Write-up

Hypothesis

Merge sort will outperform insertion sort for all but very small lists, with the difference in speed becoming particularly apparent for very large lists. As for just how small the list will have to get for insertion sort to outpace merge sort, I'd guess low double digits.

Methods

The code I used to test these algorithms can be found at https://github.com/khuldraeseth/cse431hw4. Clone the repository, move into p1, and stack run -- --output wherever-you-want-the-results.html to run these tests yourself. Compiles with -threaded.

These are my own implementations of insertion sort and merge sort, written in Haskell. Merge sort does not split the list at the middle—rather, it uses the inverse of an interleave (which I have named unterleave) that takes $[a,b,c,d,e] \mapsto ([a,c,e],[b,d])$ as an example for what I expect may be a moderate speedup.

Results

See below for Criterion output. The results are summarized in this plot:



Discussion

Not quite as I had expected. For starters, merge sort is faster in these tests than insertion sort for very small lists as well as sufficiently large ones! This makes sense in retrospect—I have special cases in my merge sort code for empty and singleton cases, so these are just one pattern match from being sorted, rather than a call to foldr. Then insertion sort has its time on top, outperforming merge sort for lists of length 3, 4, and 5. The two cross somewhere between 5 and 10, and repeated tests show no consistent particular value.

Interesting is that insertion sort, after being outpaced by an order of magnitude on lists of length 35, makes a resurgence and competes with merge sort again for lengths 50, 75, and 100 before falling off again at 200 and never having a third chance. I'll attribute this to some compile-time optimization.

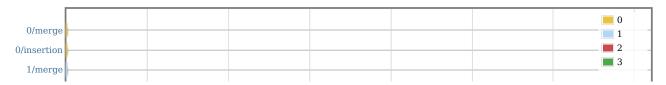
Conclusion

Under the conditions tested, merge sort was faster than insertion sort on lists of all lengths except those between 3 and 10.

criterion performance measurements

overview

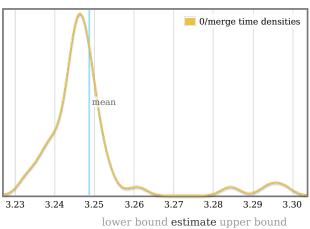
want to understand this report?

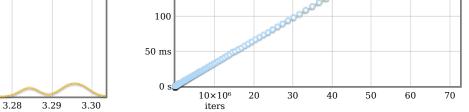




regression
0/merge times

0/merge





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250

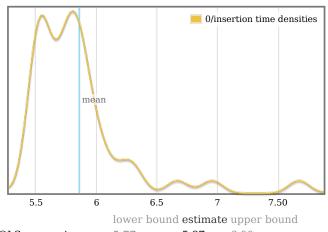
200

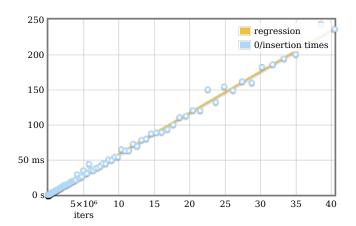
150

	lower bound	estimate	upper boun
OLS regression	3.24 ns	3.25 ns	3.26 ns
R ² goodness-of-fit	1.000	1.000	1.000
Mean execution time	3.25 ns	3.25 ns	3.25 ns
Standard deviation	8.39 ps	13.0 ps	19.2 ps

Outlying measurements have no (0.3%) effect on estimated standard deviation.

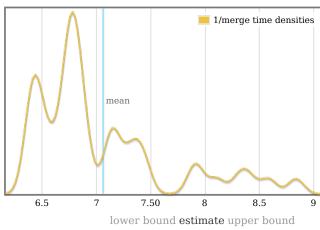
0/insertion

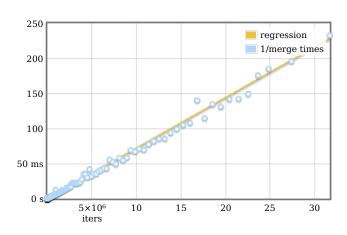




Outlying measurements have severe (85.7%) effect on estimated standard deviation.

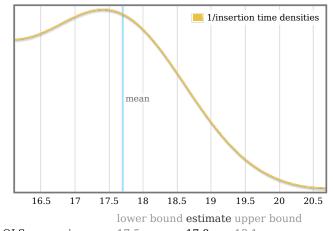
1/merge

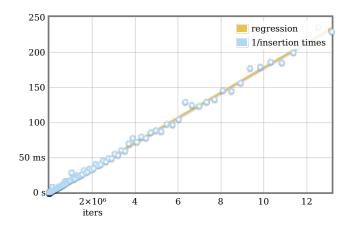




Outlying measurements have severe (90.5%) effect on estimated standard deviation.

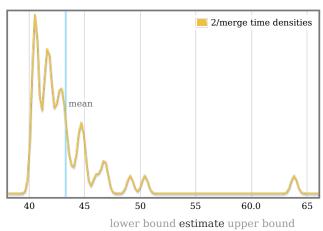
1/insertion

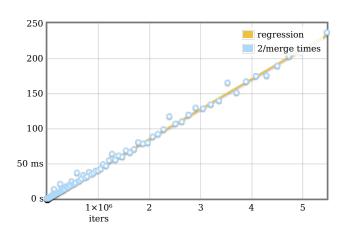




Outlying measurements have severe (75.1%) effect on estimated standard deviation.

2/merge

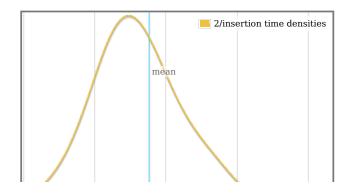


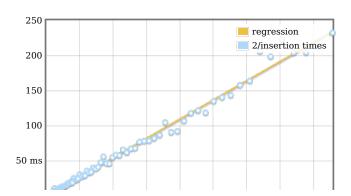


OLS regression 42.1 ns 42.6 ns 43.3 ns R^2 goodness-of-fit 0.996 0.998 0.999 Mean execution time 42.5 ns 43.3 ns 45.6 ns Standard deviation 2.16 ns 3.93 ns 6.72 ns

Outlying measurements have severe (89.7%) effect on estimated standard deviation.

2/insertion

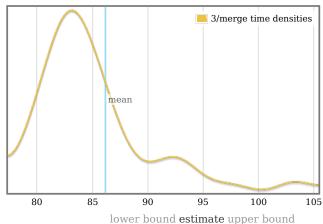


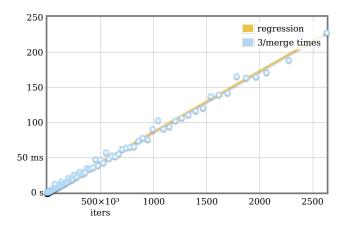


lower bound estimate upper bound

Outlying measurements have severe (85.2%) effect on estimated standard deviation.

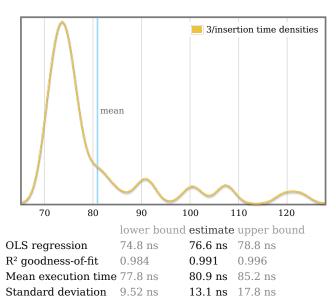
3/merge

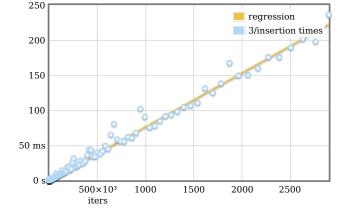




Outlying measurements have severe (82.7%) effect on estimated standard deviation.

3/insertion

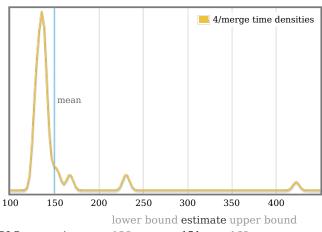


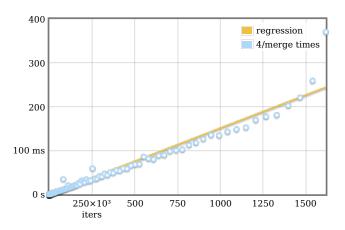


criterion report

Outlying measurements have severe (96.4%) effect on estimated standard deviation.

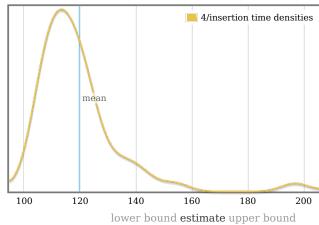
4/merge

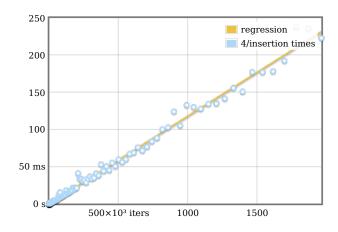




Outlying measurements have severe (99.1%) effect on estimated standard deviation.

4/insertion

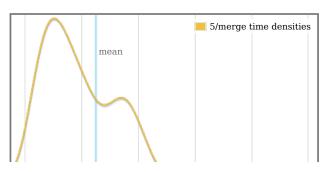


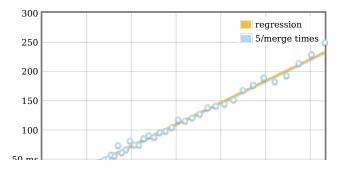


OLS regression 114 ns 118 ns 122 ns R^2 goodness-of-fit 0.991 0.993 0.997 Mean execution time 117 ns 120 ns 127 ns Standard deviation 9.62 ns 15.7 ns 28.0 ns

Outlying measurements have severe (94.6%) effect on estimated standard deviation.

5/merge



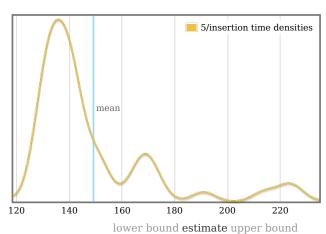


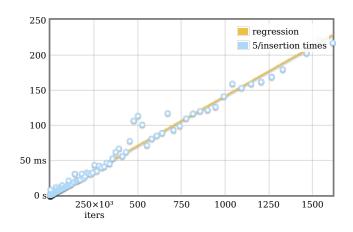
LOMATOR	haiind	estimate	unnar	hound

OLS regression	181 ns	184 ns	188 ns
R ² goodness-of-fit	0.997	0.998	0.999
Mean execution time	180 ns	182 ns	187 ns
Standard deviation	7.58 ns	10.2 ns	13.9 ns

Outlying measurements have severe (74.1%) effect on estimated standard deviation.

5/insertion



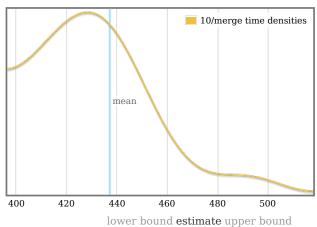


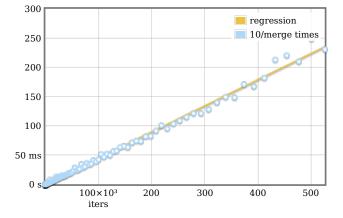
OLS regression 138 ns **141 ns** 146 ns R² goodness-of-fit 0.975 0.987 0.996 149 ns 157 ns Mean execution time 143 ns

Standard deviation 15.6 ns 24.0 ns 32.5 ns

Outlying measurements have severe (96.2%) effect on estimated standard deviation.

10/merge



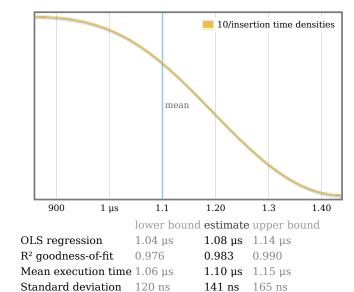


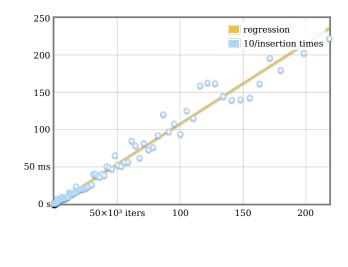
OLS regression 433 ns **447** ns 459 ns R² goodness-of-fit 0.994 0.996 0.999 Mean execution time 430 ns 437 ns 446 ns Standard deviation 19.9 ns 26.1 ns 32.7 ns

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Outlying measurements have severe (75.2%) effect on estimated standard deviation.

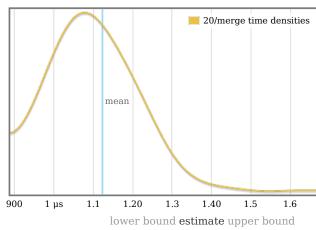
10/insertion

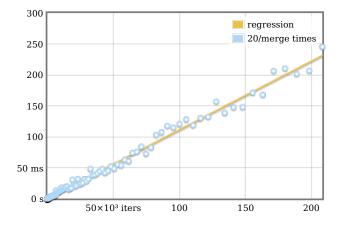




Outlying measurements have severe (92.9%) effect on estimated standard deviation.

20/merge

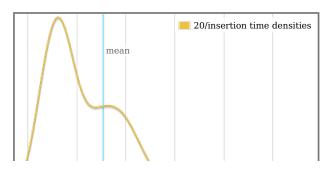


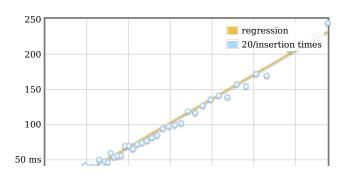


OLS regression 1.07 μ s 1.11 μ s 1.15 μ s R² goodness-of-fit 0.990 0.992 0.995 Mean execution time 1.08 μ s 1.12 μ s 1.16 μ s Standard deviation 98.7 ns 133 ns 187 ns

Outlying measurements have severe (91.9%) effect on estimated standard deviation.

20/insertion

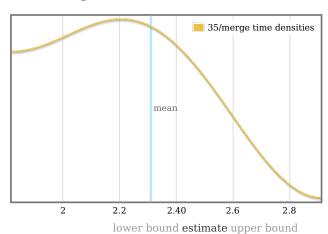


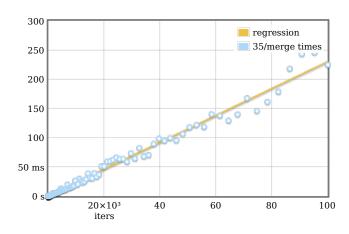


	lower bound	estimate	upper boun
OLS regression	3.32 µs	$3.42~\mu s$	3.52 µs
R ² goodness-of-fit	0.990	0.994	0.997
Mean execution time	3.30 µs	3.38 µs	3.48 µs
Standard deviation	246 ns	315 ns	409 ns

Outlying measurements have severe (85.8%) effect on estimated standard deviation.

35/merge

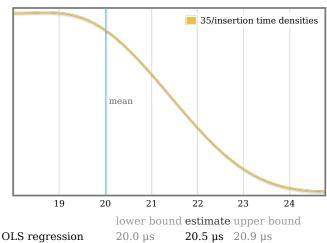




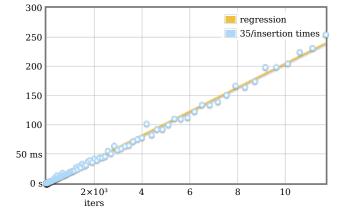
OLS regression $2.20 \, \mu s$ 2.30 μs 2.41 μs R^2 goodness-of-fit 0.987 0.993 0.983 Mean execution time $2.22 \mu s$ 2.31 μs 2.38 μs Standard deviation 225 ns 261 ns 298 ns

Outlying measurements have severe (90.7%) effect on estimated standard deviation.

35/insertion



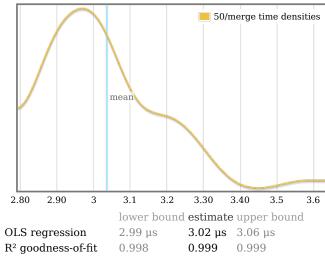
R² goodness-of-fit 0.995 0.997 0.998 Mean execution time $19.6~\mu s$ $20.0~\mu s$ $~20.4~\mu s$ 1.25 μs 1.88 μs Standard deviation $1.04 \mu s$

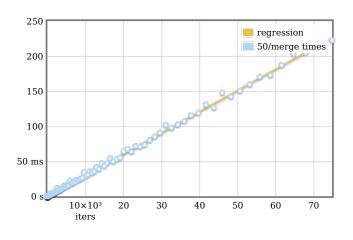


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Outlying measurements have severe (68.7%) effect on estimated standard deviation.

50/merge

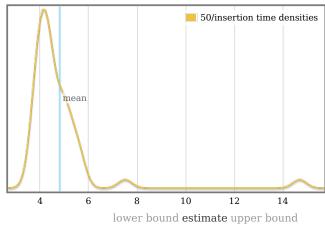


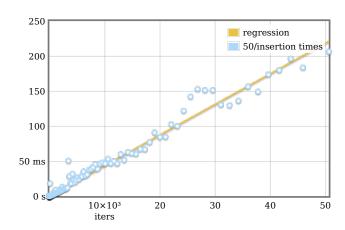


OLS regression 2.99 μ s 3.02 μ s 3.06 μ s R² goodness-of-fit 0.998 0.999 0.999 Mean execution time 3.00 μ s 3.04 μ s 3.08 μ s Standard deviation 120 ns 151 ns 204 ns

Outlying measurements have severe (63.5%) effect on estimated standard deviation.

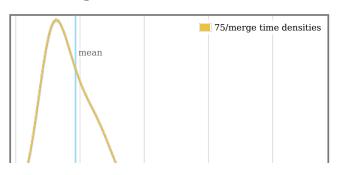
50/insertion

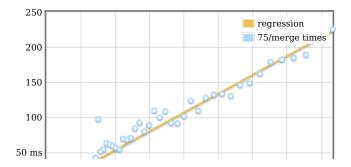




Outlying measurements have severe (98.8%) effect on estimated standard deviation.

75/merge

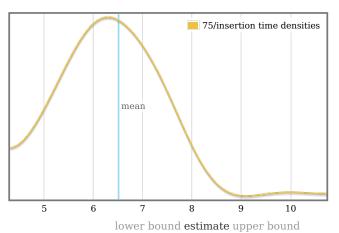


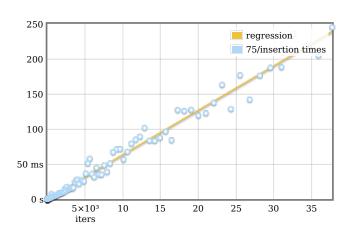


	lower bound	estimate	upper bound
OLS regression	5.25 μs	5.35 µs	5.52 μs
R ² goodness-of-fit	0.961	0.983	0.994
Mean execution time	5.59 μs	5.85 µs	6.50 μs
Standard deviation	661 ns	1.33 µs	2.54 μs

Outlying measurements have severe (97.6%) effect on estimated standard deviation.

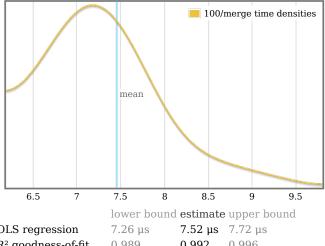
75/insertion

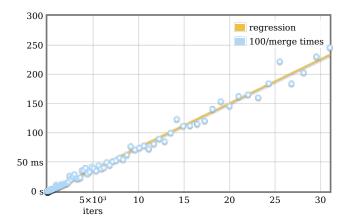




Outlying measurements have severe (95.1%) effect on estimated standard deviation.

100/merge

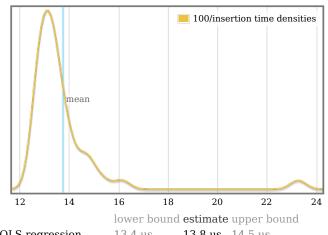


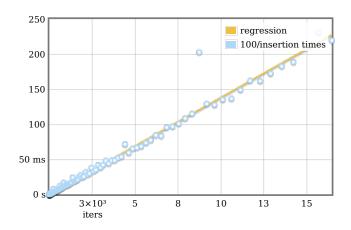


OLS regression 7.26 μ s 7.52 μ s 7.72 μ s R² goodness-of-fit 0.989 0.992 0.996 Mean execution time 7.28 μ s 7.45 μ s 7.71 μ s Standard deviation 627 ns 764 ns 936 ns

Outlying measurements have severe (87.0%) effect on estimated standard deviation.

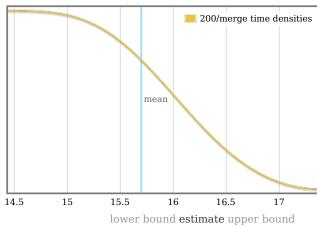
100/insertion

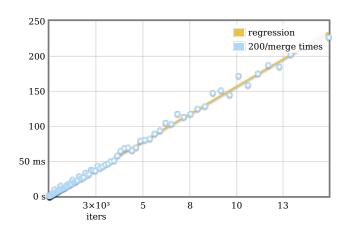




Outlying measurements have severe (90.5%) effect on estimated standard deviation.

200/merge

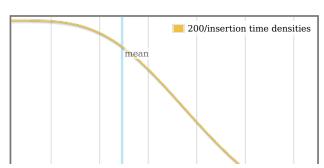


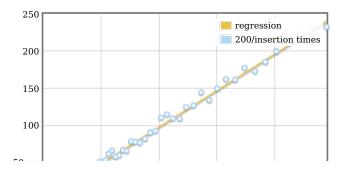


OLS regression $15.5~\mu s$ $15.7~\mu s$ $15.9~\mu s$ R^2 goodness-of-fit 0.997 0.998 0.999 Mean execution time $15.5~\mu s$ $15.7~\mu s$ $16.0~\mu s$ Standard deviation 672~n s 777~n s 897~n s

Outlying measurements have severe (58.5%) effect on estimated standard deviation.

200/insertion



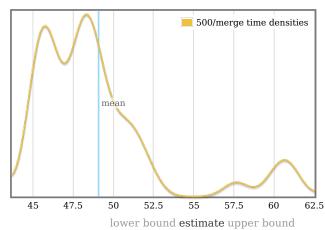


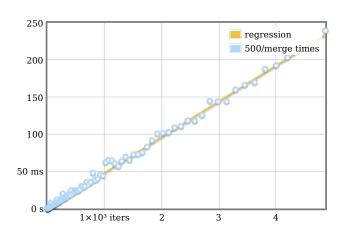
lower bound estimate upper b	ound
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OLS regression	481 μs	$489 \mu s$	498 μs
R ² goodness-of-fit	0.995	0.996	0.998
Mean execution time	482 μs	$489~\mu s$	500 μs
Standard deviation	23.3 µs	28.3 μs	34.3 µs

Outlying measurements have severe (51.4%) effect on estimated standard deviation.

500/merge



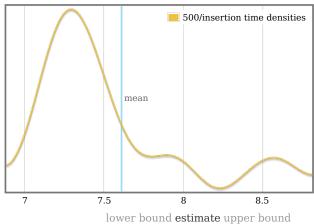


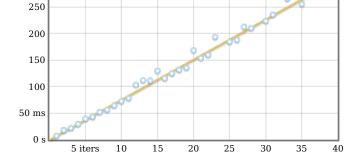
OLS regression $47.5~\mu s$ $48.1~\mu s$ $48.7~\mu s$ R^2 goodness-of-fit 0.995 0.997 0.998 Mean execution time $48.1~\mu s$ $49.1~\mu s$ $50.7~\mu s$

Standard deviation $2.72 \mu s$ $4.06 \mu s$ $5.60 \mu s$

Outlying measurements have severe (77.7%) effect on estimated standard deviation.

500/insertion





regression
500/insertion times

OLS regression 7.36 ms 7.55 ms 7.78 ms R^2 goodness-of-fit 0.984 0.991 0.996 Mean execution time 7.47 ms 7.61 ms 7.87 ms Standard deviation 382 μ s 496 μ s 617 μ s

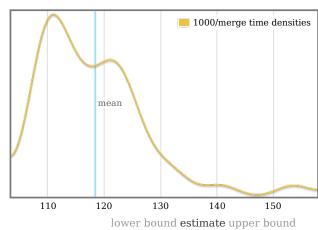
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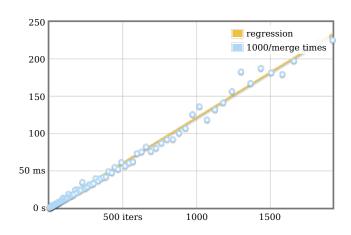
350

300

Outlying measurements have moderate (36.7%) effect on estimated standard deviation.

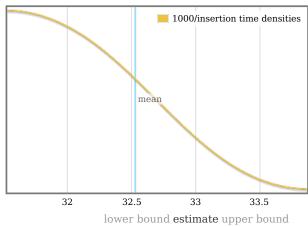
1000/merge

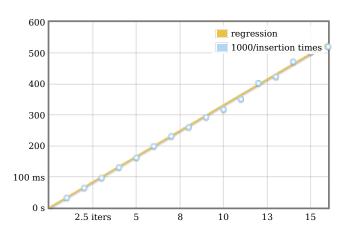




Outlying measurements have severe (73.1%) effect on estimated standard deviation.

1000/insertion

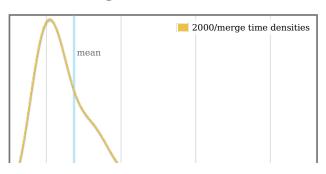


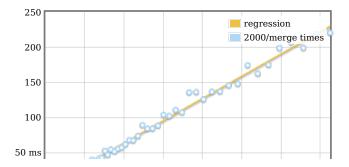


OLS regression 32.5 ms 33.2 ms 33.8 ms R^2 goodness-of-fit 0.997 0.998 0.999 Mean execution time 32.3 ms 32.5 ms 32.9 ms Standard deviation 498 μ s 661 μ s 866 μ s

Outlying measurements have slight (5.9%) effect on estimated standard deviation.

2000/merge

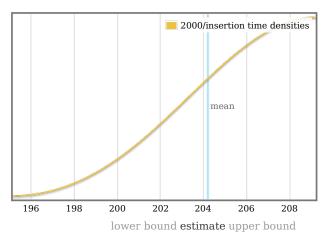


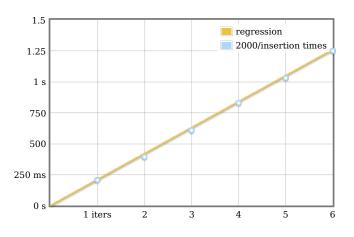


OLS regression	309 μs	316 µs	325 µs
R ² goodness-of-fit	0.992	0.994	0.997
Mean execution time	311 µs	318 µs	331 µs
Standard deviation	18.6 µs	$30.2 \mu s$	51.2 µs

Outlying measurements have severe (76.3%) effect on estimated standard deviation.

2000/insertion



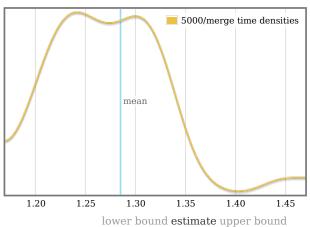


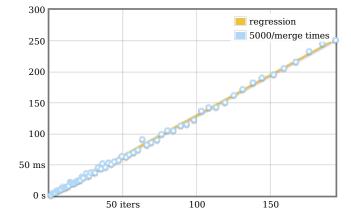
OLS regression 206 ms 210 ms 214 ms R^2 goodness-of-fit 0.998 1.000 1.000 Mean execution time 200 ms 204 ms 206 ms

Standard deviation 1.73 ms 4.30 ms 5.86 ms

Outlying measurements have moderate (13.9%) effect on estimated standard deviation.

5000/merge



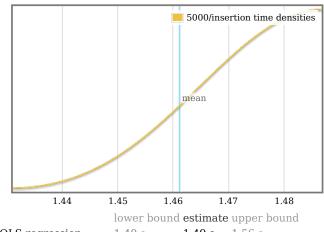


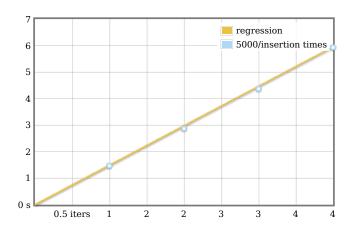
OLS regression 1.30 ms 1.31 ms 1.29 ms R² goodness-of-fit 0.997 0.998 0.999 Mean execution time 1.27 ms 1.28 ms 1.31 ms **64.4 μs** 81.9 μs Standard deviation $50.5 \mu s$

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Outlying measurements have moderate (37.9%) effect on estimated standard deviation.

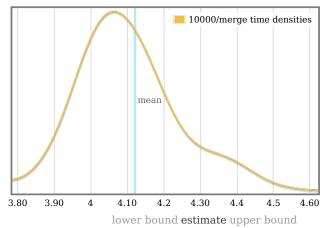
5000/insertion

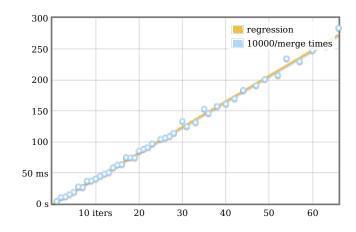




Outlying measurements have moderate (18.8%) effect on estimated standard deviation.

10000/merge

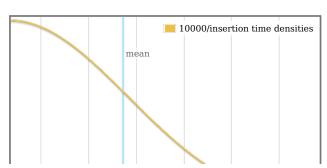


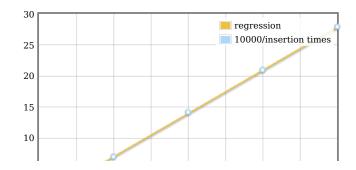


OLS regression 4.05 ms 4.14 ms 4.23 ms R^2 goodness-of-fit 0.995 0.997 0.998 Mean execution time 4.08 ms 4.12 ms 4.18 ms Standard deviation 118 μs 152 μs 213 μs

Outlying measurements have moderate (19.4%) effect on estimated standard deviation.

10000/insertion





lower bound estimate upper bound

Outlying measurements have moderate (18.8%) effect on estimated standard deviation.

understanding this report

In this report, each function benchmarked by criterion is assigned a section of its own. The charts in each section are active; if you hover your mouse over data points and annotations, you will see more details.

- The chart on the left is a kernel density estimate (also known as a KDE) of time measurements. This graphs the probability of any given time measurement occurring. A spike indicates that a measurement of a particular time occurred; its height indicates how often that measurement was repeated.
- The chart on the right is the raw data from which the kernel density estimate is built. The *x* axis indicates the number of loop iterations, while the *y* axis shows measured execution time for the given number of loop iterations. The line behind the values is the linear regression prediction of execution time for a given number of iterations. Ideally, all measurements will be on (or very near) this line.

Under the charts is a small table. The first two rows are the results of a linear regression run on the measurements displayed in the right-hand chart.

- *OLS regression* indicates the time estimated for a single loop iteration using an ordinary least-squares regression model. This number is more accurate than the *mean* estimate below it, as it more effectively eliminates measurement overhead and other constant factors.
- R² goodness-of-fit is a measure of how accurately the linear regression model fits the observed measurements. If the measurements are not too noisy, R² should lie between 0.99 and 1, indicating an excellent fit. If the number is below 0.99, something is confounding the accuracy of the linear model.
- Mean execution time and standard deviation are statistics calculated from execution time divided by number of iterations.

We use a statistical technique called the bootstrap to provide confidence intervals on our estimates. The bootstrap-derived upper and lower bounds on estimates let you see how accurate we believe those estimates to be. (Hover the mouse over the table headers to see the confidence levels.)

A noisy benchmarking environment can cause some or many measurements to fall far from the mean. These outlying measurements can have a significant inflationary effect on the estimate of the standard deviation. We calculate and display an estimate of the extent to which the standard deviation has been inflated by outliers.

colophon

This report was created using the <u>criterion</u> benchmark execution and performance analysis tool.

Criterion is developed and maintained by Bryan O'Sullivan.