REPORT

Introduction to Focus Areas - Advanced Algorithms - Group 8 "C run ch time : O(T-deadline)"

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Abstract

Goal of the project: Implement and compare different methods for finding exact matches of a pattern within a reference. Analyzation of the overall runtime and memory consumption.

Methods used in the project: Implemented a naive search, binary search, an mlr-heuristic approach and a *longest common prefix*-based approach in C++20.

Main results of the project: A huge improvement in runtime comes with a relatively small cost of memory consumption when choosing a suffixarray based approach over a naive search [O(m*n) vs. $O(log_2(n*m))]$.

Possible improvements: Overload the '< />' operators for customized comparisons of dna5-types. Use a more efficient or better sorted map, so the lookup time remains constant. Find the left and right borders in parallel threads, or restrict the search for the right borders to the intervall from the previous found left border to references' end.

Keywords: C++; Exact matching; MLR; LCP; Benchmark; Weekly homework

1 Introduction

In the field of bionforamtics, pattern matching algorithms are frequently used to search for specific sequences of nucleotides or amino acids within DNA or protein sequences. It is often referred to as matching a query or pattern to a reference. Optimizing these searches is a crucial task to reduce time consumption, required computational power and memory consumption. Unlike methods and approaches that are capable of matching multiple patterns at once, like the data structure of an Aho-Corasick-Tree would allow it, we are focussing on exact, single pattern matching algorithms. One of the most often used approaches is based on a suffix-array or suffix-tree. This specific data structure requires a preparation step to store all possible suffixes of the reference with different length in a lexicographical order and at which positions they appear in the original reference. Even though this takes some time, it allows querying the data structure in a very efficient way.

1.1 Goal

With this report, we wanted to explore the different suffix-based approaches to match single queries of varying length to a reference DNA. To further explore the difference in runtime and memory consumption, we wanted to compare them to a naive and brute-force approach.

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2 Methods

2.1 Data

We were given a file in the compressed fasta.gz format which contained $\sim 101.250.000$ nucleotides of a reference chromosom. We were also given multiple fasta.gz files, each containing 100.000 reads with a different length (40, 60, 80, 100). To load the files directly as vector of dna5 data, which occupies 3 bits per nucleotide, we used the sequence_file_input function of SeqAn3 [1]. All Benchmarks were performed on the same hardware for all different query lengths and several amounts of queries.

2.2 Naive Approach

The brute force approach to find queries within the reference is a straight-forward implementation. Searching for the whole query with exact matches from the beginning of the reference allowed us to search multiple occurances by using the same function with a continuesly adjusted start-parameter, until it reaches the end of the reference. Every position of a hit gets stored in a vector.

2.3 Suffix array based approaches

To construct and sort the suffix array, we used divsufsort by Yuta Mori, which is concisley described in [2]. The suffix array contains the positions of all lexicographically ordered suffixes in regards to the reference. Because of this, all matching hits of a query should be next to each other in the suffix array, pointing to their positions in the reference.

2.3.1 Binary Search

To speed up the search for a query on the reference, a binary search was implemented. It needs an ordered data structure, so the classical binary search is conducted on the previously created and sorted suffix array. The resulting left and right borders of all hits are stored in a struct. Every stored position of the beginning of a suffix in the referenc, that lays within the range of the borders found by the binary search on the suffix array can be easily accessed and printed as all exact matches.

2.3.2 MLR-Heuristic

To speed up the search on the suffix array, the MLR-trick got applied to the binary search. This trick is used to reduce the amount of comparisons of the query with the a suffix, by quickly checking, whether the query is a prefix of the current suffix and how long it is, or not. This allows to skip the comparisons for the length of the prefix. With this, the binary search can be improved.

2.3.3 Longest Common Prefix

To improve the previous method even further, the longest common prefixes (LCP) of all suffixes can be previously computed. In our case, the LCPs are stored in a map, representing a tree. The LCP-tree stores the length of the LCPs between each pair of consecutive suffixes in the suffix array. This will automatically lead to a higher main memory consumption, but should save a lot of comparisons du to the sorted suffix array. This allows to quickly narrow down the search space and return the left and right borders of all hits. The LCP-tree gets computed and stored as a file, so it can be loaded anytime.

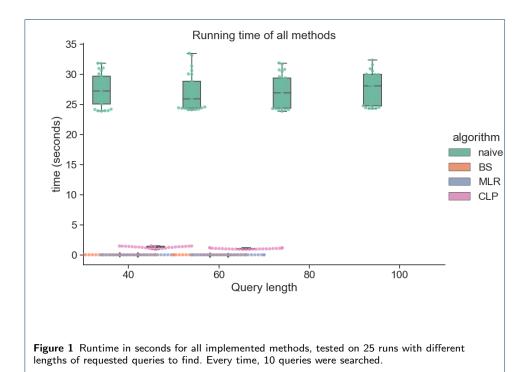
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3 Results

3.1 Runtime Benchmarks

For benchmarking the runtime of our different methods, we performed each method multiple times to be able to compare their average execution times. No additional compiler based optimization-flags were used. Construction of the suffix arrays and LPC-trees are not included in this benchmarking.

Because of the high linear runtime of the naive search, we did not perform benchmarking with more than 10 queries of different lengths, but the difference in speed for all methods is already visible when only searching 10 queries.



We only benchmarked the naive search on all different queriy lengths. As expected, the naive approach is the slowest when compared with query lengths of 40 or 60. The difference in time consumption is only barely visible for longer queries, but this might be due to the nature of the runtime of O(m*n), since m is relatively small in regards to a reference length of 100.000.000.

To compare the binary search with the MLR-Heuristic, we analyes the perfomrance of our implementations with queries of length=100 for a variety of different amounts of queries to be searched for.

It is shown in figure 2 that the MLR algorithm performs generally better than the more simple binary search. It seems like having a runtime of exponential nature.

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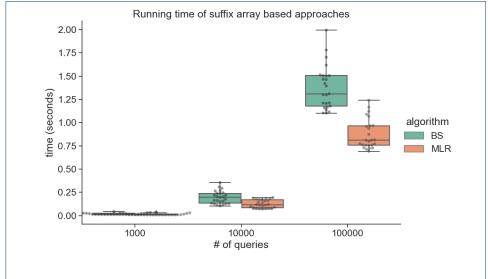


Figure 2 Runtime in seconds for Binary Search and MLR with different amounts of queries with length 100. 25 runs were performed for every method.

When comparing this to the results of the LCP-approach we implemented, it is clear, that there must be a flaw or mistake overseen in our implementation of the search with the LCP-tree.

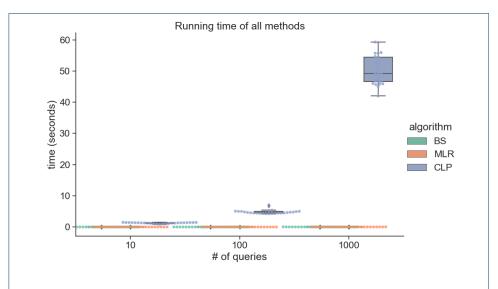


Figure 3 Runtime in seconds for all suffix array based implementations with different amounts of queries with length 40. 25 runs were performed for every method.

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3.2 Memory Consumption

The difference in memory consumption for all these methods is perceived as very straight-forward and can be measured but is also heavily depended on the implementation.

3.2.1 Naive Search

While the naive search only requires the query and reference in memory, the space consumption can be expressed as O(m+n), where m and n are the pattern and reference length.

3.2.2 Binary Search

The binary search of our implementation is performed on a suffix array. Hence, the memory consumption is O(m+2*n). This works well for all our use cases

3.2.3 MLR-Heurisitc

Since the MLR-based approach uses the same underlying functionality as our implementation of the binary search, but implents a trick for less comparisons, the overall memory consumption stays the same.

3.2.4 Longest Common Prefix

The approach with the LCP-tree uses an additional data structure to store the prefix-length of two consecutive entries in the suffix array, which leads to an an additional memory consumption of 130 * n, because the data type (dna5) used by SeqAn3 loads every nucleotide with only 3bit in the main memory, but our size_t uses 64bit and we need multiple values to store this additional information. This is due to the data structure of the LCPs, consisting of a map with pairs, i.e. std::map<std::pair<size_t, size_t>, size_t>. Consequently, we are given a consumption of O(m+134*n). This leads to an impressive consumption of over 12GB in main memory, but storing the LCPs in a file on a hard drive leads to a consumption of 4.5GB.

4 Discussion

While the naive search is the easiest of the proposed methods to understand and implement, it is also the slowest. The savings of memory consumption don't make up for the linear runtime. Our supposedly fastest implementation is by far not our most optimized version and therefore lacks a bit of speed. Also changing the structure of our LCPs could result in a way lower memory consumption. Also, beginning to search for the right boundary from the left boundary on, instead of index zero, could drastically reduce the overall runtime. The Binary search on the other hand occupies a reasonable amount of space, but is a little bit slower than the MLR-based approach, which uses the same amount of space.

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The code and data is available under:

The explanation of how to run the code can be found in a README.md file in the same repository under the paragraph "More Options?".

References

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