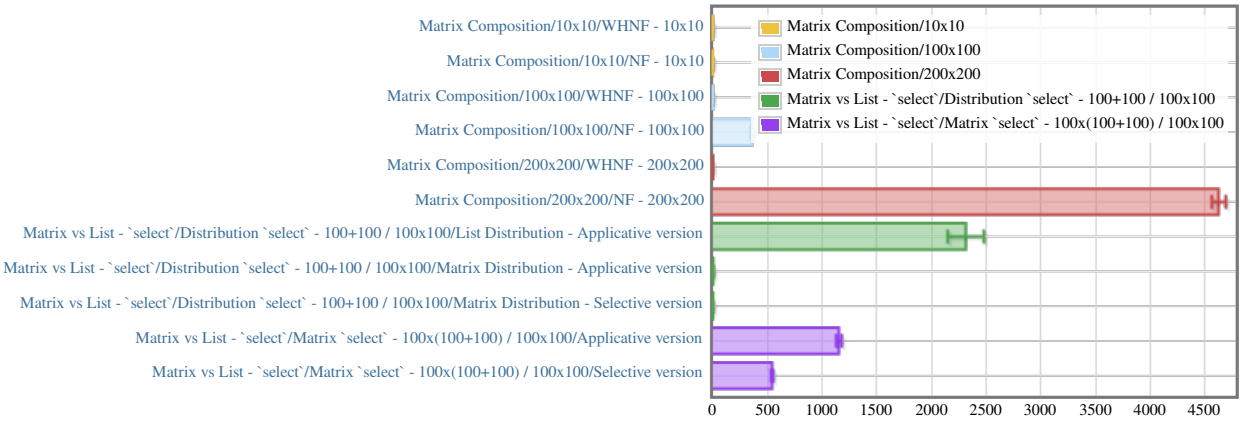
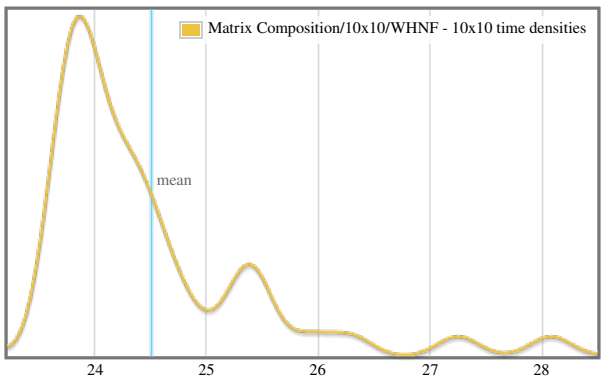
criterion performance measurements

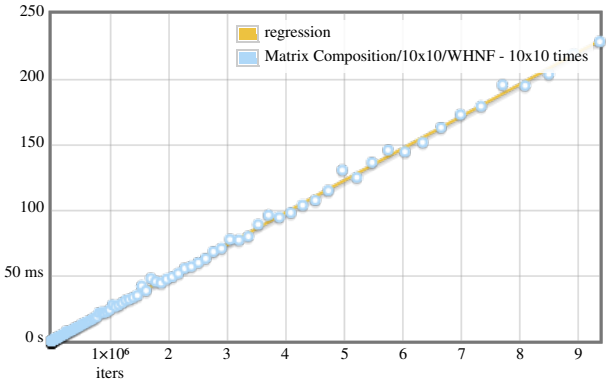
overview

want to understand this report?



Matrix Composition/10x10/WHNF - 10x10

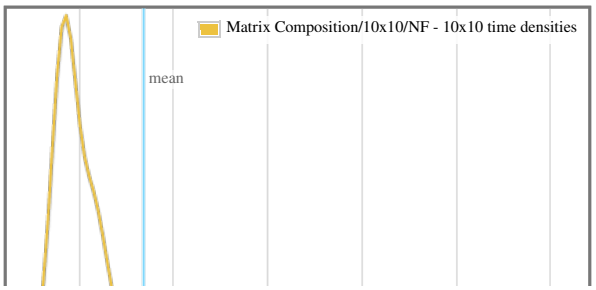


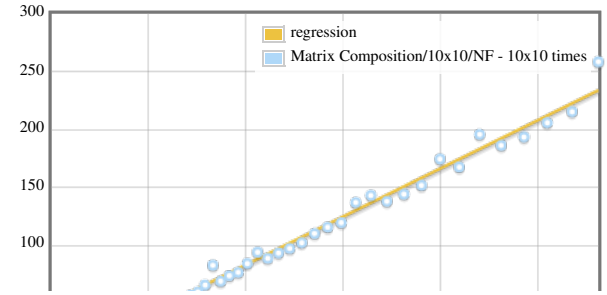


	lower bound	estimate	upper bound
OLS regression	24.3 ns	24.5 ns	24.7 ns
R² goodness-of-fit	0.999	0.999	1.000
Mean execution time	24.3 ns	24.5 ns	24.9 ns
Standard deviation	700 ps	965 ps	1.42 ns

Outlying measurements have severe (62.3%) effect on estimated standard deviation.

Matrix Composition/10x10/NF - 10x10

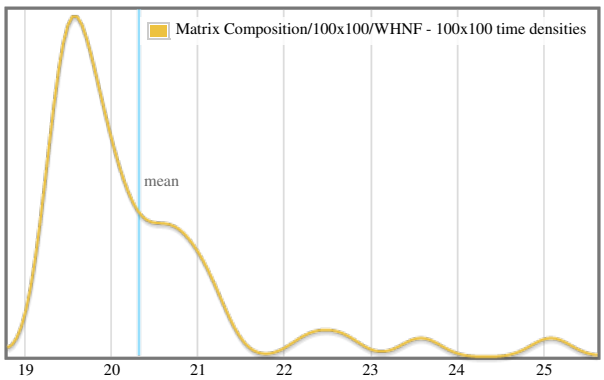


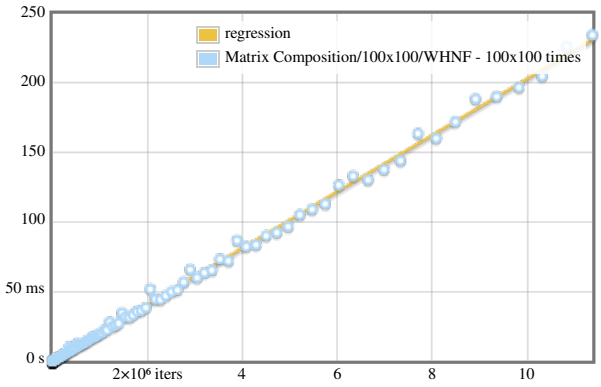


	lower bound	estimate	upper bound
OLS regression	202 µs	208 µs	215 µs
R² goodness-of-fit	0.993	0.995	0.998
Mean execution time	204 µs	207 µs	212 µs
Standard deviation	9.39 µs	12.7 µs	16.9 µs

Outlying measurements have severe (59.3%) effect on estimated standard deviation.

Matrix Composition/100x100/WHNF - 100x100

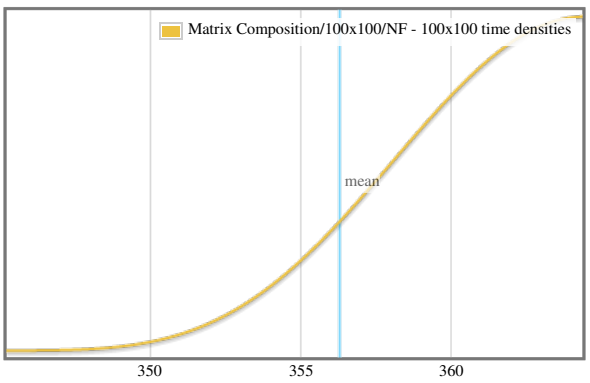


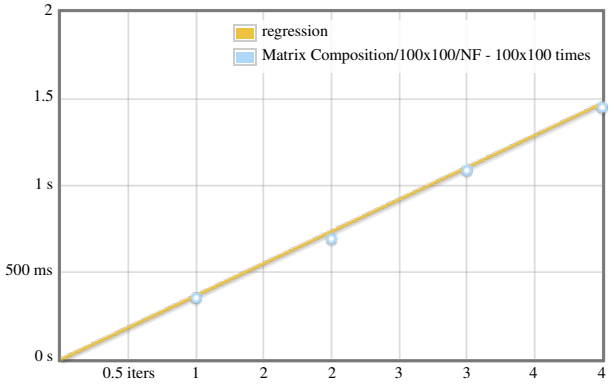


	lower bound	estimate	upper bound
OLS regression	20.1 ns	20.3 ns	20.5 ns
R² goodness-of-fit	0.998	0.999	0.999
Mean execution time	20.1 ns	20.3 ns	20.8 ns
Standard deviation	813 ps	1.17 ns	1.94 ns

Outlying measurements have severe (78.4%) effect on estimated standard deviation.

Matrix Composition/100x100/NF - 100x100

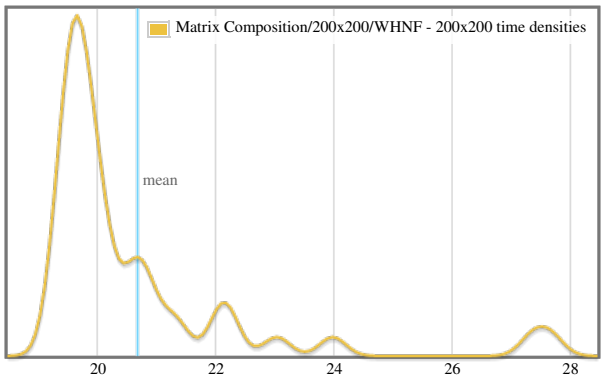


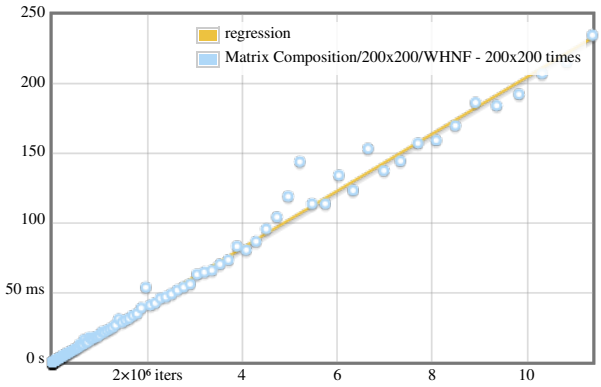


	lower bound	estimate	upper bound
OLS regression	341 ms	369 ms	394 ms
R² goodness-of-fit	0.999	0.999	1.000
Mean execution time	350 ms	356 ms	363 ms
Standard deviation	3.52 ms	7.81 ms	9.19 ms

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

Matrix Composition/200x200/WHNF - 200x200

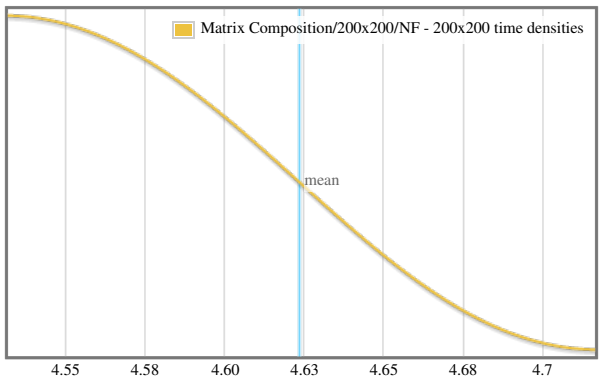


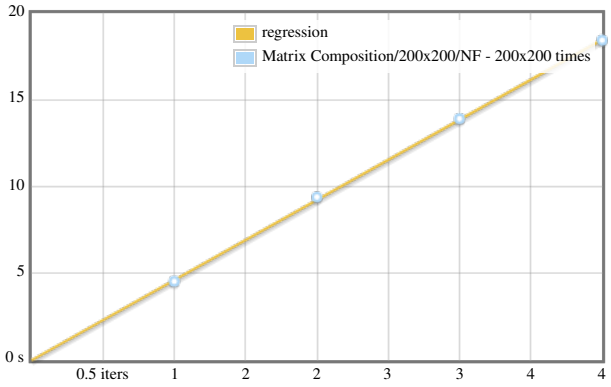


	lower bound	estimate	upper bound
OLS regression	20.1 ns	20.5 ns	21.0 ns
R² goodness-of-fit	0.988	0.994	0.999
Mean execution time	20.2 ns	20.7 ns	21.4 ns
Standard deviation	1.12 ns	1.86 ns	2.89 ns

Outlying measurements have severe (89.8%) effect on estimated standard deviation.

Matrix Composition/200x200/NF - 200x200

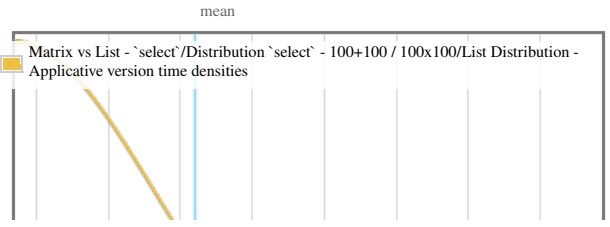


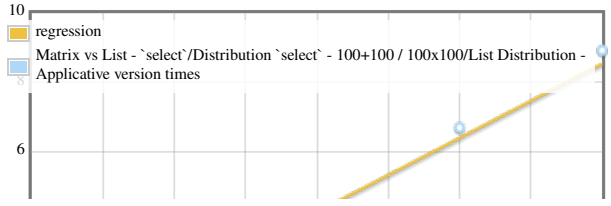


	lower bound	estimate	upper bound
OLS regression	4.51 s	4.62 s	4.86 s
R² goodness-of-fit	0.999	1.000	1.000
Mean execution time	4.56 s	4.62 s	4.67 s
Standard deviation	34.6 ms	64.1 ms	88.9 ms

Outlying measurements have moderate (18.7%) effect on estimated standard deviation.

Matrix vs List - `select`/Distribution `select` - 100+100 / 100x100/List Distribution - Applicative version

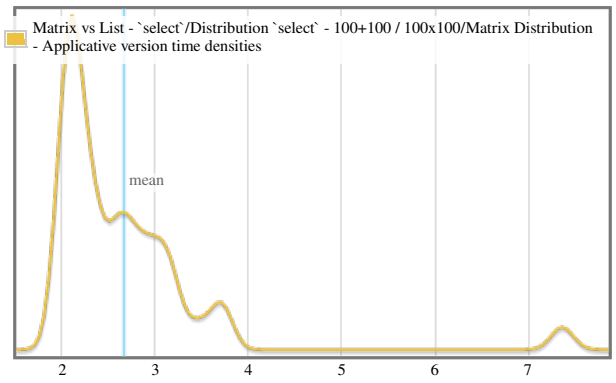


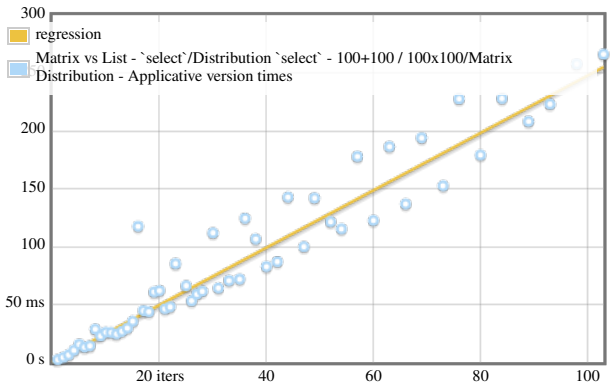


	lower bound	estimate	upper bound
OLS regression	1.88 s	2.14 s	2.26 s
R² goodness-of-fit	0.996	0.998	1.000
Mean execution time	2.22 s	2.31 s	2.48 s
Standard deviation	1.67 ms	166 ms	196 ms

Outlying measurements have moderate (20.2%) effect on estimated standard deviation.

Matrix vs List - `select`/Distribution `select` - 100+100 / 100x100/Matrix Distribution - Applicative version

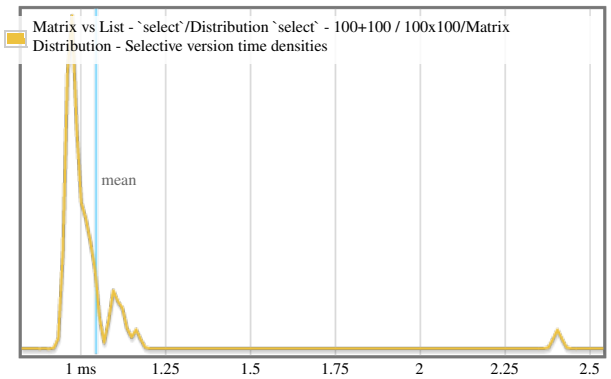


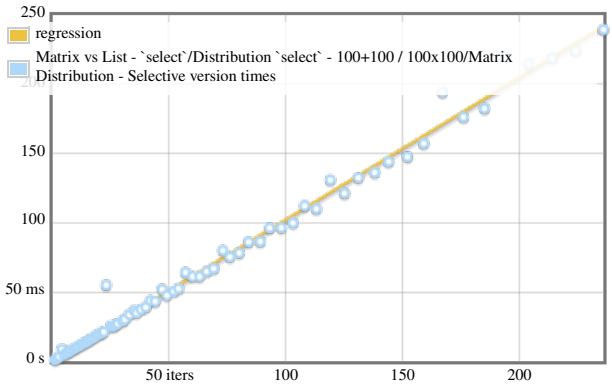


	lower bound	estimate	upper bound
OLS regression	2.31 ms	2.48 ms	2.64 ms
R² goodness-of-fit	0.862	0.922	0.964
Mean execution time	2.48 ms	2.67 ms	3.15 ms
Standard deviation	461 µs	906 µs	1.69 ms

Outlying measurements have severe (96.0%) effect on estimated standard deviation.

Matrix vs List - `select`/Distribution `select` - 100+100 / 100x100/Matrix Distribution - Selective version



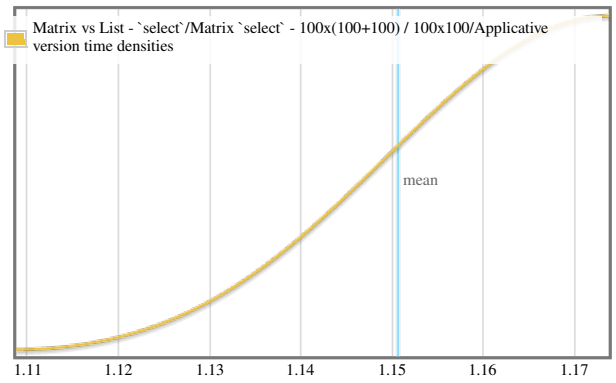


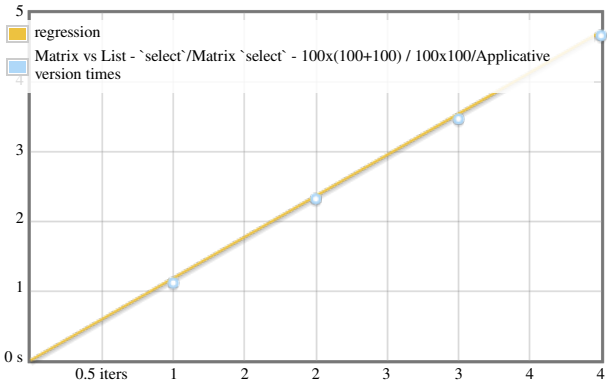
	lower bound	estimate	upper bound
OLS regression	995 µs	1.02 ms	1.06 ms

	lower bound	estimate	upper bound
R² goodness-of-fit	0.981	0.991	0.998
Mean execution time	1.01 ms	1.05 ms	1.19 ms
Standard deviation	49.4 µs	217 µs	493 µs

Outlying measurements have severe (92.4%) effect on estimated standard deviation.

Matrix vs List - `select`/Matrix `select` - 100x(100+100) / 100x100/Applicative version

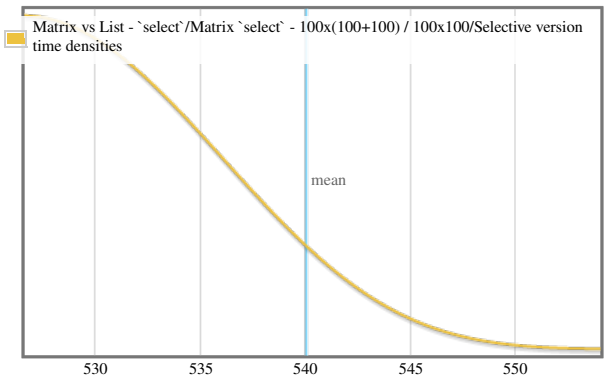


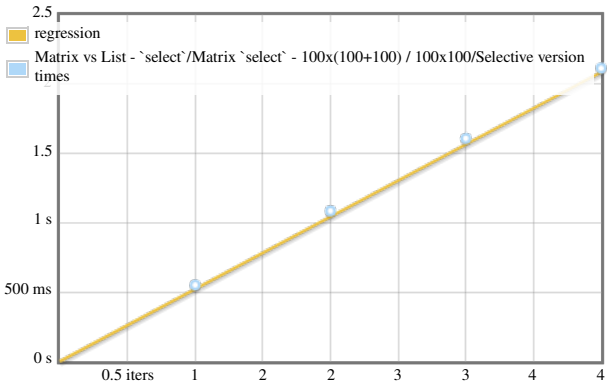


	lower bound	estimate	upper bound
OLS regression	1.15 s	1.18 s	1.21 s
R² goodness-of-fit	1.000	1.000	1.000
Mean execution time	1.11 s	1.15 s	1.16 s
Standard deviation	3.12 ms	24.7 ms	31.3 ms

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

Matrix vs List - `select`/Matrix `select` - 100x(100+100) / 100x100/Selective version





	lower bound	estimate	upper bound
OLS regression	506 ms	521 ms	534 ms
R² goodness-of-fit	1.000	1.000	1.000
Mean execution time	531 ms	540 ms	547 ms
Standard deviation	4.33 ms	9.72 ms	13.2 ms

Outlying measurements have moderate (18.7%) 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.
- R^2 *goodness-of-fit* is a measure of how accurately the linear regression model fits the observed measurements. If the measurements are not too noisy, R^2 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 criterion benchmark execution and performance analysis tool.

Criterion is developed and maintained by Bryan O'Sullivan.