Advanced Programming in Quantitative Economics

Introduction, structure, and advanced programming techniques

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Outline

Floating point numbers and rounding errors

Efficiency

System Algorithm Operators Loops Loops and conditionals Conditionals Memory

Day 2 - Afternoon

13.00L Background of computations

- Floating point numbers and rounding errors
- Compilers and CPUs
- Computing environment at Aarhus University

14.30P Tutorial

- Simulate data duration model
- Apply concepts of the day
- Think of rounding errors

16.00 End

Precision

```
Not all numbers are made equal...
Example: What is 1/3 + 1/3 + 1/3 + ...?
```

Listing 1: precision/onethird.ox

```
main()
{
    decl i, j, dD, dSum;
    dD= 1/3;
    dSum= 0.0;
    for (i= 0; i < 10; ++i)
        for (j= 0; j < 3; ++j)
            {
            print (dSum~i~(dSum-i));
            dSum+= dD; // Successively add a third
        }
}</pre>
```

See outcome: It starts going wrong after 16 digits...

Representation

- Integers are represented exactly using 4 bytes/32 bits, in range [INT_MIN, INT_MAX] = [-2147483648, 2147483647]
- ▶ Doubles are represented in 64 bits. This gives a total of $2^{64} \approx 1.84467 \times 10^{19}$ different numbers that can be represented.



Double floating point format (Graph source: Wikipedia)

Split double in

- Sign (one bit)
- Exponent (11 bits)
- Fraction or mantissa (52 bits)

Double representation

$$x = \begin{cases} (-1)^{\text{sign}} \times 2^{1-1023} \times 0.\text{mantissa} & \text{if exponent}=0x.000\\ (-1)^{\text{sign}} \times \infty & \text{if exponent}=0x.7\text{ff}\\ (-1)^{\text{sign}} \times 2^{\text{exponent}-1023} \times 1.\text{mantissa} & \text{else} \end{cases}$$

Note: Base-2 arithmetic

Sign	Expon	Mantissa	Result
0	0x.3ff	0000 0000 0000	$-10 \times 2^{(1023-1023)} \times 1.0$
0	0x.3ff	0000 0000 0001	$ = 0 -1^0 \times 2^{(1023 - 1023)} \times 1.00000000000000222 $
			= 1.0000000000000222
0	0x.400	0000 0000 0000	$-1^0 imes 2^{(1024-1023)} imes 1.0$
			= 2
0	0x.400	0000 0000 0001	$-1^{0} \times 2^{(1024-1023)} \times 1.00000000000000222$
			= 2.0000000000000444

Bit weird...

Consequence: Addition

Let's work in Base-10 arithmetic, assuming 4 significant digits:

Sign	Exponent	Mantissa	Result	х	
+	4	0.1234	$0.1234 imes 10^{4}$	1234	
+	3	0.5670	0.5670×10^{3}	567	

What is the sum?

Consequence: Addition

Let's work in Base-10 arithmetic, assuming 4 significant digits:

Sign	Exponent	Mantissa	Result	x	
+	4	0.1234	0.1234×10^{4}	1234	
+	3	0.5670	0.5670×10^{3}	567	

What is the sum?

Sign	Exponent	Mantissa	Result	Х	
+	4	0.1234	$0.1234 imes10^4$	1234	
+	4	0.0567	$0.0567 imes10^4$	567	
+	4	0.1801	$0.1801 imes 10^4$	1801	

Shift to same exponent, add mantissas, perfect

-Floating point numbers and rounding errors

Consequence: Addition II

Let's use dissimilar numbers:

Sign	Exponent	Mantissa	Result	Х
+	4	0.1234	$0.1234 imes10^4$	1234
+	1	0.5670	$0.5670 imes 10^1$	5.67

What is the sum?

Floating point numbers and rounding errors

Consequence: Addition II

Let's use dissimilar numbers:

Sign	Exponent	Mantissa	Result	Х
+	4	0.1234	$0.1234 imes10^4$	1234
+	1	0.5670	$0.5670 imes10^1$	5.67

What is the sum?

Sign	Exponent	Mantissa Result	X
+	4	$0.1234 0.1234 imes 10^4$	1234
+	4	$0.0005 \frac{67}{67}$ $0.05 imes 10^4$	5
+	4	$0.1239 0.1239 imes 10^4$	1239

Shift to same exponent, add mantissas, loose precision...

Further consequence:

Add numbers of similar size together, preferably! In Ox/C/Java/Matlab/Octave/Gauss: 16 digits (\approx 52 bits) available instead of 4 here

Consequence: Addition III

Check what happens in practice:

Listing 2: precision/accuracy.ox

```
main()
{
    decl dA, dB, dC;
    dA= 0.123456 * 10^0;
    dB= 0.471132 * 10^15;
    dC= -dB;
    println ("a: ", dA, ", b: ", dB, ", c: ", dC);
    println ("a + b + c: ", dA+dB+dC);
    println ("(a + b) + c); ", dA+(dB+dC);
}
```

Consequence: Addition III

Check what happens in practice:

Listing 3: precision/accuracy.ox

```
main()
{
    decl dA, dB, dC;
    dA= 0.123456 * 10^0;
    dB= 0.471132 * 10^15;
    dC= -dB;
    println (<u>"a: ", dA, ", b: ", dB, ", c: "</u>, dC);
    println (<u>"a + b + c: ", dA+dB+dC);
    println ("(a + b) + c: ", dA+(dB+dC);
    println (<u>"(a + b) + c: "</u>, (dA+dB)+dC);
}</u>
```

results in

```
Dx Professional version 6.00 (Linux_64/MT) (C) J.A. Doornik, 1994-2009
a: 0.123456, b: 4.71132e+14, c: -4.71132e+14
a + b + c: 0.125
a + (b + c): 0.123456
(a + b) + c: 0.125
```

Other hints

- Adding/subtracting tends to be better than multiplying
- Hence, log-likelihood $\sum \log \mathcal{L}_i$ is better than likelihood $\prod \mathcal{L}_i$
- Use true integers when possible
- Simplify your equations, minimize number of operations
- Don't do x = exp(log(z)) if you can escape it

Other hints

- Adding/subtracting tends to be better than multiplying
- Hence, log-likelihood $\sum \log \mathcal{L}_i$ is better than likelihood $\prod \mathcal{L}_i$
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- Simplify your equations, minimize number of operations
- Don't do x = exp(log(z)) if you can escape it

(Now forget this list... use your brains, just remember that a computer is not exact...)

On architecture, algorithms, languages, and machines

Why do we program (repeat)?

To get to the results we need, in a fashion that is controllable, where we are free to implement the newest and greatest, and where we can be 'reasonably' sure of the answers

What is important here?

- 1. To get *correct* code (\approx maintainable, clear, adjustable)
- 2. To get *efficient* code (\approx quick?)

Correct code: See rest of course

Efficiency II

Efficient code: Depends on

- system (processor, memory size/structure, drives, network, operating system)
- 2. language
- 3. algorithm
- 4. more...?

Use an example:

What is the sum of all returns on the SP500 stock index, over 20 years?

$$S = \sum_{t=1}^{N} r_t$$

Efficiency: System

What do you prefer:

- 1. Old Apple II (1Mhz, 64KB)
- 2. Older Power Mac (2.5GHz, PowerPC processor)
- 3. This laptop (C2Duo 1.4Ghz, 1GB)
- 4. Home machine (I7 quadcore 2.66Ghz, 6GB)
- 5. Work machine (Nehalem dual quadcore 2.66Ghz, 12GB)
- 6. SARA Lisa supercomputer (536 dual quadcores, loads of memory)

Difference?

Efficiency: System II

Difference:

- 1. Apple does not even have the memory to store the data. Read from tape?
- Mac: PowerPC had low-level possibilities to sum vectors of numbers quickly.
- 3. C2D laptop: Not bad, fraction of a second
- 4. Core I7: Better memory management in CPU, quicker
- 5. 2xI7: Can one use 8 cores for the summation?
- 6. SARA: Can one use 536 x 2 x 4 cores for summing 1000 numbers?

In general: Recent \equiv better... More memory \equiv better... More cores/CPUs $\not\equiv$ better <code>necessarily</code>



Figure: Itanium CPU architecture (Source: Wikipedia)

- Hard drive \rightarrow memory \rightarrow CPU
- Efficiency differs per CPU architecture
- Does your data fit in memory/cache? Or should you go back to hard drive to reload it
- Communication between parts of system: Speed matters



Multi-core?

$$S = \sum_{t=1}^{N} r_t = \sum_{t=1}^{N/4} r_t + \sum_{t=N/4+1}^{N/2} r_t + \sum_{t=N/2+1}^{3N/4} r_t + \sum_{t=3N/4}^{N} r_t$$
$$= S_1 + S_2 + S_3 + S_4$$

Use 4 cores for summation?

- 1. Decide on how to split the problem
- 2. Send each core the right amount of data
- 3. Wait for the answers to come back

4. Sum
$$S = S_1 + S_2 + S_3 + S_4$$

Drawback: Overhead in communication between cores/nodes, not always faster...

(GPU, supercomputer, MPI, ...)

Algorithm

```
int i, iN; // C code
                                              C FORTRAN CODE
double dS;
                                                      INTEGER i, iN
                                                      REAL dS
iN= ...;
dS = 0.0:
                                                      iN= ...
for (i = 0; i < iN; ++i)
                                                      dS = 0.0
  dS += vR[0][i]:
                                                     DO 10 i= 1, iN
                                                        dS = dS + vR(0,i)
                                                10
                                                      CONTINUE
decl i. dS. iN: // Ox code
                                              decl dS:
                                              // Ox code, using function
iN= sizerc(vR);
                                              dS= sumc(vR);
dS = 0;
for (i = 0; i < iN; ++i)
  dS += vR[i];
```

- Fortran or C would be quickest, as compiled code can be executed directly by CPU
- Ox code is similar to C, but far slower: Each command has to be 'translated' to executable code, so even a loop takes time...
- Ox version using sumc is virtually as fast as Fortran or C

Algorithm II

Think...

Where do returns come from? First difference of log prices

$$p_t = \log P_t$$
$$r_t = p_t - p_{t-1}$$

What does this imply for the sum of the returns?

Algorithm II

Think...

Where do returns come from? First difference of log prices

$$p_t = \log P_t$$
$$r_t = p_t - p_{t-1}$$

What does this imply for the sum of the returns?

$$S = \sum_{t} r_t = \sum_{t} (p_t - p_{t-1})$$

= $(p_1 - p_0) + (p_2 - p_1) + \dots + (p_N - p_{N-1}) = p_N - p_0$

- Quicker algorithm
- This does more than quicker computer, quicker language...
- Think/measure where your program is slow: Work on bottlenecks first

Algorithm III

Good algorithms: Where to find?

- BLAS library (C/Fortran)
- LAPack library (C/Fortran)
- Within higher-level languages (Matlab/Octave/Ox/Gauss)

Examples: Regression, matrix inversion, random number generator, optimization... Easy to code little robustly, hard to get right...

Low-level 'algorithms': Operators

Not all operators are equal:

Speed	Speed Operation		Example		
Fast	Integer addition and subtraction	I+J;	I-J		
	Floating point addition and subtraction	A+B;	A-B		
	Int multiply, fl. point multiply/divide	I*J,	A*B,	A/B	
÷	Integer divide	I/J			
	Exponentiation to a positive int. constant	A^2			
	Exponentiation to a positive int. variable	A^I			
Slow	Exponentiation to a floating point variable	A^B			
What is t	he difference?				

Operators, do's and don'ts

Looking *only* at memory use and speed of operators: Don't Do A/2.0 0.5*A A^2 A * A (B - C) * (E + F)B (E + F) - C*(E+F)A^0.5 sqrt(A) B/C/D/E B/(C*D*E)T1 = B + C, T2 = D*E, A = T1 + T2 A = B+C+D*ET1 = X + Y, A = T1 + log(T1) $A = (X + Y) + \log(X + Y)$ Note: Think also of clarity of your program...

Loops and execution: Gather

Loops take most time (even empty loops take time!) Preferably use one loop, gather statements together

```
// Don't
// Do
dS= dQ= 0;
for (i= 0; i < iN; ++i)
dS+= vR1[i];
for (i= 0; i < iN; ++i)
dQ+= vR2[i];
dQ+= vR2[i];
}
/ Do
dS= dQ= 0;
for (i= 0; i < iN; ++i)
dS+= vR1[i];
dQ+= vR2[i];
}</pre>
```

Loops and execution: If-then-else

Conditional constructs are hard for CPU, take if-then-else outside of loop if possible

```
// Don't
dS= 0;
for (i= 0; i < iN; ++i)
if (iDay == 4)
dS+= vR1[i];
else
dS+= vR2[i];</pre>
```

```
// Do
dS= 0;
if (iDay == 4)
for (i= 0; i < iN; ++i)
    dS+= vR1[i];
else
for (i= 0; i < iN; ++i)
    dS+= vR2[i];</pre>
```

lf-then-else

Test for most common condition first

```
// Don't
if (iDay == 4)
dS= sumr(vR1);
else
dS= sumr(vR2);
```

```
// Do
if (iDay != 4)
  dS= sumr(vR2);
else
  dS= sumr(vR1);
```

Memory and speed

Should you care about memory?



Memory and speed

Should you care about memory? Yes and no... Make some speed difference.

Matrices
$$A = \begin{pmatrix} 0 & 1 & 2 \\ 3 & 4 & 5 \end{pmatrix}$$
 is stored:

- Column-wise: Fortran, Matlab
- Row-wise: C

0

Row-wise (with row pointers): Ox



APQE11-2b
Efficiency

Function calls

When you call a function:

- Local variables declare their memory for every function call. Can be expensive, for many/large variables. For speed: Use globals...
- It takes time: Putting small procedures in-line is quicker
- the passing of parameters also takes time. Passing pointers to variables, or using globals, is again faster.
- In general, using memory takes time (in C: new). Think about the constructs you need.
- Pointers are faster than copying data

but:

Forget this for now...

Structure!

Your time is more valuable than computer time...

Concentrate on *structured*, *readable* code. Afterwards, with *bug-free* code

- profile your code to see which routine takes most time
- think of putting mostly used data constructs global
- aforementioned tricks can shave 5-20% off execution time.
 Better algorithm can improve infinitely more...
- move from laptop to recent desktop, think of using MPI, split program in smaller tasks, etc.