

# NumPy

# What is NumPy?

- Python is a fabulous language
  - Easy to extend
  - Great syntax which encourages easy to write and maintain code
  - Incredibly large standard-library and third-party tools
- **No built-in multi-dimensional array** (but it supports the needed syntax for extracting elements from one)
- NumPy provides a **fast** built-in object (***ndarray***) which is a multi-dimensional array of a homogeneous data-type.

# Overview of NumPy

## N-D ARRAY (NDARRAY)

- N-dimensional array of rectangular data
- Element of the array can be C-structure or simple data-type or object data-type.
- Fast algorithms on machine data-types (int, float, etc.)

# Introducing NumPy Arrays

## SIMPLE ARRAY CREATION

```
>>> a = array([0,1,2,3])
>>> a
array([0, 1, 2, 3])
```

## CHECKING THE TYPE

```
>>> type(a)
<type 'array'>
```

## NUMERIC 'TYPE' OF ELEMENTS

```
>>> a.dtype
dtype('int32')
```

## BYTES PER ELEMENT

```
>>> a.itemsize # per element
4
```

## ARRAY SHAPE

```
# shape returns a tuple
# listing the length of the
# array along each dimension.
>>> a.shape
(4,)
>>> shape(a)
(4,)
```

## ARRAY SIZE

```
# size reports the entire
# number of elements in an
# array.
>>> a.size
4
>>> size(a)
4
```

# Introducing NumPy Arrays

## BYTES OF MEMORY USED

```
# returns the number of bytes  
# used by the data portion of  
# the array.
```

```
>>> a.nbytes  
12
```

## NUMBER OF DIMENSIONS

```
>>> a.ndim  
1
```

## ARRAY COPY

```
# create a copy of the array  
>>> b = a.copy()  
>>> b  
array([0, 1, 2, 3])
```

## CONVERSION TO LIST

```
# convert a numpy array to a  
# python list.
```

```
>>> a.tolist()  
[0, 1, 2, 3]
```

```
# For 1D arrays, list also  
# works equivalently, but  
# is slower.
```

```
>>> list(a)  
[0, 1, 2, 3]
```

# Setting Array Elements

## ARRAY INDEXING

```
>>> a[0]
0
>>> a[0] = 10
>>> a
[10, 1, 2, 3]
```

## FILL

```
# set all values in an array.
>>> a.fill(0)
>>> a
[0, 0, 0, 0]

# This also works, but may
# be slower.
>>> a[:] = 1
>>> a
[1, 1, 1, 1]
```



## BEWARE OF TYPE COERSION

```
>>> a.dtype
dtype('int32')
```

```
# assigning a float to into
# an int32 array will
# truncate decimal part.
```

```
>>> a[0] = 10.6
>>> a
[10, 1, 2, 3]
```

```
# fill has the same behavior
>>> a.fill(-4.8)
>>> a
[-4, -4, -4, -4]
```

# Multi-Dimensional Arrays

## MULTI-DIMENSIONAL ARRAYS

```
>>> a = array([[ 0, 1, 2, 3],
               [10,11,12,13]])
```

```
>>> a
array([[ 0, 1, 2, 3],
       [10,11,12,13]])
```

## (ROWS,COLUMNS)

```
>>> a.shape
(2, 4)
>>> shape(a)
(2, 4)
```

## ELEMENT COUNT


```
>>> a.size
8
>>> size(a)
8
```

## NUMBER OF DIMENSIONS

```
>>> a.ndim
2
```

## GET/SET ELEMENTS

```
>>> a[1,3]
13
```



```
>>> a[1,3] = -1
>>> a
array([[ 0, 1, 2, 3],
       [10,11,12,-1]])
```

## ADDRESS FIRST ROW USING SINGLE INDEX

```
>>> a[1]
array([10, 11, 12, -1])
```

# Exercise: creating array “a”

CREATING ARRAY “a” AS FOLLOWS:

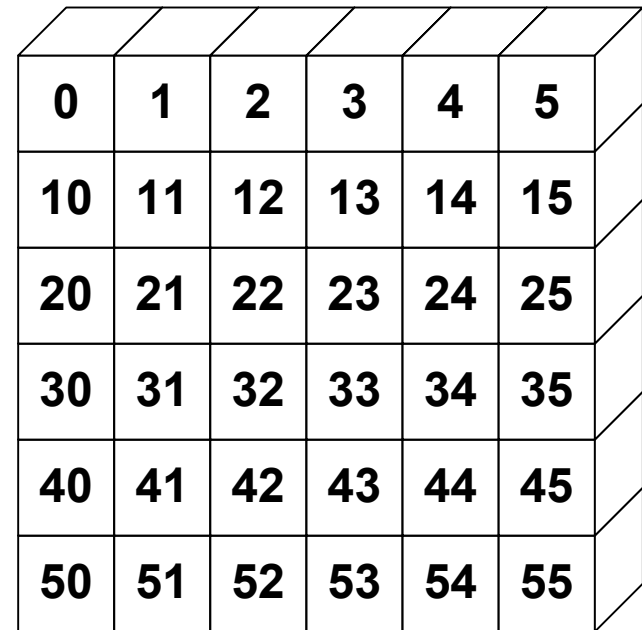
0	1	2	3	4	5
10	11	12	13	14	15
20	21	22	23	24	25
30	31	32	33	34	35
40	41	42	43	44	45
50	51	52	53	54	55



# Setting up Array “a”

## CREATING ARRAY “a” AS FOLLOWS:

```
>>> a = np.arange(0,6)
>>> a = a.repeat(6)
>>> a = a.reshape(6,6).T
>>> b = np.arange(0,6)*10
>>> a = (b+a.T).T
```



0	1	2	3	4	5
10	11	12	13	14	15
20	21	22	23	24	25
30	31	32	33	34	35
40	41	42	43	44	45
50	51	52	53	54	55

- OR -

## CREATING ARRAY “a” AS FOLLOWS:

0	1	2	3	4	5
10	11	12	13	14	15
20	21	22	23	24	25
30	31	32	33	34	35
40	41	42	43	44	45
50	51	52	53	54	55

```
a = np.ones((6,6),dtype='int32')
```

```
for row in range(6):  
    for col in range(6):  
        a[row,col] = row*10+col
```

# Array Slicing

**SLICING WORKS MUCH LIKE  
STANDARD PYTHON SLICING**

```
>>> a[0,3:5]  
array([3, 4])
```

```
>>> a[4:,4:]  
array([[44, 45],  
       [54, 55]])
```

```
>>> a[:,2]  
array([2, 12, 22, 32, 42, 52])
```

**STRIDES ARE ALSO POSSIBLE**

```
>>> a[2::2,::2]  
array([[20, 22, 24],  
       [40, 42, 44]])
```

0	1	2	3	4	5
10	11	12	13	14	15
20	21	22	23	24	25
30	31	32	33	34	35
40	41	42	43	44	45
50	51	52	53	54	55

# Slices Are References

Slices are references to memory in original array. Changing values in a slice also changes the original array. (*including strides*)

```
>>> a = array((0,1,2,3,4))
```

```
# create a slice containing only the  
# last element of a
```

```
>>> b = a[2:4]
```

```
>>> b[0] = 10
```

```
# changing b changed a!
```

```
>>> a
```

```
array([ 1,  2, 10,  3,  4])
```

# Quiz #1

What will be printed on the screen using the following code?

```
a = np.array((0,1,2,3))
b = a
b[2] = -20
b = b.reshape(2,2)
b[1,1] = -5
print a
print a.sum()
print '-----'
print b
print b.sum()
```

# Fancy Indexing in 2D

```
>>> a[(0,1,2,3,4),(1,2,3,4,5)]  
array([ 1, 12, 23, 34, 45])
```

```
>>> a[3:,[0, 2, 5]]  
array([[30, 32, 35],  
       [40, 42, 45]],  
       [50, 52, 55]])
```

```
>>> mask = array([1,0,1,0,0,1],  
                 dtype=bool)
```

```
>>> a[mask,2]  
array([2, 22, 52])
```

0	1	2	3	4	5
10	11	12	13	14	15
20	21	22	23	24	25
30	31	32	33	34	35
40	41	42	43	44	45
50	51	52	53	54	55

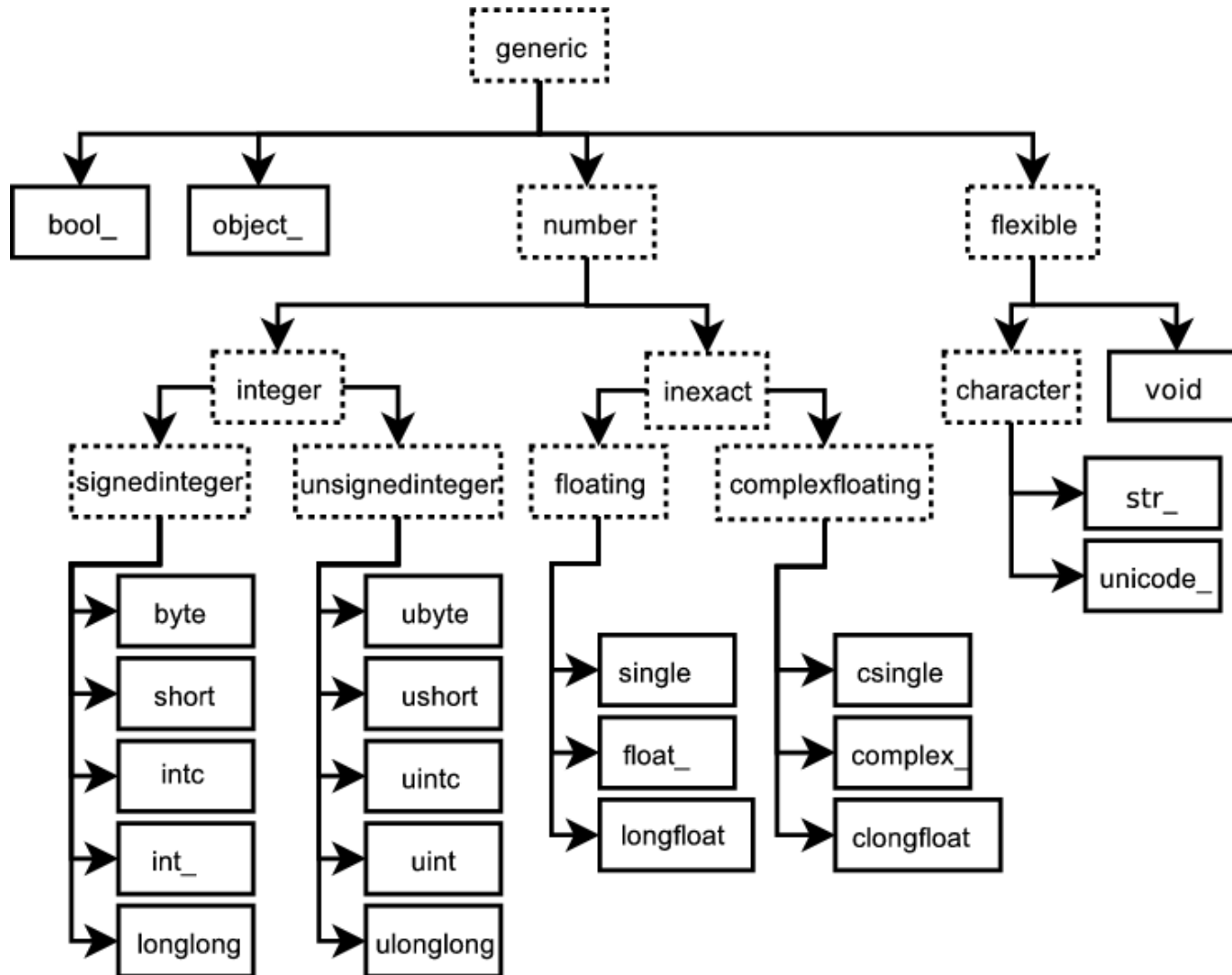


Unlike slicing, fancy indexing creates copies instead of views into original arrays.

# NumPy dtypes

Basic Type	Available NumPy types	Comments
Boolean	<code>bool</code>	Elements are 1 byte in size
Integer	<code>int8, int16, int32, int64, int128, int</code>	<code>int</code> defaults to the size of <code>int</code> in C for the platform
Unsigned Integer	<code>uint8, uint16, uint32, uint64, uint128, uint</code>	<code>uint</code> defaults to the size of unsigned <code>int</code> in C for the platform
Float	<code>float32, float64, float, longfloat,</code>	Float is always a double precision floating point value (64 bits). <code>longfloat</code> represents large precision floats. Its size is platform dependent.
Complex	<code>complex64, complex128, complex</code>	The real and complex elements of a <code>complex64</code> are each represented by a single precision (32 bit) value for a total size of 64 bits.
Strings	<code>str, unicode</code>	Unicode is always UTF32 (UCS4)
Object	<code>object</code>	Represent items in array as Python objects.
Records	<code>void</code>	Used for arbitrary data structures in record arrays.

# Built-in “scalar” types





# Defining Data-Types

- An item can include fields of different data-types.
- A field is described by a data-type object and a byte offset --- this definition allows nested records.
- The array construction command interprets tuple elements as field entries.

```
>>> my_data_type = np.dtype("int32,float32,bool")
```

```
>>> a = np.array([(1,2.35,True), (2,3.42,False)],  
dtype=my_data_type)
```

```
>>> print a['f0']  
[Hello World]
```

# Array Calculation Methods

## SUM FUNCTION

```
>>> a = array([[1,2,3],  
               [4,5,6]], float)
```

```
# Sum defaults to summing all  
# *all* array values.
```

```
>>> sum(a)  
21.
```

```
# supply the keyword axis to  
# sum along the 0th axis.
```

```
>>> sum(a, axis=0)  
array([5., 7., 9.])
```

```
# supply the keyword axis to  
# sum along the last axis.
```

```
>>> sum(a, axis=-1)  
array([6., 15.])
```

## SUM ARRAY METHOD

```
# The a.sum() defaults to  
# summing *all* array values
```

```
>>> a.sum()  
21.
```

```
# Supply an axis argument to  
# sum along a specific axis.
```

```
>>> a.sum(axis=0)  
array([5., 7., 9.])
```

## PRODUCT

```
# product along columns.
```

```
>>> a.prod(axis=0)  
array([ 4., 10., 18.])
```

```
# functional form.
```

```
>>> prod(a, axis=0)  
array([ 4., 10., 18.])
```

# Min/Max

## MIN

```
>>> a = array([2.,3.,0.,1.])
>>> a.min(axis=0)
0.
# use Numpy's amin() instead
# of Python's builtin min()
# for speed operations on
# multi-dimensional arrays.
>>> amin(a, axis=0)
0.
```

## ARGMIN

```
# Find index of minimum value.
>>> a.argmin(axis=0)
2
# functional form
>>> argmin(a, axis=0)
2
```

## MAX

```
>>> a = array([2.,1.,0.,3.])
>>> a.max(axis=0)
3.

# functional form
>>> amax(a, axis=0)
3.
```

## ARGMAX

```
# Find index of maximum value.
>>> a.argmax(axis=0)
1
# functional form
>>> argmax(a, axis=0)
1
```

# Statistics Array Methods

## MEAN

```
>>> a = array([[1,2,3],  
               [4,5,6]], float)
```

```
# mean value of each column
```

```
>>> a.mean(axis=0)  
array([ 2.5,  3.5,  4.5])  
>>> mean(a, axis=0)  
array([ 2.5,  3.5,  4.5])  
>>> average(a, axis=0)  
array([ 2.5,  3.5,  4.5])
```

```
# average can also calculate
```

```
# a weighted average
```

```
>>> average(a, weights=[1,2],  
...         axis=0)  
array([ 3.,  4.,  5.])
```

## STANDARD DEV./VARIANCE

```
# Standard Deviation
```

```
>>> a.std(axis=0)  
array([ 1.5,  1.5,  1.5])
```

```
# Variance
```

```
>>> a.var(axis=0)  
array([2.25, 2.25, 2.25])  
>>> var(a, axis=0)  
array([2.25, 2.25, 2.25])
```

# Other Array Methods

## CLIP

```
# Limit values to a range

>>> a = array([[1,2,3],
               [4,5,6]], float)

# Set values < 3 equal to 3.
# Set values > 5 equal to 5.
>>> a.clip(3,5)
>>> a
array([[ 3.,  3.,  3.],
       [ 4.,  5.,  5.]])
```

## ROUND

```
# Round values in an array.
# Numpy rounds to even, so
# 1.5 and 2.5 both round to 2.
>>> a = array([1.35, 2.5, 1.5])
>>> a.round()
array([ 1.,  2.,  2.])

# Round to first decimal place.
>>> a.round(decimals=1)
array([ 1.4,  2.5,  1.5])
```

## POINT TO POINT

```
# Calculate max - min for
# array along columns
>>> a.ptp(axis=0)
array([ 3.0,  3.0,  3.0])
# max - min for entire array.
>>> a.ptp(axis=None)
5.0
```

# Summary of (most) array attributes/methods

## BASIC ATTRIBUTES

`a.dtype` - Numerical type of array elements. `float32`, `uint8`, etc.  
`a.shape` - Shape of the array. `(m,n,o,...)`  
`a.size` - Number of elements in entire array.  
`a.itemsize` - Number of bytes used by a single element in the array.  
`a.nbytes` - Number of bytes used by entire array (data only).  
`a.ndim` - Number of dimensions in the array.

## SHAPE OPERATIONS

`a.flat` - An iterator to step through array as if it is 1D.  
`a.flatten()` - Returns a 1D copy of a multi-dimensional array.  
`a.resize(new_size)` - Change the size/shape of an array in-place.  
`a.swapaxes(axis1, axis2)` - Swap the order of two axes in an array.  
`a.transpose(*axes)` - Swap the order of any number of array axes.  
`a.T` - Shorthand for `a.transpose()`

# Summary of (most) array attributes/methods

## FILL AND COPY

`a.copy()` - Return a copy of the array.

`a.fill(value)` - Fill array with a scalar value.

## CONVERSION / COERSION

`a.tolist()` - Convert array into nested lists of values.

`a.tostring()` - raw copy of array memory into a python string.

`a.astype(dtype)` - Return array coerced to given dtype.

## COMPLEX NUMBERS

`a.real` - Return the real part of the array.

`a.imag` - Return the imaginary part of the array.

`a.conjugate()` - Return the complex conjugate of the array.

`a.conj()` - Return the complex conjugate of an array. (same as conjugate)

# Summary of (most) array attributes/methods

## SAVING

`a.dump(file)` - Store a binary array data out to the given file.  
`a.dumps()` - returns the binary pickle of the array as a string.  
`a.tofile(fid, sep="", format="%s")` Formatted ascii output to file.

## SEARCH / SORT

`a.nonzero()` - Return indices for all non-zero elements in `a`.  
`a.sort(axis=-1)` - Inplace sort of array elements along axis.  
`a.argsort(axis=-1)` - Return indices for element sort order along axis.  
`a.searchsorted(b)` - Return index where elements from `b` would go in `a`.

## ELEMENT MATH OPERATIONS

`a.clip(low, high)` - Limit values in array to the specified range.  
`a.round(decimals=0)` - Round to the specified number of digits.  
`a.cumsum(axis=None)` - Cumulative sum of elements along axis.  
`a.cumprod(axis=None)` - Cumulative product of elements along axis.



# Summary of (most) array attributes/methods

## REDUCTION METHODS

All the following methods “reduce” the size of the array by 1 dimension by carrying out an operation along the specified axis. If axis is None, the operation is carried out across the entire array.

`a.sum(axis=None)` - Sum up values along axis.

`a.prod(axis=None)` - Find the product of all values along axis.

`a.min(axis=None)` - Find the minimum value along axis.

`a.max(axis=None)` - Find the maximum value along axis.

`a.argmin(axis=None)` - Find the index of the minimum value along axis.

`a.argmax(axis=None)` - Find the index of the maximum value along axis.

`a.ptp(axis=None)` - Calculate `a.max(axis) - a.min(axis)`

`a.mean(axis=None)` - Find the mean (average) value along axis.

`a.std(axis=None)` - Find the standard deviation along axis.

`a.var(axis=None)` - Find the variance along axis.

`a.any(axis=None)` - True if any value along axis is non-zero. (or)

`a.all(axis=None)` - True if all values along axis are non-zero. (and)

# Array Operations

## SIMPLE ARRAY MATH

```
>>> a = array([1,2,3,4])
>>> b = array([2,3,4,5])
>>> a + b
array([3, 5, 7, 9])
```

Numpy defines the following constants:

```
pi = 3.14159265359
e = 2.71828182846
```

## MATH FUNCTIONS

```
# Create array from 0 to 10
>>> x = arange(11.)
```

```
# multiply entire array by
# scalar value
```

```
>>> a = (2*pi)/10.
```

```
>>> a
```

```
0.62831853071795862
```

```
>>> a*x
```

```
array([ 0., 0.628, ..., 6.283])
```

```
# inplace operations
```

```
>>> x *= a
```

```
>>> x
```

```
array([ 0., 0.628, ..., 6.283])
```

```
# apply functions to array.
```

```
>>> y = sin(x)
```

# Mathematic Binary Operators

$a + b \rightarrow \text{add}(a,b)$   
 $a - b \rightarrow \text{subtract}(a,b)$   
 $a \% b \rightarrow \text{remainder}(a,b)$

$a * b \rightarrow \text{multiply}(a,b)$   
 $a / b \rightarrow \text{divide}(a,b)$   
 $a ** b \rightarrow \text{power}(a,b)$

## MULTIPLY BY A SCALAR

```
>>> a = array((1,2))
>>> a*3.
array([3., 6.] )
```

## ELEMENT BY ELEMENT ADDITION

```
>>> a = array([1,2])
>>> b = array([3,4])
>>> a + b
array([4, 6])
```

## ADDITION USING AN OPERATOR FUNCTION

```
>>> add(a,b)
array([4, 6])
```

## IN PLACE OPERATION

```
# Overwrite contents of a.
# Saves array creation
# overhead
>>> add(a,b,a) # a += b
array([4, 6])
>>> a
array([4, 6])
```

# Comparison and Logical Operators

<code>equal</code>	<code>(==)</code>	<code>not_equal</code>	<code>(!=)</code>	<code>greater</code>	<code>(&gt;)</code>
<code>greater_equal</code>	<code>(&gt;=)</code>	<code>less</code>	<code>(&lt;)</code>	<code>less_equal</code>	<code>(&lt;=)</code>
<code>logical_and</code>		<code>logical_or</code>		<code>logical_xor</code>	
<code>logical_not</code>					

## 2D EXAMPLE

```
>>> a = array(((1,2,3,4) , (2,3,4,5)))
>>> b = array(((1,2,5,4) , (1,3,4,5)))
>>> a == b
array([[True, True, False, True],
       [False, True, True, True]])
# functional equivalent
>>> equal(a,b)
array([[True, True, False, True],
       [False, True, True, True]])
```

# Bitwise Operators

<code>bitwise_and</code>	<code>(&amp;)</code>	<code>invert</code>	<code>(~)</code>	<code>right_shift(a,shifts)</code>
<code>bitwise_or</code>	<code>( )</code>	<code>bitwise_xor</code>		<code>left_shift (a,shifts)</code>

## BITWISE EXAMPLES

```
>>> a = array((1,2,4,8))
>>> b = array((16,32,64,128))
>>> bitwise_or(a,b)
array([ 17,  34,  68, 136])
```

```
# bit inversion
```

```
>>> a = array((1,2,3,4), uint8)
>>> invert(a)
array([254, 253, 252, 251], dtype=uint8)
```

```
# left shift operation
```

```
>>> left_shift(a,3)
array([ 8, 16, 24, 32], dtype=uint8)
```

# Trig and Other Functions

## TRIGONOMETRIC

<code>sin(x)</code>	<code>sinh(x)</code>
<code>cos(x)</code>	<code>cosh(x)</code>
<code>arccos(x)</code>	<code>arccosh(x)</code>
<code>arctan(x)</code>	<code>arctanh(x)</code>
<code>arcsin(x)</code>	<code>arcsinh(x)</code>
<code>arctan2(x,y)</code>	

## OTHERS

<code>exp(x)</code>	<code>log(x)</code>
<code>log10(x)</code>	<code>sqrt(x)</code>
<code>absolute(x)</code>	<code>conjugate(x)</code>
<code>negative(x)</code>	<code>ceil(x)</code>
<code>floor(x)</code>	<code>fabs(x)</code>
<code>hypot(x,y)</code>	<code>fmod(x,y)</code>
<code>maximum(x,y)</code>	<code>minimum(x,y)</code>

`hypot(x,y)`

Element by element distance  
calculation using  $\sqrt{x^2 + y^2}$

# Broadcasting

When there are multiple inputs, then they all must be “broadcastable” to the same shape.

- All arrays are promoted to the same number of dimensions (by prepending 1's to the shape)
- All dimensions of length 1 are expanded as determined by other inputs with non-unit lengths in that dimension.

```
a = np.array([0,1,2])  
b = np.array([[0],[10],[20],[30]])  
print a+b
```

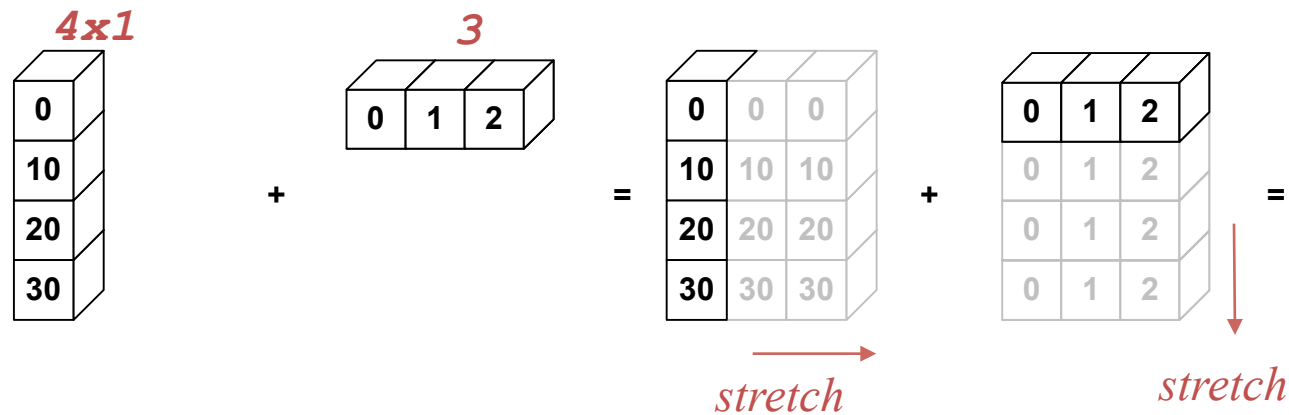
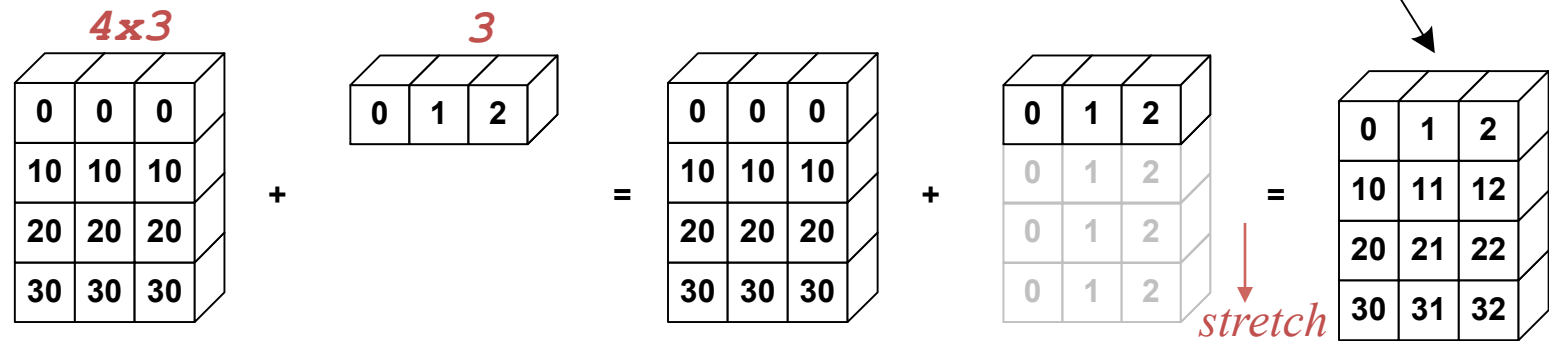
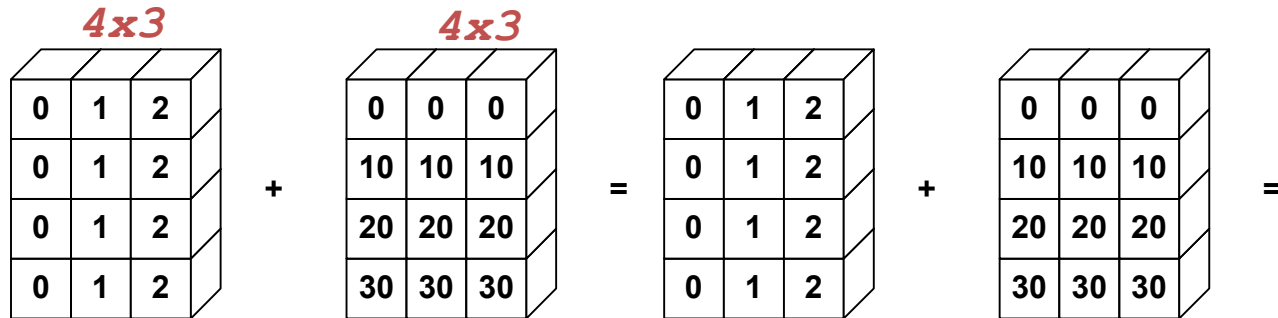
```
[[ 0  1  2]  
 [10 11 12]  
 [20 21 22]  
 [30 31 32]]
```

a has shape (3,) the ufunc sees it as having shape (1,3)

b has shape (4,1)

The ufunc result has shape (4,3)

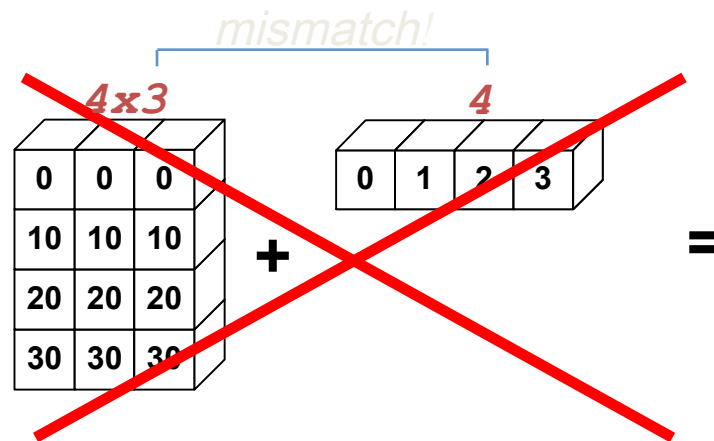
# Array Broadcasting





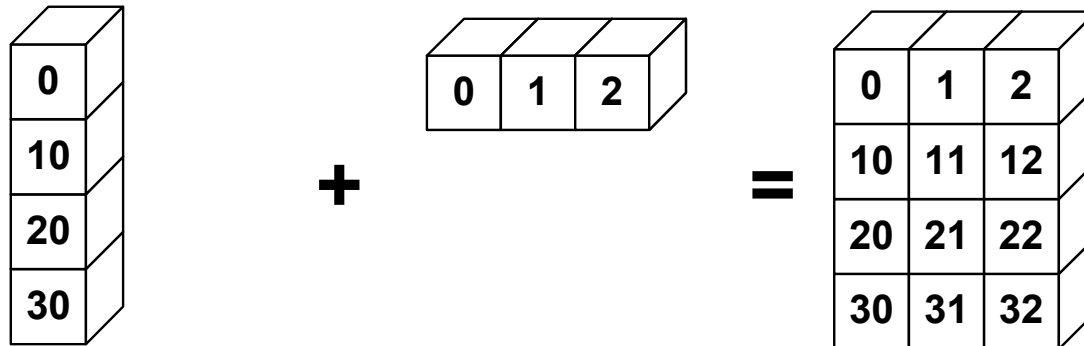
# Broadcasting Rules

The *trailing* axes of both arrays must either be 1 or have the same size for broadcasting to occur. Otherwise, a “`ValueError: frames are not aligned`” exception is thrown.



# Broadcasting in Action

```
>>> a = array((0,10,20,30))  
>>> b = array((0,1,2))  
>>> y = a[:, None] + b
```



# Vectorizing Functions

## VECTORIZING FUNCTIONS

### Example

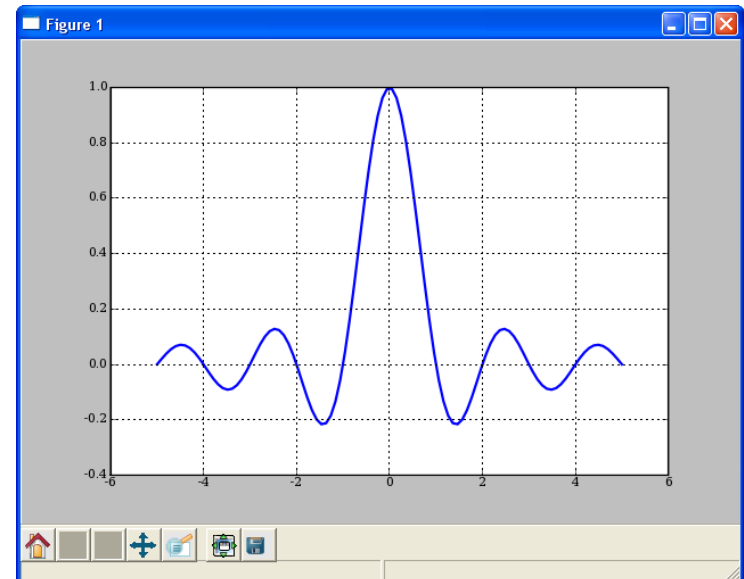
```
def sinc(x):  
    if x == 0.0:  
        return 1.0  
    else:  
        w = pi*x  
        return sin(w) / w
```

# attempt

```
>>> sinc([1.3,1.5])  
TypeError: can't multiply  
sequence to non-int  
>>> x = r_[-5:5:100j]  
>>> y = vsinc(x)  
>>> plot(x, y)
```

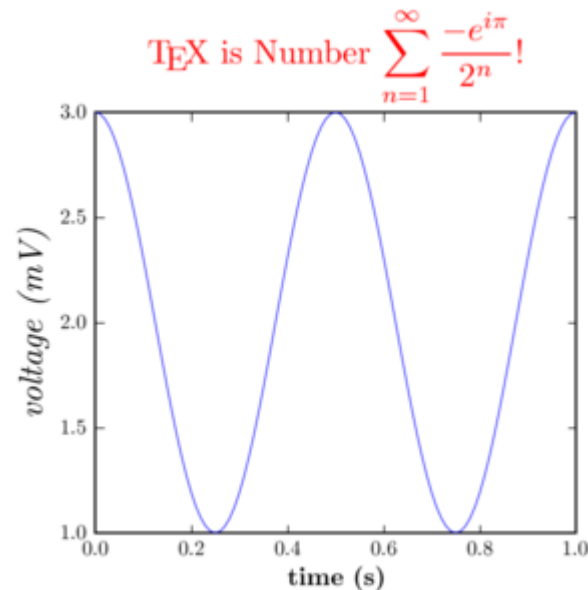
## SOLUTION

```
>>> from numpy import vectorize  
>>> vsinc = vectorize(sinc)  
>>> vsinc([1.3,1.5])  
array([-0.1981, -0.2122])
```



# Matplotlib

- Requires NumPy extension. Provides powerful plotting commands.
- <http://matplotlib.sourceforge.net>



# Recommendations

- **Matplotlib** for day-to-day data exploration.

Matplotlib has a large community, tons of plot types, and is well integrated into ipython. It is the de-facto standard for 'command line' plotting from ipython.

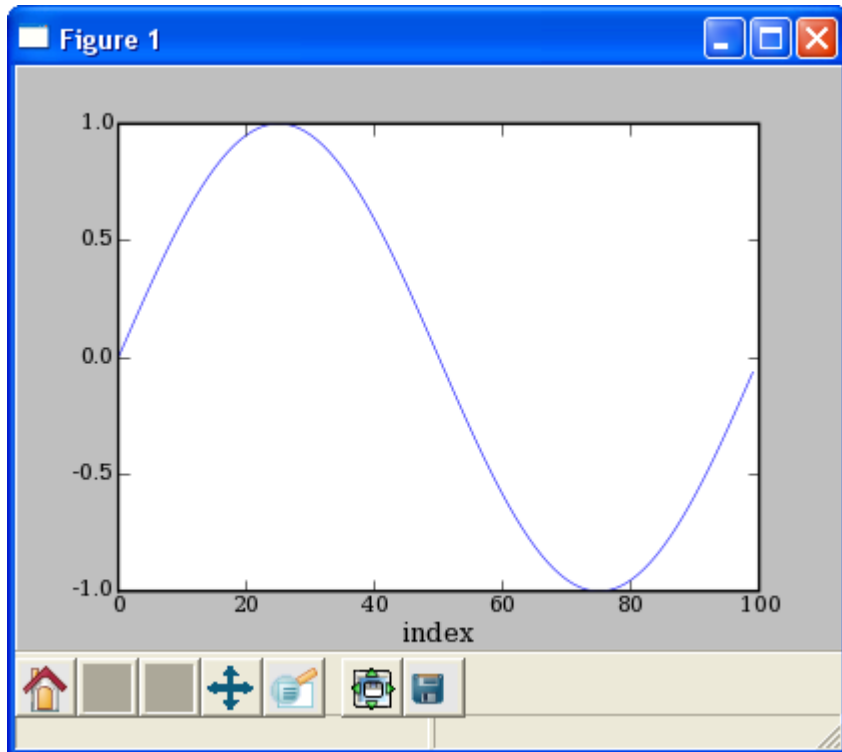
- **Chaco** for building interactive plotting applications

Chaco is architected for building highly interactive and configurable plots in python. It is more useful as plotting toolkit than for making one-off plots.

# Line Plots

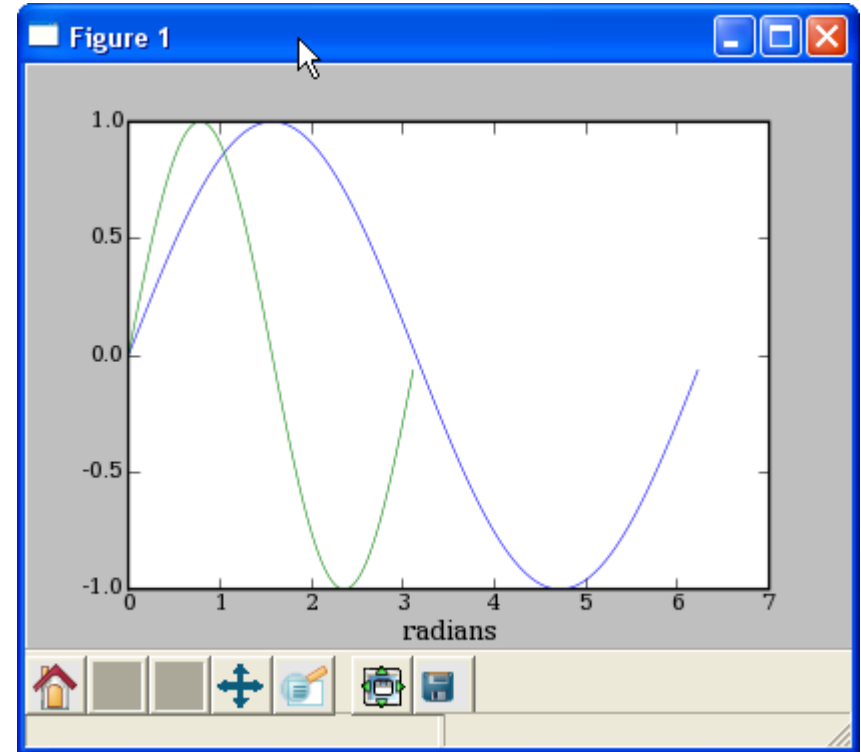
## PLOT AGAINST INDICES

```
>>> x = arange(50)*2*pi/50.  
>>> y = sin(x)  
>>> plot(y)  
>>> xlabel('index')
```



## MULTIPLE DATA SETS

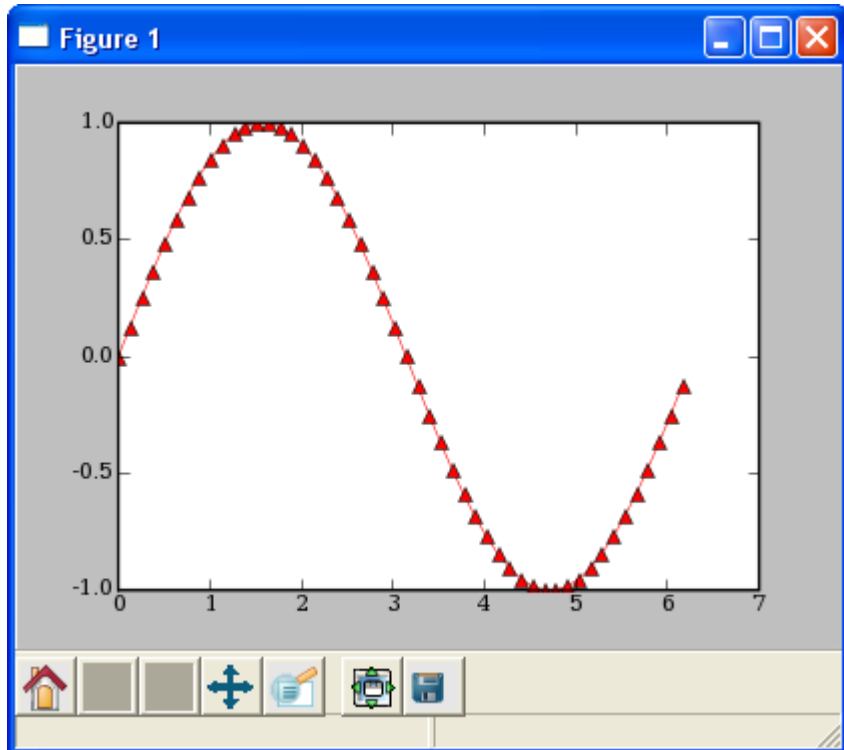
```
>>> plot(x,y,x2,y2)  
>>> xlabel('radians')
```



# Line Plots

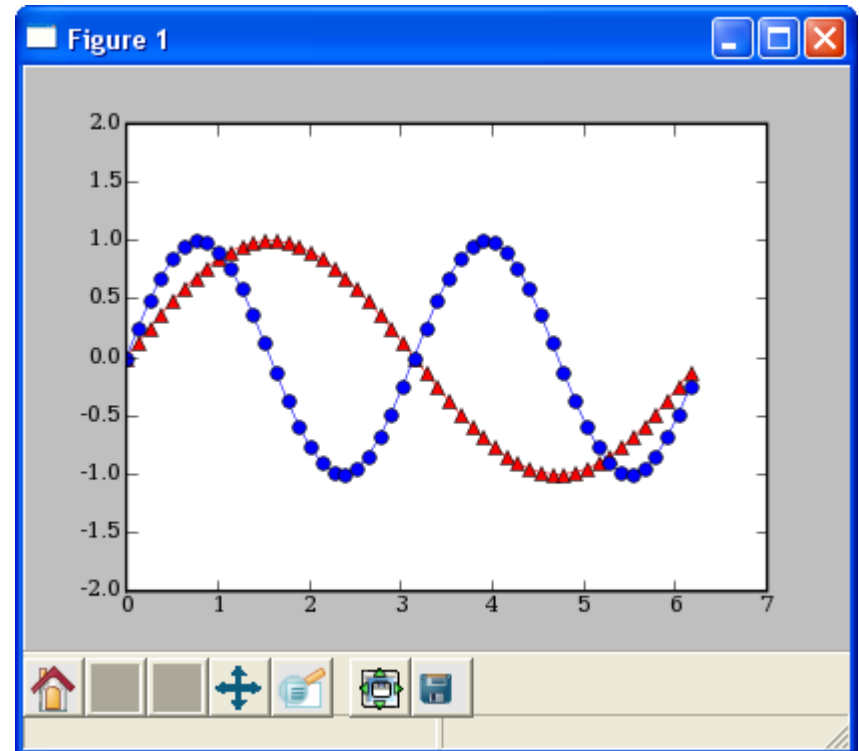
## LINE FORMATTING

```
# red, dot-dash, triangles  
>>> plot(x,sin(x), 'r-^')
```



## MULTIPLE PLOT GROUPS

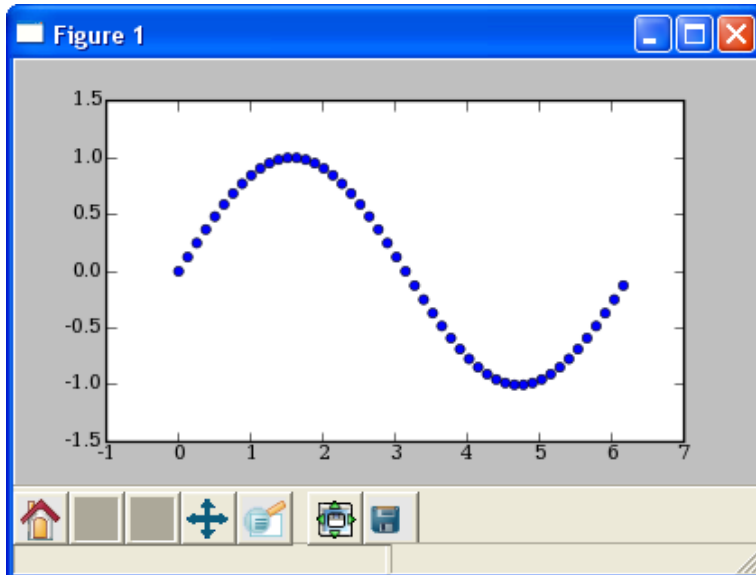
```
>>> plot(x,y1,'b-o', x,y2), r-^')  
>>> axis([0,7,-2,2])
```



# Scatter Plots

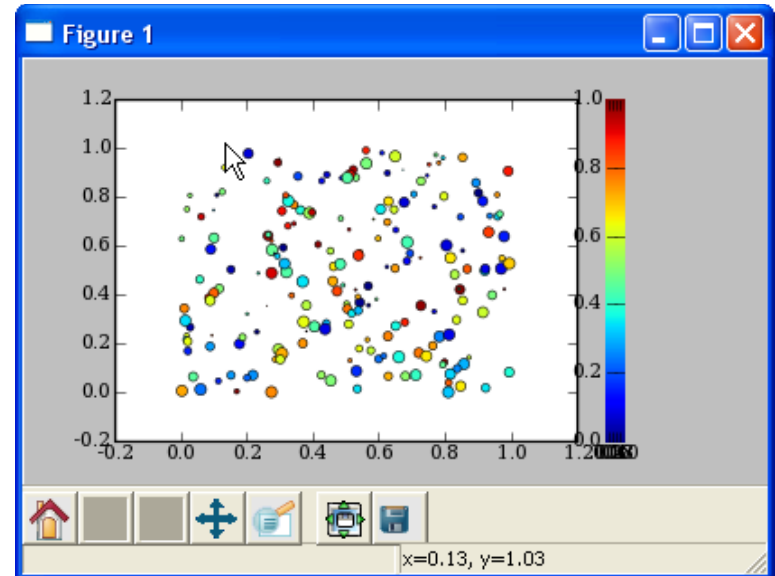
## SIMPLE SCATTER PLOT

```
>>> x = arange(50)*2*pi/50.  
>>> y = sin(x)  
>>> scatter(x,y)
```



## COLORMAPPED SCATTER

```
# marker size/color set with data  
>>> x = rand(200)  
>>> y = rand(200)  
>>> size = rand(200)*30  
>>> color = rand(200)  
>>> scatter(x, y, size, color)  
>>> colorbar()
```

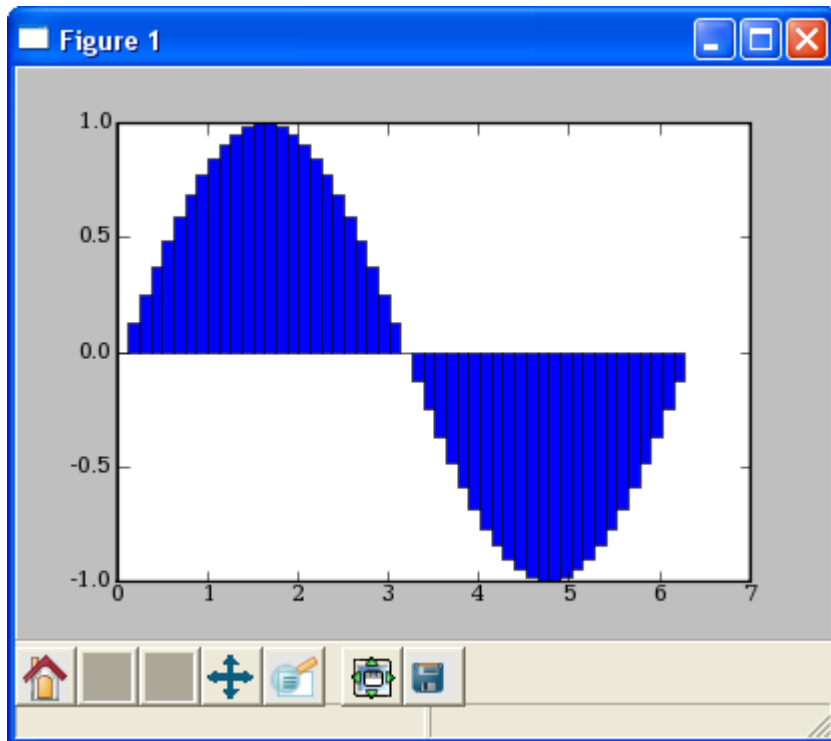




# Bar Plots

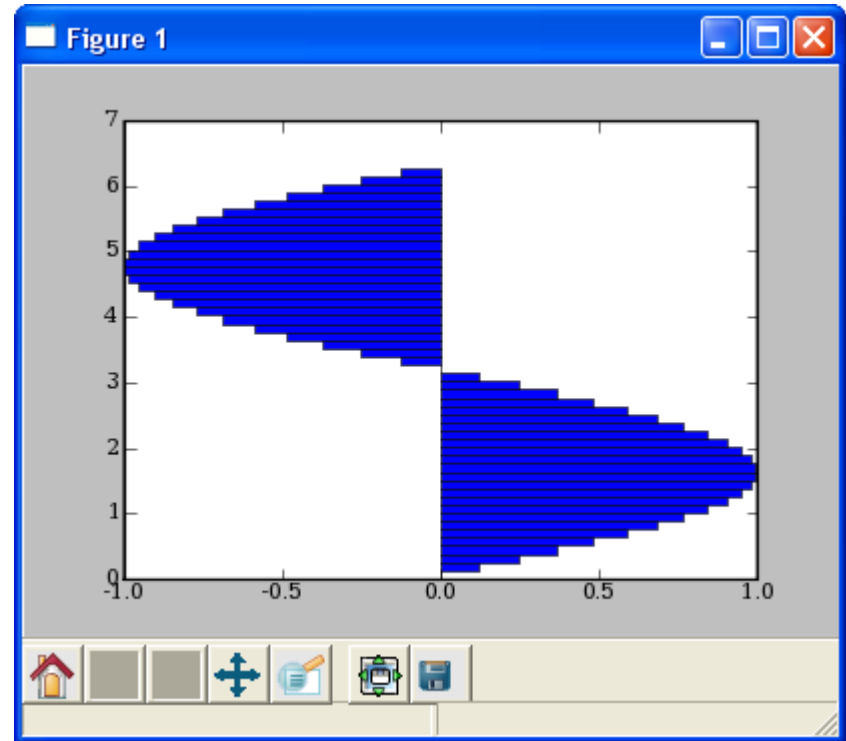
## BAR PLOT

```
>>> bar(x,sin(x),  
...      width=x[1]-x[0])
```



## HORIZONTAL BAR PLOT

```
>>> hbar(x,sin(x),  
...       height=x[1]-x[0],  
...       orientation='horizontal')
```



# Bar Plots

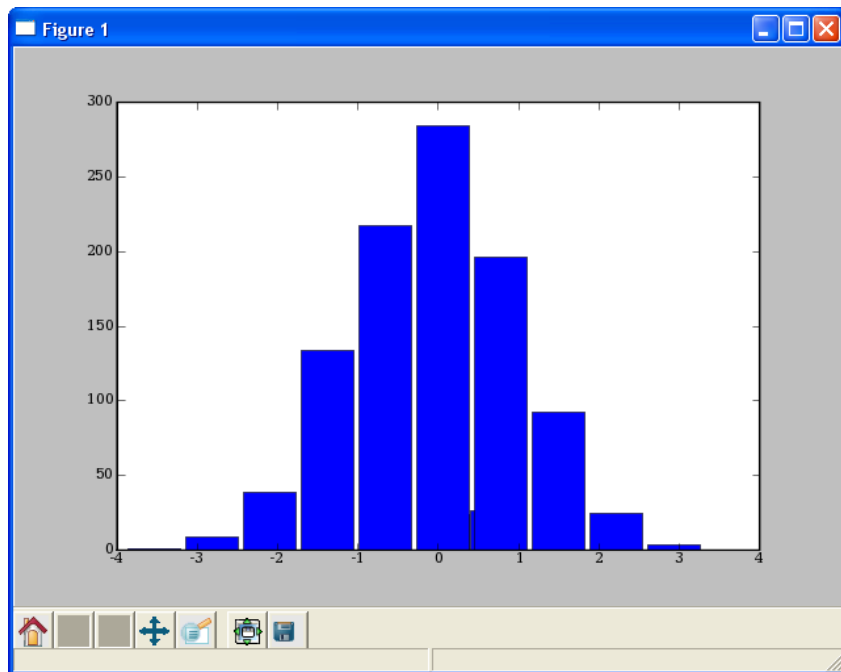
DEMO/MATPLOTLIB\_PLOTTING/BARCHART\_DEMO.PY



# HISTOGRAMS

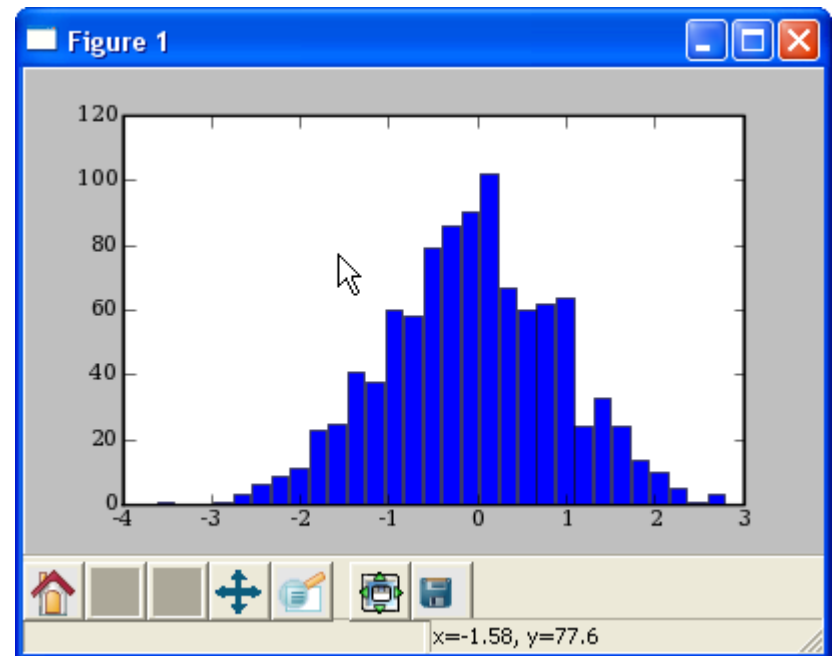
## HISTOGRAM

```
# plot histogram  
# default to 10 bins  
>>> hist(randn(1000))
```



## HISTOGRAM 2

```
# change the number of bins  
>>> hist(randn(1000), 30)
```



**Questions? :-)**