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In [1]: import numpy as np

In [2]: # Now we will look at some commonly used functions in numpy

In [3]: # Arrays
        # asarray() assigns data with same structure
        a = np.asarray([1,2,3], dtype=np.float32)
        print(a[1],a[2])

2.0 3.0

In [4]: # linspace() creates sequences
        a=np.linspace(0, 10, num=5)
        a

Out[4]: array([ 0. ,  2.5,  5. ,  7.5, 10. ])

In [5]: # Some basic numerical operations/functions
        np.mean(a)

Out[5]: 5.0

In [6]: np.var(a)

Out[6]: 12.5

In [7]: np.exp(1)

Out[7]: 2.7182818284590451

In [8]: np.sin(np.pi/2)

Out[8]: 1.0

In [9]: np.cumsum(a)

Out[9]: array([ 0. ,  2.5,  7.5, 15. , 25. ])

In [10]: # Linear algebra
         # To construct a matrix
         m=np.matrix([[1,2],[3,4]])
         m

Out[10]: matrix([[1, 2],
                [3, 4]])

In [11]: # Linear algebra is inside package linalg
         from numpy import linalg

In [12]: # Determinant of matrix
         linalg.det(m)

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Out[12]: -2.0000000000000004
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In [13]: # Invert of matrix  
         linalg.inv(m)
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Out[13]: matrix([[ -2. ,  1. ],  
                 [ 1.5, -0.5]])
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In [14]: # Matrix production  
         linalg.inv(m)*m  
         # notice that different from R, * here is not element-by-element production
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Out[14]: matrix([[ 1.00000000e+00,  4.44089210e-16],  
                 [ 0.00000000e+00,  1.00000000e+00]])
```

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In [15]: # Element-by-element multiplication  
         np.multiply(linalg.inv(m),m)
```

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Out[15]: matrix([[ -2. ,  2. ],  
                 [ 4.5, -2. ]])
```

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In [16]: # eigenvalues and eigenvectors  
         linalg.eig(m)
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Out[16]: (array([-0.37228132,  5.37228132]), matrix([[ -0.82456484, -0.41597356],  
                                                    [ 0.56576746, -0.90937671]]))
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