### **Introduction to High Performance Scientific Computing**

**Autumn, 2018** 

Lecture 7

### **Notes**

#### Panopto:

- uploaded lecture 4, part 2
- Lecture 6 did not record, will provide notes on Neural networks early next week

#### MLC VMs

- Meeting with ICT today, will post announcement afterwards
- Fine to use Windows/Anaconda for coursework, labs 2 and 3 (but talk to me if you're unsure how to use git)

# **Working with Python**

- Can be difficult to develop code within functions (can't easily check the values variables are taking on)
- Instead, can initially develop small sections of code in the terminal, or in a script – then have access to all variables used (see Newton's method/sqrt demo on Panopto)
- When functions become long or unwieldy, can define new functions (inside or outside the long ones)
- If using atom, install the linter-pylint package (Preferences → Packages within atom)
  - Also useful for Fortran: language-fortran, linter-gfortran
  - Linters flag errors as you develop code

Last time: unconstrained optimization:

Find  $\mathbf{x}$  so that  $f(\mathbf{x})$  is minimized

- Last time: sketched out basic ideas of classical unconstrained optimization
- Based on Newton's method:  $H|_{\mathbf{x}_0}\mathbf{h} = -\nabla f|_{\mathbf{x}_0}$
- Assumes smooth cost function, however BFGS often works well with noisy or non-smooth data

Last time: unconstrained optimization:

Find  $\mathbf{x}$  so that  $f(\mathbf{x})$  is minimized

- Last time: sketched out basic ideas of classical unconstrained optimization
- Based on Newton's method:  $H|_{\mathbf{x}_0}\mathbf{h} = -\nabla f|_{\mathbf{x}_0}$
- Assumes smooth cost function, however BFGS often works well with noisy or non-smooth data
- scipy.optimize.minimize provides access to a number of powerful optimization methods
  - Have to provide functions defining cost function, and if feasible, the gradient (and for some methods, the Hessian)

Lab 3: constrained optimization:

```
Find \mathbf{x} so that f(\mathbf{x}) is minimized with \mathbf{x} restricted by: equality constraints: g_1(\mathbf{x}) = 0, g_2(\mathbf{x} = 0), ... and/or inequality constraints: h_1(\mathbf{x}) \geq 0, h_2(\mathbf{x} \geq 0), ...
```

- Can apply simple bounds on parameters using L-BFGS and TNC methods
- Nonlinear constraints require either COBYLA or SLSQP
  - Classified as either inequality or equality constraints
  - Functions must be provided defining each constraint

- Several methods available which do not use/require
  derivative: simulated annealing (basin hopping), Powell, Nelder-Mead, genetic algorithms (differential\_evolution)
- Newton-type methods find *local* minima simulated annealing, genetic algorithms, and "brute" search for global minima – no guarantee of success!

# **Python libraries**

- We have looked at scipy.optimize and scikit-learn
- However, there is much, much more out there
- Just in scipy:
- scipy.special
- scipy.integrate
- scipy.fftpack
- scipy.signal
- Try tab completion: scipy. <tab>
   import scipy. <tab>

### **Neural networks**

- There are several libraries for neural networks (not all Python-specific)
- Scikit-learn is a good place to start, but other packages typically used for large networks
  - Keras, Torch, Tensorflow, ...
  - We will take a quick look at Tensorflow towards the end of term

### Python notes

- We have focused on Python as tool for 'simulations'
- Also, very useful for data analysis
  - Can easily read most common data formats into Numpy ndarrays (using e.g. np.loadtxt)
  - Many useful commands for exploring/analyzing 2-D arrays
    - min, max, mean, var, argmin, argmax, ... see numpy reference on course webpage
  - For datasets with many variables, should consider Pandas package for data analysis
    - Provides mix of database and numpy tools
    - Not very intuitive or easy to learn though!

### Fortran intro

- Fortran is a compiled language (like C++) designed for scientific computing (like Matlab)
- Fortran has evolved substantially from F66 to F77, F90, Fortran 2008.
- Fortran 77 was the dominant standard, but is now outdated, clumsy.
  - But, python, matlab and other software rely on fortran 77 libraries (especially lapack)
- Fortran 90 is a powerful, completely modern programming language.
- Typically, F77 codes have *.f* extension, F90 codes use *.f90*

### Interpreted vs compiled languages

Determines how code is converted into machine instructions

#### Interpreter:

- Goes through code line-by-line, translates into machine language, and executes
- Allows for "interactive" programming as in Matlab and Python
- However, cannot optimize over blocks of code (e.g. a for loop)

### Interpreted vs compiled languages

Determines how code is converted into machine instructions

#### Interpreter:

- Goes through code line-by-line, translates into machine language, and executes
- Allows for "interactive" programming as in Matlab and Python
- However, cannot optimize over blocks of code (e.g. a for loop)

### **Compiler:**

- Programs stored in file(s) called source code
- Compiler analyzes the source code, optimizes where possible, and generates *object* files
- A linker converts object files into an executable file



Interactive programming is not possible, but code may run much faster.

### **Basic code structure**

```
! Basic Fortran 90 code structure
!1. Header
program template
    !2. Variable declarations (e.g. integers, real numbers,...)
    !3. basic code: input, loops, if-statements, subroutine calls
    print *, 'template code'
!4. End program
end program template
! To compile this code:
! $ gfortran —o template.exe template.f90
! To run the resulting executable: $ ./template.exe
```

Note: Indentation is optional, but *highly* recommended (makes code readable).

# See template.f90 Imperial College London

Compute *sin(i)*, i=1,2,3, ..., N

#### **Declare a few variables:**

!1. Header:

```
!2. Variable declarations:
   implicit none !means all variables in code must be declared
   integer :: i1,j1,N
   real(kind=8) :: var1, var2
   real(kind=8), dimension(10) :: array1
```

Compute *sin(i)*, i=1,2,3, ..., N

#### Read data:

!3. basic code: input, loops, if—statements, subroutine call

```
!read data from data.in
open(unit=10, file='data.in')
    read(10,*) N
close(10)
```

Compute *sin(i)*, i=1,2,3, ..., N

#### Main code:

```
!check that N is smaller than size of array1:
   if (N <= size(array1)) then</pre>
```

Compute *sin(i)*, i=1,2,3, ..., N

#### Main code:

```
!check that N is smaller than size of array1:
   if (N <= size(array1)) then

!compute sin(x) where x = 1,2,3,...,N
   do i1 = 1,N !loop from 1 to N
      var1 = dble(i1) !convert integer to double-prec number
      array1(i1) = sin(var1)
   end do</pre>
```

```
Compute sin(i), i=1,2,3, ..., N
Main code:
!check that N is smaller than size of array1:
   if (N <= size(array1)) then</pre>
       !compute sin(x) where x = 1,2,3,...,N
       do i1 = 1,N !loop from 1 to N
           var1 = dble(i1) !convert integer to double-prec number
          array1(i1) = sin(var1)
       end do
       !print 1st N elements of array
       print *, 'array1=',array1(1:N)
   else
       print *, 'N must be smaller than', size(array1)
        ST<sub>0</sub>P
   end if
```

Compute *sin(i)*, i=1,2,3, ..., N

#### Compile and run:

'Cleaner' code: move loop to a subroutine:

#### 3. Main code:

```
!check that N is smaller than size of array1:
    if (N <= size(array1)) then
        !compute sin(x) where x = 1,2,3,...,N
        call calculations(N,array1)

        !print 1st N elements of array
        print *, 'array1=',array1(1:N)
    else
        print *, 'N must be smaller than', size(array1)
    end if</pre>
```

Need subroutine, calculations, which take N as input and returns array1

```
!subroutine calculations
subroutine calculations(N,array)
    implicit none
    integer, intent(in) :: N
    real(kind=8), dimension(10), intent(out) :: array
    integer :: i1
    real(kind=8) :: var1
    do i1 = 1,N !loop from 1 to N
          var1 = dble(i1) !convert integer to real number
          array(i1) = sin(var1)
    end do
end subroutine calculations
```

See f90example2.f90

### Floating point numbers in Fortran

Single precision: 7 significant figures (4 bytes), not often used

```
!single precision
real :: var1
real(kind=4) :: var2
real*4 :: var3
```

 Double precision: 15 significant figures (8 bytes), almost always want double precision in scientific computing

```
!double precision
real(kind=8) :: dvar1
real*8 :: dvar2
double precision :: dvar3
```

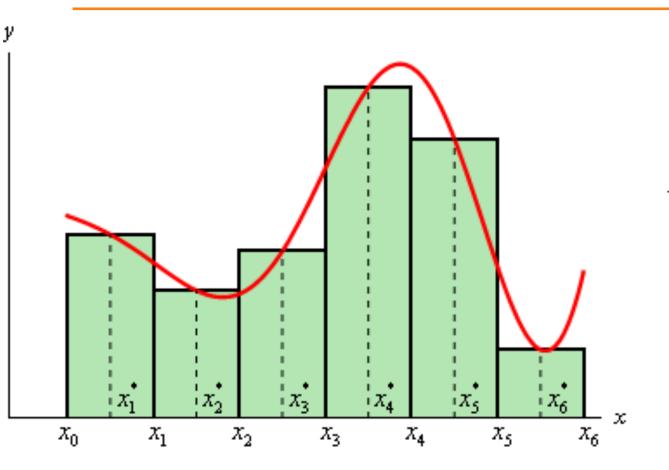
• All three double precision variable declarations are equivalent, but the *real(kind=)* syntax is "more standard" than the others

### Floating point numbers in Fortran

- Use dble to convert integer to ensure double precision
- Write numbers with "d" after decimal for double precision 2.d0 or 3.2d0
- Can also include a flag when compiling to force singleprecision numbers to be treated as double precision.

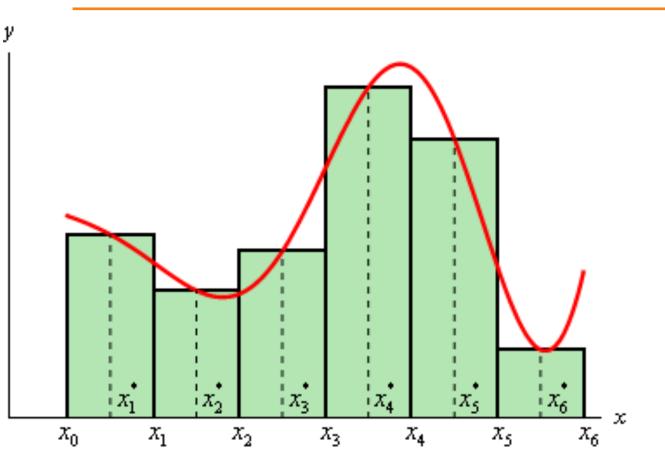
#### In gfortran:

\$ gfortran -freal-4-real-8



Estimate integral with midpoint rule,

$$I = \int_{x_0}^{x_6} f(x) dx$$

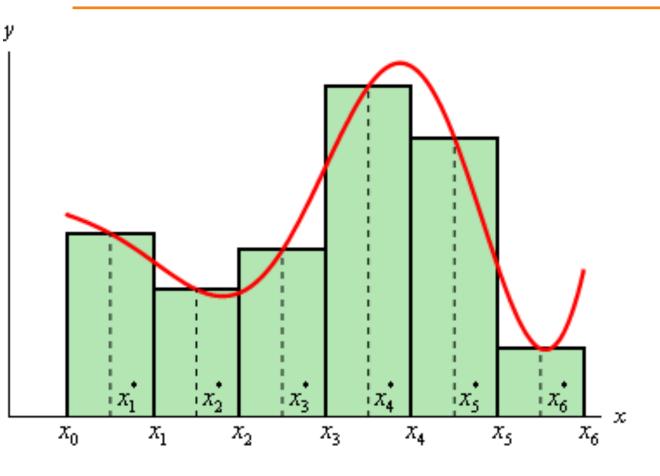


Estimate integral with midpoint rule,

$$I = \int_{x_0}^{x_6} f(x) dx$$

1. Compute:

$$f(x_1^*), f(x_2^*), \dots$$



Estimate integral with midpoint rule,

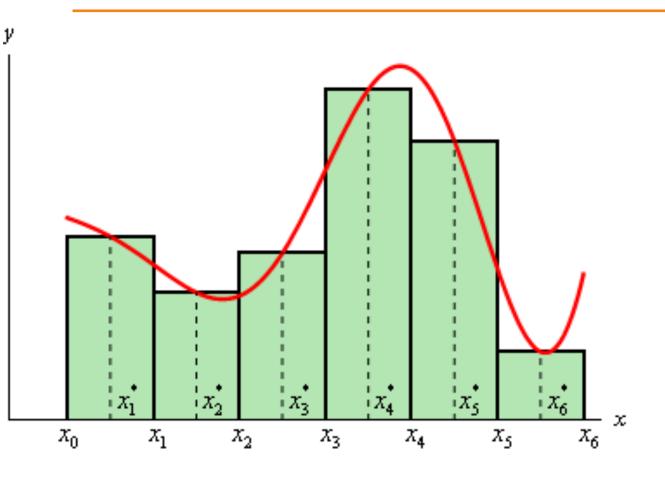
$$I = \int_{x_0}^{x_6} f(x) dx$$

1. Compute:

$$f(x_1^*), f(x_2^*), \dots$$

2. Compute areas of rectangles:

$$I_1 = (x_1 - x_0) * f(x_1^*)$$



Estimate integral with midpoint rule,

$$I = \int_{x_0}^{x_6} f(x) dx$$

1. Compute:

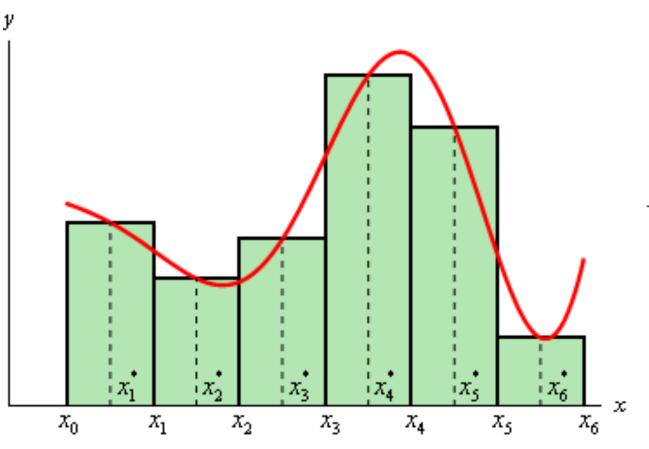
$$f(x_1^*), f(x_2^*), \dots$$

2. Compute areas of rectangles:

$$I_1 = (x_1 - x_0) * f(x_1^*)$$

3. Sum areas:

$$I \approx I_1 + I_2 + I_3 + \dots$$



Estimate integral with midpoint rule,

$$I = \int_0^1 \frac{4}{1+x^2} dx$$

#### **Basic steps:**

- 1. Read in number of intervals, N
- 2. Compute interval size, dx = 1.d0/N
- 3. Loop over the N intervals, within each interval:
  - 1. compute the midpoint,  $x_m$
  - 2. evaluate  $4/(1+x^2)$  at midpoint
  - 3. compute area of i<sup>th</sup> rectangle:  $sum_i = dx*f(x_m)$

See midpoint.f90

#### **Basic steps:**

1. Read in number of intervals, N

```
!read data from data.in
  open(unit=10, file='data.in')
     read(10,*) N
  close(10)
```

#### **Basic steps:**

- 1. Read in number of intervals, N
- 2. Compute interval size, dx = 1.0/N

```
!read data from data.in
  open(unit=10, file='data.in')
    read(10,*) N
  close(10)

dx = 1.d0/dble(N) !interval size
```

#### **Basic steps:**

- 3. Loop over the N intervals, within each interval:
  - 1. compute the midpoint,  $x_m$
  - 2. evaluate 4/(1+x<sup>2</sup>) at midpoint
  - 3. compute area of i<sup>th</sup> rectangle:  $sum_i = dx*f(x_m)$

```
!loop over intervals computing each interval's contribution to integral
    do i1 = 1,N
        xm = dx*(dble(i1)-0.5d0) !midpoint of interval i1
        call integrand(xm,f)
        sum_i = dx*f
        sum = sum + sum_i !add contribution from interval to total
integral
    end do
```

Here, integrand, is a subroutine which evaluates 4/(1+x²) at xm

Here, integrand, is a subroutine which evaluates 4/(1+x²) at xm

## Fortran reference: do loops

```
integer :: i1,start,finish,step
!do-loop structure
    do i1 = start,finish,step
     !commands which depend on i1 in some way
    end do
```

- do-loop index must be an integer
- Use exit to break a do-loop

### Fortran reference: if-then

```
!if-then structure
   if (boolean expression here) then
        !some commands
   elseif (another boolean) then
        !more commands
   elseif (another boolean) then
        leven more commands
   else
        !more commands
   end if
```

- Can have arbitrary number of elseif blocks
  - Can also have just the 1st if statement

### Fortran reference: if-then

### Relational operators:

```
less than
.lt. or
                   less than or equal
.le. or
         <=
                   equal
.eq. or
                   greater than or equal
.ge. or
             >=
                   greater than
             >
.gt. or
             /=
                   not equal
.ne. or
.not.
                   not
                   and
.and.
                   inclusive or
.or.
```