

RLPBWT

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Outline

- 1 The Old Idea
- 2 The New Idea
- 3 BitVector Version
- 4 Matching Statistics

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Durbin's Algorithm

Algorithm 1 Algorithm 5 from Durbin's paper

function FIND_SET_MAXIMAL_MATCHES_FROM_Z(z)

for $k \leftarrow 0$ **to** N **do**

$e, f, g \leftarrow \text{Update_Z_Matches}(k, z, e, f, g)$

function UPDATE_Z_MATCHES(k, z, e, f, g)

$f' \leftarrow w(k, f, z[k])$

▷ a_k , d_k and y_i^k as in Durbin's paper

$g' \leftarrow w(k, g, z[k])$

if $f' < g'$ **then**

▷ if k is $N - 1$ report matches from e_k to $N - 1$

$e' \leftarrow e_k$

else

▷ report matches from e_k to k

$e' \leftarrow d_{k+1}[f'] - 1$

if $z[e'] = 0$ **and** $f' > 0$ **then**

$f' \leftarrow g' - 1$

while $z[e' - 1] = y_{f'}^{k+1}[e' - 1]$ **do** $e' \leftarrow e' - 1$

while $d_{k+1}[f'] \leq e'$ **do** $f' \leftarrow f' - 1$

else

$g' \leftarrow f' + 1$

while $z[e' - 1] = y_{f'}^{k+1}[e' - 1]$ **do** $e' \leftarrow e' - 1$

while $g' < M$ **and** $d_{k+1}[g'] \leq e'$ **do** $g' \leftarrow g' + 1$

return e', f', g'

Durbin's Algorithm Example

PBWT	00	01	02	03	04	05	06	07	08	09	10	11	12	13	14
00	1	1	0	1	1	0	0	0	1	0	0	1	1	1	1
01	1	1	0	1	1	0	0	0	1	0	0	1	1	1	1
02	1	1	0	1	1	1	0	0	0	1	1	1	0	1	1
03	1	1	0	1	1	0	0	0	1	0	0	1	1	0	1
04	0	1	0	1	0	1	0	0	1	0	0	1	1	0	1
05	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
06	0	1	0	1	0	1	0	0	0	0	0	1	0	0	0
07	0	1	0	1	1	1	0	0	0	0	0	0	1	0	1
08	0	1	0	0	1	0	0	0	1	0	0	0	1	0	1
09	0	1	0	1	0	0	0	0	1	0	0	0	0	0	1
10	0	1	0	1	1	0	0	0	0	0	0	1	1	0	1
11	0	1	0	0	1	0	1	1	0	0	0	1	0	0	1
12	0	1	0	0	1	0	0	1	0	0	0	0	0	0	1
13	0	1	0	0	0	0	0	1	0	0	0	0	0	0	1
14	0	1	0	0	0	0	0	0	0	0	0	0	0	0	1
15	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1
16	0	0	0	1	0	0	0	0	0	0	0	1	0	0	1
17	1	0	1	0	0	0	0	0	0	0	1	1	0	0	1
18	0	0	1	0	0	0	0	0	0	0	1	1	0	0	1
19	0	1	0	0	0	0	0	0	0	0	1	1	0	0	1
z	0	1	0	0	1	0	1	0	0	0	1	1	1	0	1



Durbin's Algorithm Example

X	00	01	02	03	04	05	06	07	08	09	10	11	12	13	14	15
00	0	4	0	0	8	14	14	14	14	0	0	0	7	1	18	11
01	1	5	1	1	11	15	15	15	15	16	16	16	19	9	4	18
02	2	6	2	2	12	17	0	0	0	8	11	18	1	10	5	4
03	3	7	3	3	13	0	9	9	9	11	18	17	14	18	6	5
04	4	8	4	4	14	4	10	10	10	18	17	4	15	4	2	6
05	5	9	5	5	15	5	16	16	16	17	4	5	9	5	3	2
06	6	10	6	6	17	6	8	8	8	4	5	6	10	6	11	3
07	7	11	7	7	18	7	11	11	11	5	6	7	0	2	12	12
08	8	12	8	8	19	9	12	12	12	6	7	19	16	3	13	13
09	9	13	9	9	0	10	13	13	13	7	19	1	18	11	8	8
10	10	14	10	10	1	16	18	18	18	19	1	2	17	12	7	7
11	11	15	11	11	2	8	19	1	17	1	2	3	4	13	19	19
12	12	16	12	12	3	11	1	2	4	2	3	14	5	8	14	14
13	13	18	13	13	4	12	2	3	5	3	14	15	6	7	15	15
14	14	19	14	14	5	13	3	17	6	14	15	9	2	19	0	0
15	15	0	15	15	6	18	17	4	7	15	9	10	3	14	16	16
16	16	1	16	16	7	19	4	5	19	9	10	11	11	15	17	17
17	17	2	18	17	9	1	5	6	1	10	12	12	12	0	1	1
18	18	3	19	18	10	2	6	7	2	12	13	13	13	16	9	9
19	19	17	17	19	16	3	7	19	3	13	8	8	8	17	10	10

Durbin's Algorithm Example

X	00	01	02	03	04	05	06	07	08	09	10	11	12	13	14
00	1	0	0	1	0	0	0	0	0	0	0	1	1	0	1
01	1	0	0	1	1	0	0	1	0	0	0	0	0	1	1
02	1	0	0	1	1	0	0	1	0	0	0	1	0	0	1
03	1	0	0	1	1	0	0	1	0	0	0	1	0	0	1
04	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
05	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
06	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
07	0	1	0	1	0	1	0	0	0	0	0	0	1	0	1
08	0	1	0	0	1	0	0	0	0	1	1	1	0	0	1
09	0	1	0	1	0	0	0	0	1	0	0	0	0	1	1
10	0	1	0	1	0	0	0	0	1	0	0	0	0	1	1
11	0	1	0	0	1	0	0	0	0	0	1	1	0	0	0
12	0	1	0	0	1	0	0	0	1	0	1	1	0	0	1
13	0	1	0	0	1	0	0	0	1	0	1	1	0	0	1
14	0	1	0	0	0	0	0	0	1	0	0	0	1	0	1
15	0	1	0	0	0	0	0	0	1	0	0	0	1	0	1
16	0	1	0	1	0	0	0	0	0	0	0	1	1	0	1
17	1	1	0	0	0	1	0	0	0	0	0	1	1	0	1
18	0	1	1	0	1	0	0	0	0	0	0	1	0	0	1
19	0	1	1	0	1	0	1	0	0	0	0	0	1	0	1
z	0	1	0	0	1	0	1	0	0	0	1	1	1	0	1

$pre_e = 6$
at 7

$pre_e = 8$ $pre_e = 13$
at 11 at 13

X	00	01	02	03	04	05	06	07	08	09	10	11	12	13	14
00	1	0	0	1	0	0	0	0	0	0	0	1	1	0	1
01	1	0	0	1	1	0	0	1	0	0	0	0	0	1	1
02	1	0	0	1	1	0	0	1	0	0	0	1	0	0	1
03	1	0	0	1	1	0	0	1	0	0	0	1	0	0	1
04	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
05	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
06	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
07	0	1	0	1	0	1	0	0	0	0	0	0	1	0	1
08	0	1	0	0	1	0	0	0	0	1	1	1	0	0	1
09	0	1	0	1	0	0	0	0	1	0	0	0	0	1	1
10	0	1	0	1	0	0	0	0	1	0	0	0	0	1	1
11	0	1	0	0	1	0	0	0	0	0	1	1	0	0	0
12	0	1	0	0	1	0	0	0	1	0	1	1	0	0	1
13	0	1	0	0	1	0	0	0	1	0	1	1	0	0	1
14	0	1	0	0	0	0	0	0	1	0	0	0	1	0	1
15	0	1	0	0	0	0	0	0	1	0	0	0	1	0	1
16	0	1	0	1	0	0	0	0	0	0	0	1	1	0	1
17	1	1	0	0	0	1	0	0	0	0	0	1	1	0	1
18	0	1	1	0	1	0	0	0	0	0	0	1	0	0	1
19	0	1	1	0	1	0	1	0	0	0	0	0	1	0	1
z	0	1	0	0	1	0	1	0	0	0	1	1	1	0	1

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New Idea I

Let's take a step back and get closer to Durbin's original idea.

At the moment we save:

- the position of every head of a run
- a boolean to mark if the first run is composed by zeros or ones
- the c value of the column
- a single value for u and v (that are as in Durbin)
- the whole *divergence array*, actually the *LCP array*, (**WIP**)

New Idea I

```
column: 5
start with 0? yes, c: 15
0    0
2    2
3    1
4    3
8    5
0 5 4 1 3 5 5 5 5 5 5 0 5 5 5 2 5 1 5 5
```

Figura: Example, column 5: 00101111000000000000

New Idea II

uv values trick

Values u and v increase alternately in the biallelic case so we save every time the only value that increase at the head of a run.

The, with a simple *If/Else* selection based on the first element of the column and the index of the run and on being even or odd of the index we can extract both u and v values. Infact the two values are, alternatively, saved in the current index and in the previous one.

$w(i, \sigma)$ function

We can use the same *LF-mapping* as in Durbin but we have to consider sometimes an *offset* between the position of the head of the run, that's i , which contains the “virtual” index, and the index itself.

$$w(i, \sigma) = \begin{cases} u[i] + \text{offset} & \text{if } \sigma = 0 \\ c + v[i] + \text{offset} & \text{if } \sigma = 1 \end{cases}$$

New Idea III

External haplotype matches

Then we proceed as in Durbin, updating f and g using $w(i, \sigma)$.

Every time we “virtually” use indexes over the whole column but actually run heads plus offsets are used.

In case we update e using f and the *divergence/LCP array* (**WIP**).

In order to update e we should in theory follow the line indicated in $i + 1$ by f in the original panel which we have not memorized. So, at most at a cost of $O(r)$ for every column, we proceed to reverse the use of u and v to move backwards between the columns virtually following a row of the original panel.

Then we use the *divergence/LCP array* (**WIP**) to update f and g depending on the case.

After detect a match, we can know the cardinality of the lines that match but not what they are,

New Idea IV

WIP

At the moment *divergence array* is saved as an `sds1::int_vector<>` on which it's used `sds1::util::bit_compress()` in order to save space. The original idea of thresholds seems to me absolutely not applicable but maybe we can think of storing only a subset of the *divergence/LCP array* and I'm thinking how to do it.

Testing

Benchmark

At the moment I'm testing the implementation using the sample data (VCF files) at:

https://github.com/ZhiGroup/Syllable-PBWT/tree/master/sample_data
(the panel is 900×500 with 100 queries).

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The column data I

We save, for every column:

- a bit vector of the same length of the dense PBWT column, with 1 in every position before a head of a run
- 2 bit vector that can be queried to obtain u and v value. So we have a bitvector for zeros that contains 1 in position i iff we have i zeros at point in the column when we change value anche similar for ones(**to be explained better**)
- the c value and a bool to indicate how a column start

The column data II

example

If we have a column:

$$c = 00001110001111110001111111$$

We save:

$$h = 000100100100000100010000000(1)$$

$$u = 00010010001 \text{ (to represent 4,3 and 4 zeros)}$$

$$v = 0010000010000001 \text{ (to represent 3,6 and 7 ones)}$$

The column data III

example

If we want to obtain, for example, the number of zeros before and index i , for example $i = 18$:

- we *rank* h to obtain the run that contain i , $rank_h(18) = 4$
- we know that the column start with 0 so we are in a run of zeros and we have $\lfloor \frac{4}{2} \rfloor = 2$ run's of zeros before
- we know that the number of zeros in the previous complete runs is $select_u(2) + 1 = 7$
- remain zeros in the run to compute are given by $i - (select_h(4) + 1)$

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Matching Statistics and RLPBWT

Some definitions

- for every run in a column of the PBTW with define the **threshold** as the index of the minimum *lcp value* (and we save it as a sparse bitvector of the same length of a column with 1 in these positions)
- for every run in a column of the PBTW with define the **prefix array samples** as the prefix array value at the begin of the run and at the end
- we define matching statistics as a two components (row, len) vector MS of the same length of the query such that $\forall k \in [0, |z|)$:
 - $panel_{row}[k - (len+1)..k] = z[k - (len + 1)..k]$
 - $z[k - (len + 1)..k + 1]$ doesn't match with any row in the panel
- we compute pos and len in two different scans
- we need random access to the panel. At this time we are using a bitvector for every column of the panel but we can use a data structure, called *SLP*, with a very low memory cost but with high access times

Example

PBWT	00	01	02	03	04	05	06	07	08	09	10	11	12	13	14
00	1	1	0	1	1	0	•0	0	1	0	0	1	1	1	1
01	1	1	0	1	1	0	0	0	1	0	0	1	1	1	1
02	1	1	0	1	1	1	0	0	0	1	1	1	0	1	1
03	1	1	0	1	1	0	0	0	1	0	0	1	1	0	1
04	0	1	0	1	0	1	0	0	1	0	0	1	1	0	1
05	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
06	0	1	0	1	•0	1	0	0	0	0	0	1	0	0	0
07	0	1	0	1	1	1	0	0	0	0	0	0	1	0	1
08	0	1	0	0	1	0	0	0	1	0	0	0	1	0	1
09	0	1	0	1	0	0	•0	0	1	0	•0	0	0	0	1
10	0	1	0	1	1	0	0	0	0	0	0	1	1	0	1
11	0	1	0	0	1	0	1	1	0	0	0	1	0	0	1
12	0	1	0	0	1	0	0	1	0	0	0	0	0	0	1
13	0	1	0	0	0	0	0	1	0	0	•0	0	0	0	1
14	0	1	0	0	0	0	0	0	0	0	0	0	0	0	1
15	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1
16	0	0	0	1	0	0	0	0	0	0	0	1	•0	0	1
17	1	0	•1	0	0	0	0	0	0	0	1	1	0	0	1
18	0	0	1	0	0	0	0	0	0	0	1	1	0	0	1
19	0	1	0	0	0	0	0	0	0	0	1	1	0	0	1
z	0	1	0	0	1	0	1	0	0	0	1	1	1	0	1
POS	19	19	17	17	13	13	19	19	19	19	11	11	17	17	17
LEN	1	2	2	3	5	6	4	5	6	7	4	5	2	3	4

Example

k	00	01	02	03	04	05	06	07	08	09	10	11	12	13	14
z	0	1	0	0	1	0	1	0	0	0	1	1	1	0	1
row	19	19	17	17	13	13	19	19	19	19	11	11	17	17	17
len	1	2	2	3	5	6	4	5	6	7	4	5	2	3	4

Matching Statistics using LCE queries

- using an SLP we can make **Longest Common Extensions (LCE) queries**. Given two row of the panel and a column we can compute the longest common suffix as far as the given column.
- as for the case of the thresholds study we starts from the bottom-left symbol in the panel and we move left to right using LF-mapping
- if we have a match we continue adding 1 every time to the *len* of *MS*
- every time we have a mismatch we look for the previous and the next good symbol in the column. Using the prefix array samples we have the original-row index of the end of the previous run and the original-row index of the begin of the next run (if they exists)
- if we are in a column with only the wrong symbol we restart the computation from the last row of the original panel
- we compute the two *LCE* and we take the longest
- next *pos* in the *MS* will be the index of the row with the maximum *LCE* and the *len* of the *MS* will be the length of the *LCE* plus 1

Example

PBWT	00	01	02	03	04	05	06	07	08	09	10	11	12	13	14
00	1	1	0	1	1	0	0	0	1	0	0	1	1	1	1
01	1	1	0	1	1	0	0	0	1	0	0	1	1	1	1
02	1	1	0	1	1	1	0	0	0	1	1	1	0	1	1
03	1	1	0	1	1	0	0	0	1	0	0	1	1	0	1
04	0	1	0	1	0	1	0	0	1	0	0	1	1	0	1
05	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
06	0	1	0	1	0	1	0	0	0	0	0	1	0	0	0
07	0	1	0	1	1	1	0	0	0	0	0	0	1	0	1
08	0	1	0	0	1	0	0	0	1	0	0	0	1	0	1
09	0	1	0	1	0	0	0	0	1	0	0	0	0	0	1
10	0	1	0	1	1	0	0	0	0	0	0	1	0	0	1
11	0	1	0	0	1	0	0	1	0	0	0	1	0	0	1
12	0	1	0	0	1	0	0	1	0	0	0	0	0	0	1
13	0	1	0	0	0	0	0	1	0	0	0	0	0	0	1
14	0	1	0	0	0	0	0	0	0	0	0	0	0	0	1
15	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1
16	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1
17	1	0	1	0	0	0	0	0	0	0	1	1	0	0	1
18	0	0	1	0	0	0	0	0	0	0	1	1	0	0	1
19	0	1	0	0	0	0	0	0	0	0	1	1	0	0	1
z	0	1	0	0	1	0	1	0	0	0	1	1	1	0	1
POS	19	19	16	15	13	13	19	19	19	19	11	11	17	17	17
LEN	1	2	3	4	5	6	4	5	6	7	4	5	2	3	4

Example

X	00	01	02	03	04	05	06	07	08	09	10	11	12	13	14
00	1	0	0	1	0	0	0	0	0	0	0	1	1	0	1
01	1	0	0	1	1	0	0	1	0	0	0	0	0	1	1
02	1	0	0	1	1	0	0	1	0	0	0	1	0	0	1
03	1	0	0	1	1	0	0	1	0	0	0	1	0	0	1
04	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
05	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
06	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
07	0	1	0	1	0	1	0	0	0	0	0	0	1	0	1
08	0	1	0	0	1	0	0	0	0	1	1	1	0	0	1
09	0	1	0	1	0	0	0	0	1	0	0	0	0	1	1
10	0	1	0	1	0	0	0	0	1	0	0	0	0	1	1
11	0	1	0	0	1	0	0	<u>0</u>	<u>0</u>	<u>0</u>	1	<u>1</u>	0	0	0
12	0	1	0	0	1	0	0	0	1	<u>0</u>	1	1	0	0	1
13	<u>0</u>	<u>1</u>	0	<u>0</u>	1	0	0	0	1	0	1	1	0	0	1
14	0	1	0	0	0	0	0	0	1	0	0	0	1	0	1
15	<u>0</u>	<u>1</u>	<u>0</u>	<u>0</u>	0	0	0	0	1	0	0	0	1	0	1
16	<u>0</u>	<u>1</u>	0	1	0	0	0	0	0	0	0	1	1	0	1
17	1	<u>1</u>	0	0	0	1	0	0	0	0	0	<u>1</u>	1	0	1
18	0	1	1	<u>0</u>	1	0	0	0	0	0	0	1	0	0	1
19	<u>0</u>	<u>1</u>	1	<u>0</u>	<u>1</u>	<u>0</u>	1	<u>0</u>	<u>0</u>	<u>0</u>	0	0	1	0	1
z	0	1	0	0	1	0	1	0	0	0	1	1	1	0	1

Example

k	00	01	02	03	04	05	06	07	08	09	10	11	12	13	14
z	0	1	0	0	1	0	1	0	0	0	1	1	1	0	1
row	19	19	16	15	13	13	19	19	19	19	11	11	17	17	17
len	1	2	3	4	5	6	4	5	6	7	4	5	2	3	4