



Cheatsheets (/)

by QuantEcon

[Numerical \(index.html\)](#)[Python \(python-cheatsheet.html\)](#)[Julia \(julia-cheatsheet.html\)](#)[Statistics \(stats-cheatsheet.html\)](#)[\(https://github.com/QuantEcon/QuantEcon.cheatsheet\)](https://github.com/QuantEcon/QuantEcon.cheatsheet)

MATLAB–Python–Julia cheatsheet

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

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]`

Column vector: size (n, 1)

`A = [1; 2; 3]``A = np.array([1, 2, 3]).reshape(3, 1)``A = [1 2 3]'`

1d array: size (n,)

Not possible

`A = np.array([1, 2, 3])``A = [1; 2; 3]`

or

`A = [1, 2, 3]`

MATLAB

PYTHON

JULIA

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, k)
```

```
A = np.linspace(1, 5, k)
```

```
A = range(1, 5,  
length = k)
```

Creating Matrices

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)
```

2 x 2 identity matrix

```
A = eye(2, 2)
```

```
A = np.eye(2)
```

```
A = I # will adopt  
# 2x2 dims if demanded  
# by  
# neighboring matrices
```

Diagonal matrix

```
A = diag([1 2 3])
```

```
A = np.diag([1, 2, 3])
```

```
A = Diagonal([1, 2,  
3])
```

MATLAB

PYTHON

JULIA

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
```

```
from scipy.sparse import  
coo_matrix  
  
A = coo_matrix(([4, 1],  
([0, 1],  
[1, 1])),  
  
shape=(2, 2))
```

```
using SparseArrays  
A = spzeros(2, 2)  
A[1, 2] = 4  
A[2, 2] = 1
```

Tridiagonal Matrices

```
A = [1 2 3 NaN;  
     4 5 6 7;  
     NaN 8 9 0]  
spdiags(A',[ -1 0 1], 4,  
4)
```

```
import sp.sparse as sp  
diagonals = [[4, 5, 6,  
7], [1, 2, 3], [8, 9,  
10]]  
sp.diags(diagonals, [0,  
-1, 2]).toarray()
```

```
x = [1, 2, 3]  
y = [4, 5, 6, 7]  
z = [8, 9, 10]  
Tridiagonal(x, y, z)
```

Manipulating Vectors and Matrices¶

MATLAB

PYTHON

JULIA

Transpose

```
A.'
```

```
A.T
```

```
transpose(A)
```

MATLAB

PYTHON

JULIA

Complex conjugate transpose (Adjoint)

A'

A.conj()

A'

Concatenate horizontally

A = [[1 2] [1 2]]

or

A = horzcat([1 2], [1 2])

B = np.array([1, 2])

A = np.hstack((B, B))

A = [[1 2] [1 2]]

or

A = hcat([1 2], [1 2])

Concatenate vertically

A = [[1 2]; [1 2]]

or

A = vertcat([1 2], [1 2])

B = np.array([1, 2])

A = np.vstack((B, B))

A = [[1 2]; [1 2]]

or

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)

reverse(A, dims = 2)

Flip up/down

flipud(A)

np.flipud(A)

reverse(A, dims = 1)

MATLAB

PYTHON

JULIA

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

```
x = rand(10)
y = zeros(size(x, 1),
           size(x, 2))
```

N/A similar type

```
x = np.random.rand(3, 3)
y = np.empty_like(x)

# new dims
y = np.empty((2, 3))
```

```
x = rand(3, 3)
y = similar(x)
# new dims
y = similar(x, 2, 2)
```

Broadcast a function over a collection/matrix/vector

```
f = @(x) x.^2
g = @(x, y) x + 2 + y.^2
x = 1:10
y = 2:11
f(x)
g(x, y)
```

Functions broadcast directly

```
def f(x):
    return x**2
def g(x, y):
    return x + 2 + y**2
x = np.arange(1, 10, 1)
y = np.arange(2, 11, 1)
f(x)
g(x, y)
```

Functions broadcast directly

```
f(x) = x^2
g(x, y) = x + 2 + y^2
x = 1:10
y = 2:11
f.(x)
g.(x, y)
```

Accessing Vector/Matrix Elements¶

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, :]

MATLAB

PYTHON

JULIA

Access specific columns

A(:, 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 matrix

[nrow ncol] = size(A)

nrow, ncol = np.shape(A)

nrow, ncol = size(A)

Mathematical Operations

MATLAB

PYTHON

JULIA

Dot product

dot(A, B)

np.dot(A, B) or A @ B

dot(A, B)

A · B # \cdot <TAB>

Matrix multiplication

A * B

A @ B

A * B

Inplace matrix multiplication

Not possible

```
x = [1, 2]
A = [1 2; 3 4]
y = similar(x)
mul!(y, A, x)
```

MATLAB**Element-wise multiplication**

A .* B

PYTHON

```
x = np.array([1,
2]).reshape(2, 1)
A = np.array(([1, 2],
[3, 4]))
y = np.empty_like(x)
Ap.matmul(A, x, y)
```

JULIA

A .* B

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

Inverse

inv(A)

np.linalg.inv(A)

inv(A)

or

or

A^(-1)

A^(-1)

Determinant

det(A)

np.linalg.det(A)

det(A)

Eigenvalues and eigenvectors

[vec, val] = eig(A)

```
val, vec =
np.linalg.eig(A)
```

val, vec = eigen(A)

Euclidean norm

norm(A)

np.linalg.norm(A)

norm(A)

Solve linear system $Ax = b$ (when A is square)

A\b

np.linalg.solve(A, b)

A\b

MATLAB

PYTHON

JULIA

Solve least squares problem $Ax = b$ (when A is rectangular)

`A\b`

`np.linalg.lstsq(A, b)`

`A\b`

Sum / max / min

MATLAB

PYTHON

JULIA

Sum / max / min of each column

`sum(A, 1)`
`max(A, [], 1)`
`min(A, [], 1)`

`sum(A, 0)`
`np.amax(A, 0)`
`np.amin(A, 0)`

`sum(A, dims = 1)`
`maximum(A, dims = 1)`
`minimum(A, dims = 1)`

Sum / max / min of each row

`sum(A, 2)`
`max(A, [], 2)`
`min(A, [], 2)`

`sum(A, 1)`
`np.amax(A, 1)`
`np.amin(A, 1)`

`sum(A, dims = 2)`
`maximum(A, dims = 2)`
`minimum(A, dims = 2)`

Sum / max / min of entire matrix

`sum(A(:))`
`max(A(:))`
`min(A(:))`

`np.sum(A)`
`np.amax(A)`
`np.amin(A)`

`sum(A)`
`maximum(A)`
`minimum(A)`

Cumulative sum / max / min by row

`cumsum(A, 1)`
`cummax(A, 1)`
`cummin(A, 1)`

`np.cumsum(A, 0)`
`np.maximum.accumulate(A, 0)`
`np.minimum.accumulate(A, 0)`

`cumsum(A, dims = 1)`
`accumulate(max, A, dims = 1)`
`accumulate(min, A, dims = 1)`

Cumulative sum / max / min by column

`cumsum(A, 2)`
`cummax(A, 2)`
`cummin(A, 2)`

MATLAB**PYTHON****JULIA****Programming**

```
np.cumsum(A, 1)
np.maximum.accumulate(A,
1)
np.minimum.accumulate(A,
1)
```

```
cumsum(A, dims = 2)
accumulate(max, A, dims
= 2)
accumulate(min, A, dims
= 2)
```

MATLAB**PYTHON****JULIA****Comment one line**

```
% This is a comment
```

```
# This is a comment
```

```
# This is a comment
```

Comment block

```
%{
Comment block
%}
```

```
# Block
# comment
# following PEP8
```

```
#=#
Comment block
=#
```

For loop

```
for i = 1:N
    % do something
end
```

```
for i in range(n):
    # do something
```

```
for i in 1:N
    # do something
end
```

While loop

```
while i <= N
    % do something
end
```

```
while i <= N:
    # do something
```

```
while i <= N
    # do something
end
```

If

```
if i <= N
    % do something
end
```

```
if i <= N:
    # do something
```

```
if i <= N
    # do something
end
```

If / else

MATLAB

```
if i <= N
    % do something
else
    % do something else
```

Print text and variable**PYTHON**

```
if i <= N:
    # do something
else:
    # do something else
```

JULIA

```
if i <= N
    # do something
else
    # do something else
end
```

```
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

```
function out = f(x)
    out = x^2
end
```

```
def f(x):
    return x**2
```

```
function f(x)
    return x^2
end

f(x) = x^2 # not anon!
```

Tuples

```
t = {1 2.0 "test"}
t[1]
```

```
t = (1, 2.0, "test")
t[0]
```

```
t = (1, 2.0, "test")
t[1]
```

Can use cells but watch
performance

Named Tuples/ Anonymous Structures

```
m.x = 1
m.y = 2

m.x
```

```
from collections import
namedtuple

mdef = namedtuple('m',
'x y')
m = mdef(1, 2)

m.x
```

```
# vanilla
m = (x = 1, y = 2)
m.x

# constructor
using Parameters
mdef = @with_kw (x=1,
y=2)
m = mdef() # same as
above
m = mdef(x = 3)
```

MATLAB**PYTHON****JULIA****Closures**

```
a = 2.0
f = @(x) a + x
f(1.0)
```

```
a = 2.0
def f(x):
    return a + x
f(1.0)
```

```
a = 2.0
f(x) = a + x
f(1.0)
```

Inplace Modification

No consistent or simple syntax to achieve this (<https://blogs.mathworks.com/loren/2007/03/22/in-place-operations-on-data/>)

```
def f(x):
    x **= 2
    return

x = np.random.rand(10)
f(x)
```

```
function f!(out, x)
    out .= x.^2
end

x = rand(10)
y = similar(x)
f!(y, x)
```

Credits

This cheat sheet was created by Victoria Gregory (<https://github.com/vgregory757>), Andrij Stachurski (<http://drdrij.com/>), Natasha Watkins (<https://github.com/natashawatkins>) and other collaborators on behalf of QuantEcon (<http://quantecon.org/>).

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