

# Numpy commands

July 18, 2019

## 0.1 Numpy Commands

This file consists of useful Numpy commands from Random, linear algebra libraries and also some other functions

```
[132]: import numpy as np
import matplotlib.pyplot as plt
```

### 1. Random library

```
[133]: np.random.rand(3,2) #create a 3x2 array of random values between 0 and 1
```

```
[133]: array([[0.38734262, 0.65746878],
            [0.8160697 , 0.90042571],
            [0.65609903, 0.62370866]])
```

```
[134]: np.random.randn(3,3) #creates a random 3x3 array of values from standard normal  
→distribution
```

```
[134]: array([[ 0.10549219,  0.79660127,  0.12826328],
            [ 1.79133473,  1.5245318 ,  0.81386083],
            [ 0.59409159,  1.51377016, -0.49635222]])
```

```
[135]: np.random.randint(1,5,size=(2,2)) #create array of size 10 with values between  
→1(inclusive) to 5(exclusive)
```

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

```
[136]: np.random.random() #returns a random sample between 0.0(exclusive) and 1.  
→0(inclusive)  
np.random.sample() #returns a random sample between 0.0(exclusive) and 1.  
→0(inclusive)
```

```
[136]: 0.31876138609234017
```

```
[137]: df = np.array([22,34,56])  
np.random.choice(df,1,p=[0.1,0.6,0.3]) #choose 1 number from df where p is  
→probabilites of there occurences
```

```
[137]: array([56])
```

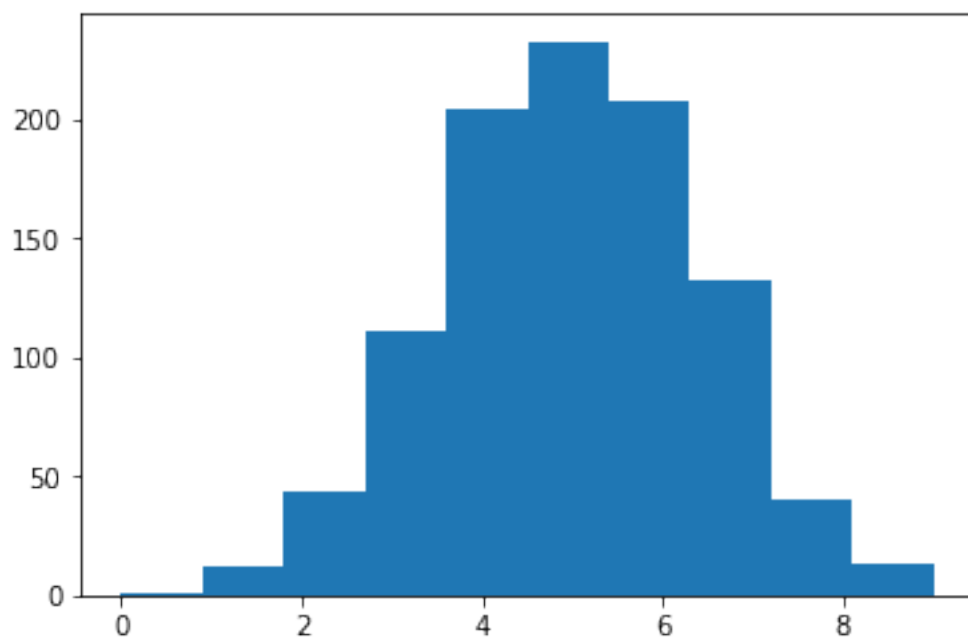
```
[138]: np.random.choice(4,3,replace = True) #choose 3 number between 0-4(exclusive)
      ↪where repition is allowed
```

```
[138]: array([1, 1, 3])
```

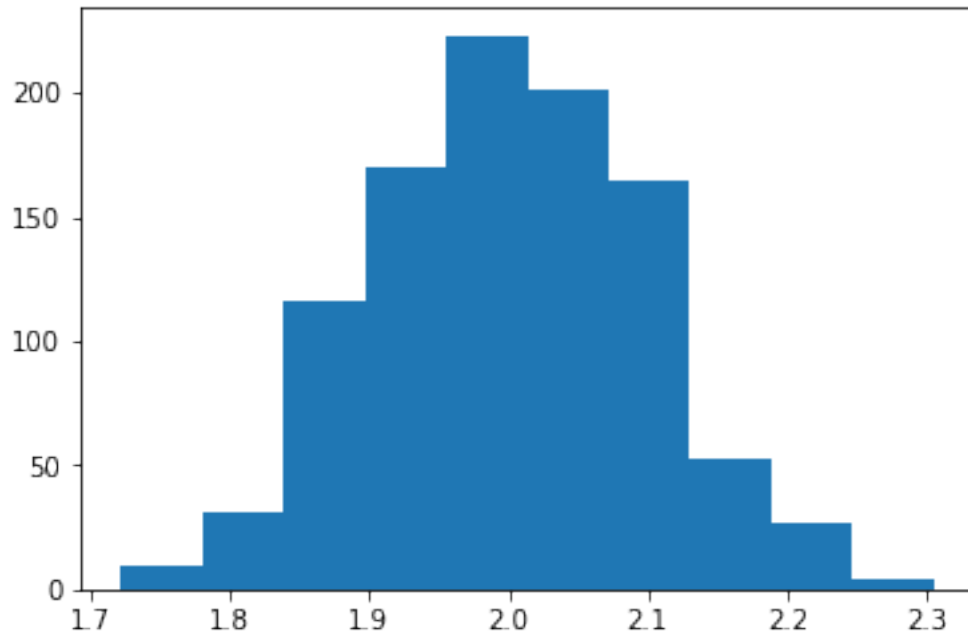
```
[139]: np.random.choice(4,3,replace = False) #choose 3 number between 0-4(exclusive)
      ↪where repition is not allowed
```

```
[139]: array([3, 0, 1])
```

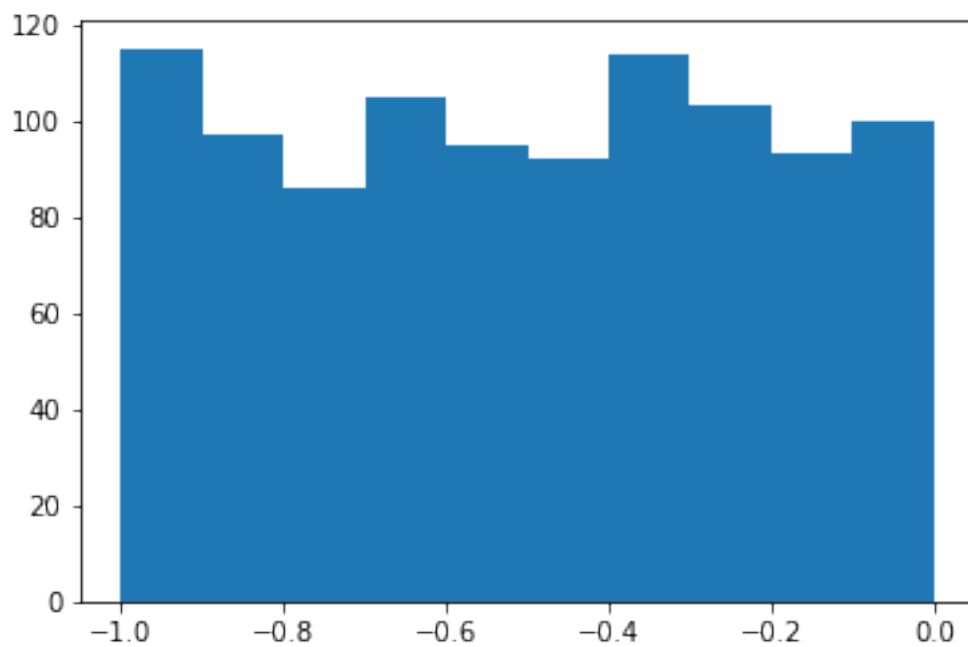
```
[140]: n, p = 10, 0.5 # number of trials, probability of each trial
      s = np.random.binomial(n, p, 1000)
      # result of flipping a coin 10 times, tested 1000 times.
      plt.hist(s);
```



```
[141]: mean, std = 2, 0.1 # mean and standard deviation
      s = np.random.normal(mean, std, 1000)
      plt.hist(s);#binomial distribution is discrete and normal idistribution is
      ↪continous.
```



```
[142]: s = np.random.uniform(-1,0,1000)
plt.hist(s);#uniform distribution of between -1 and 0 form 1000 times
```



```
[143]: np.random.permutation(10) #randomly arrange array of numbers between 0-9.
```

```
[143]: array([3, 2, 9, 0, 1, 7, 8, 4, 6, 5])
```

```
[144]: np.random.permutation([1, 4, 9, 12, 15]) #random arrange the mentioned numbers  
      →in that array.
```

```
[144]: array([ 9,  1, 15,  4, 12])
```

## 2. Linear Algebra

```
[145]: np.dot(3,5) #dot product of two values
```

```
[145]: 15
```

```
[146]: a = np.array([[1,2],[3,4]])  
      b = np.array([[4,5],[6,7]])  
      np.dot(a,b) #returns dot product of two matrices
```

```
[146]: array([[16, 19],  
            [36, 43]])
```

```
[147]: np.dot([2j, 3j], [2j, 3j]) #dotproduct of conjugates
```

```
[147]: (-13+0j)
```

```
[148]: a = np.array([[1,2],[3,4]])  
      b = np.array([[4,5],[6,7]])  
      c=np.array([[2,3],[3,4]]) #the dot product of two or more arrays in a single  
      →function call,  
      np.linalg.multi_dot([a,b,c]) #while automatically selecting the fastest  
      →evaluation order.
```

```
[148]: array([[ 89, 124],  
            [201, 280]])
```

```
[149]: a = [[1, 0], [0, 1]]  
      b = [[4, 1], [2, 2]]  
      np.matmul(a, b) #matrix multiplication of two arrays.
```

```
[149]: array([[4, 1],  
            [2, 2]])
```

```
[150]: b = [[4, 1], [2, 2]]  
      np.linalg.matrix_power(b,2) #Raise a square matrix to the (integer) power 2.
```

```
[150]: array([[18,  6],  
            [12,  6]])
```

```
[151]: A = np.array([[1,-2j],[2j,5]])  
      L = np.linalg.cholesky(A) #Cholesky decomposition of a matrix.  
      L
```

```
[151]: array([[1.+0.j, 0.+0.j],  
            [0.+2.j, 1.+0.j]])
```

```
[152]: a = np.array([[1,2,3],[4,5,6],[7,8,9]])
np.linalg.eigvals(a) #returns eigen values of matrix 'a'
```

```
[152]: array([ 1.61168440e+01, -1.11684397e+00, -9.75918483e-16])
```

```
[153]: a = np.array([[1,2,3],[4,5,6],[7,8,9]])
w,v = np.linalg.eig(a) # computes eigen values and right eigen vectors of
    ↳ square array. w is eigen values, v is eigen vector
print(w)
print(' ')
print(v)
```

```
[ 1.61168440e+01 -1.11684397e+00 -9.75918483e-16]
```

```
[[-0.23197069 -0.78583024  0.40824829]
 [-0.52532209 -0.08675134 -0.81649658]
 [-0.8186735   0.61232756  0.40824829]]
```

```
[154]: b = [[4, 1], [2, 2]]
np.linalg.matrix_rank(b) #find the rank of a matrix.
```

```
[154]: 2
```

```
[155]: b = [[4, 1], [2, 2]]
np.linalg.det(b) #find the determinant of an array
```

```
[155]: 6.0
```

```
[156]: b = [[4, 1], [2, 2]]
np.linalg.inv(b) #finds the multiplicative inverse of a matrix.
```

```
[156]: array([[ 0.33333333, -0.16666667],
           [-0.33333333,  0.66666667]])
```

### 3. other useful commands

```
[157]: #1. numpy where for conditional search
a = np.array([6,6,7,8,4,5,3,6,7,1,2,2,2])
index_gt_5 = np.where(a > 5) #index where a > 5
a.take(index_gt_5) #get values in those indexes
```

```
[157]: array([[6, 6, 7, 8, 6, 7]])
```

```
[158]: #2.Position of maximum and minumum element in array
a = np.array([61,43,555,6,7,8,99])
print(np.argmax(a)) #index of maximum element(index starts from 0)
np.argmin(a) #indexof minimum element
```

2

```
[158]: 3
```

```
[159]: #3.concatination of two arrays
a = np.zeros([4, 4])
b = np.ones([4, 4])
#vertical concatintion
print(np.concatenate([a, b], axis=0))
print(' ')
print(np.vstack([a,b]))
np.r_[a,b] #all the three do same work
```

```
[[0. 0. 0. 0.]
 [0. 0. 0. 0.]
 [0. 0. 0. 0.]
 [0. 0. 0. 0.]
 [1. 1. 1. 1.]
 [1. 1. 1. 1.]
 [1. 1. 1. 1.]
 [1. 1. 1. 1.]]
```

```
[[0. 0. 0. 0.]
 [0. 0. 0. 0.]
 [0. 0. 0. 0.]
 [0. 0. 0. 0.]
 [1. 1. 1. 1.]
 [1. 1. 1. 1.]
 [1. 1. 1. 1.]
 [1. 1. 1. 1.]]
```

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

```
[160]: #horizontal concatination
print(np.concatenate([a, b], axis=1))
print(' ')
print(np.hstack([a,b]))
np.c_[a,b] #all the three do same work
```

```
[[0. 0. 0. 0. 1. 1. 1. 1.]
 [0. 0. 0. 0. 1. 1. 1. 1.]
 [0. 0. 0. 0. 1. 1. 1. 1.]
 [0. 0. 0. 0. 1. 1. 1. 1.]]
```

```

[[0. 0. 0. 0. 1. 1. 1. 1.]
 [0. 0. 0. 0. 1. 1. 1. 1.]
 [0. 0. 0. 0. 1. 1. 1. 1.]
 [0. 0. 0. 0. 1. 1. 1. 1.]]

```

```

[160]: array([[0., 0., 0., 0., 1., 1., 1., 1.],
             [0., 0., 0., 0., 1., 1., 1., 1.],
             [0., 0., 0., 0., 1., 1., 1., 1.],
             [0., 0., 0., 0., 1., 1., 1., 1.]])

```

```

[161]: #4.sorting a array in different ways
arr = np.random.randint(1,6, size=[8, 4])
print('arr')
print(arr)
print('each column sort')
print(np.sort(arr,axis =0)) #sort each column
print('each row sort')
np.sort(arr,axis =1) #sort each row

```

```

arr
[[4 3 4 5]
 [4 3 3 1]
 [1 2 5 5]
 [1 3 2 3]
 [5 4 1 1]
 [2 3 3 1]
 [1 3 2 5]
 [4 3 3 2]]
each column sort
[[1 2 1 1]
 [1 3 2 1]
 [1 3 2 1]
 [2 3 3 2]
 [4 3 3 3]
 [4 3 3 5]
 [4 3 4 5]
 [5 4 5 5]]
each row sort

```

```

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

```

```
[163]: sort_acc_1stcol = arr[:, 0].argsort()  
# Sort 'arr' by *first column* without disturbing the content in rows  
arr[sort_acc_1stcol]
```

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