

Accelerating DNA Sequence Alignment using Altera DE2-115

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ARTICLE INFO	ABSTRACT
Article history: Received 4 January 2024 Received in revised form 28 August 2024 Accepted 30 August 2024 Available online 1 October 2024 <i>Keywords:</i> FPGA; Smith-Waterman algorithm; DNA	DNA sequence alignment is a technique for discovering information between two base sequences which the Smith-Waterman algorithm is the accurate method that provides a precise result for alignment compared to others. However, the performance was influence by size of dataset and a long DNA base sequence which resulted the time required for the alignment process is much longer in relation to the number of DNA sequence samples. There are many ways to accelerate DNA sequence alignment, and Field Programmable Gate Array (FPGA) is a good choice due to its parallel processing and cost efficiency. Although FPGA acceleration approaches are not new, this work investigates a purely software-based FPGA acceleration using the Altera Cyclone IV EP4CE115F29C7N FPGA as the target device. The SW algorithm was developed using the C language in Quartus II version 18.1 and the Nios II software build tools for Eclipse. The development starts with setting up the QSys architecture before developing the code in Eclipse to determine the computational performance. The result shows the computational timing and speed of the implementation, with the highest speed achieved being 198.76 cells per millisecond. To summarise, the computational performance ultimately depends on the maximum matrix size of the FPGA, which is also
sequence alignment	influenced by the DNA-based pair length and able to complete using low-cost FPGA.

1. Introduction

The human body stored the information of itself in nucleic acid sequences called the human genome. The genome consists of 23 chromosomes, a dense strand of deoxyribonucleic acid or DNA that coded all the information [1]. Four bases made up DNA; adenine(A), cytosine(C), guanine(G), and thymine(T). A human has billions of bases [2,3], and the sequence of the bases made each person different from each other, even from its close relatives. Based on "The 1000 Genomes Project Consortium" [4], the typical difference between two persons estimated at around 0.6% or 20 million base pairs. The variations in human DNA cause different features in a person like tanned skin,

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different eyes, and hair colour. The bases differences can be found by comparing two DNA samples and the process called DNA sequence alignment. In more detail, the DNA sequence alignment is a process of comparing and determining the order of nucleic acid and the bases sequence; adenine, cytosine, guanine, and thymine.

This method is also indispensable in biological research and practical applications such as medical diagnosis, criminal forensics, pharmaceutical development, and virology. In forensics science, it can be used to determine the gender of the criminal and check the similarity between DNA samples. As most of the viruses are small, the virus study can use the approach, and it also allows researchers to predict future virus outbreaks as the viral sequencing can be used by measuring its mutation rate. Other than that, DNA sequencing can provide us with specific antibiotics and reduce the risk of giving antibiotics that are not effective to certain bacteria or bodies [5]. Hence, creating a faster way to align the DNA bases made the applications more effective as it is allowing better discovery of applications, faster research development, and guaranteed precision.

Generally, there are two main methods of DNA sequencing, that is heuristic and dynamic programming as shown in Figure 1. Word or heuristic methods such as FASTA and BLAST produce outcomes that are not guaranteed in optimal alignment, but it is faster than dynamic programming [6–10]. Heuristic methods are suitable to speed up database searching since it is not searching for the details similarity. The dynamic approach is a more precise method of alignment, but the speed decreasing according to the size of data.



Fig. 1. The approaches used for DNA sequence alignment

Over the past few decades, most research for DNA sequence alignment has emphasized using the general computer as the computational medium. In 1969, Saul B. Needleman and Christian D. Wunsch developed an adaptable computer method that finds the similarities in the amino acid sequences between two proteins [11]. In the study, the algorithm proposed applying the dynamic programming to use global alignment, which executes the sequence comparison from end-to-end of the sample. Later, the Smith-Waterman algorithm[12] was introduced as the previous Needleman-Wunsch algorithm alteration, which implements a local alignment and different scoring methods.

Using a general computer for DNA sequence alignment requires a high cost and hinders DNA sequencing technology development due to computational sequence execution issues. As a result, the FPGA should be more suitable for this proposed DNA sequence alignment application due to faster speed [6,13], low cost [6,13[14]-15], and high cost-effectiveness [15]. In [16], the researcher presents the performance comparison between the FPGA, Graphics Processor Units (GPU), and IBM's Cell Broadband Engine (Cell BE) for the Smith-Waterman pairwise sequence alignment algorithm. The result shows that FPGA has outperformed the GPU and Cell BE in performance per dollar and performance per watts. In conclusion FPGA offers a more cost-effective and reliable solution for this design process. Additionally, FPGA technology has been widely adopted [17]-[19].

In achieving the speed and precision of DNA sequence alignment, it is important to improve the speed of this Smith-Waterman algorithm since it already has good precision. Numerous approaches have been proposed to accelerate this Smith-Waterman algorithm using FPGA, including be it in matrix fill-up, systolic array, FPGA & server, sub-function implementation, affine gap or path reconstruction approach; [2,13,15,20-32]. Based on the previous work proposed on FPGA, in this paper, the focus will be more on implementing the Smith-Waterman algorithm on the FPGA and examining the influence of the design over the frequency and processing runtime.

1.1 Smith-Waterman Algorithm

The Smith-Waterman algorithm is known as local sequence alignment, where it is suitable for more divergent sequencing, while Needleman-Wunsch is a global sequence alignment and is more suitable for closely related sequences. However, both of these dynamic sequences are guaranteed to find the optimal local alignment. The Smith-Waterman algorithm is divided into three basics steps that are:1) Initialization, 2) Matrix filling, and 3) Traceback.

1.1.1 Initialization

In the Smith-Waterman algorithm, the maximum match can be determined by creating an editdistance matrix [33] that compares the short sample of query sequences (Q) and target sequences (T) while giving information regarding the similarity of compared sequences. In identifying the pairings of segments with a high degree of similarity, matrix H is set up with (M+1)×(N+1) dimensions. The first steps of the Smith Waterman are initialization, which creates a zero region at H_(0,j) and H_(i,0) as shown in Table 1.

	$H_{i,\theta}$	Q_1	Q_2	••••	Q _M
$H_{ heta,j}$	0	0	0	••••	0
T_1	0				
T_2	0				
T_N	0				

 Table 1

 Initialization of the matrix

The region involved during the initialization process can be summarized by Eq. (1).

 $H_{i,0} = 0; \ 0 \le i \le m$ $H_{0,j} = 0; \ 0 \le j \le n$

(1)

1.1.2 Matrix filling

$$H_{i,j} = max \begin{cases} 0 \\ H_{i-1,j-1} + s_{i,j} \\ H_{i-1,j-d} \\ H_{i,j-1-d} \end{cases}$$
(2)

The matrix filling is the next stage where the remaining cell will be filled. Given the sequences Q and T with the length of the sample is M and N respectively, the matrix indicates the number of changes in terms of substitutions, insertions, and deletions that performed to align Q with T. In the matrix filling, the scoring for the sequence will be filled on $a_{i,j}$, $b_{i,j}$, and the rest by comparing the bases on Q_M and T_N . The score was based on the Q_M and T_N either it is a match or not and the scoring for the column on the upper, left and diagonal upper left, shown in Figure 2. The three separate score is calculated and the maximum score between these columns will be assigned into the cell.

	<i>H</i> _{<i>i</i>,0}	Q_1	Q_2		Q_M
<i>H</i> _{0,j}		0	0	0	0
<i>T</i> ₁	0-	a _{i,j}	b _{i,j}		
<i>T</i> ₂	0	$d_{i,j}$	C _{i,j}		
	0			e _{i,j}	
T_N	0				

Fig. 2. Edit-distance matrix; query (q) with *m* length and target (t) with *n* length

The scoring for match, mismatch and gap is +10, -5, -10 respectively which the negative value become 0 when filled on a new column (Eq. (3) and (4)).

$$Matching Q and T; \begin{cases} (top cell) - 10 \\ (left cell) - 10 \\ (diagonal cell) + 10 \end{cases}$$
(3)
$$Mismatch Q and T; \begin{cases} (top cell) - 10 \\ (left cell) - 10 \\ (diagonal cell) - 5 \end{cases}$$
(4)

1.1.3 Traceback

The traceback is the final step, and it starts after all the cell is being filled. The trace will begin from the highest value of score gain in the matrix table to form an optimal sequence alignment path and end when the similarity score drops. Table 2 shows the traceback pathway in grey.

Table 2

Traceback pathway in grey					
	$H_{i,\theta}$	G	A	Т	С
H _{0,j}	0	0	0	0	0
Т	0	0	0	10	0
Α	0	0	10	0	5
Т	0	0	0	20	10
С	0	0	0	10	30

In Table 2, the Q sequences (GATC) and T sequences (TATC) are being aligned, and the matrix has been filled with the scoring. This traceback step starts at the highest score cell, usually located at the back of the matrix. Later, it moves to the next highest neighboring cell to form an optimal sequence alignment as shown in Figure 3.

ATC	
ATC	

Fig. 3. The optimal sequence alignment that been formed from Q sequences (GATC) and T sequences (TATC)

2. Methodology

The SW algorithm was developed using C language in Quartus II version 18.1, and Nios II Software Build Tools for Eclipse, targeted to Altera Cyclone IV EP4CE115F29C7N FPGA. The results for the timing of the program execution will be compiled as the result of this project. Figure 4 shows in this implementation, the project is created in Quartus Prime 18.1 version, which is used to build an embedded sub-system in SOPC builder. The function of this builder is to integrate the sub-system in the FPGA to become one complete system according to the user's need. The SOPC was created using the Platform Designer, and the Qsys file created interconnected logic will be used to program the FPGA. The Nios II software was later used to run the C codes that have been developed with the input of the DNA sequence data in the text file.

In reading the text file on Nios II, the BSP setting needs to be setting up by including the software packages for HAL's (Hardware Abstraction Layer). This software package provides generic device models for classes of peripherals command in embedded systems like timers and I/O peripherals that allow the user to use specific C command to open and read the file while mounting the text file with the program.

The code reads the text files for the query-sequence and target sequence by character and stores them in arrays. The matrix dimensions are set up, and the initialization step will take up by inserting the value of zero into the first matrix column and row. The other cell will be filled during the matrix

filling process using Eq. (3) and (4) depends on its condition, match, or mismatch. The traceback later will be performed to find the highest scoring matrix from the cells.



Fig. 4. Block diagram flow of SW implementation

3. Results

In examining the effect of the clock on the runtime speed, the clock speeds 50MHz, 100MHz, and 200MHz are used to aligning two DNA sequences with identical lengths from 1 to 128 DNA sequence characters. The operating system of EP4CE115F29C7N has been compile using Nios II, with the data from the text file is mounted into the program. Each runtime result of the implementation was recorded and summarised in Table 3.

Table 3					
The computation	The computational time (ticks) for s-w implementation				
	Runtime (ticks)				
Numbers of Cells	50 MHz	100 MHz	200 MHz		
	(50M ticks/s)	(100M ticks/s)	(200M ticks/s)		
1	5 227	5 189	5 212		
4	6 591	6 384	6 395		
8	14 119	13 862	13 211		
64	64 557	64 320	64 401		
256	417 875	418 669	418 522		
1024	3 209 344	3 253 244	3 252 114		
4096	26 314 309	26 373 011	26 372 481		
1638/	211 270 /27	217 547 260	217 554 241		

The timestamp that is obtained from the Nios II operation is in ticks and ticks per second. To change the value of data into second, Eq. (5) is use.

Runtime (ms) = $\frac{Ticks (runtime)}{Ticks per second} x 1 second$

(5)

Table 4 shows the entire computational time for the S-W implementation in a millisecond (ms) interval.

Table 4				
The computational time for s-w implementation				
Number of Cells	Runtime (ms)			
	50 MHz	100 MHz	200 MHz	
1	0.105	0.052	0.026	
4	0.132	0.064	0.032	
8	0.282	0.139	0.066	
64	1.291	0.643	0.322	
256	8.358	4.187	2.093	
1024	64.19	32.53	16.26	
4096	526.28	263.73	131.86	
16384	4225.59	2175.47	1087.77	

As highlighted in Table 4, the runtime is increasing proportionally to the number of cells, which the number of cells can be determined based on m+1 and n+1 dimensions. For example, when the Q sequences and T sequences have 64 bases characters, the dimension of the matrix is (64+1) x (64+1) = 4226. The additional rows are the row that formed during the initialization step.

Figure 5 shows the runtime versus the number of cell graph. Based on the results, when the clock speed is doubled, the runtime for each alignment is 50% less compared to the previous speed. As tabulated in Table 4, when the clock speed increases from 100 MHz to 200MHz for the 4 cells, the runtime decreases by half from 0.064ms to 0.032ms.



Fig. 5. Runtime versus the number of cell graph

In Figure 5, the line graph shows the runtime for each Smith-Waterman alignment that varies for each number of cells and clock speed. This graph was plotted based on the result recorded in Table 4. This graph clearly shows that the increment of the runtime is proportional to the increment number of cells. Meanwhile, the size of the memory been shown in Table 5.

Table 5				
The size of memory for every implementation				
Number of Cells Memory (kB)				
1	77			
4	77			
8	77			
64	77			
256	78			
1024	82			
4096	93			
16384	178			

The increase in the runtime is cause by the increase size of the memory for the program after the compile process. This shows that the increase of memory for the DNA sequence alignment, increase the runtime for the S-W implementation. Figure 6 shows the speed that each alignment obtained from the time taken to align a cell.



Fig. 6. Speed of each SW alignment completion

Moreover, the peak runtime speed was recorded when aligning 64 number cells compared to the rest with the runtime speed per cell trend increasing for single cells until 64 cells before decreasing after that. The 200MHz clock speed has recorded the fastest average speed compared to the others with the highest during 64 cells with 198.76 cells per milliseconds.

4. Conclusions

This paper presents the DNA sequence alignment accelerator using Quartus II version 18.1, and Nios II Software Build Tools for Eclipse, targeted to Altera Cyclone IV EP4CE115F29C7N Field Programmable Gate Array (FPGA) as to overcome the performance and computational complexity issue. Twenty-four series of tests comprising different DNA sequence-based pair data ranging from single cells up to 16384 cells and three different computational clock speeds have been used to measure the implementation performance. The highest speed recorded is during the 200MHz clock speed with 64 cells, which is 198.76 cells per milliseconds. Not only the computational clock speed but the number of memories also influenced the overall computational runtime. Finally, it can be

concluded that the computational performance ultimately depends on the DNA-based pair's length and computational clock speed.

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