Report of experiments

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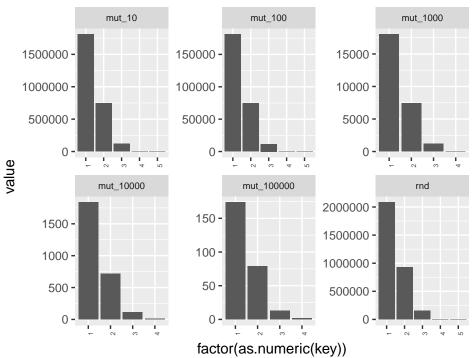
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1 Input properties

For various types of inputs ("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) run the MS algorithm and count the number of

- consecutive parent() calls during the runs construction.
- consecutive wl() calls during the ms construction.
- the number of 1s in the runs bit vector
- double rank calls that fail (i.e the search down the WT is interrupted prior to reaching a leaf)
- the number of maximal repeats

Counting consecutive parent() calls



Counting consecutive parent() calls

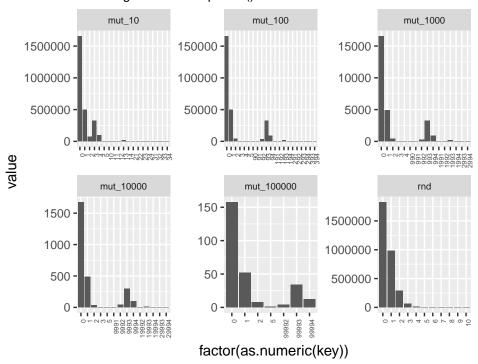


Table 1: Double rank iterations that fail for various input types.

b_path	fail	nofail	perc
rnd_100Ms_5Mt	690911	1818974	37.98
mut_100Ms_5Mt_10	1494544	2323907	64.31
mut_100Ms_5Mt_100	4649474	4732340	98.25
mut_100Ms_5Mt_1000	4965082	4973326	99.83
mut_100Ms_5Mt_10000	4996520	4997320	99.98
$mut_100Ms_5Mt_100000$	4999634	4999731	100.00

Table 2: Composition of the runs vector for various input types.

inp_type	one	zero	zero_perc
mut_10	2323907	2676093	115.15
mut_100	4732340	267660	5.66
mut_1000	4973326	26674	0.54
mut_10000	4997320	2680	0.05
mut_100000	4999731	269	0.01
rnd	1818974	3181026	174.88

Table 3: Composition of the B vector (containing ends of maximal repeats) for various input types.

inp_type	maximal	non_maximal	non_maximal_perc
mut_10	34894073	982212	2.81
mut_100	34893276	981720	2.81

inp_type	maximal	non_maximal	non_maximal_perc
mut_1000	34891214	980977	2.81
mut_10000	34894098	982309	2.82
mut_100000	34897894	980817	2.81
rnd	50053548	5525912	11.04

2 Current performance

Table 4: Run time in seconds, on random input with $|\mathbf{s}|=1\mathrm{MB},\,|\mathbf{t}|=5\mathrm{MB}$

lazy	fail	maxrep	$total_s$	$maxrep_s$	$tot_minus_maxrep_s$
0	1	0	90.351	NA	90.351
1	1	0	90.817	NA	90.817
1	1	1	92.478	2.024	90.454
1	0	0	92.844	NA	92.844
0	0	0	92.850	NA	92.850
0	1	1	93.020	2.020	91.000
0	0	1	94.044	1.987	92.057
1	0	1	94.092	1.992	92.100

3 Double vs. single rank

3.1 Code

// RANK(idx) const uint64_t* p = m_basic_block.data() + ((idx>>8)&0xFFFFFFFFFFFFFEULL); return *p + ((*(p+1)>>(63 - 9*((idx&0x1FF)>>6)))&0x1FF) +(idx&0x3F ? trait_type::word_rank(m_v->data(), idx) : 0); // DOUBLE RANK OD(i, j) $if((i>>8) == (j>>8)){$ 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 { 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); } return res

The single rank and double rank implementations in sdsl: rank support v.hpp link

The FC implementation worked better. Data are saved in rank_timing.csv.fc and rank_timing.csv.od but not shown below.

(i&0x3F ? trait_type::word_rank(m_v->data(), i) : 0),

(j&0x3F ? trait_type::word_rank(m_v->data(), j) : 0));

3.2 Performance

// DOUBLE RANK FC(i, j)

const uint64_t* b = m_basic_block.data();

return (*pi + ((*(pi+1)>>(63 - 9*((i&0x1FF)>>6)))&0x1FF) +

pj + (((pj+1)>>(63 - 9*((j&0x1FF)>>6)))&0x1FF) +

item	${\rm avg_time}$	sd_time
double double_fail single	551.5556 559.0000 583.7778	4.666667 0.000000 8.613620

4 Maxrep

4.1 Maxrep construction

Applying the first optimization we get 8% improvement on a (ran of a 1MB input string).

```
# EXISTING CODE
denas@denas-osx:$ for i in 1 2 3 4 5; \
do compute_maxrep -answer 0 -load_cst 0 -s_path datasets/synthetic/rnd_1Ms_5Mt.s; \
done 2>&1 | grep mill
 * computing MAXREP DONE ( 1098 milliseconds)
 * computing MAXREP DONE ( 1116 milliseconds)
 * computing MAXREP DONE ( 1120 milliseconds)
 * computing MAXREP DONE ( 1094 milliseconds)
 * computing MAXREP DONE ( 1100 milliseconds)
denas@denas-osx:$
# OPTIMIZED CODE
denas@denas-osx:$ for i in 1 2 3 4 5; \
do compute_maxrep -answer 0 -load_cst 0 -s_path datasets/synthetic/rnd_1Ms_5Mt.s; \
done 2>&1 | grep mill
 * computing MAXREP DONE ( 1020 milliseconds)
 * computing MAXREP DONE ( 1023 milliseconds)
 * computing MAXREP DONE ( 999 milliseconds)
 * computing MAXREP DONE ( 1020 milliseconds)
 * computing MAXREP DONE ( 1015 milliseconds)
```

4.2 Performance

The figure below shows 8 runs of the program with and without the use of the maxrep (or B) vector. The plot shows times (in seconds) for the construction of the ms bitvector. The table below that, shows the time (in seconds) to construct the maxrep vector. The input data is random and has |s|=100MB and |t|=5MB.

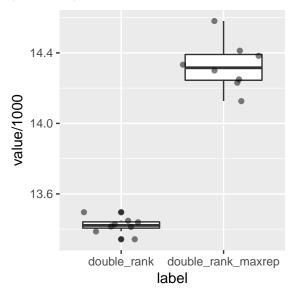


Table 6: Time to build the maxrep bit vector

label	avg	se
double_rank_maxrep	113	8

5 Lazy vs non-lazy

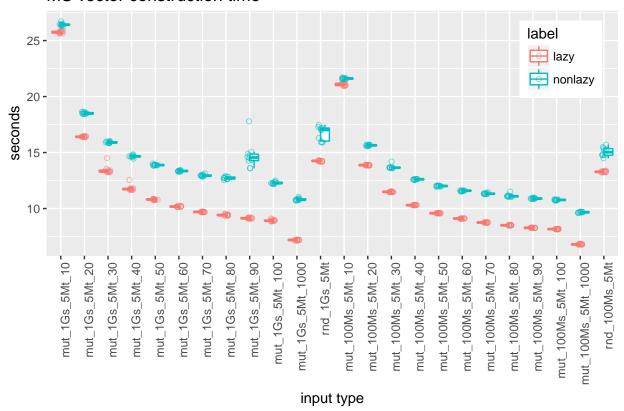
5.1 Code

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++;
        }
    }
}
```

5.2 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.

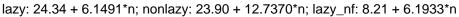
5.3 Sandbox timing

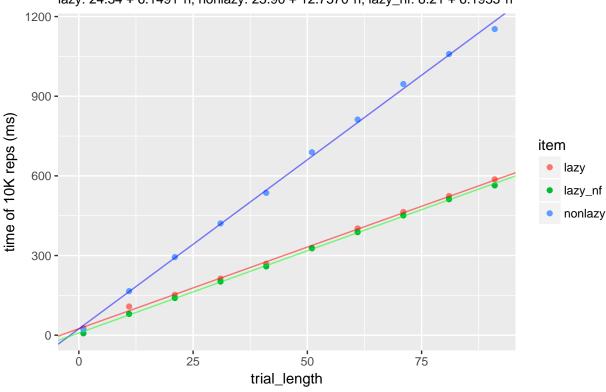
Measure the time of 10k repetitions of

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

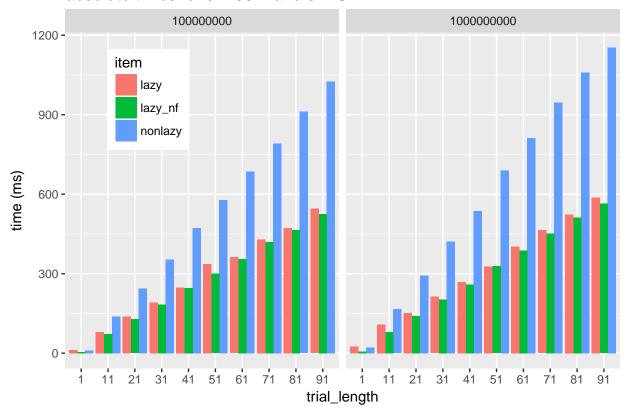
```
// lazy
for(size_type i = 0; i < trial_length; i++)
    v = st.lazy_wl(v, s_rev[k--]);
if(h_star > h_star_prev) // // we must have called lazy_wl(). complete the node
    st.lazy_wl_followup(v);
...
// non-lazy
for(size_type i = 0; i < trial_length; i++)
    v = st.wl(v, s_rev[k--]);
...
// lazy_nf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



5.4 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

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

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

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 $(corr(\delta t, \delta \hat{l}) = 0.71)$.

6 Double rank and fail

6.1 Code

```
// Given subtree_double_rank(v, i, j) -> (a.first, a.second) -- to simplify code
// DOUBLE RANK: int i, int j, char c
p = bit_path(c)
result_i, result_j = i, j;
node_type v = m_tree.root();
for (1 = 0; 1 < path_len; ++1, p >>= 1) {
 a = subtree_double_rank(v, m_tree.bv_pos(v) + result_i, m_tree.bv_pos(v) + result_j);
  if(p&1){ // left child
      if(result_i > 0) result_i = a.first;
      if(result_j > 0) result_j = a.second;
  } else { // right child
      if(result_i > 0) result_i -= a.first;
      if(result_j > 0) result_j -= a.second;
 v = m_tree.child(v, p&1); // goto child
return(result_i, result_j)
// DOUBLE RANK AND FAIL
p = bit_path(c)
result_i, result_j = i, j;
node_type v = m_tree.root();
for (1 = 0; 1 < path_len; ++1, p >>= 1) {
  a = subtree_double_rank(v, m_tree.bv_pos(v) + result_i, m_tree.bv_pos(v) + result_j);
  if(p&1){ // left child
      if(result_i > 0) result_i = a.first;
      if(result_j > 0) result_j = a.second;
  } else { // right child
      if(result_i > 0) result_i -= a.first;
      if(result_j > 0) result_j -= a.second;
  if(result_i == result_j) // Weiner Link call will fail
   return(0, 0)
  v = m_tree.child(v, p&1); // goto child
return(result_i, result_j)
```

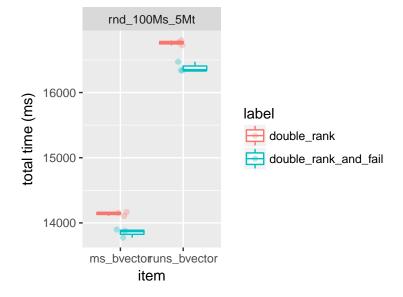
6.2 Performance

Table 8: 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	label	b_path	avg_time	sd_time
ms_bvector	double_rank	rnd_100Ms_5Mt	14142.00	30.27
$ms_bvector$	double_rank_and_fail	rnd_100Ms_5Mt	13850.33	66.16
$runs_bvector$	$double_rank$	rnd_100Ms_5Mt	16763.67	37.69
$runs_bvector$	$double_rank_and_fail$	rnd_100Ms_5Mt	16384.00	76.22

Table 9: Single vs. double rank. Absolute (double / single) and relative (100 * |double - single| / single) ratios of average times.

item	double_rank	double_rank_and_fail	abs_ratio	rel_ratio
ms_bvector	14142.00	13850.33	0.98	2.06
runs bvector	16763.67	16384.00	0.98	2.26



7 Parallelization

7.1 Code

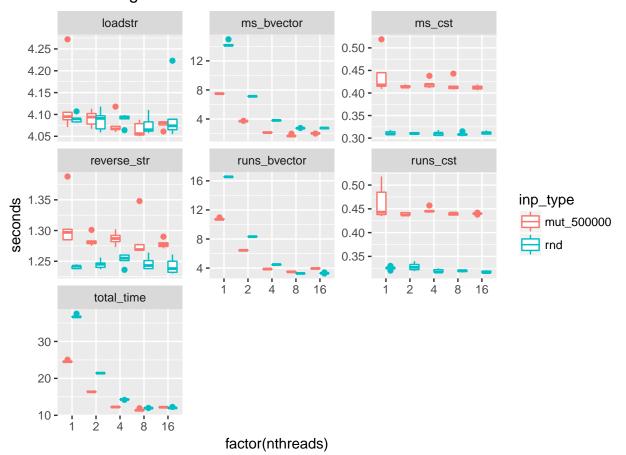
See the pseudo-code in the repo (link)

7.2 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.

Time usage



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.

