



# Numerical Error Analysis for Statistical Software on Multi-Core Systems

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# Outline

1. Numerical Errors and Accuracy Control
2. A Method for Accuracy Control
  - e.g. Result : 9.676383250285714\*10<sup>3</sup>  
  
Accurate 12 digits      Contaminated digits  
due to rounding errors  
Accuracy: 12 digits
3. Applied to a Statistical Software – R
4. Parallelization on Multi-Core Systems
5. Conclusion

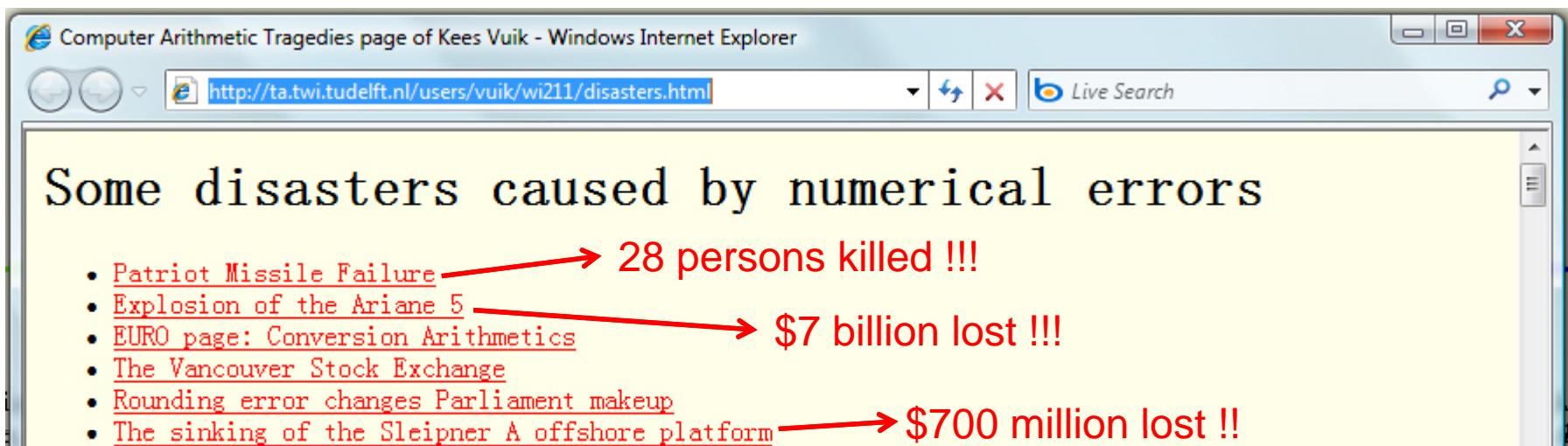


# Disasters Caused by Numerical Errors

**“Numerical precision is the very soul of science.”**

-- D'Arcy Wentworth

<http://ta.twi.tudelft.nl/users/vuik/wi211/disasters.html>



The screenshot shows a Windows Internet Explorer window with the title "Computer Arithmetic Tragedies page of Kees Vuik - Windows Internet Explorer". The address bar contains the URL <http://ta.twi.tudelft.nl/users/vuik/wi211/disasters.html>. The main content area has a yellow background and displays the following text:

Some disasters caused by numerical errors

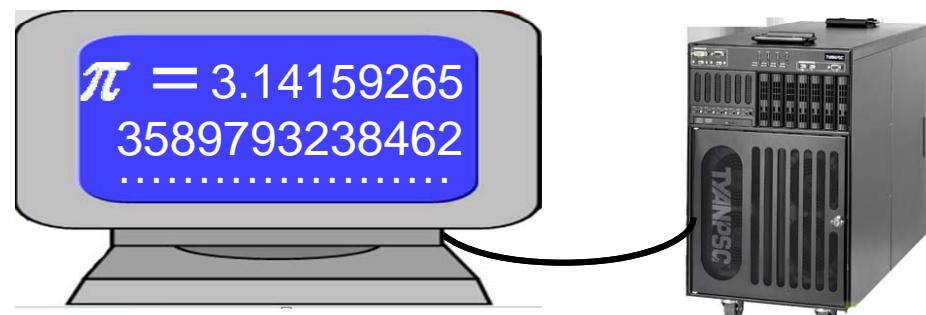
- [Patriot Missile Failure](#) → 28 persons killed !!!
- [Explosion of the Ariane 5](#) → \$7 billion lost !!!
- [EURO page: Conversion Arithmetics](#)
- [The Vancouver Stock Exchange](#)
- [Rounding error changes Parliament makeup](#)
- [The sinking of the Sleipner A offshore platform](#) → \$700 million lost !!



# Schemes to Evaluate Numerical Quality

- Compare the result with a high precision reference
  - Reference is obtained by repeating the computation in arithmetic of increasing precision.

$\pi = 3.14$





# Schemes to Evaluate Numerical Quality

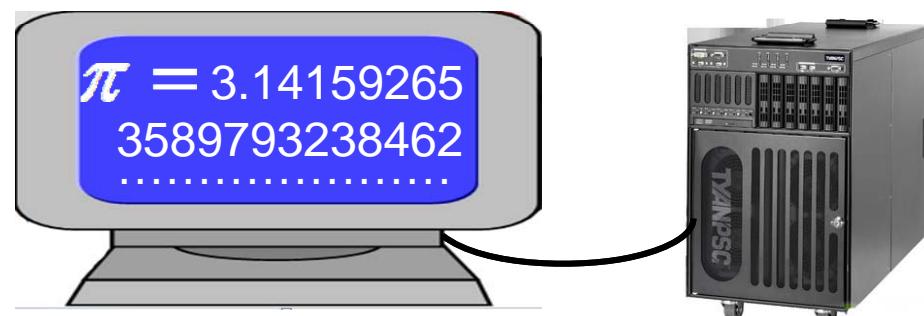
- Compare the result with a high precision reference
  - How many digits are sufficient?
  - Low performance:
    - Single precision performance >> Double precision (e.g. GPU, CellBE, etc.).
    - Arbitrary high precision is several orders of magnitude times slower.
  - Hard to modify source codes to generate high precision reference:
    - millions of lines of source code.
    - Unknown libraries → source code available (readable) ?
    - Mismatch in memory:
      - Fortran: common
      - C/C++ : union



# Schemes to Evaluate Numerical Quality

- Compare the result with a high precision reference
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$$\pi = 3.14$$



- Interval arithmetic
  - e.g. Your result is in  $(-\infty, +\infty)$  guaranteed
- Probabilistic error analysis



# Probability Distribution of the Rounding Error

- The calculated result in a program

$$\hat{R} = r + \varepsilon_{\text{rounding}}$$

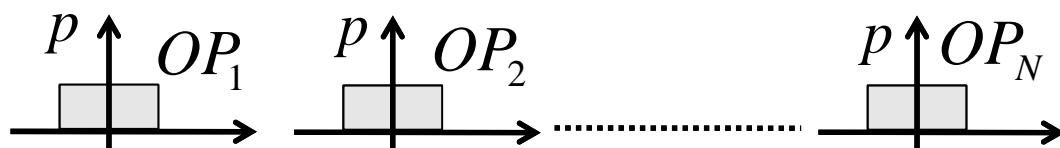
$r$  : exact result

$\hat{R}$  : approximation of  $r$  due to rounding error.

$\varepsilon_{\text{rounding}}$  : rounding error.

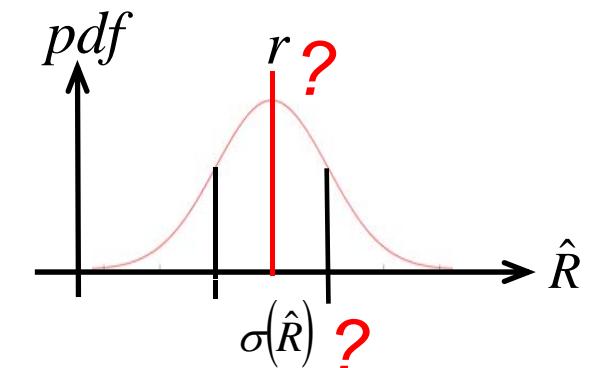
- First order model ([1]) :  $\varepsilon_{\text{rounding}} = \sum_{k=1}^n g_i(d) \cdot 2^{-p} \alpha_i + O(2^{-2p})$

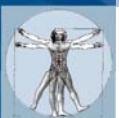
Assume a probability distribution for rounding error of elementary operation



Central Limit  
Theorem

Gaussian Distribution





# Probabilistic Rounding Error Analysis (PREA)

- Consider a sequence of computations providing an exact result  $r$ .
- Perform computations  $N$  times with random rounding (randomly choosing rounding to  $+\infty$  or  $-\infty$ ),  $N$  results  $\hat{R}_i, i = 1, \dots, N$  are obtained.
- The computed result is

$$\bar{R} = \frac{1}{N} \sum_{i=1}^N \hat{R}_i$$

- Instead of  $\sigma(\bar{R})$ , we have sample variance:

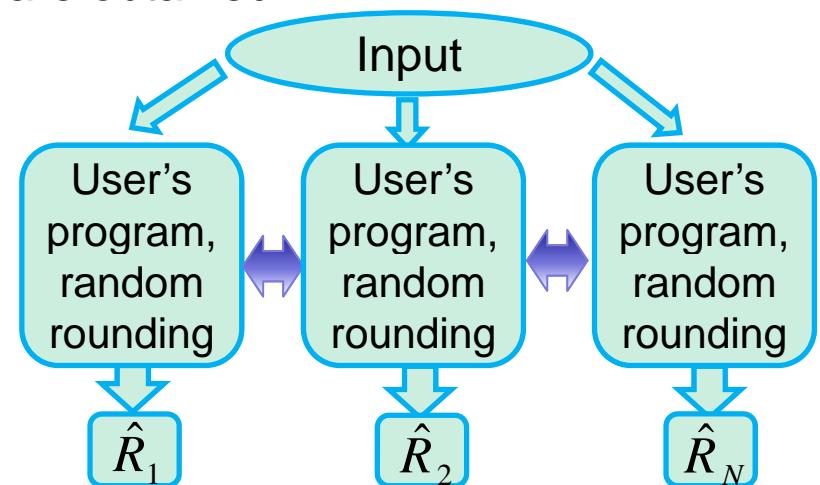
$$s^2 = \frac{1}{N-1} \sum_{i=1}^N (\hat{R}_i - \bar{R})^2$$

- Define

$$T = \frac{\bar{R} - r}{s/\sqrt{N}} \quad \text{sample from Student's t-distribution}$$

- Probability density function

$$f(T = t) = \frac{\Gamma\left(\frac{\nu+1}{2}\right)}{\sqrt{\nu\pi} \cdot \Gamma(\nu/2)} \left(1 + \frac{t^2}{\nu}\right)^{-\left(\frac{\nu+1}{2}\right)}$$





# Number of Significant Digits

- Confidence Interval

$$\Pr\left(-\tau_\beta < \frac{\bar{R} - r}{s/\sqrt{N}} < \tau_\beta\right) = \beta$$

- With probability  $\beta$ , the number of the significant digits (i.e. accurate digits) of  $\bar{R}$  is

$$N_{\text{Significant Digit of } \bar{R}} \geq \log_{10}\left(\frac{\sqrt{N} \cdot |\bar{R}|}{\tau_\beta \cdot s}\right)$$

- Two Hypothesis should hold ([1, 2]) :

- Random rounding error is Gaussian distributed.  
-- not important because of robustness of Student's t-test.
- The first order approximation is legitimate.
  - The operands of multiplication must be significant.
  - The divisor of division must be significant.

} Self validation



# Applied to Complex Statistical Software

*R* (v2.10.1): a software for statistical computing and graphics.

Benchmarks: NIST StRD (a collection of data sets and certified values)

Accuracy Estimation  
using PREA :

$$\log_{10} \left( \frac{\sqrt{N} \cdot |R|}{\tau_\beta \cdot s} \right)$$

Certified values  
from StRD

Reference (true accuracy):

$$LRE = -\log_{10} \left( \frac{\bar{R} - r}{r} \right)$$

|             |           | PREA | True | PREA               | True | PREA            | True | PREA | True |
|-------------|-----------|------|------|--------------------|------|-----------------|------|------|------|
| UNIV        | benchmark | Mean |      | Standard Deviation |      | Autocorrelation |      | -    |      |
| <i>mean</i> | Mavro     | 15   | 15   | 13                 | 13   | 13              | 14   | -    | -    |
| <i>sd</i>   | Numacc3   | 15   | 15   | 10                 | 10   | 11              | 10   | -    | -    |
| <i>acf</i>  |           |      |      |                    |      |                 |      |      |      |



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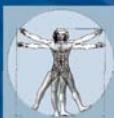
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Number of significant Digits



# Applied to Complex Statistical Software

|       |           | PREA                     | True                     | PREA  | True  | PREA  | True  | PREA        | True  |
|-------|-----------|--------------------------|--------------------------|-------|-------|-------|-------|-------------|-------|
| ANOVA | benchmark | SST                      |                          | SSE   |       | MSE   |       | F-statistic |       |
|       |           | SiRstv                   | 12 12                    | 12 13 | 12 13 | 12 12 | 12 12 | 2           | 2     |
| aov   | SmLs08    | 4 4                      | 2 2                      | 2 2   | 2 2   | 2 2   | 2 2   | 2 2         | 2 2   |
|       | LINR      | Coefficient              |                          | RSD   |       | $R^2$ |       | F-statistic |       |
|       |           | Norris                   | 12 13                    | 14 14 | 15 15 | 15 15 | 13 14 | 13 14       | 13 14 |
| NLINR | Wampler5  | 5 6                      | 15 15                    | 14 14 | 14 14 | 14 15 | 14 15 | 14 15       | 14 15 |
|       | Nls       | Coefficient <sub>i</sub> | Coefficient <sub>j</sub> | RSS   |       | RSD   |       |             |       |
|       |           | Lanczos2                 | 5 5                      | 6 6   | 7 8   | 7 8   | 8 8   | 8 8         | 8 8   |
|       | Bennet5   | 4 4                      | 4 5                      | 9 9   | 9 9   | 9 9   | 9 9   | 9 9         | 9 9   |

|                    | Exact estimation | Underestimation by 1 digit | Overestimation by 1 digit |
|--------------------|------------------|----------------------------|---------------------------|
| Relative Frequency | 67%              | 29%                        | 4%                        |

If underestimation by 1 digit is tolerable:

→ reliable estimation in 96% of the cases

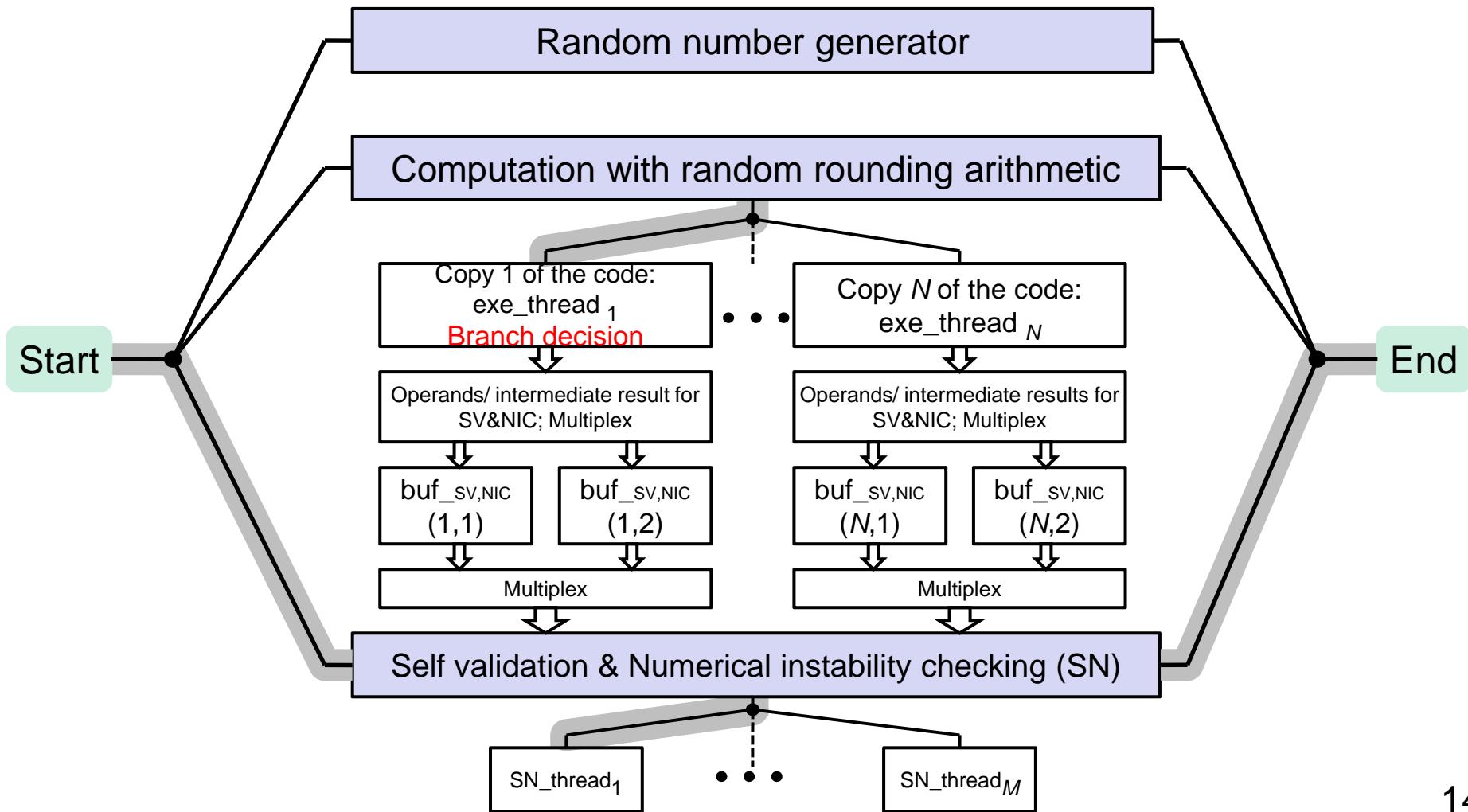


# Parallelization for Acceleration

- Why do we need parallel computing?
  - The multiple runs of the code are intensive in computation time.
  - Multi-cores are the mainstream hardware architecture.
  - To harvest the performance potential of multi-core processors.
- Parallel execution
  - To get  $N$  random rounding result, the user's code can be executed concurrently on different CPU cores.
  - Communication between threads is necessary to:
    - make a unitive decision for instructions such as “ IF() THEN”.
    - perform Self Validation (SV)
    - perform Numerical Instabilities Checking (NIC).  
i.e. sudden accuracy loss due to cancellation in ‘+’ or ‘-’ ).
  - How to minimize communication & synchronization overhead?

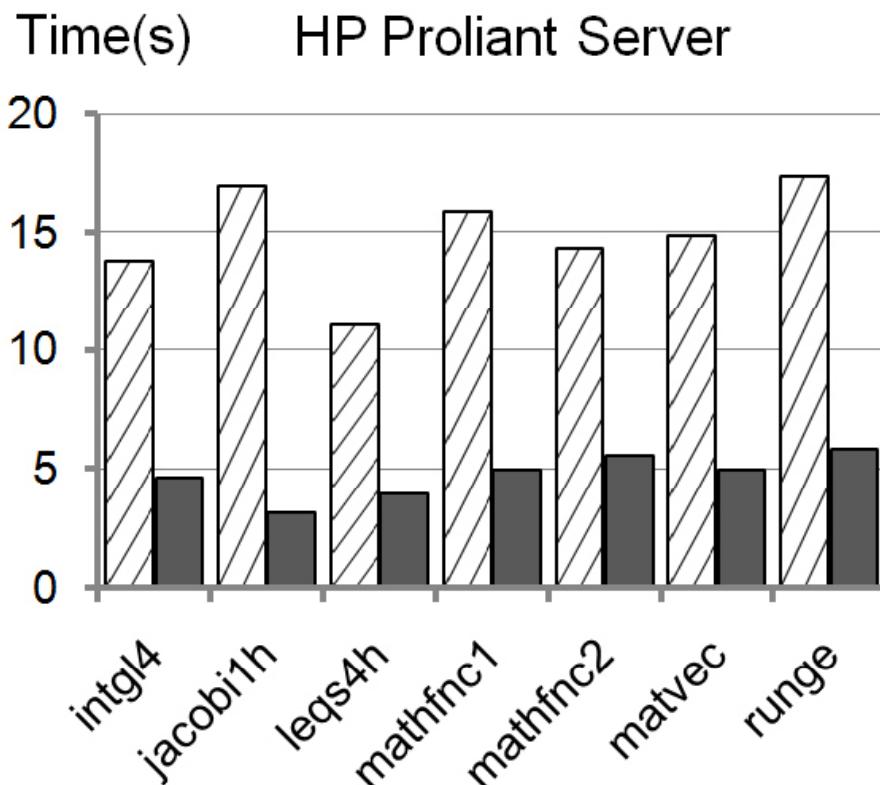


# Parallelization with Asynchronous SV & NIC

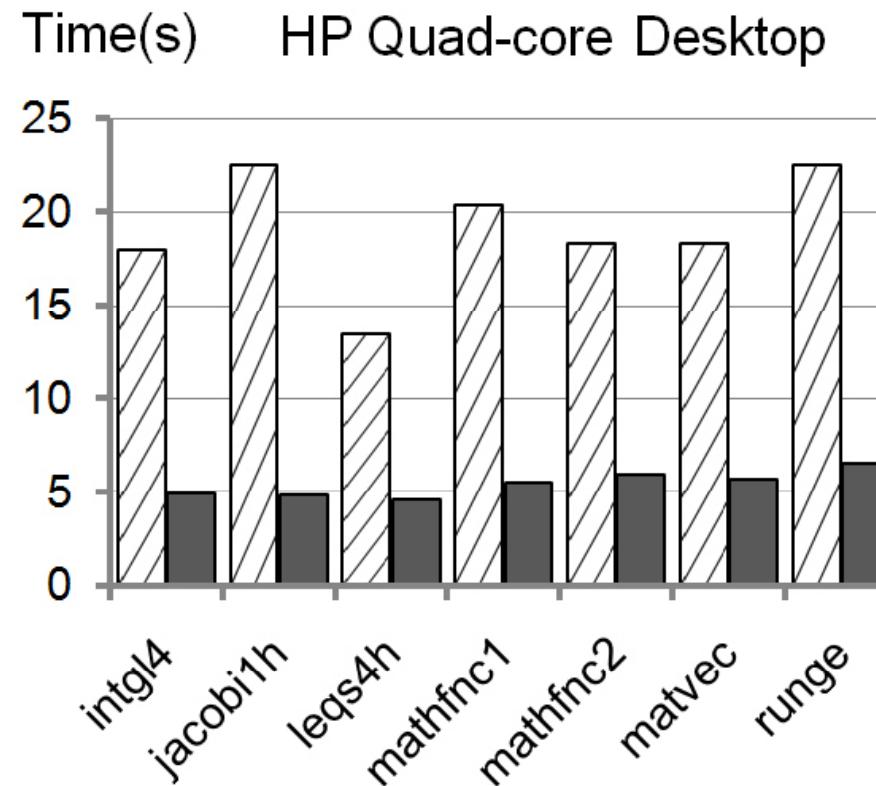




# Performance of Parallelization with Asynchronous SV & NIC



▨ Sequential CADNA

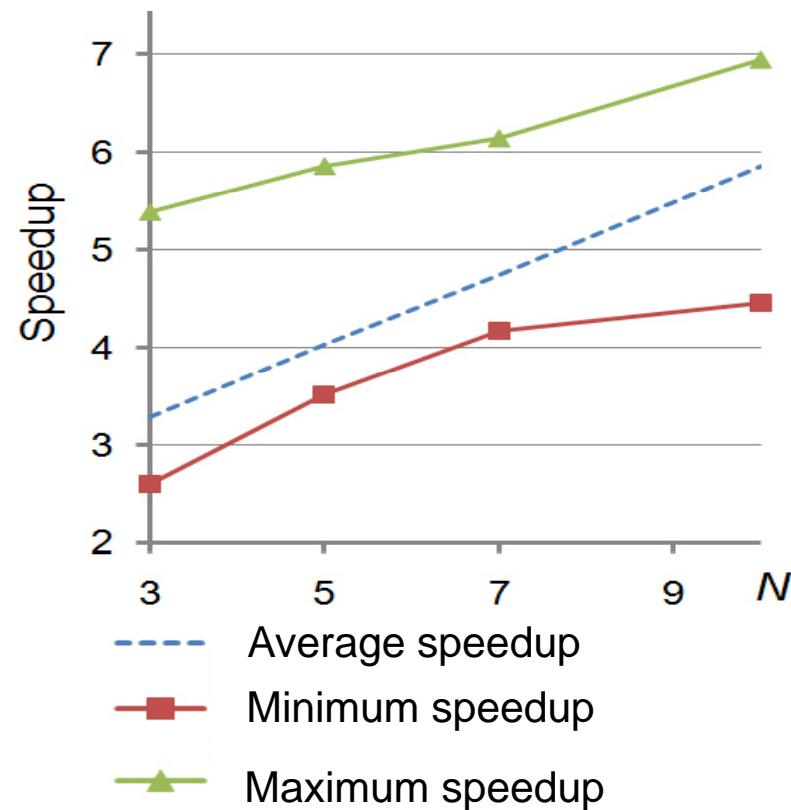
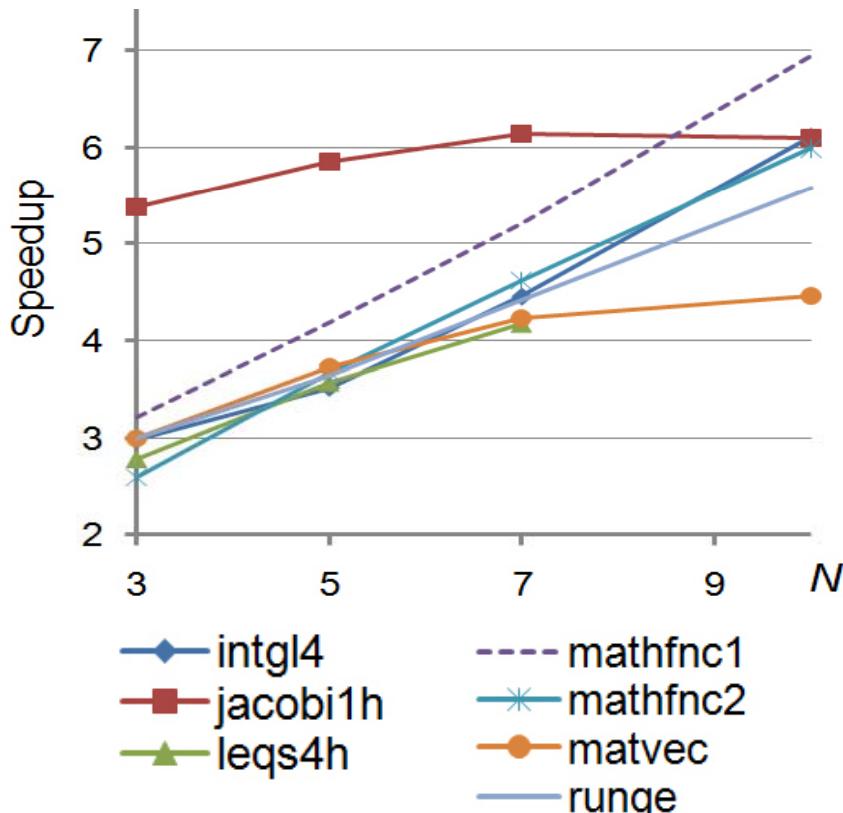


■ Parallelization with asynchronous SV & NIC



# Performance of Parallelization with Asynchronous SV & NIC

This parallelization approach achieves an almost linear scalability.





# Conclusion

1. Probabilistic error analysis method is very robust and provide tight rounding error estimation.
2. Numerical instability can be localized.

Line 10034 → 9.676383250285714 $\cdot 10^3$

Line 10035 → 2.786786213387928 $\cdot 10^3$



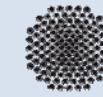
Sudden increase  
in numerical error

3. Parallelization approaches significantly accelerate the proposed method, and shows an almost linear scalability.



# Reference

- www.simtech.uni-stuttgart.de
1. J. –M. Chesneaux, J.-M , *Study of the computing accuracy by using probabilistic approach; Contribution to Computer Arithmetic and Self-Validating Numerical Methods*, ed. C. Ulrich (J.C. Baltzer) 1990, 19-30.
  2. J. Vignes, *Discrete stochastic arithmetic for validating results of numerical software. Numerical Algorithms*, Vol. 37, p. 377-390
  3. Douglas C. Montgomery, *Applied Statistics and Probability for Engineers*. page 257.
  4. N. Tajima, FORTRAN benchmark tests. [http : //www:apphy:fukui-u.ac.jp/tajima/bench/index\\_e.html](http://www:apphy:fukui-u.ac.jp/tajima/bench/index_e.html)



# Thanks!

Questions?

Comment?