#### Resources

```
M. Scott Shell,
http://www.engr.ucsb.edu/%7eshell/che210d/numpy.pdf
```

http://wiki.scipy.org/NumPy\_for\_Matlab\_Users

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http://mathesaurus.sourceforge.net/matlab-numpy.html

## Numpy arrays

- import numpy as np
- Numpy provides class ndarray, called "array"
- Create array from a list

```
>>> x = np.array([3.0,5,7,5])
>>> x
array([ 3., 5., 7., 5.])
```

- If appear to be integers in list, need "float"
- 2D arrays

```
>>> A = np.array([[8.,1.,6.],[3.,5.,7.],[4.,9.,2.]])
array([[ 8.,   1.,  6.],
       [ 3.,  5.,  7.],
       [ 4.,  9.,  2.]])
```

- Use brackets to denote subscripts
- Start counting at 0

```
>>> x[0]
3.0
>>> A[1,2]
7.0
```

Colons work, be careful of last value!

```
>>> x[0:1]
array([ 3.0])
```

- Use brackets to denote subscripts
- Start counting at 0

```
>>> x[0]
3.0
>>> A[1,2]
7.0
```

Colons work, be careful of last value!

```
>>> x[0:1]
array([ 3.0])
>>> x[0:2]
array([ 3.0, 5.])
```

- Use brackets to denote subscripts
- Start counting at 0

```
>>> x[0]
3.0
>>> A[1,2]
7.0
```

Colons work, be careful of last value!

```
>>> x[0:1]
array([ 3.0])
>>> x[0:2]
array([ 3.0, 5.])
>>> A[:,2]
array([ 6., 7., 2.])
```

- Use brackets to denote subscripts
- Start counting at 0

```
>>> x[0]
3.0
>>> A[1,2]
7.0
```

Colons work, be careful of last value!

```
>>> x[0:1]
array([ 3.0])
>>> x[0:2]
array([ 3.0, 5.])
>>> A[:,2]
array([ 6., 7., 2.])
```

Negative indices count from end

```
>>> x[-1] 5.0
```

#### **Attributes**

```
>>> A.shape
(3, 3)
>>> A.flatten()
array([ 8., 1., 6., 3., 5., 7., 4., 9., 2.])
>>> B=A.copy()
>>> B[1,1]=-1
>>> A[1,1]
5.0
>>> B[1,1]
-1.0
>>> A.transpose()
array([[ 8., 3., 4.],
       [ 1., 5., 9.],
       [6., 7., 2.]])
```

#### Methods

```
>>> x=np.arange(24)
                                  # array-range
>>> y=x.reshape([4,6]).copy()
                                  # turn into 4 X 6 matrix
>>> v
array([[ 0, 1, 2, 3, 4, 5],
       [6, 7, 8, 9, 10, 11],
       [12, 13, 14, 15, 16, 17],
       [18, 19, 20, 21, 22, 23]])
>>> np.sum(y)
                                  # sum all of v
276
>>> y.sum()
                                  # sum all of y
276
>>> y.sum(0)
                                  # sum columns
array([36, 40, 44, 48, 52, 56])
>>> v.sum(axis=0)
                                  # sum columns
array([36, 40, 44, 48, 52, 56])
>>> np.sum(y,axis=0)
                                  # sum columns
array([36, 40, 44, 48, 52, 56])
>>> np.sum(y[:,0])
                                  # sum only first column
36
>>> np.sum(v[:,1])
                                  # sum only second column
40
                                  # sum along rows
>>> y.sum(1)
array([ 15, 51, 87, 123])
>>> v.sum(axis=1)
                                  # sum along rows
array([ 15, 51, 87, 123])
>>> np.sum(y[0,:])
                                  # sum first row
15
```

### Useful attributes

- ightharpoonup  $\pi = np.pi, e = np.e$
- 1d array of zeros x=np.zeros (N)
- 2d array of zeros x=np.zeros([N,M])
- Array of zeros same size as: y=np.zeros\_like(x)
- np.ones, np.ones\_like
- np.empty, np.empty\_like

### Functions like MATLAB

- ▶ np.diag
- ▶ np.random.rand
- ▶ np.tril, np.triu

## Operations on arrays

Suppose **x** is the array  $[x_i]$ , **y** is the array  $[y_i]$ , **A** is the 2d array  $[a_{ij}]$  and **B** is the 2d array  $[b_{ij}]$ :

- All operations are elementwise.
- ▶ np.sin(x) is the array  $[sin(x_i)]$ , same for matrices
- ▶ 3\*x is the array  $[3x_i]$ , same for matrices
- **x+y** is the array  $[x_i + y_i]$ , same for matrices
- **x**\*y is the array  $[x_iy_i]$ , same for matrices
- **x/y** is the array  $[x_i/y_i]$ , same for matrices
- ▶ np.dot (x, y) =  $\sum_i x_i y_i$  (usual dot product)
- ▶ np.dot (A, B) =  $\left[\sum_{k} a_{ik} b_{kj}\right]$  (usual matrix multiplication)

## Array selection

It is sometimes convenient to select particular elements from arrays.

- np.max(x) finds the maximum value.
- np.argmax(x) finds the locaion of the maximum value.
- ► Can use axis=n
- Also np.min, np.argmin
- Boolean selection

```
>>> x=np.array([[5.,3.],[4.,9.]])
>>> x[x>=5]
array([ 5.,  9.])
>>> x[np.logical_and(x>=5,x<8)]
array([ 5.])</pre>
```

# Array selection by subscript

```
One-dimensional arrays
  >>> x=2*np.arange(9.)
 >>> x
  array([ 0., 2., 4., 6., 8., 10., 12., 14., 16.])
 >>> i=np.array([0,0,1,5,2])
 >>> i
 array([0, 0, 1, 5, 2])
 >>> x[i]
  array([ 0., 0., 2., 10., 4.])
Two-dimensional arrays: selector for each axis.
  >>> A=x.reshape([3,3])
  >>> A
 array([[ 0., 2., 4.],
         [ 6., 8., 10.],
         [ 12. , 14., 16.]])
  # construct minor associated with A[1,0]
  >>> i=[[0,0],[2,2]]  # 2-axis means result will be 2-axis
  >>> j=[[1,2],[1,2]]
  >>> A[i,j]
  array([[ 2., 4.],
         [ 14., 16.]])
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```

## Higher-level linear algebra

- import scipy.linalg as la
- Built on BLAS and Lapack
- ▶ la.norm: vector and matrix norm
- la.det: determinant
- la.solve: solve system of equations
- la.inv: construct inverse matrix
- la.eig: eigenvalues and eigenvectors
- la.eigvals: eigenvalues only
- fenics does not use this package!

## **Plotting**

- import matplotlib.pyplot as plt
- Must use plt.show() to see your plot!
- ▶ Plot like MATLAB

```
>>> x=np.linspace(0,4*np.pi,1000)
>>> plt.plot(x, np.exp(-x), x, np.sin(x))
[<matplotlib.lines.Line2D object at 0x4306a90>, <matr
>>> plt.legend(('exp','sin'))
<matplotlib.legend.Legend object at 0x45df890>
>>> plt.show()
```

plt.semilogy also works



