### MATLAB-Python-Julia cheatsheet

- MATLAB-Python-Julia cheatsheet
  - Dependencies and Setup
  - Creating Vectors
  - Creating Matrices
  - Manipulating Vectors and Matrices
  - Accessing Vector/Matrix Elements
  - Mathematical Operations
  - Sum / max / min
  - Programming

### Dependencies and Setup

In the Python code we assume that you have already run import numpy as np
In the Julia, we assume you are using **v1.0.2 or later** with Compat **v1.3.0 or later** and have run using LinearAlgebra,
Statistics, Compat

### **Creating Vectors**

Operation	MATLAB	Python	Julia
Row vector: size (1, n)	$A = [1 \ 2 \ 3]$	A = np.array([1, 2, 3]).reshape(1, 3)	A = [1 2 3]

Operation	MATLAB	Python	Julia
Column vector: size (n, 1)	A = [1; 2; 3]	A = np.array([1, 2, 3]).reshape(3, 1)	A = [1 2 3]'
			A = [1; 2; 3]
1d array: size (n, )	Not possible	A = np.array([1, 2, 3])	or
			A = [1, 2, 3]
Integers from j to n with step size k	A = j:k:n	A = np.arange(j, n+1, k)	A = j:k:n
Linearly spaced vector of k points	A = linspace(1, 5,	A = np.linspace(1, 5, k)	A = range(1, 5,
Emeany spaced vector of k points	k)	inp. i i inopace (i , j , k)	length = k)

<b>Creating Matrices</b>			
Operation	MATLAB	Python	Julia
Create a matrix	$A = [1 \ 2; \ 3 \ 4]$	A = np.array([[1, 2], [3, 4]])	$A = [1 \ 2; \ 3 \ 4]$
2 x 2 matrix of zeros	A = zeros(2, 2)	A = np.zeros((2, 2))	A = zeros(2, 2)
2 x 2 matrix of ones	A = ones(2, 2)	A = np.ones((2, 2))	A = ones(2, 2)
			A = I # will adopt # 2x2 dims if
2 x 2 identity matrix	A = eye(2, 2)	A = np.eye(2)	demanded by # neighboring matrices

Operation	MATLAB	Python	Julia
Diagonal matrix	A = diag([1 2 3])	A = np.diag([1, 2, 3])	A = Diagonal([1, 2, 3])
Uniform random numbers	A = rand(2, 2)	A = np.random.rand(2, 2)	A = rand(2, 2)
Normal random numbers	A = randn(2, 2)	A = np.random.randn(2, 2)	A = randn(2, 2)
Sparse Matrices	A = sparse(2, 2) A(1, 2) = 4 A(2, 2) = 1	<pre>from scipy.sparse import coo_matrix  A = coo_matrix(([4, 1],</pre>	<pre>using SparseArrays A = spzeros(2, 2) A[1, 2] = 4 A[2, 2] = 1</pre>
Tridiagonal Matrices		<pre>import sp.sparse as sp diagonals = [[4, 5, 6, 7], [1, 2, 3], [8, 9, 10]] sp.diags(diagonals, [0, -1, 2]).toarray()</pre>	x = [1, 2, 3] y = [4, 5, 6, 7] z = [8, 9, 10] Tridiagonal(x, y, z)

Manipulating Vectors and Matrices			
Operation	MATLAB	Python	Julia
Transpose	A.'	A.T	transpose(A)
Complex conjugate transpose (Adjoint)	A '	A.conj()	A'
Concatenate horizontally	$A = [[1 \ 2] \ [1 \ 2]]$	B = np.array([1, 2	$A = [[1 \ 2] \ [1 \ 2]]$

Operation	MATLAB	Python	Julia
	or	A = np.hstack((B,	or
	$A = horzcat([1 \ 2], [1$	B))	A = hcat([1 2],
	2])		[1 2])
	$A = [[1 \ 2]; [1 \ 2]]$	B = np.array([1, 2])	A = [[1 2]; [1 2]]
Concatenate vertically	or	A = np.vstack((B,	or
oonoutonate vertically	A = vertcat([1 2], [1		
	2])		A = vcat([1 2], [1 2])
Reshape (to 5 rows, 2 columns)	A = reshape(1:10, 5, 2)	A = A.reshape(5, 2)	A = reshape(1:10, 5, 2)
Convert matrix to vector	A(:)	A = A.flatten()	A[:]
Flip left/right	fliplr(A)	np.fliplr(A)	<pre>reverse(A, dims = 2)</pre>
Flip up/down	flipud(A)	np.flipud(A)	<pre>reverse(A, dims = 1)</pre>
Repeat matrix (3 times in the row dimension, 4 times in the column dimension)	repmat(A, 3, 4)	np.tile(A, (4, 3))	repeat(A, 3, 4)
Preallocating/Similar	<pre>x = rand(10) y = zeros(size(x, 1), size(x, 2)) N/A similar type</pre>	<pre>np.random.rand(3, 3) y = np.empty_like(x)  # new dims y = np.empty((2, 3))</pre>	<pre># new dims y = similar(x) y = similar(x, 2,</pre>

Operation	MATLAB	Python	Julia
Broadcast a function over a	$f = @(x) x.^2$ g = @(x, y) x + 2 + $y.^2$ x = 1:10	<pre>def f(x):     return x**2 def g(x, y):     return x + 2 +     y**2 x = np.arange(1, 10, 1)</pre>	$f(x) = x^2$ g(x, y) = x + 2 + 3
collection/matrix/vector	y = 2:11 f(x) g(x, y) Functions broadcast directly	<pre>y = np.arange(2, 11, 1) f(x)</pre>	y = 2:11 f. (x) g. (x, y)
		Functions broadcast directly	

# Accessing Vector/Matrix Elements

Operation	MATLAB	Python	Julia
Access one element	A(2, 2)	A[1, 1]	A[2, 2]
Access specific rows	A(1:4, :)	A[0:4, :]	A[1:4, :]
Access specific column	SA(:, 1:4)	A[:, 0:4]	A[:, 1:4]
Remove a row	A([1 2 4], :)	A[[0, 1, 3], :]	A[[1, 2, 4], :]
Diagonals of matrix	diag(A)	np.diag(A)	diag(A)
Get dimensions of matri	x[nrow ncol] = size(A	$\Lambda$ ) nrow, ncol = np.shape(A	$\Delta$ ) nrow, ncol = size(A)

Mathematical Operations			
Operation	MATLAB	Python	Julia dot(A, B)
Dot product	dot(A, B)	np.dot(A, B) or A @ B	A · B # \cdot <tab></tab>
Matrix multiplication	A * B	A @ B	A * B
Inplace matrix multiplication	Not possible	<pre>x = np.array([1, 2]).reshape(2, 1) A = np.array(([1, 2], [3, 4])) y = np.empty_like(x) np.matmul(A, x, y)</pre>	x = [1, 2] A = [1 2; 3 4] y = similar(x) mul!(y, A, x)
Element-wise multiplication	А.* В	A * B	А.* В
Matrix to a power	A^2	np.linalg.matrix_power(A, 2)	A^2
Matrix to a power, elementwise	A.^2	A**2	A.^2
	inv(A)		inv(A)
Inverse	or	np.linalg.inv(A)	or
	A^(-1)		A^(-1)
Determinant	det(A)	np.linalg.det(A)	det(A)
Eigenvalues and eigenvectors	<pre>[vec, val] = eig(A)</pre>	<pre>val, vec = np.linalg.eig(A)</pre>	val, vec = eigen(A)

## Operation

Euclidean norm

Solve linear system  $\(Ax=b\)$  (when  $\(A\)$  is square) Solve least squares problem  $\(Ax=b\)$  (when  $\(A\)$  is rectangular)

MATLAB	Python	Julia
norm(A)	np.linalg.norm(A)	norm(A)
A\b	np.linalg.solve(A, b)	A\b
A\b	np.linalg.lstsq(A, b)	A\b

Sum / max / min			
Operation	MATLAB	Python	Julia
Sum / max / min of each column	<pre>sum(A, 1) max(A, [], 1) min(A, [], 1)</pre>	<pre>sum(A, 0) np.amax(A, 0) np.amin(A, 0)</pre>	<pre>sum(A, dims = 1) maximum(A, dims = 1) minimum(A, dims = 1)</pre>
Sum / max / min of each row		<pre>sum(A, 1) np.amax(A, 1) np.amin(A, 1)</pre>	<pre>sum(A, dims = 2) maximum(A, dims = 2) minimum(A, dims = 2)</pre>
Sum / max / min of entire matrix		np.sum(A) np.amax(A) np.amin(A)	<pre>sum(A) maximum(A) minimum(A)</pre>
Cumulative sum / max / min by row	<pre>cumsum(A, 1) cummax(A, 1)</pre>	np.cumsum(A, 0)	cumsum(A, dims = 1)

Operation	MATLAB	Python	Julia
	cummin(A, 1)	np.maximum.accumulate(A,	<pre>accumulate(max, A, dims = 1)</pre>
		np.minimum.accumulate(A, 0)	accumulate(min, A, dims = 1)
		np.cumsum(A, 1)	cumsum(A, dims = 2)
Cumulative sum / max / min by	<pre>cumsum(A, 2) cummax(A, 2)</pre>	np.maximum.accumulate(A, 1)	<pre>accumulate(max, A, dims = 2)</pre>
column	cummin(A, 2)	np.minimum.accumulate(A,	<pre>accumulate(min, A, dims = 2)</pre>

Programming						
Operation	MATLAB	Python	Julia			
Comment one line	% This is a comment	# This is a comment	# This is a comment			
Comment block	<pre>%{ Comment block %}</pre>	<pre># Block # comment # following PEP8</pre>	#= Comment block =#			
For loop	<pre>for i = 1:N</pre>	<pre>for i in range(n):     # do something</pre>	<pre>for i in 1:N     # do something end</pre>			
While loop	<pre>while i &lt;= N</pre>	<pre>while i &lt;= N:     # do something</pre>	<pre>while i &lt;= N     # do something end</pre>			

Operation	MATLAB	Python	Julia
If	<pre>if i &lt;= N</pre>	<pre>if i &lt;= N:     # do something</pre>	<pre>if i &lt;= N     # do something end</pre>
If / else	<pre>if i &lt;= N</pre>	<pre>if i &lt;= N:     # do something else:     # so something else</pre>	<pre>if i &lt;= N     # do something else     # do something else end</pre>
Print text and variable	$x = 10$ fprintf('x = %d \n', x)	x = 10 print(f'x = {x}')	x = 10 println(" $x = $x$ ")
Function: anonymous	$f = @(x) x^2$	f = lambda x: x**2	f = x -> x^2 # can be rebound
Function	<pre>function out = f(x)   out = x^2 end</pre>	<pre>def f(x):     return x**2</pre>	<pre>function f(x)     return x^2 end  f(x) = x^2 # not anon!</pre>
Tuples	<pre>t = {1 2.0 "test"} t{1}  Can use cells but watch performance</pre>	t = (1, 2.0, "test") t[0]	t = (1, 2.0, "test") t[1]

Operation	MATLAB	Python	Julia
			# $vanilla$ m = $(x = 1, y = 2)$
Named Tuples/ Anonymous Structures		from collections import namedtuple	$m \cdot (x - 1), y - 2$
	m.x = 1 $m.y = 2$	<pre>mdef = namedtuple('m', 'x y') m = mdef(1, 2)</pre>	<pre># constructor using Parameters mdef = @with_kw</pre>
	111 • 22	m.x	<pre>(x=1, y=2) m = mdef() # same as _above m = mdef(x = 3)</pre>
		0.0	m macr(x 3)
Closures	a = 2.0 f = @(x) a + x f(1.0)	<pre>a = 2.0 def f(x):     return a + x f(1.0)</pre>	a = 2.0 f(x) = a + x f(1.0)
Inplace Modification	No consistent or simple syntax to achieve this	<b>def</b> f(x): x **=2	<pre>function f! (out, x) out .= x.^2</pre>
		return	<b>end</b> x = rand(10)
		x = np.random.rand(10) f(x)	y = similar(x) f!(y, x)

**Credits** This cheat sheet was created by Victoria Gregory, Andrij Stachurski, Natasha Watkins and other collaborators on behalf of QuantEcon.

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