

NumPy Package

What is NumPy?

NumPy is the fundamental package for scientific computing with Python. It contains among other things:

- a powerful N-dimensional array object
- sophisticated (broadcasting) functions
- tools for integrating C/C++ and Fortran code
- useful linear algebra, Fourier transform, and random number capabilities

NumPy for Matlab users

NumPy	Matlab
Python uses 0 (zero) based indexing. The initial element of a sequence is found using <code>a[0]</code> .	MATLAB® uses 1 (one) based indexing. The initial element of a sequence is found using <code>a(1)</code> .
Uses multidimensional array *	Uses multidimensional Matrice
2x3 matrix literal <code>array([[1.,2.,3.], [4.,5.,6.]])</code>	2x3 matrix literal <code>[1 2 3; 4 5 6]</code>
access element in second row, fifth column <code>a[1,4]</code>	access element in second row, fifth column <code>a(2,5)</code>

For more information, visit: <https://docs.scipy.org/doc/numpy-dev/user/numpy-for-matlab-users.html>

* There is a **matrix** type in NumPy for 2-dim linear algebra. But we prefer array

Let's Start Programming

NumPy for Matlab users

General Purpose Equivalents

MATLAB	numpy	Notes
<code>help func</code>	<code>info(func)</code> or <code>help(func)</code> or <code>func?</code> (in Ipython)	get help on the function <i>func</i>
<code>which func</code>	see note HELP	find out where <i>func</i> is defined
<code>type func</code>	<code>source(func)</code> or <code>func??</code> (in Ipython)	print source for <i>func</i> (if not a native function)
<code>a && b</code>	<code>a and b</code>	short-circuiting logical AND operator (Python native operator); scalar arguments only
<code>a b</code>	<code>a or b</code>	short-circuiting logical OR operator (Python native operator); scalar arguments only
<code>1*i, 1*j, 1i, 1j</code>	<code>1j</code>	complex numbers
<code>eps</code>	<code>np.spacing(1)</code>	Distance between 1 and the nearest floating point number.
<code>ode45</code>	<code>scipy.integrate.ode(f).set_integrator('dopri5')</code>	integrate an ODE with Runge-Kutta 4,5
<code>ode15s</code>	<code>scipy.integrate.ode(f).set_integrator('vode', method='bdf', order=5)</code>	integrate an ODE with BDF method

MATLAB	NumPy	Notes
<code>ndims(a)</code>	<code>ndim(a)</code> or <code>a.ndim</code>	get the number of dimensions of an array
<code>numel(a)</code>	<code>size(a)</code> or <code>a.size</code>	get the number of elements of an array
<code>size(a)</code>	<code>shape(a)</code> or <code>a.shape</code>	get the "size" of the matrix
<code>size(a,n)</code>	<code>a.shape[n-1]</code>	get the number of elements of the n-th dimension of array <code>a</code> . (Note that MATLAB® uses 1 based indexing while Python uses 0 based indexing, See note INDEXING)
<code>[1 2 3; 4 5 6]</code>	<code>array([[1.,2.,3.], [4.,5.,6.]])</code>	2x3 matrix literal
<code>[a b; c d]</code>	<code>vstack([hstack([a,b]), hstack([c,d])])</code> or <code>bmat('a b; c d').A</code>	construct a matrix from blocks <code>a</code> , <code>b</code> , <code>c</code> , and <code>d</code>
<code>a(end)</code>	<code>a[-1]</code>	access last element in the 1xn matrix <code>a</code>
<code>a(2,5)</code>	<code>a[1,4]</code>	access element in second row, fifth column
<code>a(2,:)</code>	<code>a[1]</code> or <code>a[1,:]</code>	entire second row of <code>a</code>
<code>a(1:5,:)</code>	<code>a[0:5]</code> or <code>a[:5]</code> or <code>a[0:5,:]</code>	the first five rows of <code>a</code>
<code>a(end-4:end,:)</code>	<code>a[-5:]</code>	the last five rows of <code>a</code>
<code>a(1:3,5:9)</code>	<code>a[0:3][:,4:9]</code>	rows one to three and columns five to nine of <code>a</code> . This gives read-only access.
<code>a([2,4,5],[1,3])</code>	<code>a[ix_([1,3,4],[0,2])]</code>	rows 2,4 and 5 and columns 1 and 3. This allows the matrix to be modified, and doesn't require a regular slice.
<code>a(3:2:21,:)</code>	<code>a[2:21:2,:]</code>	every other row of <code>a</code> , starting with the third and going to the twenty-first
<code>a(1:2:end,:)</code>	<code>a[::2,:]</code>	every other row of <code>a</code> , starting with the first
<code>a(end:-1:1,:)</code> or <code>flipud(a)</code>	<code>a[::-1,:]</code>	<code>a</code> with rows in reverse order
<code>a([1:end 1],:)</code>	<code>a[r_[:len(a),0]]</code>	<code>a</code> with copy of the first row appended to the end

<code>a.'</code>	<code>a.transpose()</code> or <code>a.T</code>	transpose of <code>a</code>
<code>a'</code>	<code>a.conj().transpose()</code> or <code>a.conj().T</code>	conjugate transpose of <code>a</code>
<code>a * b</code>	<code>a.dot(b)</code>	matrix multiply
<code>a .* b</code>	<code>a * b</code>	element-wise multiply
<code>a./b</code>	<code>a/b</code>	element-wise divide
<code>a.^3</code>	<code>a**3</code>	element-wise exponentiation
<code>(a>0.5)</code>	<code>(a>0.5)</code>	matrix whose i,jth element is <code>(a_ij > 0.5)</code> . The Matlab result is an array of 0s and 1s. The NumPy result is an array of the boolean values <code>False</code> and <code>True</code> .
<code>find(a>0.5)</code>	<code>nonzero(a>0.5)</code>	find the indices where <code>(a > 0.5)</code>
<code>a(:,find(v>0.5))</code>	<code>a[:,nonzero(v>0.5)[0]]</code>	extract the columns of <code>a</code> where vector <code>v > 0.5</code>
<code>a(:,find(v>0.5))</code>	<code>a[:,v.T>0.5]</code>	extract the columns of <code>a</code> where column vector <code>v > 0.5</code>
<code>a(a<0.5)=0</code>	<code>a[a<0.5]=0</code>	<code>a</code> with elements less than 0.5 zeroed out
<code>a .* (a>0.5)</code>	<code>a * (a>0.5)</code>	<code>a</code> with elements less than 0.5 zeroed out
<code>a(:) = 3</code>	<code>a[:] = 3</code>	set all values to the same scalar value
<code>y=x</code>	<code>y = x.copy()</code>	numpy assigns by reference
<code>y=x(2,:)</code>	<code>y = x[1,:].copy()</code>	numpy slices are by reference
<code>y=x(:)</code>	<code>y = x.flatten()</code>	turn array into vector (note that this forces a copy)
<code>1:10</code>	<code>arange(1.,11.)</code> or <code>r_[1.:11.]</code> or <code>r_[1:10:10j]</code>	create an increasing vector (see note RANGES)
<code>0:9</code>	<code>arange(10.)</code> or <code>r_[:10.]</code> or <code>r_[:9:10j]</code>	create an increasing vector (see note RANGES)
<code>[1:10]'</code>	<code>arange(1.,11.)[:, newaxis]</code>	create a column vector
<code>zeros(3,4)</code>	<code>zeros((3,4))</code>	3x4 two-dimensional array full of 64-bit floating point zeros

<code>zeros(3,4,5)</code>	<code>zeros((3,4,5))</code>	3x4x5 three-dimensional array full of 64-bit floating point zeros
<code>ones(3,4)</code>	<code>ones((3,4))</code>	3x4 two-dimensional array full of 64-bit floating point ones
<code>eye(3)</code>	<code>eye(3)</code>	3x3 identity matrix
<code>diag(a)</code>	<code>diag(a)</code>	vector of diagonal elements of <code>a</code>
<code>diag(a,0)</code>	<code>diag(a,0)</code>	square diagonal matrix whose nonzero values are the elements of <code>a</code>
<code>rand(3,4)</code>	<code>random.rand(3,4)</code>	random 3x4 matrix
<code>linspace(1,3,4)</code>	<code>linspace(1,3,4)</code>	4 equally spaced samples between 1 and 3, inclusive
<code>[x,y]=meshgrid(0:8,0:5)</code>	<code>mgrid[0:9.,0:6.]</code> or <code>meshgrid(r_[0:9.],r_[0:6.])</code>	two 2D arrays: one of x values, the other of y values
	<code>ogrid[0:9.,0:6.]</code> or <code>ix_(r_[0:9.],r_[0:6.])</code>	the best way to eval functions on a grid
<code>[x,y]=meshgrid([1,2,4],[2,4,5])</code>	<code>meshgrid([1,2,4],[2,4,5])</code>	
	<code>ix_([1,2,4],[2,4,5])</code>	the best way to eval functions on a grid
<code>repmat(a, m, n)</code>	<code>tile(a, (m, n))</code>	create m by n copies of <code>a</code>
<code>[a b]</code>	<code>concatenate((a,b),1)</code> or <code>hstack((a,b))</code> or <code>column_stack((a,b))</code> or <code>c_[a,b]</code>	concatenate columns of <code>a</code> and <code>b</code>
<code>[a; b]</code>	<code>concatenate((a,b))</code> or <code>vstack((a,b))</code> or <code>r_[a,b]</code>	concatenate rows of <code>a</code> and <code>b</code>
<code>max(max(a))</code>	<code>a.max()</code>	maximum element of <code>a</code> (with <code>ndims(a)<=2</code> for matlab)
<code>max(a)</code>	<code>a.max(0)</code>	maximum element of each column of matrix <code>a</code>
<code>max(a,[],2)</code>	<code>a.max(1)</code>	maximum element of each row of matrix <code>a</code>
<code>max(a,b)</code>	<code>maximum(a, b)</code>	compares <code>a</code> and <code>b</code> element-wise, and returns the maximum value from each pair

<code>norm(v)</code>	<code>sqrt(dot(v,v))</code> or <code>np.linalg.norm(v)</code>	L2 norm of vector <code>v</code>
<code>a & b</code>	<code>logical_and(a,b)</code>	element-by-element AND operator (NumPy ufunc) See note LOGICOPS
<code>a b</code>	<code>logical_or(a,b)</code>	element-by-element OR operator (NumPy ufunc) See note LOGICOPS
<code>bitand(a,b)</code>	<code>a & b</code>	bitwise AND operator (Python native and NumPy ufunc)
<code>bitor(a,b)</code>	<code>a b</code>	bitwise OR operator (Python native and NumPy ufunc)
<code>inv(a)</code>	<code>linalg.inv(a)</code>	inverse of square matrix <code>a</code>
<code>pinv(a)</code>	<code>linalg.pinv(a)</code>	pseudo-inverse of matrix <code>a</code>
<code>rank(a)</code>	<code>linalg.matrix_rank(a)</code>	matrix rank of a 2D array / matrix <code>a</code>
<code>a\b</code>	<code>linalg.solve(a,b)</code> if <code>a</code> is square; <code>linalg.lstsq(a,b)</code> otherwise	solution of $a x = b$ for x
<code>b/a</code>	Solve $a.T x.T = b.T$ instead	solution of $x a = b$ for x
<code>[U,S,V]=svd(a)</code>	<code>U, S, Vh = linalg.svd(a), V = Vh.T</code>	singular value decomposition of <code>a</code>
<code>chol(a)</code>	<code>linalg.cholesky(a).T</code>	cholesky factorization of a matrix (<code>chol(a)</code> in matlab returns an upper triangular matrix, but <code>linalg.cholesky(a)</code> returns a lower triangular matrix)
<code>[V,D]=eig(a)</code>	<code>D,V = linalg.eig(a)</code>	eigenvalues and eigenvectors of <code>a</code>
<code>[V,D]=eig(a,b)</code>	<code>V,D = np.linalg.eig(a,b)</code>	eigenvalues and eigenvectors of <code>a, b</code>
<code>[V,D]=eigs(a,k)</code>		find the <code>k</code> largest eigenvalues and eigenvectors of <code>a</code>
<code>[Q,R,P]=qr(a,0)</code>	<code>Q,R = scipy.linalg.qr(a)</code>	QR decomposition
<code>[L,U,P]=lu(a)</code>	<code>L,U = scipy.linalg.lu(a)</code> or <code>LU,P=scipy.linalg.lu_factor(a)</code>	LU decomposition (note: <code>P(Matlab) == transpose(P(numpy))</code>)

<code>conjgrad</code>	<code>scipy.sparse.linalg.cg</code>	Conjugate gradients solver
<code>fft(a)</code>	<code>fft(a)</code>	Fourier transform of <code>a</code>
<code>ifft(a)</code>	<code>ifft(a)</code>	inverse Fourier transform of <code>a</code>
<code>sort(a)</code>	<code>sort(a)</code> or <code>a.sort()</code>	sort the matrix
<code>[b,I] = sortrows(a,i)</code>	<code>I = argsort(a[:,i]), b=a[I,:]</code>	sort the rows of the matrix
<code>regress(y,X)</code>	<code>linalg.lstsq(X,y)</code>	multilinear regression
<code>decimate(x, q)</code>	<code>scipy.signal.resample(x, len(x)/q)</code>	downsample with low-pass filtering
<code>unique(a)</code>	<code>unique(a)</code>	
<code>squeeze(a)</code>	<code>a.squeeze()</code>	

Table Reference:

<https://docs.scipy.org/doc/numpy-dev/user/numpy-for-matlab-users.html>

How to debug?

- Google it!
 - StackOverflow
 - Scipy.Org