, 1.04696188],
       [0.17595271, -1.13639865, -0.54922125, 0.70725439]])
  arr.mean()
-0.08943400646176203
  np.mean(arr)
```

-0.08943400646176203

```
arr.sum()
-1.7886801292352406
  arr.mean(axis = 1) # columns
array([-0.66142312, 0.06547274, 0.02635139, 0.32303215, -0.2006032])
  arr.sum(axis = 1)
array([-2.64569248, 0.26189098, 0.10540556, 1.29212861, -0.8024128])
  arr = np.array([0,1,2,3,4,5,6,7])
  arr.cumsum()
array([ 0, 1, 3, 6, 10, 15, 21, 28])
  arr = np.array([[0,1,2], [3, 4, 5], [6, 7, 8]])
  arr
array([[0, 1, 2],
       [3, 4, 5],
       [6, 7, 8]])
  \# arr.cumsum(axis = 0 ) computes the cumulative sum along rows
  # arr.sumsum (axis= 1) computes the sum along columns
  arr.cumsum(axis = 0)
array([[ 0, 1, 2],
       [3, 5, 7],
       [ 9, 12, 15]])
```

```
arr.cumsum(axis = 1)
array([[ 0, 1, 3],
       [3, 7, 12],
       [ 6, 13, 21]])
methods for boolean arrays
  arr = rng.standard_normal(100)
  (arr> 0).sum()
41
  (arr <= 0).sum() #all non-po
59
Sorting
  arr = rng.standard_normal(6)
  arr
array([ 0.81272428, -0.67629236, 0.09344394, -0.20621744, 0.10364886,
        0.70966403])
  arr.sort()
  arr
```

array([-0.67629236, -0.20621744, 0.09344394, 0.10364886, 0.70966403,

0.81272428])

```
arr = rng.standard_normal((5, 3))
  arr
array([[-1.58684863, -0.1143117, 2.38420916],
       [-0.64811009, 1.31931176, 0.01123432],
      [-0.90663373, -0.96531814, 0.46431808],
       [0.52164015, -0.08486576, -0.98397298],
       [ 0.09054187, -1.08417551, -0.48832961]])
  arr.sort (axis = 0) #sorts the values across columns
  arr
array([[-1.58684863, -1.08417551, -0.98397298],
       [-0.96531814, -0.90663373, -0.48832961],
       [-0.64811009, -0.1143117, 0.01123432],
       [-0.08486576, 0.09054187, 0.46431808],
       [ 0.52164015, 1.31931176, 2.38420916]])
  arr.sort (axis = 1)
  arr
array([[-1.58684863, -1.08417551, -0.98397298],
       [-0.96531814, -0.90663373, -0.48832961],
       [-0.64811009, -0.1143117, 0.01123432],
       [-0.08486576, 0.09054187, 0.46431808],
       [ 0.52164015, 1.31931176, 2.38420916]])
  arr2 = np.array([5, -10, 7, 1, 0, -3])
  sorted_arr2 = np.sort(arr2)
  sorted_arr2
array([-10, -3, 0, 1, 5, 7])
```

unique and other set logic

```
names
array(['bob', 'joe', 'will', 'zhou', 'lu', 'wei', 'rajwinder'],
      dtype='<U9')
  np.unique(names)
array(['bob', 'joe', 'lu', 'rajwinder', 'wei', 'will', 'zhou'],
      dtype='<U9')
  np.append(names, 'lu')
array(['bob', 'joe', 'will', 'zhou', 'lu', 'wei', 'rajwinder', 'lu'],
      dtype='<U9')
  # we've 'lu' twice now, let's see now unique
  # sorting done aswell
  np.unique(names)
array(['bob', 'joe', 'lu', 'rajwinder', 'wei', 'will', 'zhou'],
      dtype='<U9')
  # python alternative
  sorted(set(names))
['bob', 'joe', 'lu', 'rajwinder', 'wei', 'will', 'zhou']
```

array set operations

```
# numpy.in1d for testing memebership of the values in one array
  values = np.array([6, 0,0,0,3,2])
  np.in1d(values, [1,2,3])
array([False, False, False, False, True, True])
file input and output
  arr = np.arange(10)
  np.save('some_array', arr)
  np.load('some_array.npy')
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
  # save multiple arrays using np.savez
  np.savez('array_archive.npz', a = arr, b=arr)
  arch = np.load("array_archive.npz")
  arch['b']
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
  # saving in compressed format
  np.savez_compressed('arrays_compressed.npz', a= arr, b= arr)
```

Linear Alzerba

```
x = np.array([[1, 2, 3], [4, 5, 6]])
  y = np.array([[6, 23], [-1, 7], [8,9]])
  X
array([[1, 2, 3],
       [4, 5, 6]])
  у
array([[ 6, 23],
       [-1, 7],
       [8, 9]])
  x.dot(y)
array([[ 28, 64],
       [ 67, 181]])
  # equivalent to
  np.dot(x, y)
array([[ 28, 64],
       [ 67, 181]])
  # product of 1d and 2d array
  x @ np.ones(3)
array([ 6., 15.])
```

```
# numpy.linalg (matrix decompositions)
  from numpy.linalg import inv, qr
  X = rng.standard_normal((5, 5))
  mat = X.T @ X
  mat.
array([[ 5.79511464, -3.30831545, -2.66542844, -0.61858429, -4.34315368],
       [-3.30831545, 6.04913293, 1.09484984, -0.88187098, 3.79344801],
      [-2.66542844, 1.09484984, 3.59693921, -0.10949232, 1.50109261],
      [-0.61858429, -0.88187098, -0.10949232, 0.68764721, 0.24806815],
      [-4.34315368, 3.79344801, 1.50109261, 0.24806815, 4.09980802]])
  inv(mat)
array([[ 1.95391205, 0.4259796 , 0.86161239, 1.99396982, 1.23962108],
      [0.4259796, 1.84110512, 0.55359754, 3.43225775, -1.66263314],
      [0.86161239, 0.55359754, 0.79117237, 1.60608307, 0.01366661],
      [1.99396982, 3.43225775, 1.60608307, 8.69084511, -2.17736422],
      [ 1.23962108, -1.66263314, 0.01366661, -2.17736422, 3.22224774]])
  mat @ inv(mat)
array([[ 1.00000000e+00, 7.37690538e-17, 8.63526934e-17,
        2.45602532e-16, 5.57110698e-16],
      [ 3.59505366e-17, 1.00000000e+00, -1.43602651e-16,
        1.56181454e-15, -8.26684003e-16],
       [-2.51848975e-16, -9.56323491e-18, 1.00000000e+00,
       -7.81952475e-16, -4.32942875e-16],
      [-1.22081410e-16, 5.77266093e-17, -3.23653576e-16,
        1.00000000e+00, -9.26541377e-18],
      [-4.93401745e-16, 1.63171237e-15, -7.64319458e-17,
        1.26774536e-15, 1.00000000e+00]])
```

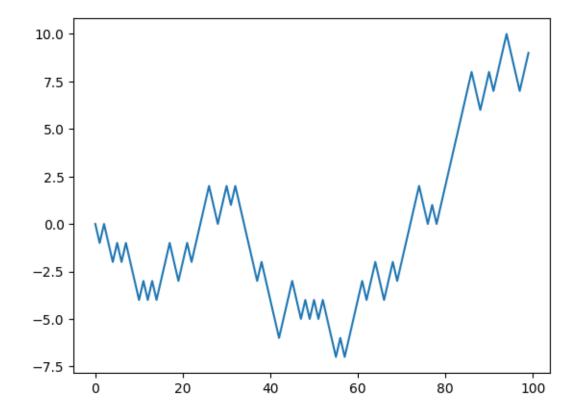
Random walks

```
# with python

import random
position = 0
walk = [position]
nsteps = 1000

for _ in range(nsteps):
    step = 1 if random.randint(0, 1) else -1
    position += step
    walk.append (position)

plt.plot(walk[:100])
```



```
# with numpy
  nsteps = 1000
  rng = np.random.default_rng (seed = 12345)
  draws = rng.integers(0, 2, size= nsteps)
  steps = np.where(draws == 0, 1, -1)
  walk = steps.cumsum()
  walk.min()
-8
  walk.max()
50
  (np.abs(walk) >= 10).argmax()
155
  # simulting many random walks at once with numpy
  nwalks = 5000
  nsteps = 1000
  draws = rng.integers(0, 2, size = (nwalks, nsteps))
  steps = np.where(draws > 0, 1, -1)
  walks = steps.cumsum(axis = 1)
  walks
```

```
array([[ 1,
              2, 1, ..., -24, -25, -26],
       [ -1,
              0, -1, \ldots, -2, -1, 0],
       [ 1,
              0, 1, \ldots, -22, -23, -24],
                  1, ..., 0,
       [ 1,
              Ο,
                                  1,
             -2, -3, ..., 78, 77, 78],
              2, 1, ..., -42, -41, -40]])
  walks.max()
143
  walks.min()
-125
  # any method to check for details
  hits30 = (np.abs(walks) >= 30).any(axis = 1)
  hits30
array([ True, False, True, ..., False, True, True])
  hits30.sum()
3314
  crossing_times = (np.abs(walks[hits30]) >= 30).argmax(axis = 1)
  crossing_times
array([897, 187, 607, ..., 497, 363, 337], dtype=int64)
  # average minn
```