

# Python Short Course

## Lecture 2: Numerical Python

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# NumPy Modules

- NumPy has many of the features of Matlab, in a free, multiplatform program. It also allows you to do intensive computing operations in a simple way
- Numeric Module: Array Constructors
  - ones, zeros, identity
  - arange
- LinearAlgebra Module: Solvers
  - Singular Value Decomposition
  - Eigenvalue, Eigenvector
  - Inverse
  - Determinant
  - Linear System Solver



# Simple Numeric Constructors

- Arrays are slightly different from lists. They can only contain one type of data structure, and they are much faster to work with numerically.

```
>>> from Numeric import *
>>> s = arange(0,2*pi,0.1) # "arange" also
>>> print s
[0., 0.1, ... 6.2]
>>> sin(s) #Numeric.sin maps onto arrays
[0., 0.099833, ... -0.0830894]
```



# Shape and reshape

```
>>> a = zeros((3,3),Float)
>>> print a
[[0.,0.,0.],
 [0.,0.,0.],
 [0.,0.,0.]]
>>> print a.shape
(3,3)
>>> reshape(a,(9,)) # could also use a.flat
>>> print a
[0.,0.,0.,0.,0.,0.,0.,0.,0.]
```



# Arrays and Constructors

```
>>> a = ones((3,3),Float)
>>> print a
[[1., 1., 1.],
 [1., 1., 1.],
 [1., 1., 1.]]
>>> b = zeros((3,3),Float)
>>> b = b + 2.*identity(3) #"+" is overloaded
>>> c = a + b
>>> print c
[[3., 1., 1.],
 [1., 3., 1.],
 [1., 1., 3.]]
```



# Overloaded operators

```
>>> b = 2.*ones((2,2),Float) #overloaded
>>> print b
[[2.,2.],
 [2.,2.]]
>>> b = b+1      # Addition of a scalar is
>>> print b      #   element-by-element
[[3.,3.],
 [3.,3.]]
>>> c = 2.*b     # Multiplication by a scalar is
>>> print c      #   element-by-element
[[6.,6.],
 [6.,6.]]
```



# More on overloaded operators

```
>>> c = 6.*ones((2,2),Float)
>>> a = identity(2)
>>> print a*c
[[6.,0.],          # ARGH! element-by-element!
 [0.,6.]]
>>> matrixmultiply(a,c)
[[6.,6.],
 [6.,6.]]
```



# Array functions

```
>>> from LinearAlgebra import *
>>> a = zeros((3,3),Float) + 2.*identity(3)
>>> print inverse(a)
[[0.5, 0., 0.],
 [0., 0.5, 0.],
 [0., 0., 0.5]]
>>> print determinant(inverse(a))
0.125
>>> print diagonal(a)
[0.5,0.5,0.5]
>>> print diagonal(a,1)
[0.,0.]
```

- transpose(a), argsort(), dot()





# Eigenvalues

```
>>> from LinearAlgebra import *
>>> val = eigenvalues(c)
>>> val, vec = eigenvectors(c)
>>> print val
[1., 4., 1.]
>>> print vec
[[0.816, -0.408, -0.408],
 [0.575, 0.577, 0.577],
 [-0.324, -0.487, 0.811]]
```

- also solve\_linear\_equations, singular\_value\_decomposition, etc.



# Least Squares Fitting

- Part of Hinsen's Scientific Python module

```
>>> from LeastSquares import *
>>> def func(params,x): #  $y=ax^2+bx+c$ 
    return params[0]*x*x + params[1]*x +
        params[2]

>>> data = []
>>> for i in range(10):
    data.append((i,i*i))
>>> guess = (3,2,1)
>>> fit_params, fit_error =
    leastSquaresFit(func,guess,data)
>>> print fit_params
[1.00,0.000,0.00]
```



# FFT

```
>>> from FFT import *
>>> data = array((1,0,1,0,1,0,1,0))
>>> print fft(data).real
[4., 0., 0., 0., 4., 0., 0., 0.]
```

- Also note that the FFTW package ("fastest Fourier transform in the West") has a python wrapper. See notes at the end



# Example: Particle in a Box

```
N = 100
```

```
T = get_kinetic_energy(N)
```

```
V = get_particle_box_potential(N)
```

```
H = T + V
```

```
val, vec = eigenvectors(H)
```

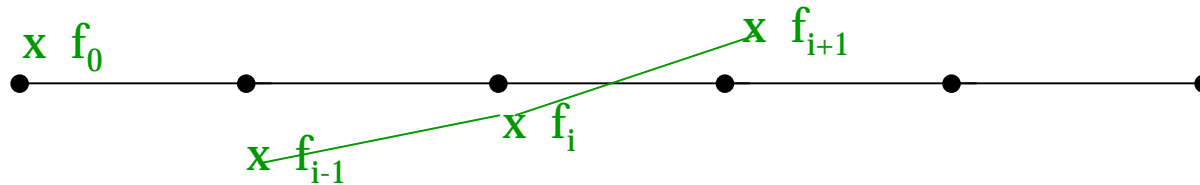
```
val, vec = ev_sort(val, vec)
```

```
plot_results(vec[:2])
```



# Finite Difference Approximation

- Consider a set of functional values on a grid



- We can calculate forward and backwards derivatives by simple differences
  - $df^- = (f_i - f_{i-1})/h$
  - $df^+ = (f_{i+1} - f_i)/h$
- We can take differences of these to get an approximation to the second derivative
  - $ddf = (df^+ - df^-)/h$
  - $ddf = (f_{i-1} - 2f_i + f_{i+1})/h^2$



# get\_kinetic\_energy function

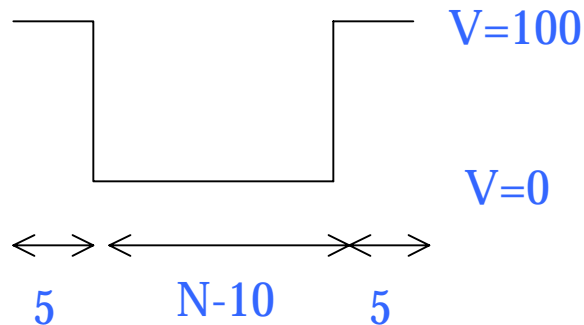
```
def get_kinetic_energy(N):  
    T = zeros((N,N),Float) + identity(N)  
    for i in range(N-1):  
        T[i,i+1] = T[i+1,i] = -0.5  
    return T
```

1.0	-0.5		
-0.5	1.0	-0.5	
	-0.5	1.0	-0.5
		-0.5	1.0



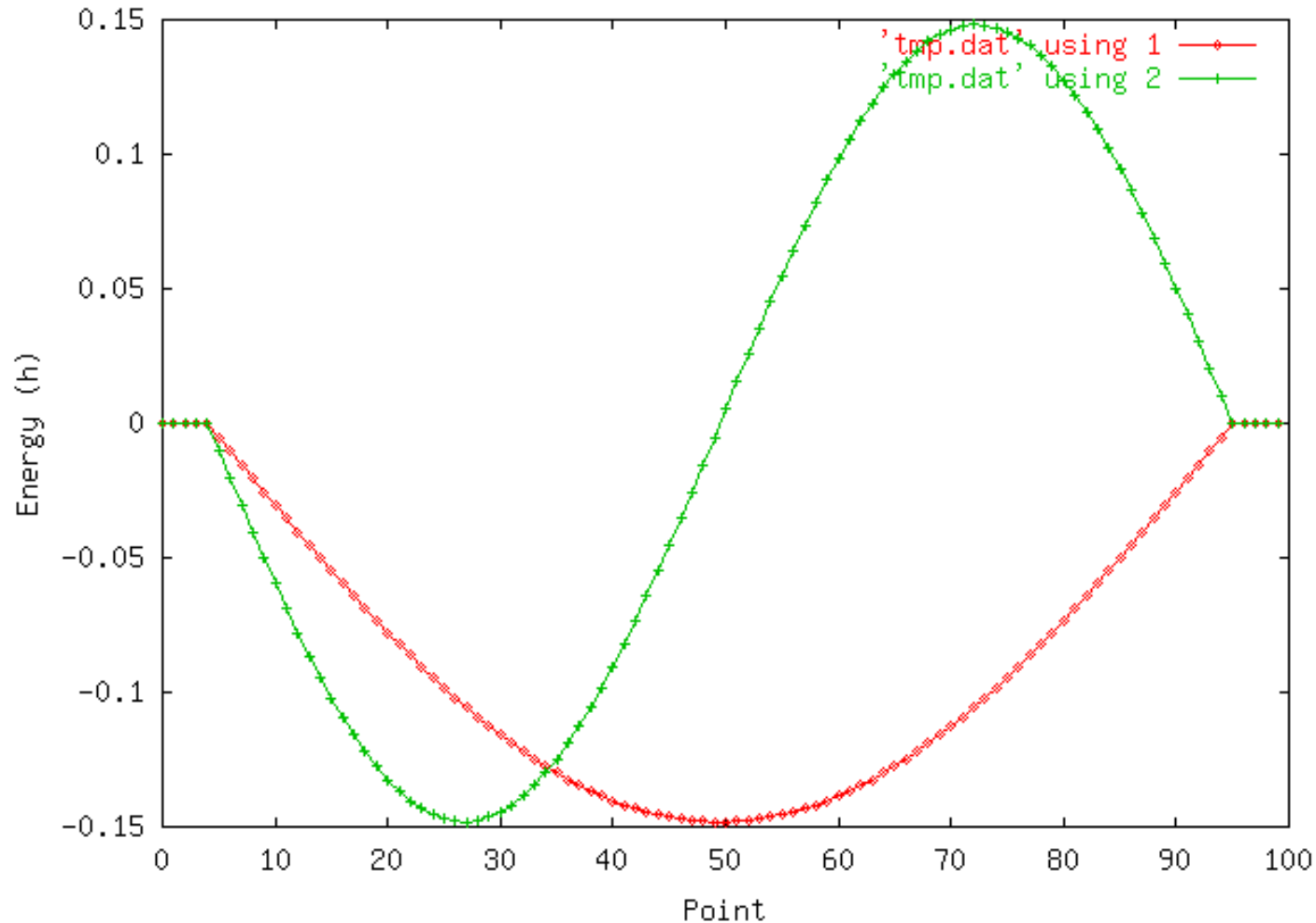
# get\_particle\_box\_potential function

```
def get_particle_box_potential(N):  
    border = 5  
    V = zeros((N,N),Float)  
    for i in range(border):  
        V[i,i] = V[N-1-i,N-1-i] = 100.  
    return V
```



# Particle in a Box Wave Function

Particle in a Box Eigenvectors





# Example: Harmonic Oscillator

```
N = 100
```

```
T = get_kinetic_energy(N)
```

```
V = get_harmonic_oscillator_potential(N)
```

```
H = T + V
```

```
val, vec = eigenvectors(H)
```

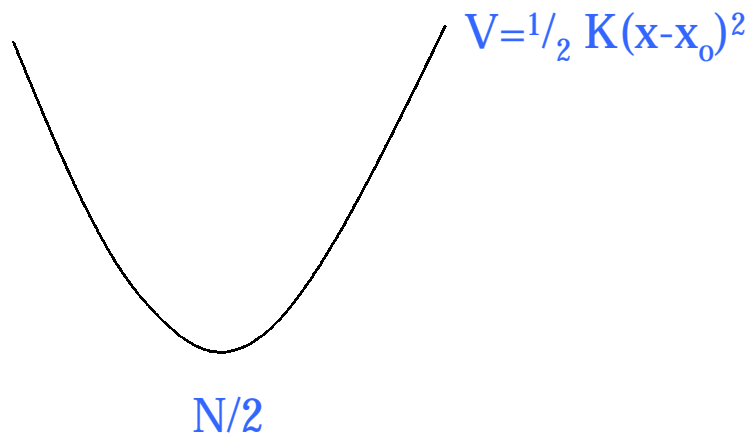
```
val, vec = ev_sort(val, vec)
```

```
plot_results(vec[:2])
```



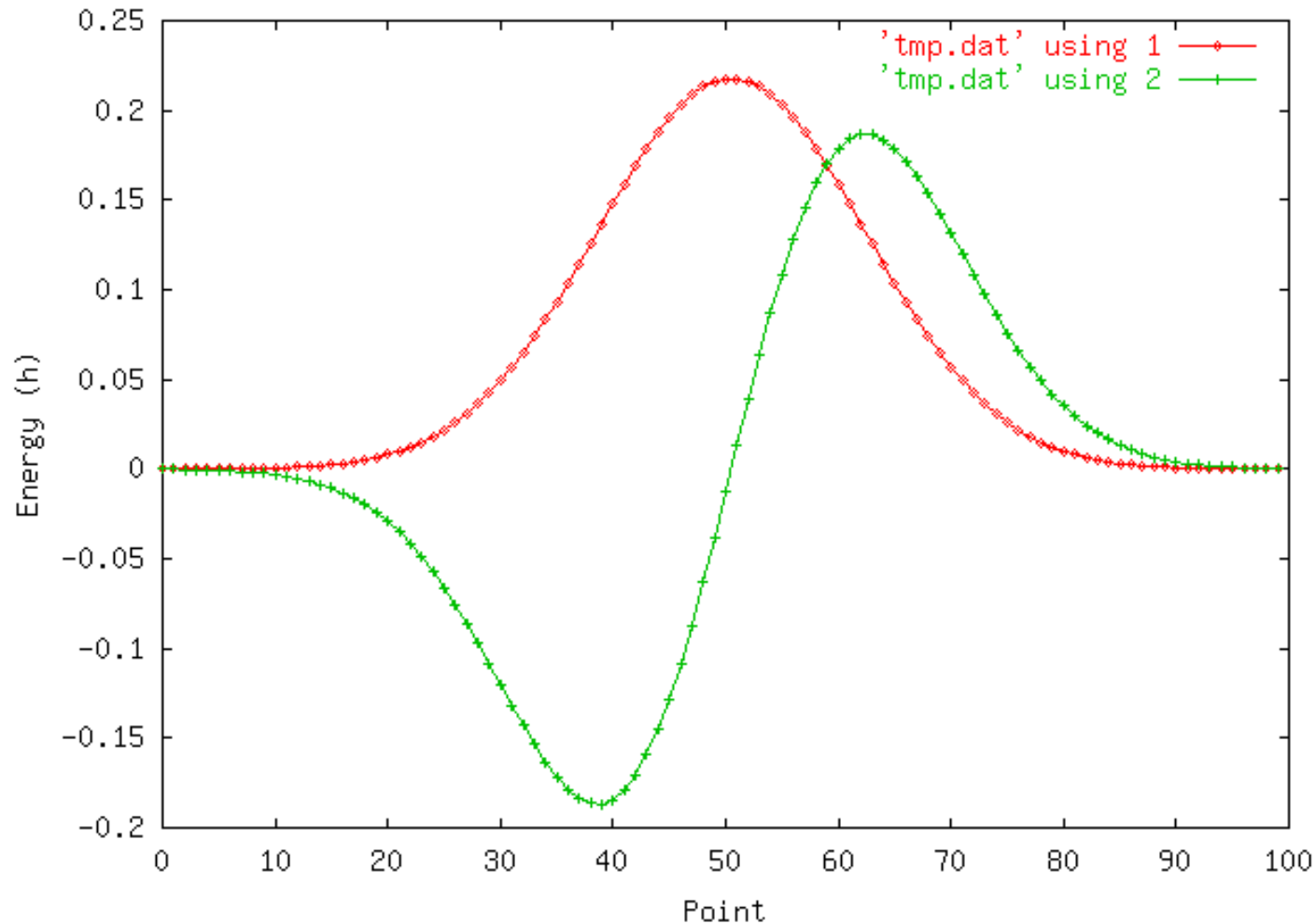
# get\_harmonic\_oscillator\_potential

```
def get_harmonic_oscillator_potential(N):  
    midpoint = N/2 + 0.5  
    K = 0.5/(N*N) # independent of N  
    V = zeros((N,N),Float)  
    for i in range(N):  
        delx = i - midpoint  
        V[i,i] = 0.5*K*delx*delx  
    return V
```



# Harmonic Oscillator Wave Function

Harmonic Oscillator Eigenvectors



# Example: One-D Hydrogen

```
N = 100
```

```
T = get_kinetic_energy(N)
```

```
V = get_oned_hydrogen_potential(N)
```

```
H = T + V
```

```
val, vec = eigenvectors(H)
```

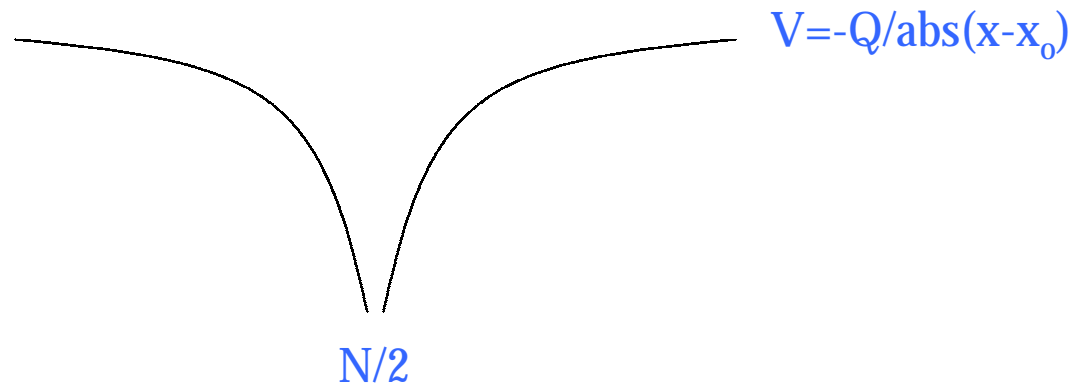
```
val, vec = ev_sort(val, vec)
```

```
plot_results(vec[:2])
```

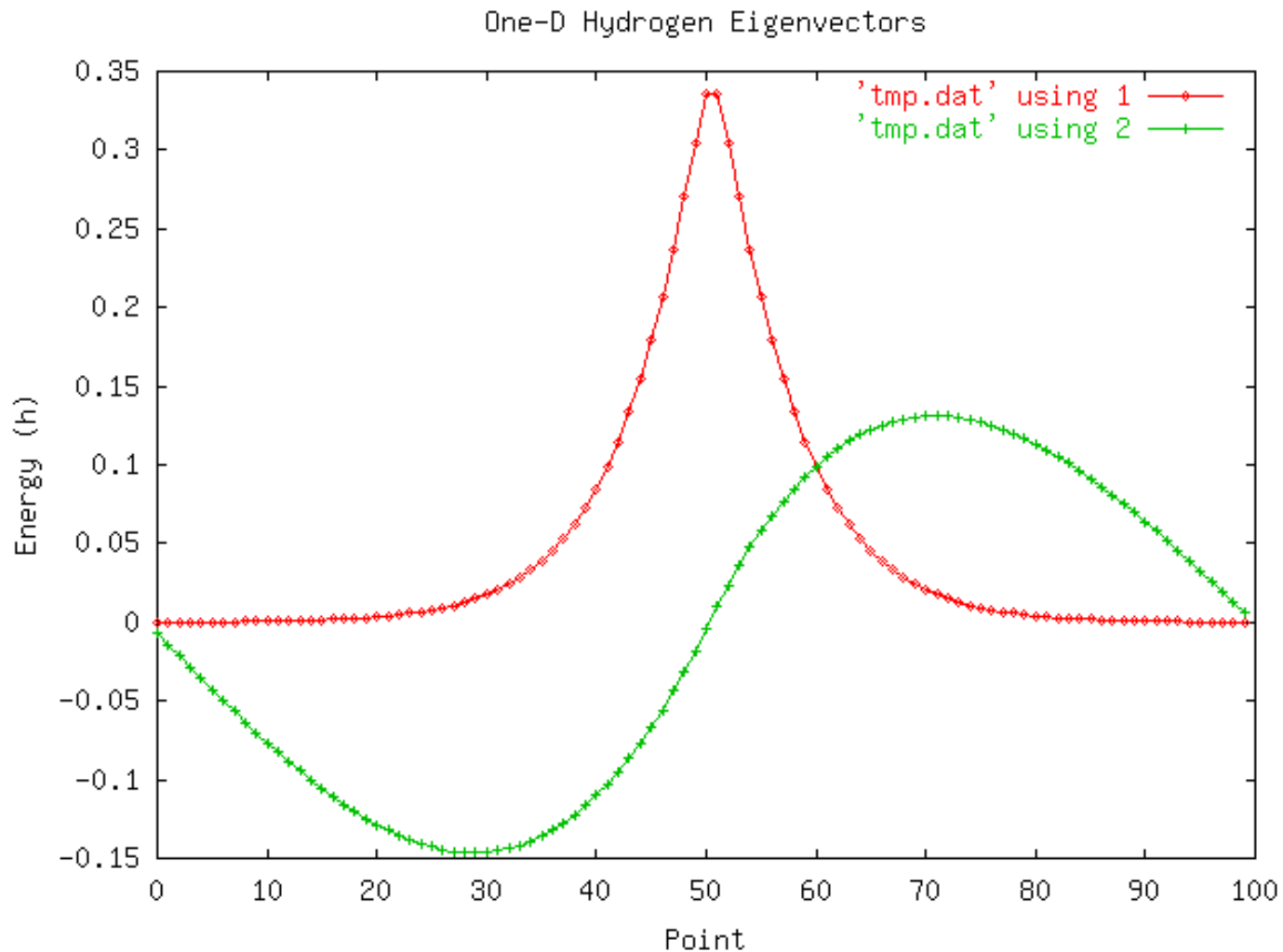


# get\_oned\_hydrogen\_potential

```
def get_oned_hydrogen_potential(N):  
    midpoint = N/2 + 0.5  
    Qeff = 3./N # independent of N  
    V = zeros((N,N),Float)  
    for i in range(N):  
        delx = i - midpoint  
        V[i,i] = -Qeff/abs(delx)  
    return V
```



# One-D Hydrogen Eigenvectors



# Command Line Arguments

- We would like to be able to choose the different potentials via **command-line flags**, i.e.,

```
% one_d_hamiltonian.py -b -n 50  
(Box wave function with 50 points)
```

```
% one_d_hamiltonian.py -s  
(Spring wave function with 100 points)
```

```
% one_d_hamiltonian.py -h -n 200  
(One-D H wave function with 200 points)
```



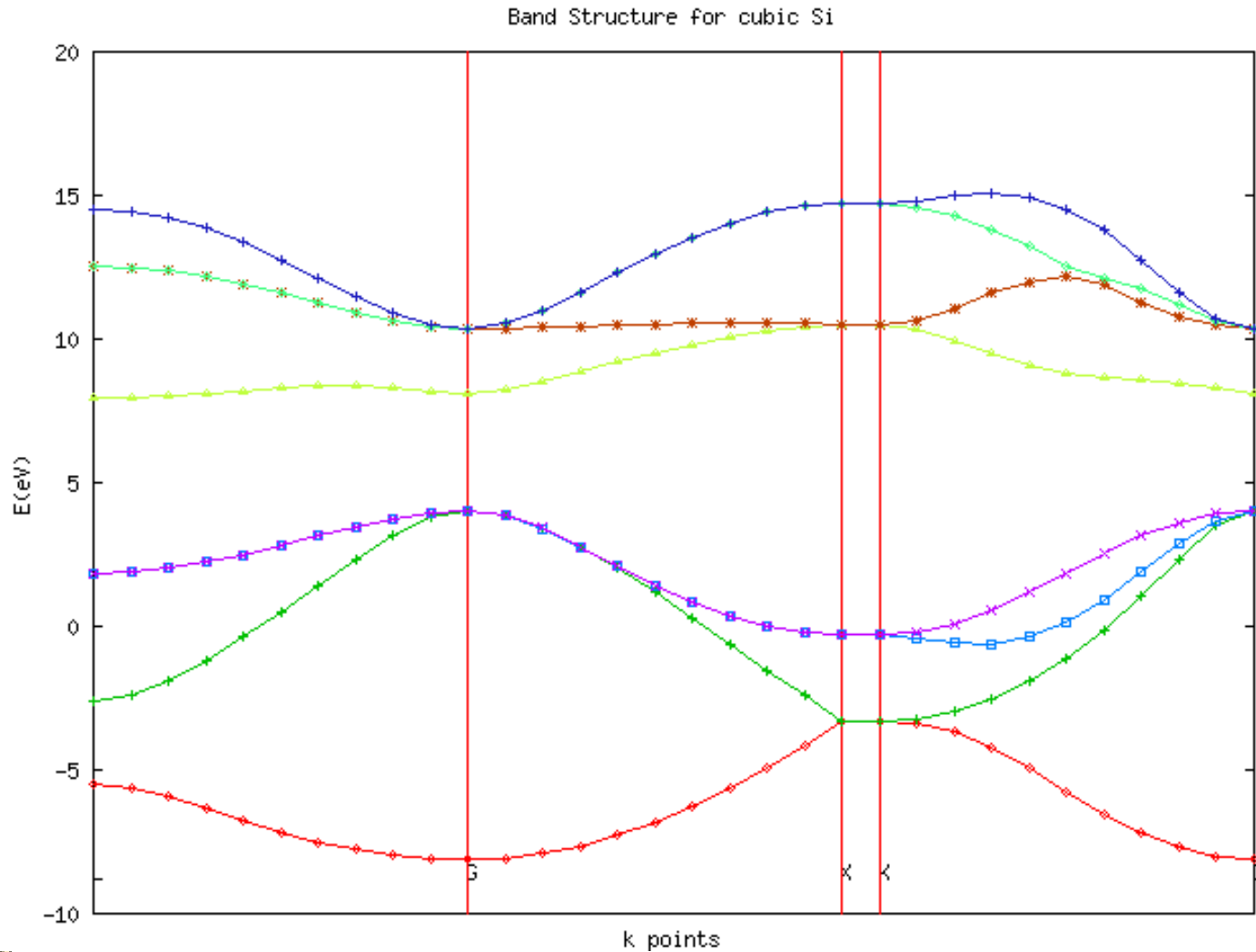
# getopt module

```
opts, args = getopt.getopt(sys.argv[1:], 'bshn:')
for opt in opts:
    key,value = opt
    if key == '-b':
        potential == 'box'
    elif key == '-s':
        potential == 'spring'
    elif key == '-h':
        potential == 'hydrogen'
    elif key == '-n':
        N = eval(value)
```





# Example: Tight Binding Band Structure of Semiconductors



# Tight Binding Theory

- Extended-Huckel treatment of electronic structure
  - Diagonal elements of H have a self term.
  - Off-diagonal elements of H have a term related to the coupling
- Include periodic boundary conditions
- k-points sample different space group symmetries
- Look at a sample Hamiltonian
  - Diamond, cubic Zincblend structures
  - 2 atoms per unit cell (cation and anion)
  - Minimal basis ( $s, p_x, p_y, p_z$ ) functions on each atom
  - Off-diagonal coupling is modulated by phase factors



# Harrison Hamiltonian

	$s^c$	$s^a$	$p_x^c$	$p_y^c$	$p_z^c$	$p_x^a$	$p_y^a$	$p_z^a$
$s^c$	$E_s^c$	$E_{ss}g_0$				$E_{sp}g_1$	$E_{sp}g_2$	$E_{sp}g_3$
$s^a$	$E_{ss}g_0$	$E_s^a$	$-E_{sp}g_1^*$	$-E_{sp}g_2^*$	$-E_{sp}g_3^*$			
$p_x^c$		$-E_{sp}g_1$	$E_p^c$			$E_{xx}g_0$	$E_{xy}g_3$	$E_{xy}g_2$
$p_y^c$		$-E_{sp}g_2$		$E_p^c$		$E_{xy}g_3$	$E_{xx}g_0$	$E_{xy}g_1$
$p_z^c$		$-E_{sp}g_3$			$E_p^a$	$E_{xy}g_2$	$E_{xy}g_1$	$E_{xx}g_0$
$p_x^a$	$E_{sp}g_1^*$		$E_{xx}g_0^*$	$E_{xy}g_3^*$	$E_{xy}g_2^*$	$E_p^a$		
$p_y^a$	$E_{sp}g_2^*$		$E_{xy}g_3^*$	$E_{xx}g_0^*$	$E_{xy}g_1^*$		$E_p^a$	
$p_z^a$	$E_{sp}g_3^*$		$E_{xy}g_2^*$	$E_{xy}g_1^*$	$E_{xx}g_0^*$			$E_p^a$



# Harrison Parameters

- C and A refer to *cation* and *anion*
  - Ga,N for a 3,5 semiconductor in Cubic Zincblende form
  - Si,Si for a pure semiconductor in Diamond form
- $E_s$ ,  $E_{sp}$ ,  $E_{xx}$ ,  $E_{xy}$  are fit to experiment
- $g_0$ ,  $g_1$ ,  $g_2$ ,  $g_3$  are functions of the *k-vectors*
  - k-vectors are phase factors in reciprocal space
  - show how band varies in different space group symmetries
  - $g_0(\mathbf{k}) = e^{-ikd_1} + e^{-ikd_2} + e^{-ikd_3} + e^{-ikd_4}$
  - Tetrahedral directions:  $d_1 = [111]a/4$ ,  $d_2 = [1-1-1]a/4$



# Outline of TB Program

```
kpoints = get_k_points(N)
energy_archive = []
for kpoint in kpoints:
    H = get_TB_Hamiltonian(kpoint)
    energies = eigenvalues(H)
    energies = ev_sort(energies)
    energy_archive.append(energies)
gnuplot_output(energy_archive)
```



# get\_k\_points function

```
def get_k_points(N):  
    kpoints = []  
    kx,ky,kz = 0.5,0.5,0.5 #L Point  
    k_points.append((kx,ky,kz))  
    step = 0.5/float(N)  
    for i in range(N): # Move to Gamma (0,0,0)  
        kx,ky,kz = kx-step,ky-step,kz-step  
        k_points.append((kx,ky,kz))  
    # Similar steps for X (1,0,0) & K (1,1,0)  
    return kpoints
```

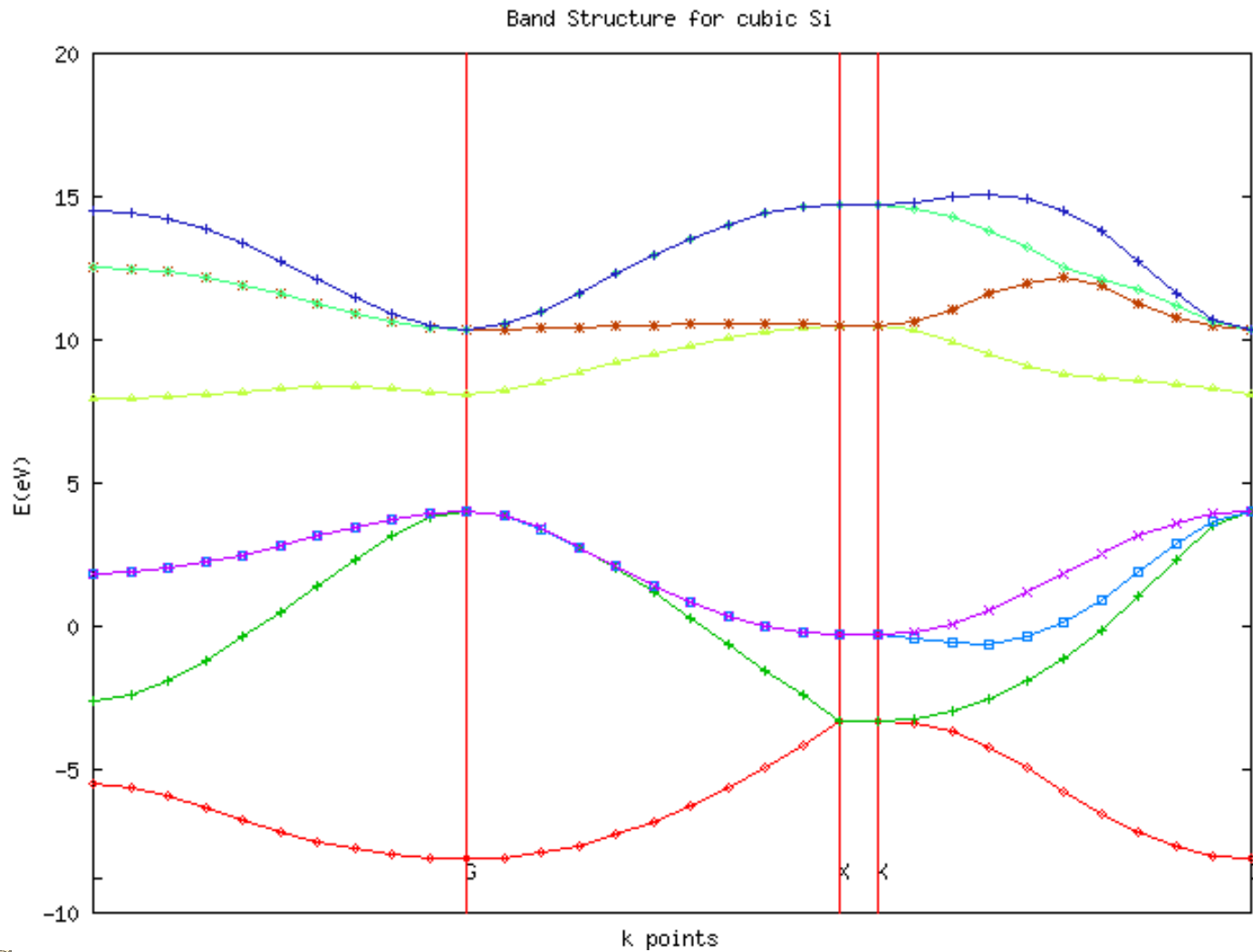


# get\_TB\_Hamiltonian function

```
def get_TB_Hamiltonian(kpoint):  
    phase_factors = get_phases(kpoint)  
    H = zeros((8,8),Complex)  
    H = set_diag_values(H)  
    H = set_off_diag_values(H,phase_factors)  
    return H
```



# Band Structure Output





# Numeric Python References

- <http://numpy.sourceforge.net> NumPy Web Site
- <http://numpy.sourceforge.net/numpy.pdf> NumPy Documentation
- <http://starship.python.net/crew/hinsen/scientific.html> Konrad Hinsen's Scientific Python page, a set of Python modules useful for scientists, including the LeastSquares package.
- <http://starship.python.net/crew/hinsen/MMTK/> Konrad Hinsen's Molecular Modeling Tool Kit, a biological molecular modeling kit written using Numerical Python.
- <http://oliphant.netpedia.net/> Travis Oliphant's Python Pages, including: FFTW, Sparse Matrices, Special Functions, Signal Processing, Gaussian Quadrature, Binary File I/O



# Tight Binding References

W. A. Harrison. *Electronic Structure and the Properties of Solids*.  
Dover (New York, 1989).

D. J. Chadi and M. L. Cohen. "Tight Binding Calculations of the  
Valence Bands of Diamond and Zincblende Crystals." *Phys. Stat.*  
*Solids*. **B68**, 405 (1975).

<http://www.wag.caltech.edu/home/rpm/projects/tight-binding/>

My tightbinding programs. Includes one that reproduces  
Harrison's method, and one that reproduces Chadi and Cohen's  
methods (the parameterization differs slightly).

[http://www.wag.caltech.edu/home/rpm/python\\_course/tb.py](http://www.wag.caltech.edu/home/rpm/python_course/tb.py)

Simplified (and organized) TB program.

