

SFB 876 Verfügbarkeit von Information durch Analyse unter Ressourcenbeschränkung



A Memory-Efficient Data Structure for Pattern Matching in DNA with Backward Search

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Since backward search was introduced by Ferragina and Manzini, it became a standard index-based linear-time exact pattern search technique [2]. Due to the inherently high memory usage of its auxiliary tables, we developed a data structure that provides at least a 20-fold data reduction of memory usage without increasing computation time significantly.

Backward search

Improved backward search

Given: A text T with n = |T| and pattern P with m = |P| over a finite alphabet Σ . **Task:** Find all occurences of *P* in *T* in $\mathcal{O}(m)$ time.

Example

T = GCTATGATAGTCAT

Backward search uses trick by exploiting sorted order of characters in Burrows-Wheeler transform [1] of T.

 $T_{\rm bwt} = {\rm TTCGTTGT}AAACGA$

Auxiliary tables

Two auxiliary tables must be pre-computed.

- •less: table of size $|\Sigma|$, where less [c] is the number of characters in T lexicographically smaller than c
- occ: $|\Sigma| \times n$ matrix containing the number of cs in the BWT up to (and including) index d

С	less[c]	occ[<i>c</i>]														
Α	1	0,	0,	0,	0,	0,	0,	0,	0,	0,	1,	2,	3,	3,	3,	4
С	5	0,	0,	1,	1,	1,	1,	1,	1,	1,	1,	1,	1,	2,	2,	2
G	7	0,	0,	0,	1,	1,	1,	2,	2,	2,	2,	2,	2,	2,	3,	3
Т	10	1,	2,	2,	2,	3,	4,	4,	5,	5,	5,	5,	5,	5,	5,	5
$T_{\rm bwt}$	=	Т	Т	С	G	Т	Т	G	Т	\$	А	А	Α	С	G	Α

Idea: Lots of information in occ table are redundant. Difference of following entries is at most 1. Store information where character appears bitwise in app table.

С	app[c]														
Α	0,	0,	0,	0,	0,	0,	0,	0,	0,	1,	1,	1,	0,	0,	1
С	0,	0,	1,	0,	0,	0,	0,	0,	0,	0,	0,	0,	1,	0,	0
G	0,	0,	0,	1,	0,	0,	1,	0,	0,	0,	0,	0,	0,	1,	0
Т	1,	1,	0,	0,	1,	1,	0,	1,	0,	0,	0,	0,	0,	0,	0
$T_{\sf bwt} =$	Т	Т	С	G	Т	Т	G	Т	\$	Α	Α	А	С	G	Α

Bits are stored in blocks of long int type. **Trick:** using hardware implemented command popent to count bits in block and additionally store summed up occurences in occ table for every r-th entry. Typically the register size (long int) r = 64 is suitable.

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С		occ[c]		
Α	0,	0,	3	
С	1,	1,	1	
G	1,	2,	2	
Т	2,	5,	5	

Pattern matching

Process the pattern *backwards* character by character and update the suffix array interval [L, R] that points to the occurrences of the processed pattern suffix at each step. The update step is

> $L^+(c) = \operatorname{less}[c] + \operatorname{occ}[c][L-1],$ $R^+(c) = \operatorname{less}[c] + \operatorname{occ}[c][R] - 1.$

Interval [L, R] determines positions in suffix array where all prefixes equal P. This process takes m iteratiorns.

Bioinformatics context

Backward search is widely used in read mappers like BWA [3]. Either searching for exact seeds or error-tolerant alignment is possible. Since double-stranded DNA contains about $n \approx 6.2 \cdot 10^9$ bases, the occ table (int-typed entries of 4 byte each) reaches about

 $T_{bwt} = T T C G T T G T $ A A A C G A$

(Example: r = 4)

Determination of new interval

• Given: interval L, R from previous iteration and current char c

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• Determine long block in app table: block = L >> 6
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• Determine the i significant bits in the block: i = L \& 63
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1 block_l = 0; block_r = R >> 6
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2 if L > 0:

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block_l = (L - 1) >> 6
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appear_l = popcnt(app[c][block_l] << (63 - ((L - 1) & 63)))</pre>
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5 \text{ appear}_r = \text{popcnt}(\text{app}[c][block_r] << (63 - (R \& 63)))
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6 if block_l > 0: occur_l = occ[c][block_l - 1]
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7 | if block_r > 0: occur_r = occ[c][block_r - 1]
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9 L = less[c] + appear_l + occur_l
10 R = less[c] + appear_r + occur_r - 1
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Sampling occ table by storing only every k-th entry provides additional data reduction. Typically k = 4 is suitable. Code does not get more difficult.

Memory

Using the double-stranded DNA again we get a memory usage of app: $|\Sigma| \cdot n \cdot \frac{1}{8}$ bytes ≈ 3.1 Gbytes occ: $|\Sigma| \cdot n \cdot \frac{4}{r \cdot k}$ bytes ≈ 0.4 Gbytes pprox 3.5 Gbytes.

 $|\Sigma| \cdot n \cdot 4$ bytes ≈ 100 Gbytes.

Challenge: Reduce memory usage without increasing runtime.

References

[1] M. Burrows and D. Wheeler. A block-sorting lossless data compression algorithm. Technical Report 124, Digital Equipment Corporation, Palo Alto, CA, 1994.

[2] P. Ferragina and G. Manzini. Opportunistic data structures with applications. In Foundations of Computer Science, 2000. Proceedings. 41st Annual Symposium on, pages 390–398. IEEE, 2000.

[3] H. Li and R. Durbin. Fast and accurate short read alignment with Burrows-Wheeler transform. *Bioinformatics*, 25(14):1754–1760, 2009.