Multi-Level Repetition Benchmarking

Marek Cuchý



marek.cuchy@agents.fel.cvut.cz

B(E)4M36ESW

March 9, 2020

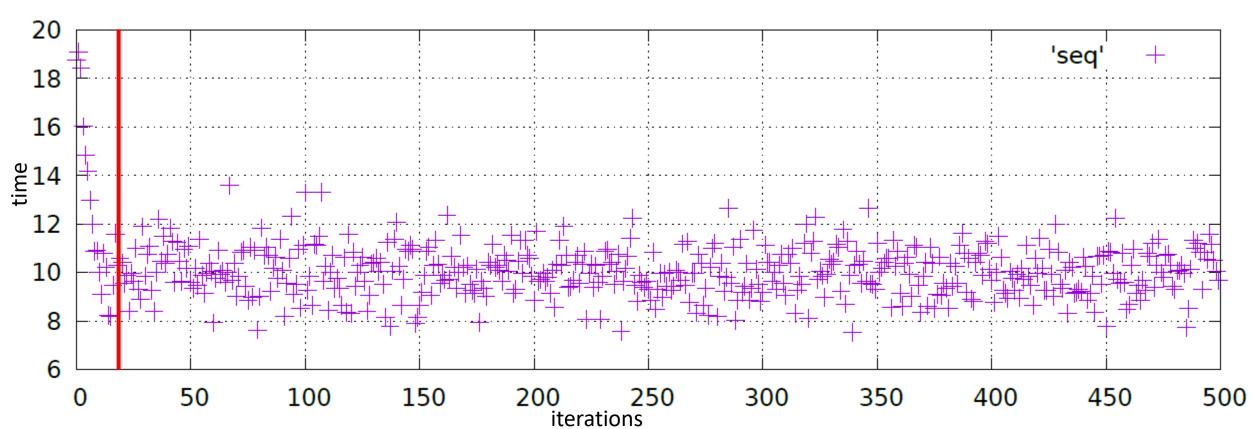


Multi-Level Repetition

- Variance in measurements may occur at higher levels
 - →we need to repeat measurements at least on the level of the variance
- The highest level is the most important one
- Levels:
 - 1. Iteration the smallest possible measurement (e.g. loop body)
 - 2. Execution running of the program
 - 3. Compilation (stable in Java)

Warm-up

- Measurements useful only after reaching a steady/independent state
- E.g. by manual inspection of sequence plot of several executions



Number of Repetitions

- Simplified to two levels (n = 2) Java
 - 1 iteration
 - 2 execution
- What is the optimal count on the lower levels to increase precision?
 - Lower levels less time needed to do more repetitions
- Run dimensioning experiment
 - e.g. 30 execution and 40 iterations excl. warm-up
- Calculate unbiased variance estimators for each level T_1^2 , T_2^2

 $T_1^2 = S_1^2$ Biased variance estimators $T_2^2 = S_2^2 - \frac{S_1^2}{r_1}$ Number of iterations used for the dimensioning experiment

Biased Variance Estimators

$$\mathbf{S_i^2} = \frac{1}{\prod_{k=i+1}^{n} \mathbf{r_k}} \frac{1}{\mathbf{r_i} - 1} \sum_{j_n=1}^{r_n} \cdots \sum_{j_i=1}^{r_i} \left(\bar{\mathbf{Y}}_{j_n \dots j_i \bullet \dots \bullet} - \bar{\mathbf{Y}}_{j_n \dots j_{i+1} \bullet \dots \bullet} \right)^2$$

- For two levels (n = 2):
 - S_1^2 mean of execution variances
 - $S_2^2 = S_n^2$ variance of execution means
- See [2] Chapter 6.1 for details and example

Number of Repetitions

$$\forall i, 1 \leq i < n, \quad r_i = \left[\sqrt{\frac{c_{i+1}}{c_i} \frac{T_i^2}{T_{i+1}^2}} \right]$$

$$T_1^2 = S_1^2$$
$$T_2^2 = S_2^2 - \frac{S_1^2}{r_1}$$

- c_1 single measurement (iteration) duration after warm-up
- c_2 execution cost (time to reach independent state warm-up)
- r_i optimal repetition count on i-th level
- What if $T_i^2 \leq 0$?
 - i-th level induces very little variance \rightarrow can be skipped
- What about r_n ???
 - More highest level repetitions always increase precision

Number of Repetitions – Example

Matrix Multiplication

Notebook

i5-7200U (2.5GHz), 8GB RAM

measureMultiply S1 = 857.2065800776355 S2 = 138.16871702761134 T1 = 857.2065800776355 T2 = 116.73855252567046 r1 = 8

measureMultiply1D		
S1	=	2187.8572504410527
s2	=	672.0203176226894
т1	=	2187.8572504410527
т2	=	617.3238863616631
r1	=	4

measureMultiplyTrans

Desktop

i5-8500 (3GHz), 32GB RAM

- measureMultiply S1 = 456.22120240936977 S2 = 73.46727479932864 T1 = 456.22120240936977 T2 = 62.061744739094394 r1 = 8 measureMultiply1D S1 = 254.87119015897457 S2 = 5.73656707647479 T1 = 254.87119015897457
- T2 = -0.6352126774995739

$$r1 = -1$$

```
measureMultiplyTrans
```

For each environment (HW, OS, ...) and benchmark (implementation) different setup is required

Execution Time + Effect Size Confidence Interval

$$\overline{Y} \quad \pm \quad t_{1-\frac{\alpha}{2},\nu} \sqrt{\frac{1}{r_n(r_n-1)} \sum_{j_n=1}^{r_n} \left(\overline{Y}_{j_n} \underbrace{\bullet \cdots \bullet}_{n-1} - \overline{Y}\right)^2}}$$
$$\overline{Y} \quad \pm \quad t_{1-\frac{\alpha}{2},\nu} \sqrt{\frac{S_n^2}{r_n}}$$

- *n* number of levels
- r_n number of repetition on the highest level
- \overline{Y} mean across all measurements (excl. warm-up)
- (1α) confidence interval (e.g. 95% confidence $\rightarrow \alpha = 0.05$)
- $t_{1-\frac{\alpha}{2},\nu} (1-\frac{\alpha}{2})$ -quantile of the *t*-distribution with $\nu = r_n 1$ degrees of freedom, can be found in a table

Speed-Up Ratios with Confidence Interval

$$\frac{\overline{Y} \cdot \overline{Y'} \mp \sqrt{\left(\overline{Y} \cdot \overline{Y'}\right)^2 - \left(\overline{Y}^2 - h^2\right) \left(\overline{Y'}^2 - h'^2\right)}}{\overline{Y}^2 - h^2}$$
$$h = \sqrt{t_{\frac{\alpha}{2},\nu}^2 \frac{S_n^2}{r_n}} \qquad h' = \sqrt{t_{\frac{\alpha}{2},\nu}^2 \frac{S_n'^2}{r_n}}$$

- \overline{Y} , $\overline{Y'}$ means of the compared implementations
- h, h' half-widths of the confidence intervals for the single implementations (you can reuse the values from prev. slide)
- S_n^2 , S'_n^2 biased variance estimator of the *n*-th level

References

[1] Kalibera, T. and Jones, R. E. (2013) Rigorous Benchmarking in Reasonable Time. In: ACM SIGPLAN International Symposium on Memory Management (ISMM 2013), 20–12 June, 2013, Seattle, Washington, USA. <u>http://kar.kent.ac.uk/33611/</u>

[2] Kalibera, T. and Jones, R. E. (2012) Quantifying performance changes with effect size confidence intervals. Technical Report 4–12, University of Kent