Report of experiments

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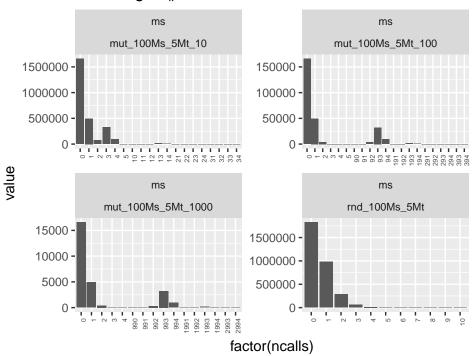
Contents

1	Input properties	2
2	Double vs. single rank	3
	2.1 Code	. 3
	2.2 Performance	. 4
3	Lazy vs non-lazy	5
	3.1 Code	. 5
	3.2 Performance	
	3.3 Sandbox timing	. 7
	3.4 Check	. 8
4	Parallelization	10
	4.1 Code	. 10
	4.2 Performance	

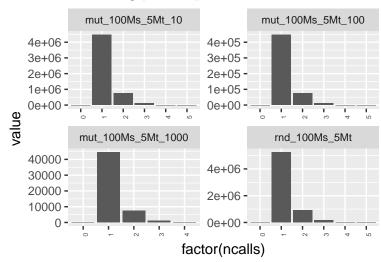
1 Input properties

For various types ("mut_XMs_YMt_Z" means s and t are random identical strings of length X, and Y million respectively with mutations inserted every Z characters. "rnd_XMs_YMt" means s and t are random strings of length X, and Y million respectively) of inputs run the MS algorithm and count the number of consecutive wl() or parent() calls during the runs or ms construction phase.

Counting wl() calls



Counting parent() calls



2 Double vs. single rank

2.1 Code

The single rank and double rank implementations in sdsl: rank support v.hpp link // RANK(idx) return *p + ((*(p+1)>>(63 - 9*((idx&0x1FF)>>6)))&0x1FF) +(idx&0x3F ? trait_type::word_rank(m_v->data(), idx) : 0); // DOUBLE RANK(i, j) $if((i>>8) == (j>>8)){$ const uint64_t* p = m_basic_block.data() + ((i>>8)&0xFFFFFFFFFFFFFEULL); res.first = *p + ((*(p+1))>(63 - 9*((i&0x1FF))>6)))&0x1FF) +(i&0x3F ? trait_type::word_rank(m_v->data(), i) : 0); res.second = *p + ((*(p+1))>(63 - 9*((j&0x1FF))>6)))&0x1FF) +(j&0x3F ? trait_type::word_rank(m_v->data(), j) : 0); } else { res.first = *p + ((*(p+1))>(63 - 9*((i&0x1FF))>6)))&0x1FF) +(i&0x3F ? trait_type::word_rank(m_v->data(), i) : 0); res.second = *p + ((*(p+1))>(63 - 9*((j&0x1FF))>6)))&0x1FF) +(j&0x3F ? trait_type::word_rank(m_v->data(), j) : 0); } return res;

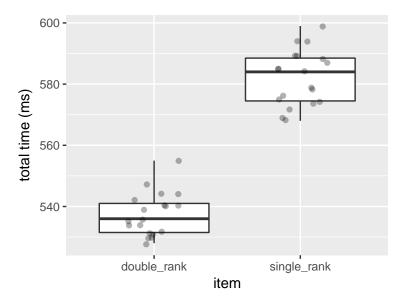
2.2 Performance

Table 1: Time (in ms) of 500K calls to wl() based on single_rank() or double_rank() methods on 100MB random DNA input; Mean/sd over 20 repetitions.

item	avg_time	sd_time
double_rank	537.47	6.96
single_rank	582.05	9.01

Table 2: Absolute (double_rank / single_rank) and relative (100 * |double_rank - single_rank| / single_rank) ratios of average times from the above table.

double_rank	$single_rank$	abs_ratio	rel_ratio	when
546.89 537.47	572.74 582.05	0.95 0.92		before after



3 Lazy vs non-lazy

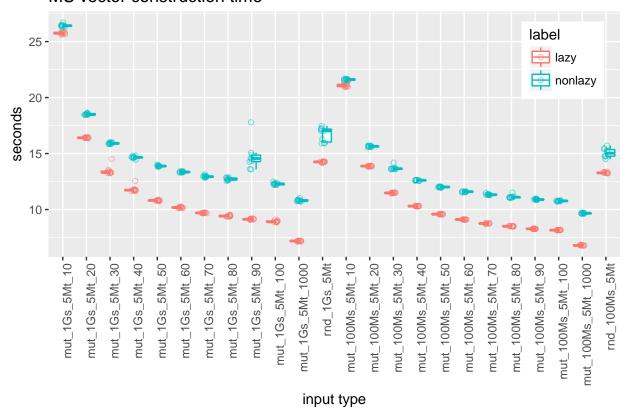
3.1 Code

The lazy and non-lazy versions differ in a couple of lines of code as follows

```
if(flags.lazy){
    for(; I.first <= I.second && h_star < ms_size; ){</pre>
        c = t[h_star];
        I = bstep_interval(st, I, c); //I.bstep(c);
        if(I.first <= I.second){</pre>
            v = st.lazy_wl(v, c);
            h_star++;
        }
    }
    if(h_star > h_star_prev) // // we must have called lazy_wl(). complete the node
        st.lazy_wl_followup(v);
} else { // non-lazy weiner links
    for(; I.first <= I.second && h_star < ms_size; ){</pre>
        c = t[h_star];
        I = bstep_interval(st, I, c); //I.bstep(c);
        if(I.first <= I.second){</pre>
            v = st.wl(v, c);
            h_star++;
        }
    }
}
```

3.2 Performance

MS vector construction time



The right panel shows the time to construct the **runs** vector. This stage is the same for both versions and is shown as a control. On the left panel it can be seen that speedup correlates positively with both the size of the indexed string and the mutation period.

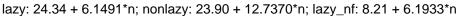
3.3 Sandbox timing

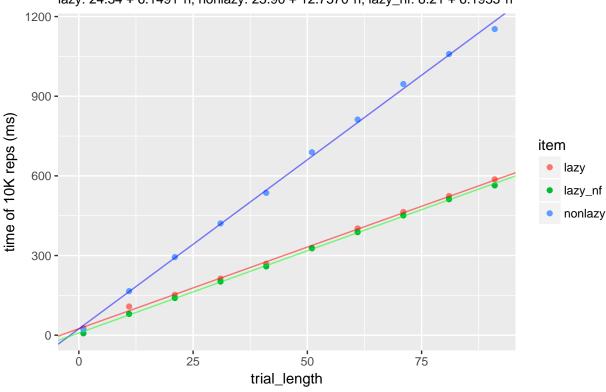
Measure the time of 10k repetitions of

- (lazy) n consecutive lazy_wl() calls followed by a lazy_wl_followup()
- (nonlazy) n consecutive wl() calls
- (lazy_nf) n consecutive lazy_wl() calls

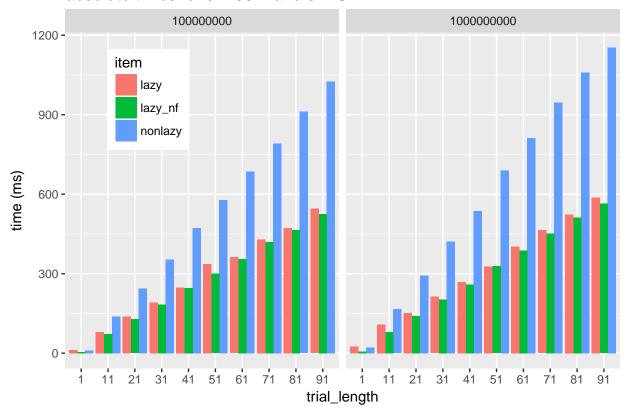
```
// lazy
for(size_type i = 0; i < trial_length; i++)
    v = st.lazy_wl(v, s_rev[k--]);
if(h_star > h_star_prev) // // we must have called lazy_wl(). complete the node
    st.lazy_wl_followup(v);
...
// non-lazy
for(size_type i = 0; i < trial_length; i++)
    v = st.wl(v, s_rev[k--]);
...
// lazy_nf
for(size_type i = 0; i < trial_length; i++)
    v = st.lazy_wl(v, s_rev[k--]);</pre>
```

indexed input size 1G





absolute times for s=100M and s=1G



3.4 Check

In the experiments above we ran the program with the "lazy" or "non-lazy" flag and measured. The total time of each experiment can be written as $t_l = l_l + a$ and $t_n = l_n + a$ for the two versions respectively; only the ts being known. Furthermore, we have \hat{l}_l and \hat{l}_n estimations – computed by combining the time / wl call with the number of with the count of wl calls in each input (Section "Input Properties"). Hence we should expect

$$\delta t = t_l - t_n = l_l + a - l_n - a = l_l - l_n \approx \delta \hat{l} = \hat{l}_l - \hat{l}_n$$

b_path	t_l1	t_n	l_l	l_n	$delta_t$	delta_l_hat
mut_100Ms_5Mt_10	21.12	21.61	8.56	6.16	-0.49	2.39
$mut_100Ms_5Mt_100$	8.16	10.77	3.36	4.33	-2.60	-0.97
mut_100Ms_5Mt_1000	6.80	9.67	2.84	4.15	-2.86	-1.31
$mut_100Ms_5Mt_20$	13.87	15.64	5.66	5.14	-1.77	0.52
$mut_100Ms_5Mt_30$	11.49	13.70	4.71	4.81	-2.21	-0.10
$mut_100Ms_5Mt_40$	10.31	12.60	4.22	4.64	-2.30	-0.41
$mut_100Ms_5Mt_50$	9.58	12.01	3.93	4.53	-2.43	-0.60
$mut_100Ms_5Mt_60$	9.11	11.58	3.74	4.47	-2.48	-0.72
$mut_100Ms_5Mt_70$	8.75	11.34	3.60	4.42	-2.59	-0.81
$mut_100Ms_5Mt_80$	8.51	11.13	3.50	4.38	-2.63	-0.88
$mut_100Ms_5Mt_90$	8.28	10.90	3.42	4.35	-2.62	-0.93
$mut_1Gs_5Mt_10$	25.75	26.43	7.57	6.65	-0.68	0.92
$mut_1Gs_5Mt_100$	8.94	12.29	3.49	4.90	-3.35	-1.41

b_path	t_1	t_n	1_1	l_n	delta_t	delta_l_hat
mut_1Gs_5Mt_1000	7.19	10.82	3.08	4.72	-3.63	-1.64
$mut_1Gs_5Mt_20$	16.42	18.52	5.30	5.68	-2.10	-0.37
$mut_1Gs_5Mt_30$	13.46	15.92	4.55	5.36	-2.46	-0.81
$mut_1Gs_5Mt_40$	11.81	14.66	4.17	5.20	-2.85	-1.02
$mut_1Gs_5Mt_50$	10.81	13.89	3.95	5.10	-3.08	-1.15
$mut_1Gs_5Mt_60$	10.19	13.36	3.80	5.03	-3.17	-1.24
$mut_1Gs_5Mt_70$	9.70	12.95	3.69	4.99	-3.26	-1.30
$mut_1Gs_5Mt_80$	9.43	12.72	3.61	4.95	-3.29	-1.35
$mut_1Gs_5Mt_90$	9.14	14.74	3.55	4.93	-5.60	-1.38
rnd_100Ms_5Mt	13.29	15.07	9.65	6.55	-1.78	3.10
rnd_1Gs_5Mt	14.25	16.72	8.20	6.92	-2.48	1.28

The numbers are not identical (process dependent factors might influence the running time of function calls), but they are correlated $(corr(\delta t, \delta \hat{l}) = 0.71)$.

4 Parallelization

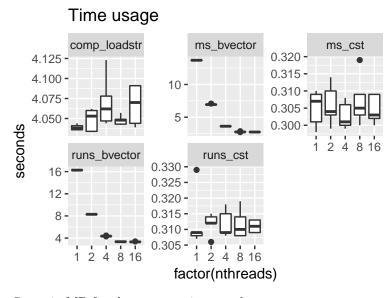
4.1 Code

See the pseudo-code in the repo (link)

4.2 Performance

Run the MS construction program on the same input (random strings s of length 100M and t of length 5M) with varying parallelization degree (nthreads = number of threads).

The time is reported over 5 runs for each fixed number of threads.



Space in MB for the same settings as above.

Each thread allocates its own ms vector with initial size |t|/nthreads then it resizes by a factor of 1.5 each time it needs to. Resizing will always result in a vector smaller than 2|t| elements.

