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

**Autumn, 2018** 

Lecture 10

### **Announcements**

- HW 2 will be posted around 8pm today
- HW1 feedback will be provided around Wednesday next week (solutions will be posted at the end of this week)
- A new VM is being installed on computer in Huxley 408 and 410. Should be substantial improvement on MLC VMs
  - Will post announcement when they are ready (hopefully sometime tomorrow!)

# **Today**

- A few comments on Fortran
- F2Py with fortran modules
- Basic computer architecture

- Do not attempt to guess Fortran syntax! look it up instead
- Do not develop Fortran code in Python using f2py get a Fortranonly code working first
- Test code (compile and run) after adding a few (4-5) lines of code.
   Don't write 100 lines and then test!
- Lecture 7: code structure, variable types, loops, if-then, subroutines
- Lecture 8: allocatable arrays, functions modules
- Lecture 9: Lapack, f2py
  - Note: On VMs, you may need to use "f2py3" instead of "f2py"

- Fortran is case insensitive, the variables a and A are treated as the same variable
- Python is case sensitive, a and A are treated as different variables
- This can cause problems when using f2py. Safe approach: use all lowercase letters for variable names
- Slicing/indexing are a little different in Fortran and Python
  - Fortran: counting starts from 1, the first n elements of A are A(1:n).

#### **Debugging code**

- If the compiler is giving you a series of errors, look at the topmost error message. It will include a line number
- If the code runs but crashes with a segmentation fault: probably a problem with array indices (e.g. trying to access the 12<sup>th</sup> element of a dimension(11) array
- If you can't tell where the code is crashing, add print statements, e.g. print \*, 1 --some code- print \*, 2
   When the code is run, if the 1 prints to screen but not the 2, you know where the problem is.
- If the code runs, but gives the wrong answer, add print statements outputting values of variables, try a small problem size where you know what values the variables can take

- Modules consist of:
  - 1. Module variables
  - 2. Module sub-programs
- Module variables are "available" throughout the module
  - They do not need to be declared within sub-programs
  - They do not need to be provided as input/output in the subprogram header
- Module variables and module sub-programs are also "available" in any program or sub-program that uses the module
- A module by itself doesn't do anything
  - There should be a "main" program which uses it
  - You can compile a module by itself: gfortran -c module.f90
  - But to generate an executable, you need a <u>program:</u> gfortran -o program.exe module.f90 program.f90 When re-compiling, first remove the previous .mod file

Compiling code that uses lapack routines:

```
$ gfortran -c program.f90
$ gfortran -o program.exe program.o -llapack
or
```

\$ gfortran -o program.exe program.f90 -llapack

#### Similarly, with f2py:

\$ f2py -llapack -c program.f90 -m module\_name

- F2Py will recognize subroutines and functions in modules
- What about variables?
  - Try f2py with circle module from last week (f2pymodule\_circle.f90)

```
$ f2py -c f2pymodule_circle.f90 -m cmod
```

```
In [10]: import cmod
In [11]: cmod.<tab>
cmod.circle cmod.so
```

Need to look at cmod.circle

#### **Need to look at** *cmod.circle*:

```
In [12]: cmod.circle?
Docstring:
'd'-scalar
initialize_pi()
Wrapper for ``initialize_pi``.
circumference = circumference(radius)
Wrapper for ``circumference``.
Parameters
radius : input float
Returns
circumference: float
                             And similar info for "area"
area = area(radius)
```

#### How do we access variables and methods in cmod.circle?

#### Can initialize pi in python:

```
In [9]: cmod.circle.pi
Out[9]: array(0.0)
In [10]: cmod.circle.pi = pi
In [11]: cmod.circle.pi
Out[11]: array(3.141592653589793)
```

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Can also initialize allocatable arrays, see f2pymodule\_circle\_array.f90:

```
!module for computing circumference, area, and "mass" of circle
module circle
   implicit none
   real(kind=8) :: pi
   real(kind=8), allocatable, dimension(:) :: weights,mass
   save
subroutine compute_mass(radius,mass)
    !compute mass = weights*area
    implicit none
    real(kind=8), intent(in) :: radius
    real(kind=8), intent(out) :: mass(:)
    mass = weights*area(radius)
end subroutine compute_mass
```

#### **Basic steps:**

1. Compile with f2py: \$ f2py -c f2pymodule\_circle\_array.f90 -m cmoda

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- 2. Import module in python: In [4]: import cmoda
- 3. Initialize variables:

```
In [15]: cmoda.circle.pi=pi
In [16]: cmoda.circle.weights=arange(5)
In [17]: cmoda.circle.pi,cmoda.circle.weights
Out[17]: (array(3.141592653589793), array([ 0., 1., 2., 3., 4.]))
```

#### **Basic steps:**

- 1. Compile with f2py: \$ f2py -c f2pymodule\_circle\_array.f90 -m cmoda
- 2. Import module in python: In [4]: import cmoda
- 3. Initialize variables:

```
In [15]: cmoda.circle.pi=pi
In [16]: cmoda.circle.weights=arange(5)
In [17]: cmoda.circle.pi,cmoda.circle.weights
Out[17]: (array(3.141592653589793), array([ 0., 1., 2., 3., 4.]))
```

#### 4. Compute mass:

```
In [18]: cmoda.circle.compute_mass(2.0)
In [19]: cmoda.circle.mass
Out[19]: array([ 0. , 12.56637061, 25.13274123, 37.69911184, 50.26548246])
```

Exercise: How does f2py treat the subroutine below with weights now provided as input? Can n be removed from the input?

```
subroutine compute_mass(n,weights,radius)
!compute mass = weights*area
implicit none
integer :: n
real(kind=8), dimension(n), intent(in) :: weights
real(kind=8), intent(in) :: radius

if (.not. allocated(mass)) allocate(mass(size(weights)))
mass = weights*area(radius)
end subroutine compute_mass
```

## Notes on compiling

• Up to now:

\$ gfortran -o program.exe program.f90 -llapack

But typically want to turn on optimization -O flag:

```
$ gfortran -03 -c program.f90
```

-O3 is highest level of optimization (can also use -O1, -O2)

## Notes on compiling

Look at midpoint\_time.f90 compiled with and without -O3:

```
$ ./mt2.exe
elapsed cpu time (seconds) = 0.32510499999999998
elapsed wall time (seconds) = 0.326231003
N= 25600
sum= 3.1415926537169634
error= 1.2717027431108363E-010
$ ./mt2 03.exe
elapsed wall time (seconds) = 0.144617006
N= 25600
sum= 3.1415926537169634
error= 1.2717027431108363E-010
```

- Optimization can make a substantial difference
- f2py uses -O3 by default

- Many built-in functions in Fortran can operate on arrays
  - e.g. with an array (or matrix), x, can compute exp(x) all at once
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- Consider a simple example: computing the sine of a large array, sin(x)

```
#compute sin(x) using loop
def sin_loop(x):
    s = np.zeros_like(x)
    for i,xi in enumerate(x):
        s[i] = np.sin(xi)
    return s

#vectorized
def sin_vec(x):
    s = np.sin(x)
    return s
```

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  - e.g. with an array (or matrix), x, can compute exp(x) all at once
- We know it is important to avoid loops in Python, does the same apply to Fortran?
- Consider a simple example: computing the sine of a large array, sin(x)
- Python: sine array.py

```
#compute sin(x) using loop
def sin_loop(x):
    s = np.zeros_like(x)
    for i,xi in enumerate(x):
        s[i] = np.sin(xi)
    return s

#vectorized
def sin_vec(x):
    s = np.sin(x)
    return s
```

```
Python: In [12]: x = np.random.randn(500000)

In [13]: timeit sin_loop(x)
808 ms ± 32.2 ms per loop (mean ± std. dev. of 7 runs, 1 loop each)

In [15]: timeit sin_vec(x)
11.2 ms ± 346 μs per loop (mean ± std. dev. of 7 runs, 100 loops each)
```

- Vectorized version is 70x faster
- The increase in speed does depend on the array size
- What about Fortran?

- What about Fortran? See sine\_array.f90
- Compile module with f2py: \$ f2py -c sine\_array.f90 -m s1

#### Import and run in python:

```
In [25]: from s1 import sine_array as sa
In [26]: sa.x = x
In [27]: n = x.size
In [28]: timeit sa.sin_loop(n)
10.5 ms ± 245 μs per loop (mean ± std. dev. of 7 runs, 100 loops each)
In [29]: timeit sa.sin_vec(n)
10.6 ms ± 192 μs per loop (mean ± std. dev. of 7 runs, 100 loops each)
```

- What about Fortran? See sine\_array.f90
  - Vectorization is essential in Python (or Matlab)
  - Much less important in Fortran
    - Provided you optimize properly when compiling!