Seminar in Computer Architecture

Meeting 2: GateKeeper

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ETH Zurich
Fall 2021
30 September 2021





Example Paper Presentation I

Let's Review This Paper [Alser+, Bioinformatics 2017]

Mohammed Alser, Hasan Hassan, Hongyi Xin, Oguz Ergin, Onur Mutlu, and Can Alkan

"GateKeeper: A New Hardware Architecture for Accelerating Pre-Alignment in

DNA Short Read Mapping"

Bioinformatics, [published online, May 31], 2017.

Source Code

Online link at Bioinformatics Journal

Bioinformatics



Article Navigation

GateKeeper: a new hardware architecture for accelerating pre-alignment in DNA short read mapping •

Mohammed Alser ™, Hasan Hassan, Hongyi Xin, Oğuz Ergin, Onur Mutlu ™, Can Alkan ™

Bioinformatics, Volume 33, Issue 21, 01 November 2017, Pages 3355-3363,

https://doi.org/10.1093/bioinformatics/btx342

Published: 31 May 2017 Article history ▼

GateKeeper: A New Hardware Architecture for Accelerating Pre-Alignment in DNA Short Read Mapping

Mohammed Alser, Hasan Hassan, Hongyi Xin, Oğuz Ergin, Onur Mutlu, Can Alkan Bioinformatics, 2017

Presented by: Mohammed Alser







Executive Summary

- Problem: Genomic similarity measurement is a computational bottleneck. Examining the similarity of highly-dissimilar genomic sequences consumes an overwhelming majority of a modern read mapper's execution time.
- Goal: Develop a fast and effective filter that can detect highly-dissimilar genomic sequences and eliminate them before invoking computationally costly alignment algorithms.
- <u>Key observation</u>: If two strings differ by *E* edits, then every pairwise match can be aligned in at most 2E shifts.

Key ideas:

- Quickly find similar sequences using Hamming Distance.
- Compute "Shifted Hamming Distance" for the rest of sequence pairs: ANDing 2E+1 Hamming vectors of two strings, to identify dissimilar sequences.
- Use only bit-parallel operations that nicely map to:
 - SIMD instructions, FPGA, Logic layer of the 3D-stacked memory, and In-memory accelerators (e.g., Ambit)

Key results:

 Provides a huge speedup of up to 130x compared to the previous state of the art software solution.

Background, Problem, & Goal

What is Genome Analysis?



Genomic analysis





Genomic analysis is the identification, measurement or comparison of genomic features such as DNA sequence, structural variation, gene expression, or regulatory and functional element annotation at a genomic scale. Methods for genomic analysis typically require high-throughput sequencing or microarray hybridization and bioinformatics.

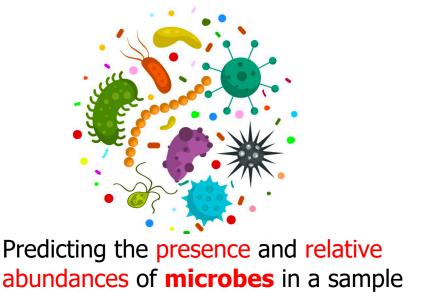
Applications of Genome Analysis



Understanding **genetic variations**



Rapid surveillance of disease outbreaks





Developing **personalized medicine**

How to Analyze a Genome?



NO

machine gives the complete sequence of genome as output



DNA Testing



Fall DNA special Just 55 CHF 89-CHF



The promotion ends today in 12 more hours!



https://www.myheritage.ch/dna



DNA Testing



Fall DNA special Just 55 CHF 89-CH



https://www.myheritage.ch/dna https://www.23andme.com/



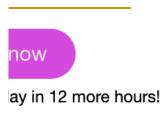
Health + Ancestry
Service

\$199

Includes everything in Ancestry +
 Traits Service

PLUS

- 10+ Health Predisposition reports*
- 5+ Wellness reports
- 40+ Carrier Status reports*





High-Throughput Sequencers



Illumina MiSeq



Illumina NovaSeq 6000



Pacific Biosciences Sequel II



Pacific Biosciences RS II





Oxford Nanopore MinION



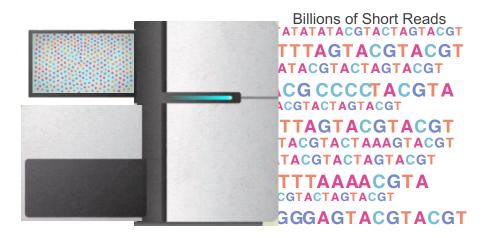
... and more! All produce data with different properties.

Genome Sequencer is a Chopper

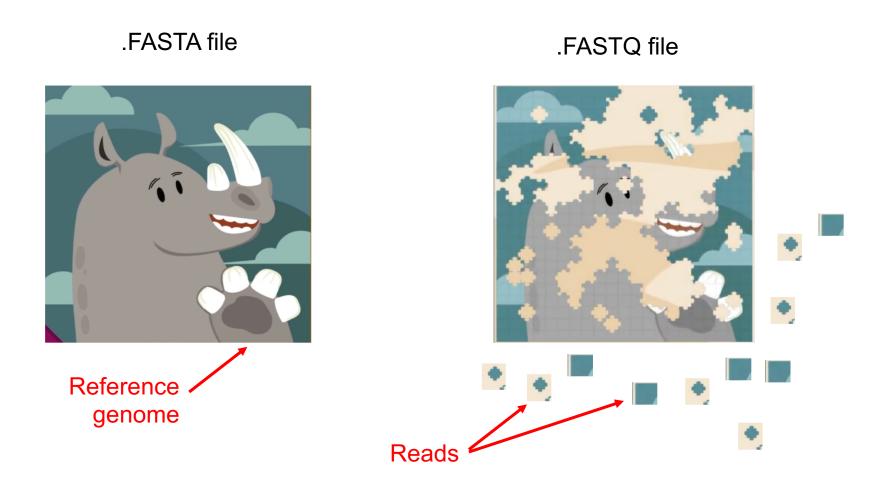
Regardless the sequencing machine,

reads still lack information about their order and location

(which part of genome they are originated from)



Solving the Puzzle



https://www.pacb.com/smrt-science/smrt-sequencing/hifi-reads-for-highly-accurate-long-read-sequencing/

Reference Genome

.FASTA file:

Obtaining the Human Reference Genome

GRCh38.p13

- Description: Genome Reference Consortium Human Build 38 patch release 13 (GRCh38.p13)
- Organism name: <u>Homo sapiens (human)</u>
- Date: 2019/02/28
- **3,099,706,404** bases
- Compressed .fna file (964.9 MB)
- https://www.ncbi.nlm.nih.gov/assembly/GCF_000001405.39

. . . .

Genomic Reads

.FASTQ file:

Obtaining .FASTQ Files

https://www.ncbi.nlm.nih.gov/sra/ERR240727



ERX215261: Whole Genome Sequencing of human TSI NA20754

1 ILLUMINA (Illumina HiSeq 2000) run: 4.1M spots, 818.7M bases, 387.2Mb downloads

Design: Illumina sequencing of library 6511095, constructed from sample accession SRS001721 for study accession SRP000540. This is part of an Illumina multiplexed sequencing run (9340 1). This submission includes reads tagged with the sequence TTAGGCAT.

Submitted by: The Wellcome Trust Sanger Institute (SC)

Study: Whole genome sequencing of (TSI) Toscani in Italia HapMap population

PRJNA33847 • SRP000540 • All experiments • All runs

Sample: Coriell GM20754

SAMN00001273 • SRS001721 • All experiments • All runs

Organism: Homo sapiens

Library:

Name: 6511095

Instrument: Illumina HiSeq 2000

Strategy: WGS Source: GENOMIC Selection: RANDOM Layout: PAIRED

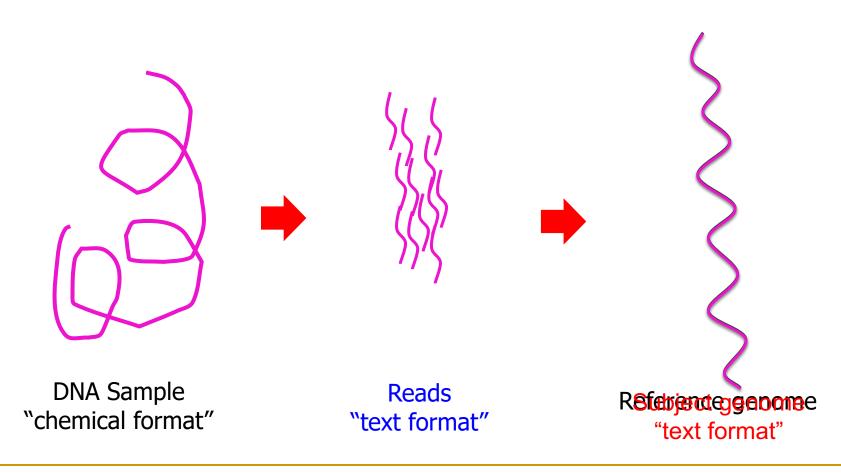
Construction protocol: Standard

Runs: 1 run, 4.1M spots, 818.7M bases, 387.2Mb

Run	# of Spots	# of Bases	Size	Published	
ERR240727	4,093,747	818.7M	387.2Mb	2013-03-22	

Read Mapping

Map reads to a known reference genome with some minor differences allowed



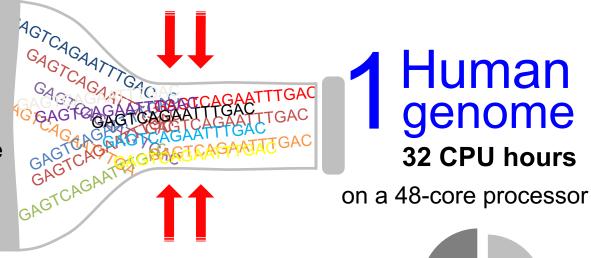
Analysis is Bottlenecked in Read Mapping!!

48 Human whole genomes

at 30 × coverage

in about 2 days

Illumina NovaSeq 6000



71%

■ Read Mapping ■ Others

What makes read mapping a bottleneck?

Let's first learn how to map a read

Matching Each Read with Reference Genome

.FASTA file:

```
>NG 008679.1:5001-38170 Homo sapiens paired box 6 (PAX6)
          TCATTGACATTTAAACTCTGGGGCAGG'
ACCCI
                                      GAACGCGGCTGTCAGATCT
CCTCCGCTCCCAGGTAACCGCC( CCCCGGCCCGGCTCGGGGCCCGCGGGGCCTCTCCGCTG
{\sf GAGGCATACAAAGATGGAAGCGAGTTACTGAGGGAGGGATAGGAAGGGGGGTGGAGGAGGGACTT}
TGCCGAGTGT
             CAAAAGTAGCA
                       CTCCTA
GAGCTGGGAGTAGGGGGGGGGGTCTGCTGCTGCTGCTGCTAAAGCCACTCGCGACCGCGAAAAATGCA
GGAGGTGGGGACGCACTTTGCATCCAGACCTCCTCTGCATCGCAGTTC
TCCGTACCCGCGCCT
           AAAGACACCCTGCCGCGGGTCGGGCGAGGTGCAGCAGAAGTTTCCC
GCGGTTGCAAAGTGCAGATGGCTGGACCGCAACAAAGTCTAGAGATGGGGTTCGTTTCTCAGAAAGACGC
```

.FASTQ file:

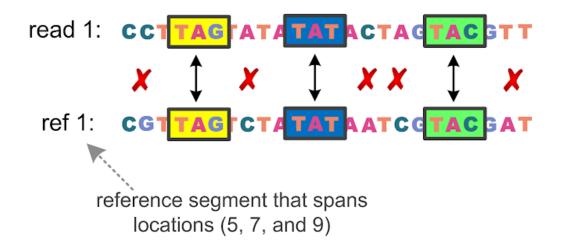
```
@HWI-EAS209_0006_FC706VJ:5:58:5894:21141#ATCACG/1

TAGATN NNNNNNNNTAG

+HWI-EAS209_0006_FC706VJ:5:58:5894:21141#ATCACG/1

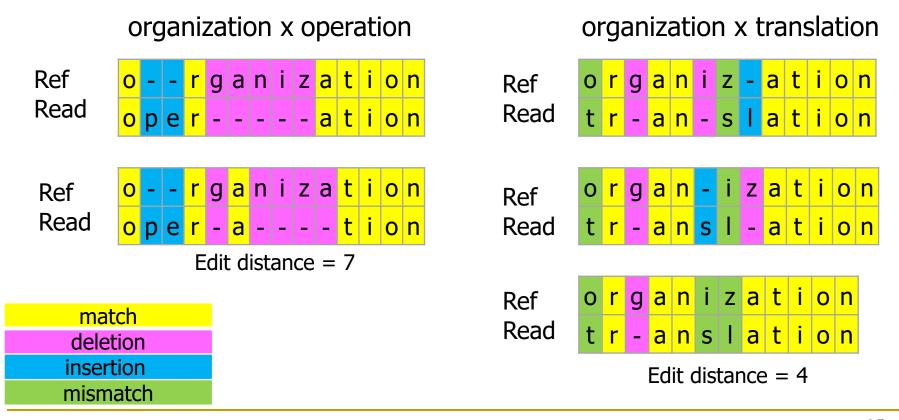
efcfffffcfeefffcfffffddf`feed]`]_Ba_^__[YBBBBBBBBBBTT
```

Base-by-Base Comparison



Sequence Alignment (Verification)

 Edit distance is defined as the minimum number of edits (i.e. insertions, deletions, or substitutions) needed to make the read exactly matches the reference segment.



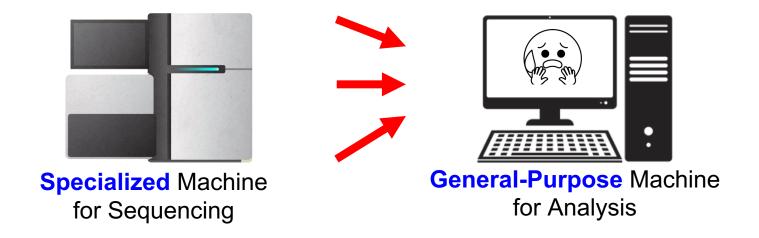
What makes read mapping a bottleneck?

A Tsunami of Sequencing Data

A Tera-scale increase in sequencing production in the past 25 years							
Genes & Operons	1990	Kilo = 1,000					
Bacterial genomes	1995	Mega = 1,000,000					
Human genome	2000	Giga = 1,000,000,000					
Human microbiome	2005	Tera = 1,000,000,000,000					
50K Microbiomes	2015	Peta = 1,000,000,000,000,000					
what is expected for the next 15 years ? (a Giga?)							
200K Microbiomes	2020	Exa = 1,000,000,000,000,000,000					
1M Microbiomes	2025	Zetta = 1,000,000,000,000,000,000,000					
Earth Microbiome	2030	Yotta = 1,000,000,000,000,000,000,000,000					

Source: <a>@kyrpides

Lack of Specialized Compute Capability

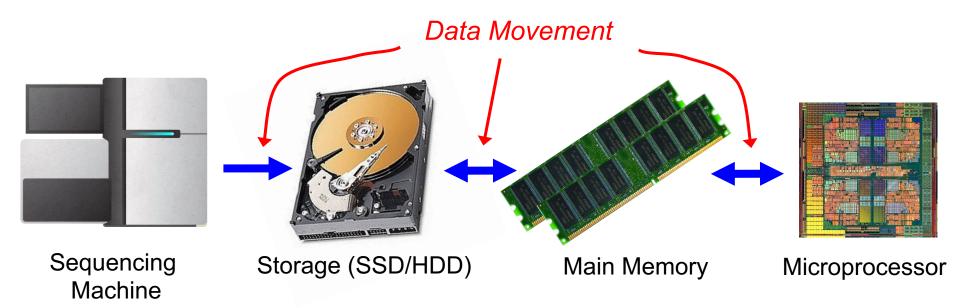


FAST

SLOW

Data Movement Dominates Performance

Data movement dominates performance and is a major system energy bottleneck (accounting for 40%-62%)



Single memory request consumes >160x-800x more energy compared to performing an addition operation

^{*} Boroumand et al., "Google Workloads for Consumer Devices: Mitigating Data Movement Bottlenecks," ASPLOS 2018

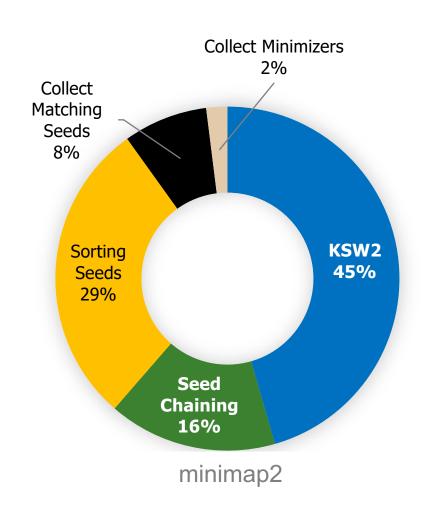
^{*}Kestor et al., "Quantifying the Energy Cost of Data Movement in Scientific Applications," IISWC 2013

[♠] Pandiyan and Wu, "Quantifying the energy cost of data movement for emerging smart phone workloads on mobile platforms," IISWC 2014

Read Mapping Execution Time

>60%

of the read mapper's execution time is spent in sequence alignment



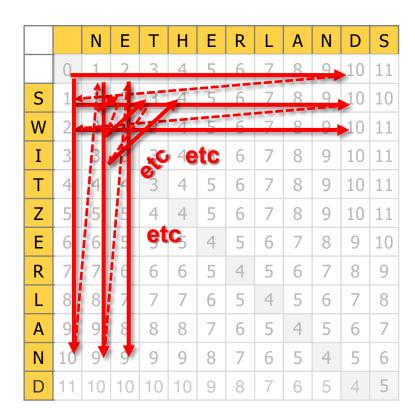
ONT FASTQ size: 103MB (151 reads), Mean length: 356,403 bp, std: 173,168 bp, longest length: 817,917 bp

Why Alignment is Computationally Costly?

Quadratic-time dynamicprogramming algorithm WHY?!

Enumerating all possible prefixes

- NETHERLANDS x SWITZERLAND
- NETHERLANDS x S
 - **NETHERLANDS x SW**
 - **NETHERLANDS x SWI**
 - **NETHERLANDS x SWIT**
 - **NETHERLANDS x SWITZ**
 - **NETHERLANDS x SWITZE**
 - **NETHERLANDS x SWITZER**
 - NETHERI ANDS x SWITZERI
 - **NETHERLANDS x SWITZERLA**
 - **NETHERLANDS x SWITZERLAN**
 - NETHERLANDS x SWITZERLAND



Why Alignment is Computationally Costly?

 Quadratic-time dynamicprogramming algorithm

Enumerating all possible prefixes

 Data dependencies limit the computation parallelism

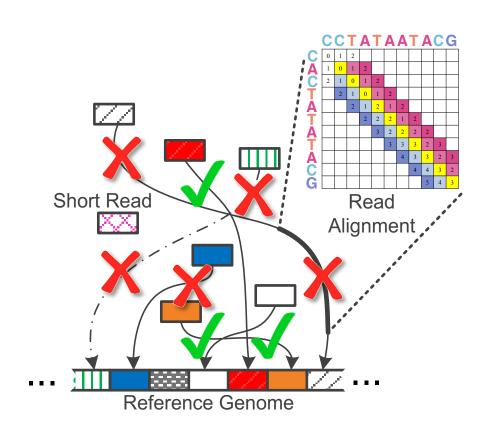
Processing row (or column) after another

 Entire matrix is computed even though strings can be dissimilar.

		N	Е	Т	Н	Е	R	L	Α	N	D	S
	0	1	2	3	4	5	6	7	8	9	10	11
S	1	1	2	3	4	5	6	7	8	9	10	10
W	2	2	2	3	4	5	6	7	8	9	10	11
I	3	3	3	3	4	5	6	7	8	9	10	11
Т	4	4	4	3	4	5	6	7	8	9	10	11
Z	5	5	5	4	4	5	6	7	8	9	10	11
			_					_	_	_		
E	6	6	5	5	5	4	5	6	7	8	9	10
R	7	7	6	6	6	5	4	5	6	7	8	9
L	8	8	7	7	7	6	5	4	5	6	7	8
Α	9	9	8	8	8	7	6	5	4	5	6	7
N	10	9	9	9	9	8	7	6	5	4	5	6
D	11	10	10	10	10	9	8	7	6	5	4	5

Number of differences is computed only at the backtracking step.

Large Search Space for Mapping Location



98%
of candidate locations
have high dissimilarity
with a given read

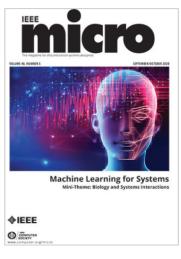
Cheng et al, BMC bioinformatics (2015) Xin et al, BMC genomics (2013)

We need intelligent algorithms and intelligent architectures that handle data well

Detailed Analysis of Tackling the Bottleneck

Mohammed Alser, Zülal Bingöl, Damla Senol Cali, Jeremie Kim, Saugata Ghose, Can Alkan, Onur Mutlu

"Accelerating Genome Analysis: A Primer on an Ongoing Journey" IEEE Micro, August 2020.





Home / Magazines / IEEE Micro / 2020.05

IEEE Micro

Accelerating Genome Analysis: A Primer on an Ongoing Journey

Sept.-Oct. 2020, pp. 65-75, vol. 40

DOI Bookmark: 10.1109/MM.2020.3013728

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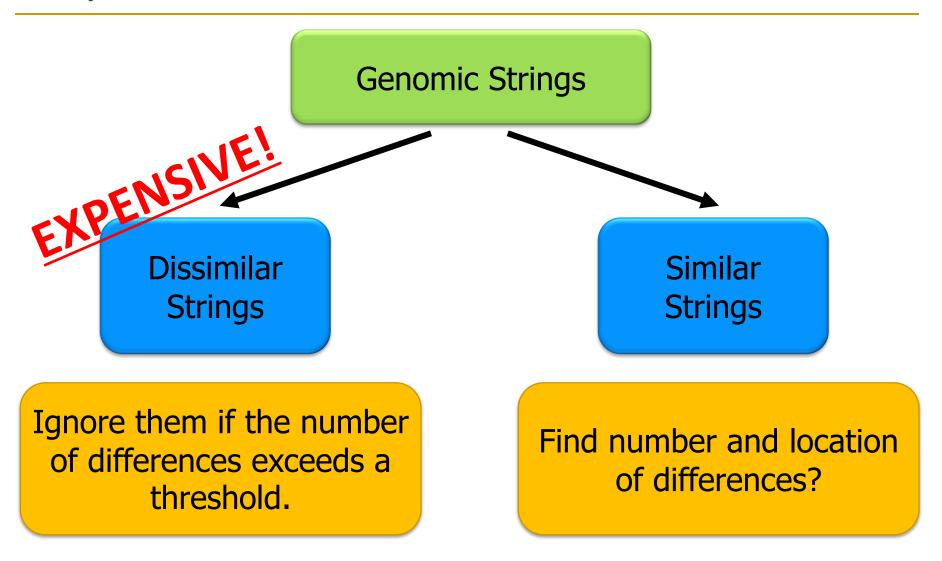
Goal: Minimizing Alignment Time

Sequence Alignment is expensive

Our goal is to accelerate read mapping by reducing the need for dynamic programming algorithms

Novelty, Key Approach, and Ideas

Key Idea

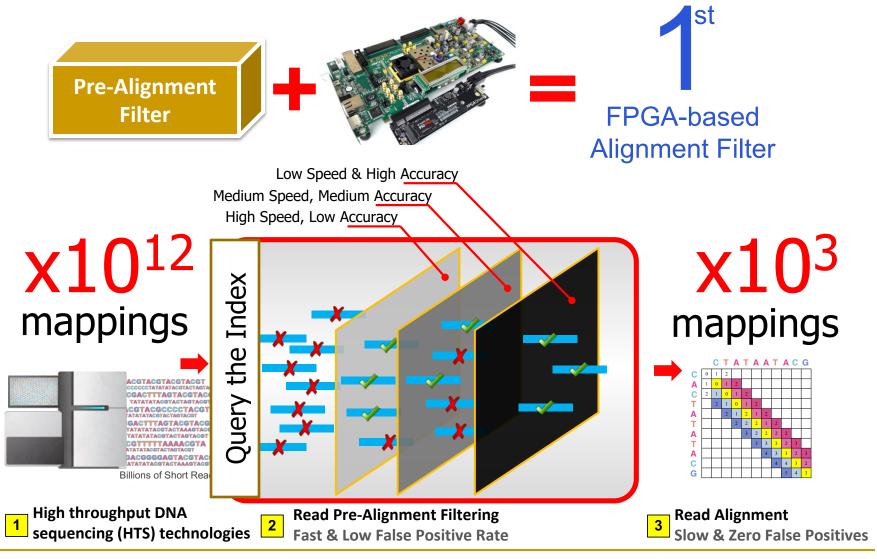


Ideal Filtering Algorithm

Step 2 Step 3 Query Read the Alignment Index

- 1. Filter out most of incorrect mappings.
- 2. Preserve all correct mappings.
- 3. Do it quickly.

Proposed Solution: GateKeeper



Mechanisms (in some detail)

GateKeeper

Key observation:

 If two strings differ by E edits, then every pairwise match can be aligned in at most 2E shifts.

Key ideas:

- Quickly find similar sequences using Hamming Distance.
- Compute "Shifted Hamming Distance": AND of 2E+1 Hamming vectors of two strings, to identify invalid mappings
- Use only bit-parallel operations that nicely map to:
 - SIMD instructions
 - FPGA
 - Logic layer of the 3D-stacked memory
 - In-memory accelerators (e.g., Ambit)

Mechanisms

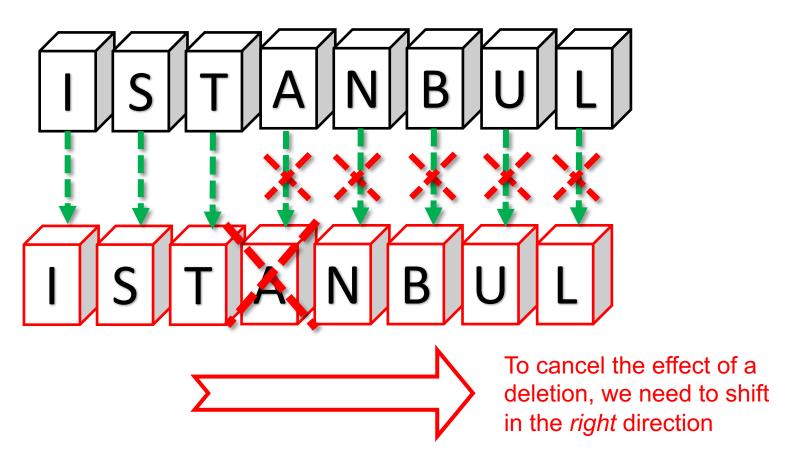
Key observation:

 If two strings differ by E edits, then every pairwise match can be aligned in at most 2E shifts.

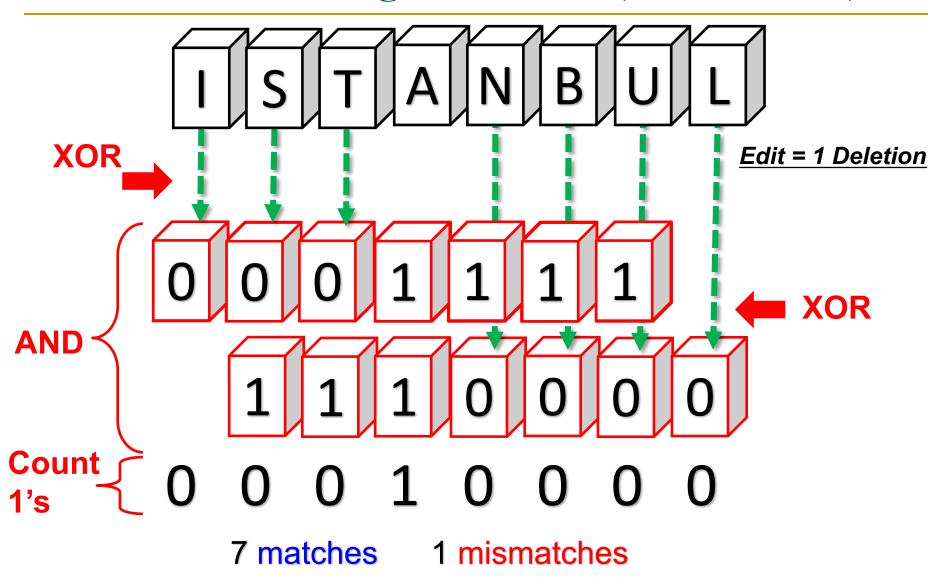
Hamming Distance $(\Sigma \oplus)$

3 matches 5 mismatches

Edit = 1 Deletion



Shifted Hamming Distance (Xin+ 2015)



Mechanisms

Key observation:

 If two strings differ by E edits, then every pairwise match can be aligned in at most 2E shifts.

Key ideas:

- Quickly find similar sequences using Hamming Distance.
- Compute "Shifted Hamming Distance": AND of 2E+1 Hamming vectors of two strings, to identify invalid mappings

GateKeeper Walkthrough

Generate 2E+1 masks

Amend random zeros: $101 \rightarrow 111 \& 1001 \rightarrow 1111$

AND all masks, ACCEPT iff number of `1' ≤ Threshold

```
:GAGAGAGATATTTAGTGTTGCAGCACTACAACACAAAAGAGGGCCAACTTACGTGTCTAAAAAGGGGGAACATTGTTGGGCCGGA
 Reference
    *GAGAGAGATAGTTAGTGTTGCAGCCACTACAACACAAAAGAGGACCAACTTACGTGTCTAAAAGGGGAGACATTGTTGGGCCGG
 000
                          110
 Our goal to track the diagonally consecutive matches in the
                          .111
                          110
         neighborhood map.
                          100
                          000
3-Ir
                          0000
```

Needleman-Wunsch Alignment

GateKeeper

Key observation:

 If two strings differ by *E* edits, then every pairwise match can be aligned in at most 2*E* shifts.

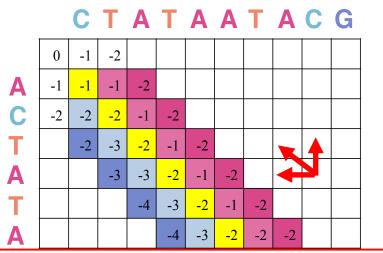
Key ideas:

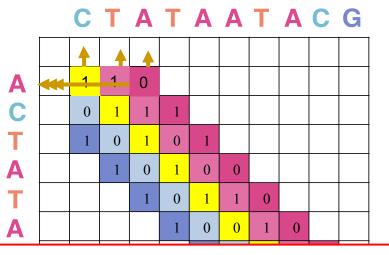
- Quickly find similar sequences using Hamming Distance.
- Compute "Shifted Hamming Distance": AND of 2E+1 Hamming vectors of two strings, to identify invalid mappings
- Use only bit-parallel operations that nicely map to:
 - SIMD instructions
 - FPGA
 - Logic layer of the 3D-stacked memory
 - In-memory accelerators (e.g., Ambit)

Alignment Matrix vs. Neighborhood Map

Needleman-Wunsch

Neighborhood Map



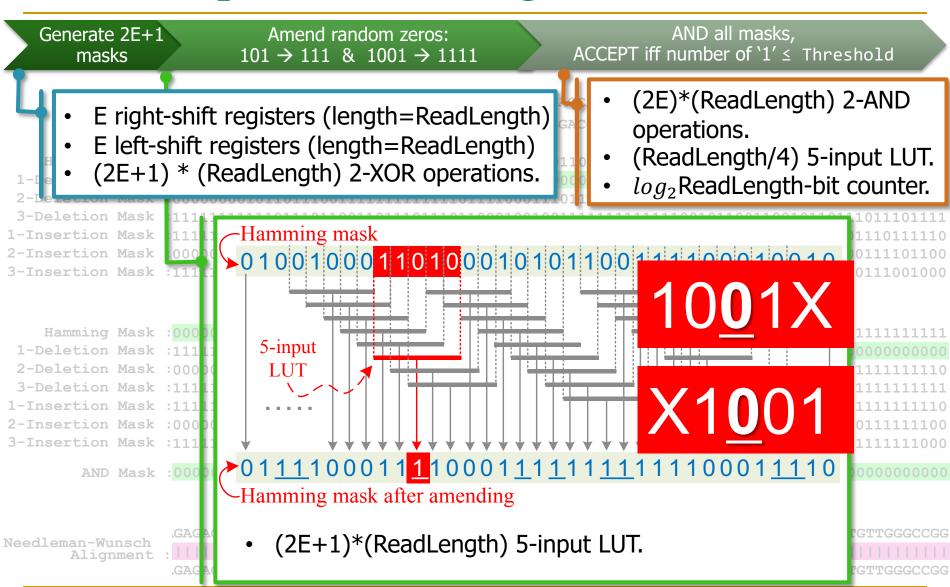


Independent vectors can be processed in parallel using hardware technologies

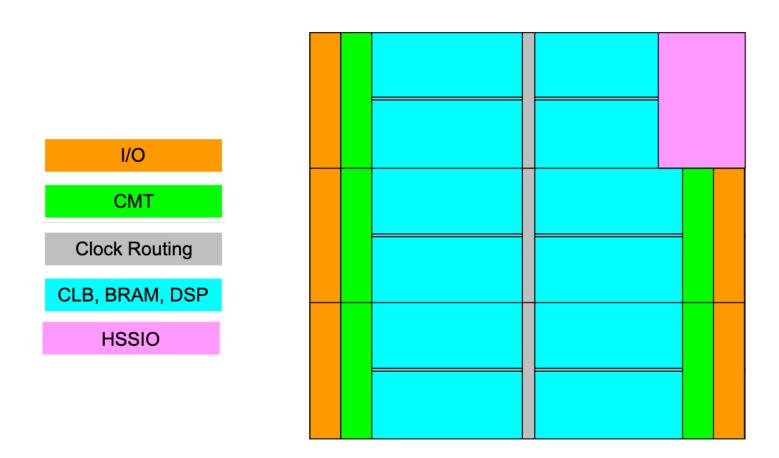


Hardware Architecture

GateKeeper Walkthrough (cont'd)



Virtex-7 FPGA Layout



Configurable logic blocks (CLBs) are the main logic resources for implementing sequential as well as combinatorial circuits

Virtex-7 FPGA Layout

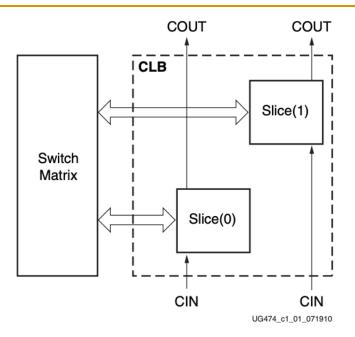


Figure 1-1: Arrangement of Slices within the CLB

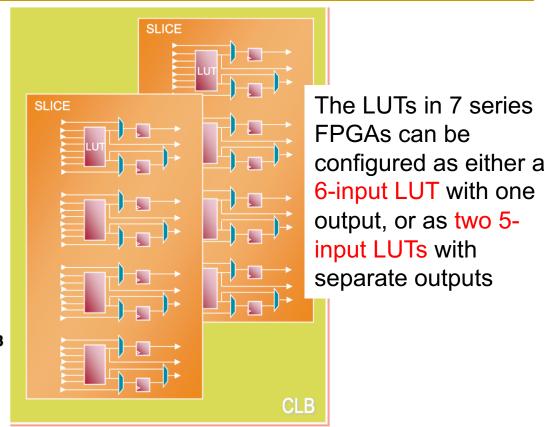


Table 2-1: Logic Resources in One CLB

Slices	LUTs	Flip-Flops	Arithmetic and Carry Chains	Distributed RAM ⁽¹⁾	Shift Registers ⁽¹⁾	
2	8	16	2	256 bits	128 bits	

Key Results: Methodology and Evaluation

Methodology

- System setup:
 - 3.6 GHz Intel i7-3820 (supports only PCIe 2.0)
 - □ Xilinx VC709 (~\$5000)
 - Architecture implementation using Vivado 2014.4 in Verilog
 - RIFFA 2.2 to perform Host-FPGA PCIe communication



- Evaluated dataset:
 - Real sequencing read set (ERR240727_1.fastq)
 - Five simulated read sets of 100 bp and 300 bp long Illuminalike reads with different type and number of edits.

Prior Work on Pre-Alignment Filtering

- Adjacency Filter (BMC Genomics, 2013)
 - Slow
 - Accepts a large number of dissimilar sequences.
- Shifted Hamming Distance (SHD) (Bioinformatics, 2015)
 - It requires the same execution time as the Adjacency Filter
 - It accepts 4X fewer dissimilar sequences compared to the Adjacency Filter.
 - It suffers from a limited sequence length (≤ 128 bp)

VC709 Resource Utilization

Theoretically:

- Up to 140 GateKeeper Processing cores on a single FPGA (E=5, 100bp)
- BUT bottlenecked by PCIe bandwidth
- Small area allows integration into FPGAs already inside of sequencers

Table 2. FPGA resource utilization for a single GateKeeper core

	Resource utilization %						
Read length	100 bp		300 bp				
Edit distance	2	5	2	5	15		
Slice LUT ^a Slice Register ^b	0.39% 0.01%	0.71% 0.01%	1.27% 0.01%	2.2% 0.01%	4.82% 0.01%		

^aLUT: look-up tables.

^bFlip-flop.

VC709 Resource Utilization

Experimentally:

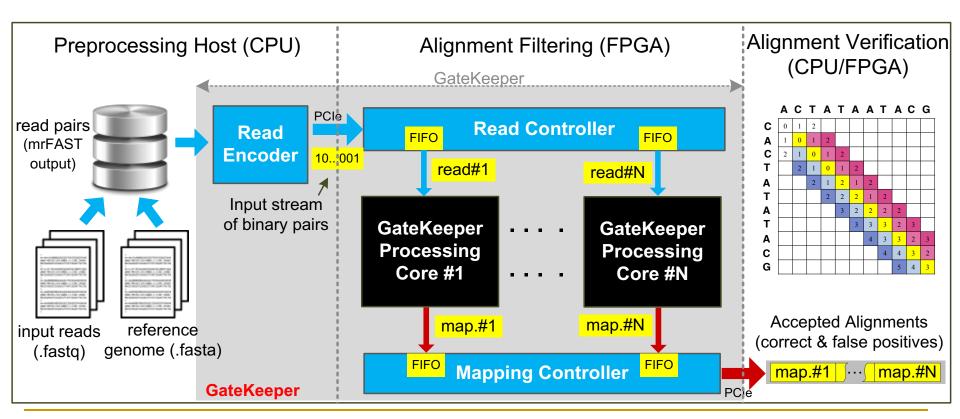
 GateKeeper aligns each read against up to 8 and 16 different reference segments in parallel, without violating the timing constraints for a sequence lengths of 300 and 100 bp, respectively.

Table 3. Overall system resource utilization under different read lengths and edit distance thresholds

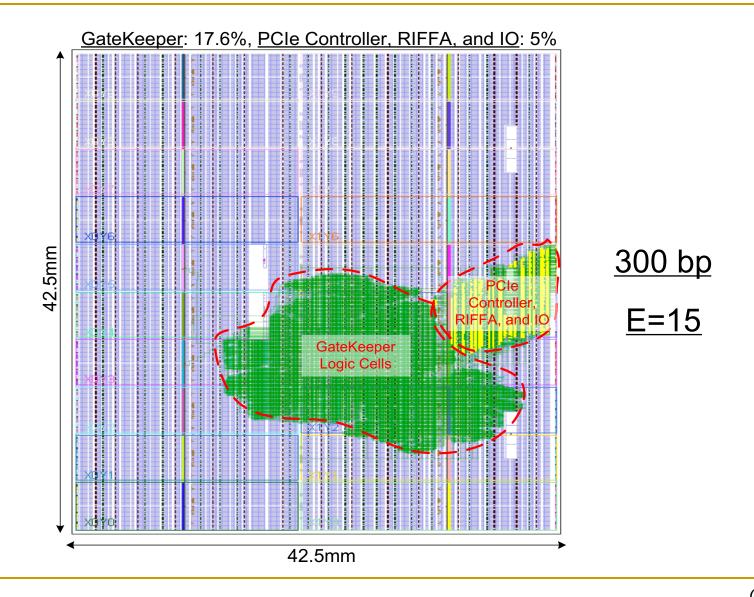
	Resource utilization %					
Read length	100 bp 16 GateKe	eper cores	300 bp <u>8 GateKee</u>	300 bp 8 GateKeeper cores		
Edit distance	2	5	2	15		
Slice LUT	32%	45%	50%	69%		
Slice register	2%	2%	17%	91%		
Block memory	2%	2%	2%	2%		

GateKeeper Accelerator Architecture

- Maximum data throughput $= \sim 13.3$ billion bases/sec
- Can examine 8 (300 bp) or 16 (100 bp) mappings concurrently at 250 MHz
- Occupies 50% (100 bp) to 91% (300 bp) of the FPGA slice LUTs and registers



FPGA Chip Layout



Speed & Accuracy Results

90x-130x faster

than SHD (Xin et al., 2015) and the Adjacency Filter (Xin et al., 2013).

Accepts 4x fewer dissimilar strings

than the Adjacency Filter (Xin et al., 2013).

10x speedup

with the addition of GateKeeper to the mrFAST mapper (Alkan et al., 2009).

Freely available online

github.com/BilkentCompGen/GateKeeper

Summary

GateKeeper Conclusions

- There is a significant performance gap between highthroughput DNA sequencers and read mapper
- Sequence alignment is computationally expensive and unavoidable
- GateKeeper is the first hardware accelerator architecture (as a pre-alignment filter) for quickly rejecting dissimilar sequences
- It provides a huge speedup of up to 130x compared to the previous state of the art software solution.

GateKeeper Conclusions

- FPGA-based pre-alignment filtering greatly speeds up read mapping
 - 10x speedup of a state-of-the-art mapper (mrFAST)

- FPGA-based pre-alignment can be integrated with the sequencer
 - It can help to hide the complexity and details of the FPGA
 - Enables real-time filtering while sequencing

More on SHD (SIMD Implementation)

- Download and test for yourself
- https://github.com/CMU-SAFARI/Shifted-Hamming-Distance

Bioinformatics, 31(10), 2015, 1553–1560 doi: 10.1093/bioinformatics/btu856 Advance Access Publication Date: 10 January 2015

Original Paper



Sequence analysis

Shifted Hamming distance: a fast and accurate SIMD-friendly filter to accelerate alignment verification in read mapping

Hongyi Xin^{1,*}, John Greth², John Emmons², Gennady Pekhimenko¹, Carl Kingsford³, Can Alkan^{4,*} and Onur Mutlu^{2,*}

More on GateKeeper

Download and test for yourself
 https://github.com/BilkentCompGen/GateKeeper

Bioinformatics



Article Navigation

GateKeeper: a new hardware architecture for accelerating pre-alignment in DNA short read mapping •

Mohammed Alser ➡, Hasan Hassan, Hongyi Xin, Oğuz Ergin, Onur Mutlu ➡, Can Alkan ➡

Bioinformatics, Volume 33, Issue 21, 01 November 2017, Pages 3355–3363,

https://doi.org/10.1093/bioinformatics/btx342

Published: 31 May 2017 Article history ▼

Alser+, "GateKeeper: A New Hardware Architecture for Accelerating Pre-Alignment in DNA Short Read Mapping", Bioinformatics, 2017.

Strengths

Strengths

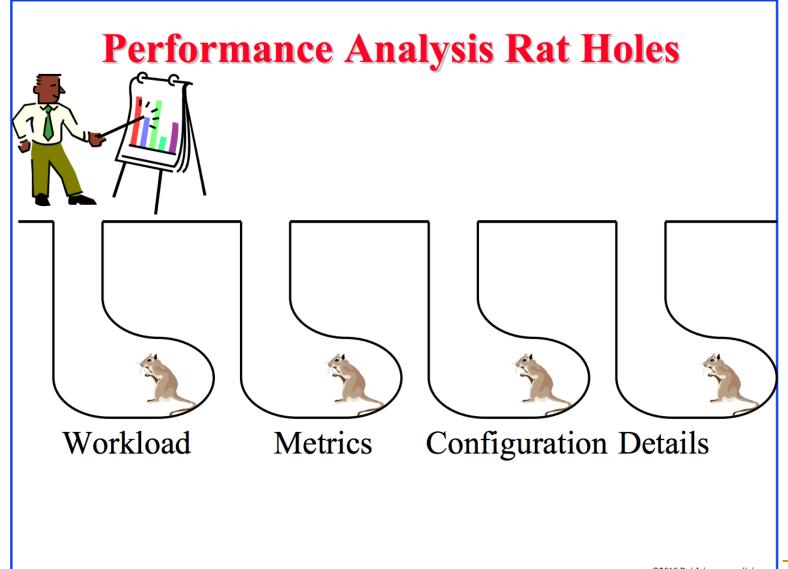
- New and simple solution to a critical problem. New algorithm and hardware architecture.
- GateKeeper does not sacrifice any of the aligner capabilities, as it does not modify or replace the alignment step.
- Design is scalable; could add more processing cores in the future.
- Some sequencers use FPGAs as well, so GateKeeper could be integrated into them.

Strengths (cont'd)

- Authors understand and highlight limitations of GateKeeper
- Greatly improves filtering speed and accuracy
- Spurred quite a few papers that build on GateKeeper
- Well-written, interesting and easy to understand paper

Weaknesses

Recall: Try to Avoid Rat Holes



Weaknesses

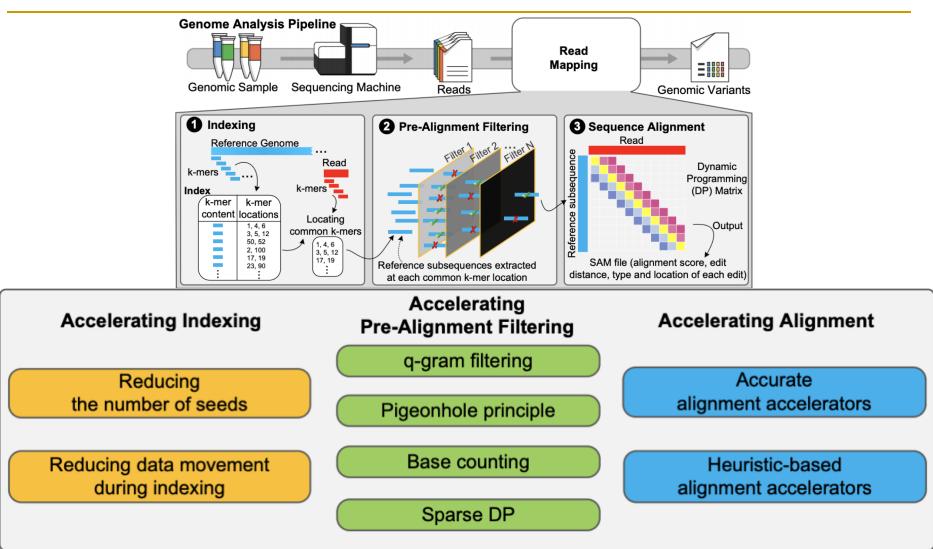
- The benefits of such a mechanism require an FPGA and advanced knowledge with computers, this may be problematic for some biologists/genomicists/geneticists
- The amendment of the random zeros is a simple "hack" to reduce the number of false positives, but there is no explanation why GateKeeper only flips the patterns 101 and 1001, what about 10001? And 10ⁿ1?
- The paper can be confusing at times due to the use of a "supplementary material" document that is constantly referred to (but understandable as there was a page limit set by the publication journal).

Weaknesses (cont'd)

- GateKeeper's accuracy degrades exponentially for E > 2%, and becomes ineffective for E > 8%.
- GateKeeper is tested using short reads
 - 3rd generation sequencing machines produce much longer reads

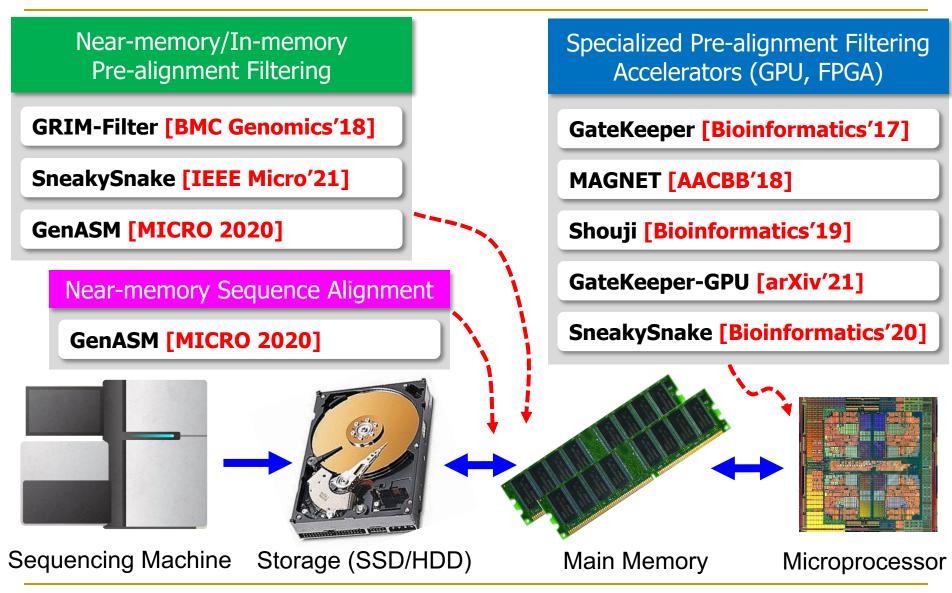
Thoughts and Ideas

Accelerating Read Mapping



Alser+, "Accelerating Genome Analysis: A Primer on an Ongoing Journey", IEEE Micro, 2020.

Our Ongoing Journey



Extensions

- Can we improve the filtering accuracy
 - Don't amend, count the number of matches accurately.
 - Yes, see <u>MAGNET paper [Alser et al. arXiv preprint 2017]</u>. But this requires large number of LUTs.

MAGNET [Alser+, arXiv 2017]

Mohammed Alser, Onur Mutlu, and Can Alkan,
 "MAGNET: Understanding and Improving the Accuracy of Genome Pre-Alignment Filtering"
 IPSI Transactions on Internet Research, July 2017.
 arXiv.org version, July 2017.
 [Source Code]

MAGNET: Understanding and Improving the Accuracy of Genome Pre-Alignment Filtering

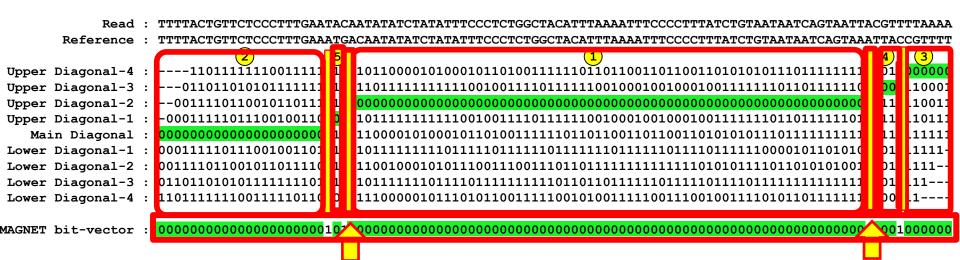
Alser, Mohammed; Mutlu, Onur; and Alkan, Can

MAGNET Walkthrough

Build Neighborhood Map

Track the Diagonally Consecutive Matches

ACCEPT iff number of '1' ≤ Threshold



Find the longest segment of consecutive zeros

Exclude the errors from the search space

Divide the problem into two subproblems and repeat

Total number of edits = number of 1's in MAGNET bit-vector

Extensions

- Can we improve the filtering accuracy
 - Don't amend, count the number of matches accurately.
 - Yes, see <u>MAGNET paper [Alser et al. arXiv preprint 2017]</u>. But this requires large number of LUTs.
- Can we improve the filtering accuracy and scalability
 - □ Yes, see Shouji paper [Alser et al. *Bioinformatics* 2019].

Shouji (障子) [Alser+, Bioinformatics 2019]

Mohammed Alser, Hasan Hassan, Akash Kumar, Onur Mutlu, and Can Alkan, "Shouji: A Fast and Efficient Pre-Alignment Filter for Sequence Alignment" Bioinformatics, [published online, March 28], 2019.

Source Code

[Online link at Bioinformatics Journal]

Bioinformatics, 2019, 1–9 doi: 10.1093/bioinformatics/btz234 Advance Access Publication Date: 28 March 2019 Original Paper



Sequence alignment

Shouji: a fast and efficient pre-alignment filter for sequence alignment

Mohammed Alser^{1,2,3,*}, Hasan Hassan¹, Akash Kumar², Onur Mutlu^{1,3,*} and Can Alkan^{3,*}

¹Computer Science Department, ETH Zürich, Zürich 8092, Switzerland, ²Chair for Processor Design, Center For Advancing Electronics Dresden, Institute of Computer Engineering, Technische Universität Dresden, 01062 Dresden, Germany and ³Computer Engineering Department, Bilkent University, 06800 Ankara, Turkey

*To whom correspondence should be addressed.

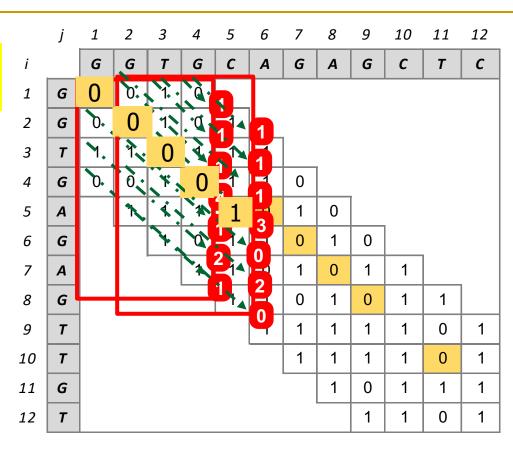
Associate Editor: Inanc Birol

Received on September 13, 2018; revised on February 27, 2019; editorial decision on March 7, 2019; accepted on March 27, 2019

Shouji Walkthrough

Building the Neighborhood Map

Finding all common subsequences (diagonal segments of consecutive zeros) shared between two given sequences.



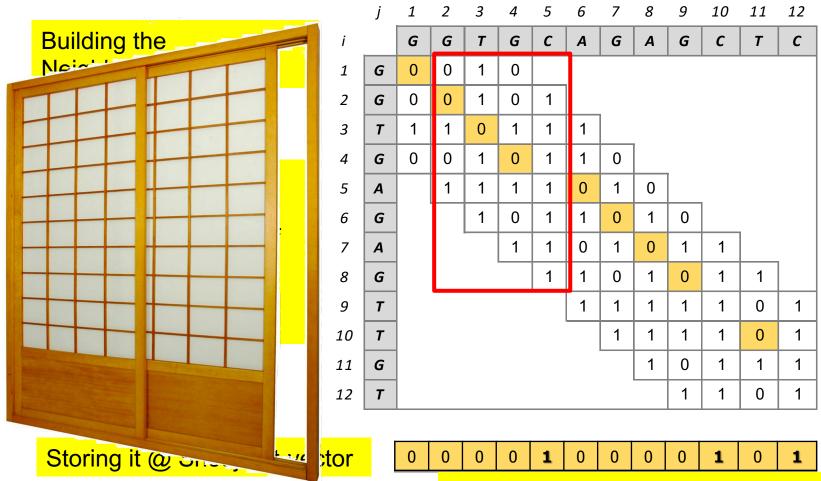
Storing it @ Shouji Bit-vector



ACCEPT iff number of '1' ≤ Threshold

Shouji: a fast and efficient pre-alignment filter for sequence alignment, *Bioinformatics* 2019, https://doi.org/10.1093/bioinformatics/btz234

Shouji Walkthrough



ACCEPT iff number of '1' ≤ Threshold

Shouji: a fast and efficient pre-alignment filter for sequence alignment, *Bioinformatics* 2019, https://doi.org/10.1093/bioinformatics/btz234

Extensions

- Can we improve the filtering accuracy
 - Don't amend, count the number of matches accurately.
 - Yes, see <u>MAGNET paper [Alser et al. arXiv preprint 2017]</u>. But this requires large number of LUTs.
- Can we improve the filtering accuracy and scalability
 - □ Yes, see Shouji paper [Alser et al. *Bioinformatics* 2019].
- Can we solve the FPGA-CPU communication bottleneck?
 - Where it makes sense: Processing-in-memory, Processingnear-storage, Processing-while-sequencing?
 - □ Yes, see GRIM-Filter [Kim et al. BMC Genomics 2018].

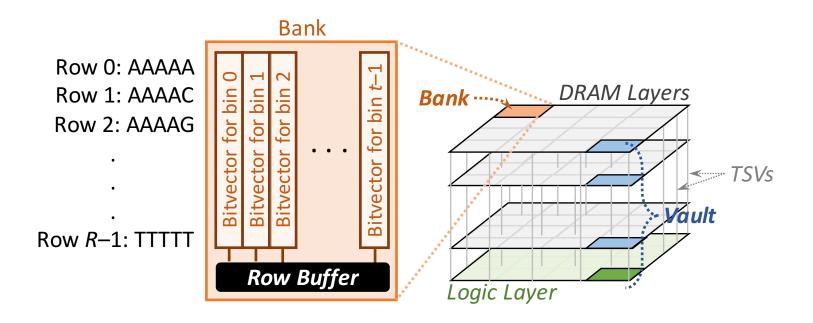
GRIM-Filter [Kim+, BMC Genomics 2018]

Jeremie S. Kim, Damla Senol Cali, Hongyi Xin, Donghyuk Lee, Saugata Ghose, Mohammed Alser, Hasan Hassan, Oguz Ergin, Can Alkan, and Onur Mutlu, "GRIM-Filter: Fast Seed Location Filtering in DNA Read Mapping Using Processing-in-Memory Technologies" to appear in <u>BMC Genomics</u>, 2018.
Proceedings of the <u>16th Asia Pacific Bioinformatics Conference</u> (APBC), Yokohama, Japan, January 2018.
arxiv.org Version (pdf)

GRIM-Filter: Fast Seed Location Filtering in DNA Read Mapping Using Processing-in-Memory Technologies

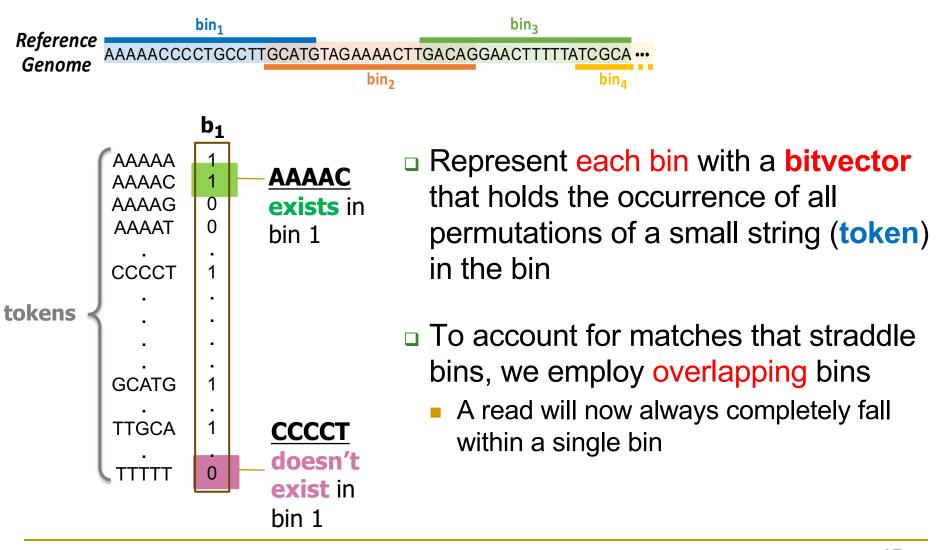
Jeremie S. Kim^{1,6*}, Damla Senol Cali¹, Hongyi Xin², Donghyuk Lee³, Saugata Ghose¹, Mohammed Alser⁴, Hasan Hassan⁶, Oguz Ergin⁵, Can Alkan^{*4}, and Onur Mutlu^{*6,1}

GRIM-Filter in 3D-Stacked DRAM

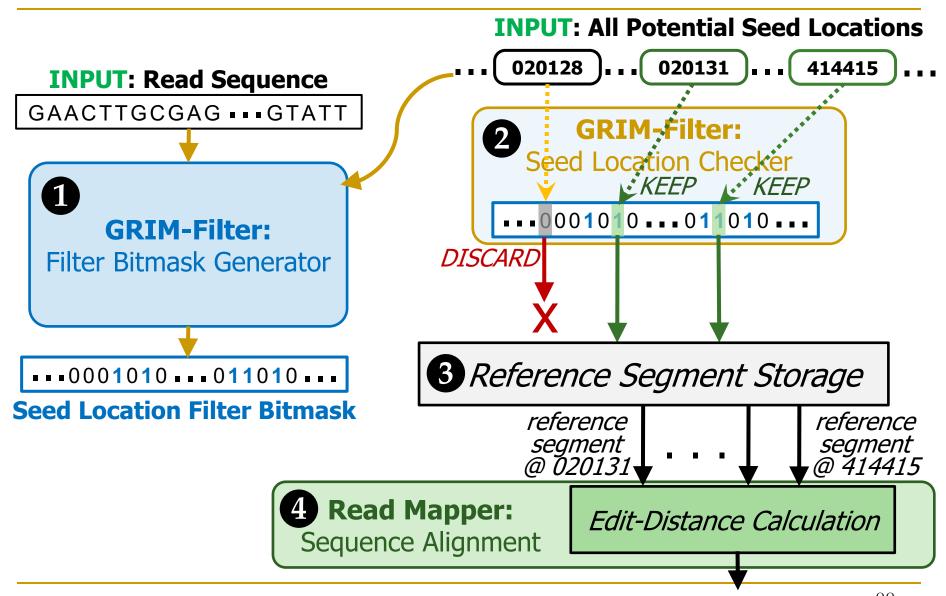


- Each DRAM layer is organized as an array of banks
 - A bank is an array of cells with a row buffer to transfer data
- The layout of bitvectors in a bank enables filtering many bins in parallel

GRIM-Filter: Bitvectors



Integrating GRIM-Filter into a Read Mapper



OUTPUT: Correct Mapping's

Faster, More Accurate, More Scalable Pre-Alignment Filtering

Specialized Hardware for Pre-alignment Filtering

Mohammed Alser, Taha Shahroodi, Juan-Gomez Luna, Can Alkan, and Onur Mutlu, "SneakySnake: A Fast and Accurate Universal Genome Pre-Alignment Filter for CPUs, GPUs, and FPGAs"

Bioinformatics, 2020.

Source Code

Online link at Bioinformatics Journal

Bioinformatics



SneakySnake: a fast and accurate universal genome prealignment filter for CPUs, GPUs and FPGAs

Mohammed Alser ™, Taha Shahroodi, Juan Gómez-Luna, Can Alkan ™, Onur Mutlu ™

Bioinformatics, btaa1015, https://doi.org/10.1093/bioinformatics/btaa1015

Published: 26 December 2020 Article history ▼

SneakySnake

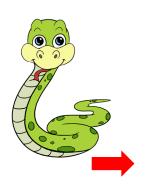
Key observation:

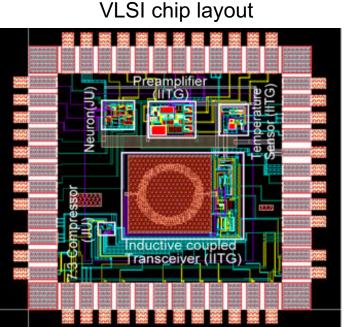
Correct alignment is a sequence of non-overlapping long matches

Key idea:

 Approximate edit distance calculation is similar to Single Net Routing problem in VLSI chip









Building Neighborhood Map

Finding the Optimal Routing Path

Examining the Snake Survival

of value '0') in its corresponding HRT. Given two genomic sequences, a reference sequence $R[1\dots m]$ and a query sequence $Q[1\dots m]$, and an edit distance threshold E, we calculate the entry Z[i,j] of the chip maze, where $1 \leq i \leq (2E+1)$ and $1 \leq j \leq m$, as follows:

$$E = 3$$

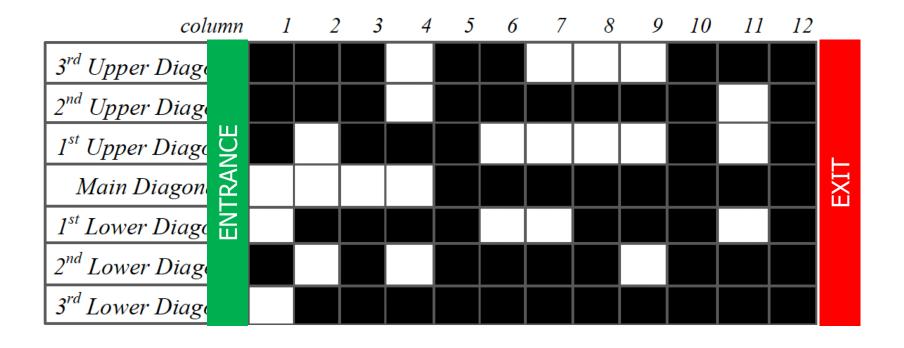
$$Z[i,j] = \begin{cases} 0, & if \ i = E+1, \ Q[j] = R[j], \\ 0, & if \ 1 \le i \le E, \ Q[j-i] = R[j], \\ 0, & if \ i > E+1, \ Q[j+i-E-1] = R[j], \\ 1, & otherwise \end{cases}$$
(1)

column	1	2	3	4	5	6	7	8	9	10	11	12
3 rd Upper Diagonal	1	1	1	0	1	1	0	0	0	1	1	1
2 nd Upper Diagonal	1	1	1	0	1	1	1	1	1	1	0	1
1 st Upper Diagonal	1	0	1	1	1	0	0	0	0	1	0	1
Main Diagonal	0	0	0	0	1	1	1	1	1	1	1	1
1 st Lower Diagonal	0	1	1	1	1	0	0	1	1	1	0	1
2 nd Lower Diagonal	1	0	1	0	1	1	1	1	0	1	1	1
3 rd Lower Diagonal	0	1	1	1	1	1	1	1	1	1	1	1

Building Neighborhood Map

Finding the Optimal Routing Path

Examining the Snake Survival

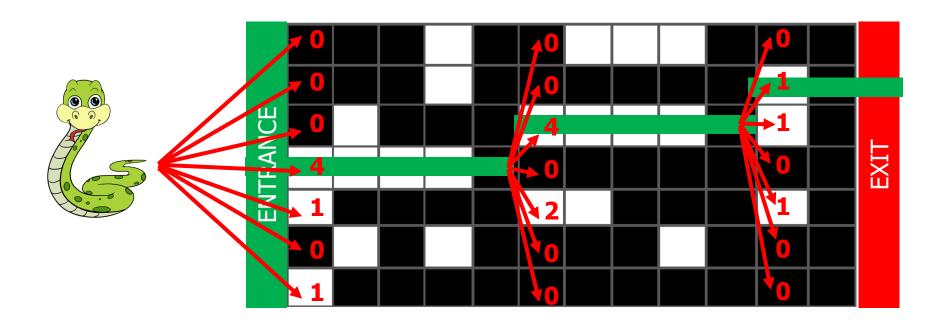


Building Neighborhood Map

Finding the Optimal Routing Path

Examining the Snake Survival





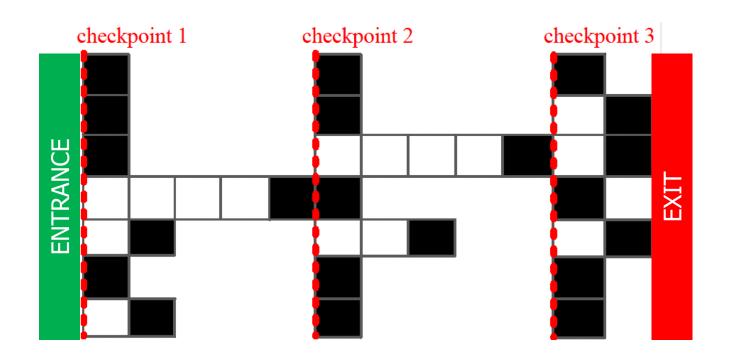
Building Neighborhood Map

Finding the Routing Travel Path

Examining the Snake Survival

This is what you actually need to build and it can be done on-the-fly!



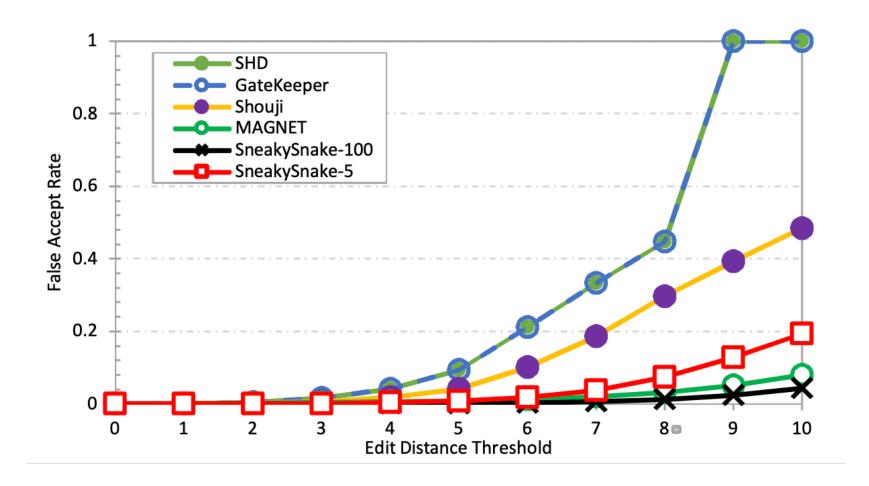


FPGA Resource Analysis

FPGA resource usage for a single filtering unit of GateKeeper,
 Shouji, and Snake-on-Chip for a sequence length of 100 and under different edit distance thresholds (E).

	<i>E</i> (bp)	Slice LUT	Slice Register	No. of Filtering Units
GateKeeper	2	0.39%	0.01%	16
	5	0.71%	0.01%	16
Shouji	2	0.69%	0.08%	16
	5	1.72%	0.16%	16
Snake-on-Chip	2	0.68%	0.16%	16
	5	1.42%	0.34%	16

Filtering Accuracy



Alser, "Accelerating the Understanding of Life's Code Through Better Algorithms and Hardware Design", arXiv preprint arXiv:1910.03936, 2019.

Key Results of SneakySnake

- SneakySnake is up to four orders of magnitude more accurate than Shouji (Bioinformatics'19) and GateKeeper (Bioinformatics'17)
- Using short reads, SneakySnake accelerates Edlib (Bioinformatics'17)
 and Parasail (BMC Bioinformatics'16) by
 - up to $37.7 \times$ and $43.9 \times$ (>12 × on average), on CPUs
 - up to 413 × and 689 × (>400 × on average) with FPGA/GPU acceleration
- Using long reads, SneakySnake accelerates Parasail and KSW2 by 140.1 × and 17.1 × on average, respectively, on CPUs

Takeaways

Key Takeaways

- A novel method to accelerate Sequence Alignment in genome analysis.
- Simple and effective
- Hardware/software cooperative
- Good potential for work building on it to extend it
 - To make things more efficient and effective
 - Multiple works have already built on the paper (see MAGNET, Shouji, GRIM-Filter, SneakySnake, GenCache)
- Easy to read and understand paper

Open Discussion

Discussion Starters (I)

- Thoughts on the previous ideas?
- Rethinking Alignment and Pre-alignment?
 - Re-use the results of the pre-alignment filter?
 - Improve the accuracy of pre-alignment filtering to achieve an optimal alignment?
- Extend the solution to longer reads, higher edit distance thresholds?
- Is this solution clearly advantageous in some cases?

Discussion Starters (II)

- Data movement is still a bottleneck. How could we try to reduce it?
 - Placing the accelerator closer to memory
 - Using newer and faster I/O
 - Closely integrate the accelerator into sequencers for real-time pre-alignment filtering
 - Offer cloud computing with access to advanced FPGA chips

Alleviating Data Movement Bottlenecks

Near-memory Pre-alignment Filtering

Gagandeep Singh, Mohammed Alser, Damla Senol Cali, Dionysios Diamantopoulos, Juan Gomez-Luna, Henk Corporaal, Onur Mutlu,

"FPGA-Based Near-Memory Acceleration of Modern Data-Intensive Applications"

IEEE Micro, 2021.

Source Code



PIEEE

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WWW.computer.org/micro



Home / Magazines / IEEE Micro / 2021.04

IEEE Micro

FPGA-Based Near-Memory Acceleration of Modern Data-Intensive Applications

July-Aug. 2021, pp. 39-48, vol. 41

DOI Bookmark: 10.1109/MM.2021.3088396

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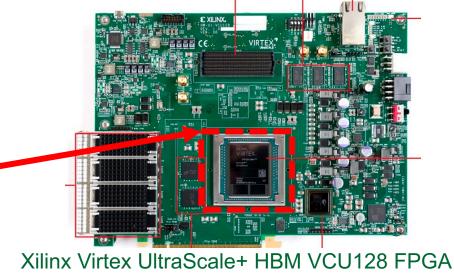
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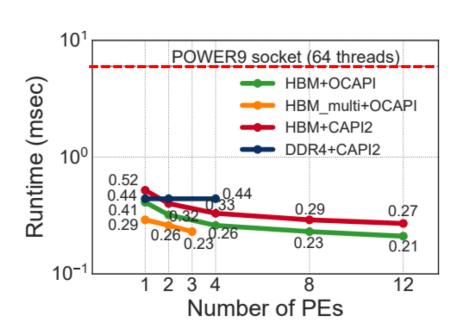
Near-memory SneakySnake

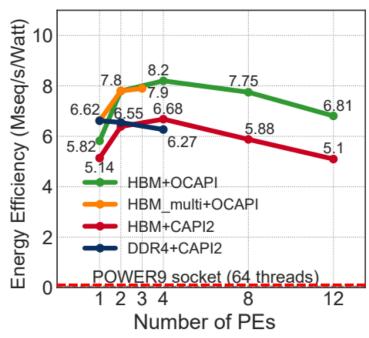
- Problem: Read Mapping is heavily bottlenecked by data movement from main memory
- Solution: Perform read mapping near where data resides (i.e., near-memory)
- We carefully redesigned the accelerator logic of SneakySnake to exploit near-memory computation capability on modern FPGA boards with high-bandwidth memory

FPGA + high-bandwidth memory on the same package substrate



Key Results of Near-memory SneakySnake





Near-memory pre-alignment filtering improves performance and energy efficiency by 27.4 × and 133 ×, respectively, over a 16-core (64 hardware threads) IBM POWER9 CPU

GenASM Framework [MICRO 2020]

Damla Senol Cali, Gurpreet S. Kalsi, Zulal Bingol, Can Firtina, Lavanya Subramanian, Jeremie S. Kim, Rachata Ausavarungnirun, Mohammed Alser, Juan Gomez-Luna, Amirali Boroumand, Anant Nori, Allison Scibisz, Sreenivas Subramoney, Can Alkan, Saugata Ghose, and Onur Mutlu, "GenASM: A High-Performance, Low-Power Approximate String Matching Acceleration Framework for Genome Sequence Analysis"

Proceedings of the 53rd International Symposium on Microarchitecture (MICRO), Virtual, October 2020.

[<u>Lightning Talk Video</u> (1.5 minutes)] [<u>Lightning Talk Slides (pptx) (pdf)</u>] [<u>Talk Video</u> (18 minutes)] [<u>Slides (pptx) (pdf)</u>]

GenASM: A High-Performance, Low-Power Approximate String Matching Acceleration Framework for Genome Sequence Analysis

Damla Senol Cali^{†™} Gurpreet S. Kalsi[™] Zülal Bingöl[▽] Can Firtina[⋄] Lavanya Subramanian[‡] Jeremie S. Kim^{⋄†} Rachata Ausavarungnirun[⊙] Mohammed Alser[⋄] Juan Gomez-Luna[⋄] Amirali Boroumand[†] Anant Nori[™] Allison Scibisz[†] Sreenivas Subramoney[™] Can Alkan[▽] Saugata Ghose^{*†} Onur Mutlu^{⋄†▽}

† Carnegie Mellon University [™] Processor Architecture Research Lab, Intel Labs [▽] Bilkent University [⋄] ETH Zürich

‡ Facebook [⊙] King Mongkut's University of Technology North Bangkok ^{*} University of Illinois at Urbana–Champaign

Near-memory GenASM Framework

- Our goal: Accelerate approximate string matching (ASM) by designing a fast and flexible framework, which can accelerate multiple steps of genome sequence analysis.
- Key ideas: Exploit the high memory bandwidth and the logic layer of 3D-stacked memory to perform highly-parallel ASM in the DRAM chip itself.
- Modify and extend Bitap^{1,2}, ASM algorithm with fast and simple bitwise operations, such that it now:
 - Supports long reads
 - Supports traceback
 - Is highly parallelizable
- Co-design of our modified scalable and memory-efficient algorithms with low-power and area-efficient hardware accelerators

Use Cases of GenASM (cont'd.)

(1) Read Alignment Step of Read Mapping

 Find the optimal alignment of how reads map to candidate reference regions

(2) Pre-Alignment Filtering for Short Reads

 Quickly identify and filter out the unlikely candidate reference regions for each read

(3) Edit Distance Calculation

- Measure the similarity or distance between two sequences
- We also discuss other possible use cases of GenASM in our paper:
 - Read-to-read overlap finding, hash-table based indexing, whole genome alignment, generic text search

Key Results of the GenASM Framework

(1) Read Alignment

- 116 × speedup, 37 × less power than Minimap2 (state-of-the-art sw)
- 111 × speedup, 33 × less power than BWA-MEM (state-of-the-art SW)
- 3.9 × better throughput, 2.7 × less power than Darwin (state-of-the-art HW)
- 1.9 × better throughput, 82% less logic power than **GenAx** (state-of-the-art **HW**)

(2) Pre-Alignment Filtering

■ 3.7 × speedup, 1.7 × less power than **Shouji** (state-of-the-art **HW**)

(3) Edit Distance Calculation

- 22-12501 × speedup, 548-582 × less power than Edlib (state-of-the-art SW)
- 9.3-400 × speedup, 67 × less power than ASAP (state-of-the-art HW)

Discussion Starters (III)

- Can you think of fields that could be similarly in need of string alignment as in read mapping in bioinformatics?
- Natural language processing
 - OCR error correction
 - Autocorrection in text-based editors or apps
 - Reconstruction of languages using the comparative method
 - Social sciences

Combining dynamic programming with filtering to solve a four-stage two-dimensional guillotine-cut bounded knapsack problem

François Clautiaux^{a,b,*}, Ruslan Sadykov^{b,a}, François Vanderbeck^{a,b}, Quentin Viaud^{a,b}

^aIMB, Université de Bordeaux, 351 cours de la Libération, 33405 Talence, France ^bINRIA Bordeaux - Sud-Ouest, 200 avenue de la Vieille Tour, 33405 Talence, France

Clautiaux+, "Combining dynamic programming with filtering to solve a four-stage two-dimensional guillotine-cut bounded knapsack problem", Discrete Optimization, 2018.

Adoption of hardware accelerators in genome analysis

Bioinformatics: Reviewer #6 (Dec. 2016)

I have a major concern with the work that is actually not a problem with the manuscript at all. Specifically, I have the concern that there has been little to no adoption of previous specialized hardware solutions related to improving the speed of alignment. While there has been considerable work in this area (which the authors do an admirable job of citing), it does not seem that these hardware-based solutions have gained any type of real traction in the community, as the vast majority of alignment is still performed on "regular" CPUs, where the extent of hardware acceleration is the adoption of specific SIMD or vectorized instructions. While I don't think that this practical concern should preclude publication of the current work, it is something worth considering (e.g. what, if any, of the proposed improvements to the SHD filter could be "back-ported" to a software-only solution).

Our Response

We see the reviewer's point, but we do not believe this should be held against the research in the area of FPGA-based acceleration of read mapping in particular or genomics in general. It always takes time to adopt a "new" or "different" hardware technology since it requires investment into the hardware infrastructure. The main challenges/barriers that limit the popularity of FPGAs in the genomics field are the high cost, design effort, and development time. Due to the fact that the deliverable of such projects is normally a hardware product, researchers tend to commercialize their research with startup companies and engage themselves with industrial collaborators, as we describe below. Today, the cost structure of FPGAs is changing because major cloud infrastructures (e.g., by Microsoft Azure and Amazon AWS) offer FPGAs as core engines of the infrastructure. Therefore, we believe the benefits of FPGA-based acceleration has become available to many more folks in the community, especially with the open-source release of such FPGA-accelerated solutions. To increase adoption, we have decided to release our source code for GateKeeper. It is available on https://github.com/BilkentCompGen/GateKeeper.

Some examples of the research groups that commercialize their research and promote FPGA-based or even cloud-based products for genomics are as follows:

http://www.timelogic.com/catalog/775

http://www.gidel.com/HPC-RC/HPC-Applications.asp

http://www.edicogenome.com/dragen_bioit_platform/the-dragen-engine-2/

http://www.bcgsc.ca/platform/bioinfo/software/XpressAlign/releases/1.0

https://www.sevenbridges.com/amazon/

http://www.falcon-computing.com/index.php/solutions/falcon-genomics-solutions/

Our Response (cont'd)

It is also important to emphasize that the necessity of designing a mapper on hardware is currently steering the field towards more personalized medicine. Hardware-accelerated mappers (using various platforms such as SIMD, GPUs, and FPGAs) are becoming increasingly popular as they can be potentially directly integrated into sequencing machines (the Illumina sequencer, for example, includes an FPGA chip inside it

https://support.illumina.com/content/dam/illumina-support/documents/downloads/software/hiseq/hcs_2-0-12/installnotes_hcs2-0-12.pdf), such that we have a single machine that can perform both sequencing and mapping (Lindner, et al., Bioinformatics 2016). This approach has two benefits. First, it can hide the complexity and details of the underlying hardware from users who are not necessarily aware about FPGAs (e.g., biologists and mathematicians). Second, it allows a significant reduction in total genome analysis time by starting read mapping while still sequencing. Hence, an end user or researcher in genomics might not directly deal with the "pre-alignment on FPGA" or "mapper on FPGA", but they might purchase a sequencer that performs pre-alignment and alignment using FPGAs inside. As such, one potential target of our research is to influence the design of more intelligent sequencing machines by integrating GateKeeper inside them.

In fact, we believe GateKeeper is very suitable to be used as part of a sequencer as it provides a complete prealignment system that includes many processing cores, where all processing cores work in parallel to provide extremely fast filtering. We believe such a fast approach can make sequencers more intelligent and attractive. Remember What We Said in the First Lecture

Dream and, they will come

- Computing landscape is very different from 10-20 years ago.
- •As applications push boundaries, computing platforms will become increasingly strained.

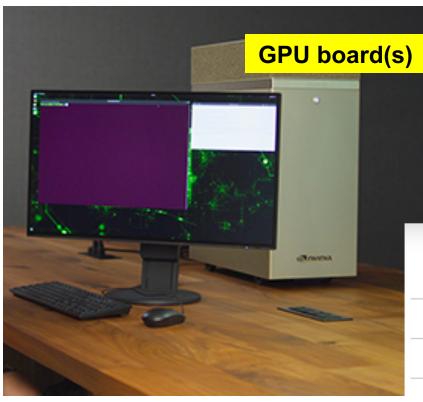
Illumina DRAGEN Bio-IT Platform (2018)

 Processes whole genome at 30x coverage in ~25 minutes with hardware support for data compression

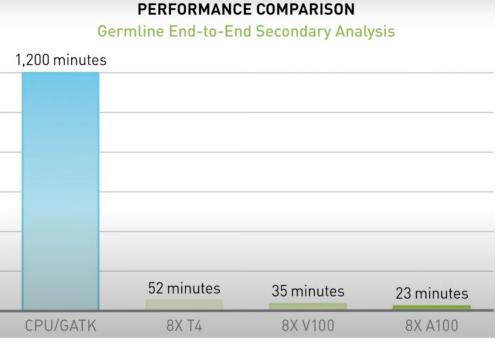


emea.illumina.com/products/by-type/informatics-products/dragen-bio-it-platform.html emea.illumina.com/company/news-center/press-releases/2018/2349147.html

NVIDIA Clara Parabricks (2020)



A University of Michigan's startup in 2018 and joined NVIDIA in 2020



Computing is Still Bottlenecked by Data Movement

Adoption Challenges of Hardware Accelerators

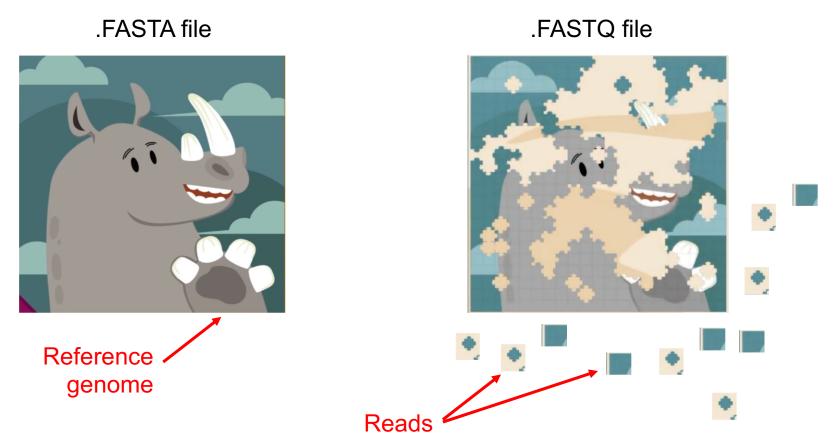
- Accelerate the entire read mapping process rather than its individual steps (Amdahl's law)
- Reduce the high amount of data movement
 - Working directly on compressed data
 - Filter out unlikely-reused data at the very first component of the compute system
- Develop flexible hardware architectures that do NOT conservatively limit the range of supported parameter values at design time
- Adapt existing genomic data formats for hardware accelerators or develop more efficient file formats

Adoption Challenges of Hardware Accelerators

- Maintaining the same (or better) accuracy/sensitivity of the output results of the software version
 - Using heuristic algorithms to gain speedup!
- High hardware cost
- Long development life-cycle for FPGA platforms

What else can be done?

What if we got a new version of the reference genome?



https://www.pacb.com/smrt-science/smrt-sequencing/hifi-reads-for-highly-accurate-long-read-sequencing/

AirLift [Kim+, arXiv 2021]

Jeremie S. Kim, Can Firtina, Meryem Banu Cavlak, Damla Senol Cali, Mohammed Alser, Nastaran Hajinazar, Can Alkan, Onur Mutlu "AirLift: A Fast and Comprehensive Technique for Translating Alignments between Reference Genomes", arXiv, 2021

[Source Code]

[Online link at arXiv]

RESEARCH

AirLift: A Fast and Comprehensive Technique for Remapping Alignments between Reference Genomes

Jeremie S. Kim¹, Can Firtina¹, Meryem Banu Cavlak², Damla Senol Cali³, Nastaran Hajinazar^{1,4}, Mohammed Alser¹, Can Alkan² and Onur Mutlu^{1,2,3*}

AirLift

- Key observation: Reference genomes are updated frequently.
 Repeating read mapping is a computationally expensive workload.
- Key idea: Update the mapping results of only affected reads depending on how a region in the old reference relates to another region in the new reference.

Key results:

- reduces number of reads that needs to be re-mapped to new reference by up to 99.99%
- reduces overall runtime to re-map reads by 6.7x, 6.6x, and
 2.8x for large (human), medium (C. elegans), and small (yeast) reference genomes

Clustering the Reference Genome Regions

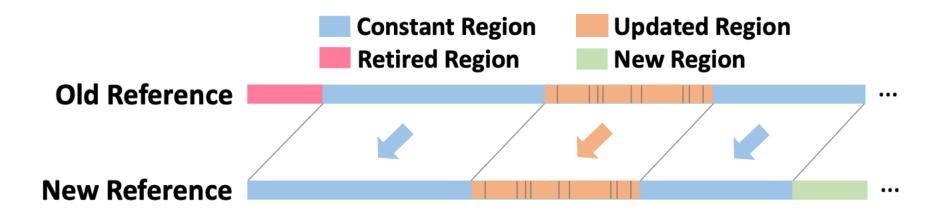
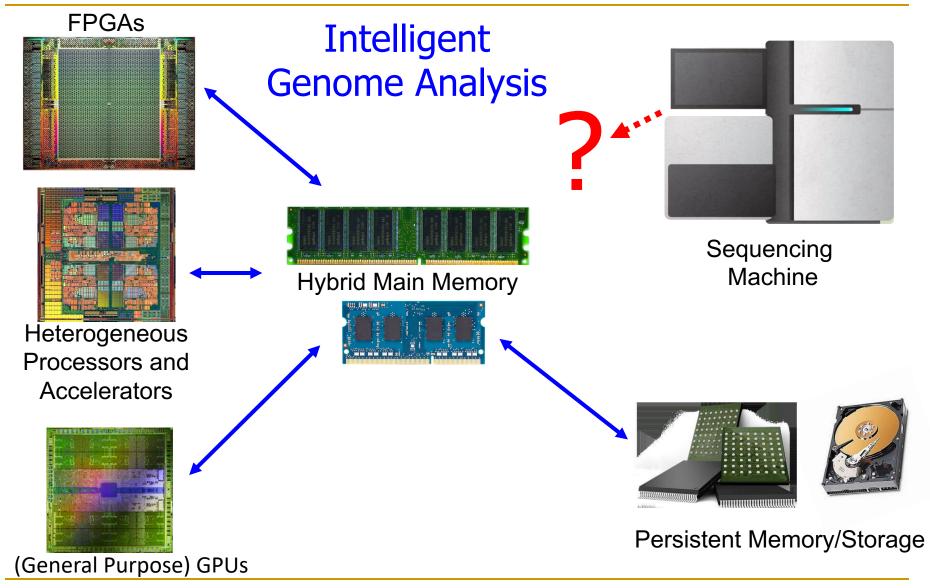


Fig. 2. Reference Genome Regions.

Processing Genomic Data Where it Makes Sense



What is Intelligent Genome Analysis?

- Fast genome analysis
 - □ Real-time analysis

Bandwidth

- Using intelligent architectures
 - □ Specialized HW with less data movement

Energy-efficiency & Latency

- DNA is a valuable asset
 - □ Controlled-access analysis

Privacy

- Population-scale genome analysis
 - □ Sequence anywhere at large scale!

Scalability

- Avoiding erroneous analysis
 - □ *E.g., your father is not your father*

Accuracy

Achieving Intelligent Genome Analysis?

How and where to enable

fast, accurate, cheap,

privacy-preserving, and exabyte scale analysis of genomic data?

Most speedup comes from parallelism enabled by novel architectures and algorithms

Read Mapping in 111 pages!

In-depth analysis of 107 read mappers (1988-2020)

Mohammed Alser, Jeremy Rotman, Dhrithi Deshpande, Kodi Taraszka, Huwenbo Shi, Pelin Icer Baykal, Harry Taegyun Yang, Victor Xue, Sergey Knyazev, Benjamin D. Singer, Brunilda Balliu, David Koslicki, Pavel Skums, Alex Zelikovsky, Can Alkan, Onur Mutlu, Serghei Mangul

"<u>Technology dictates algorithms: Recent developments in read alignment</u>" Genome Biology, 2021

Source code

Alser et al. Genome Biology (2021) 22:249 https://doi.org/10.1186/s13059-021-02443-7

Genome Biology

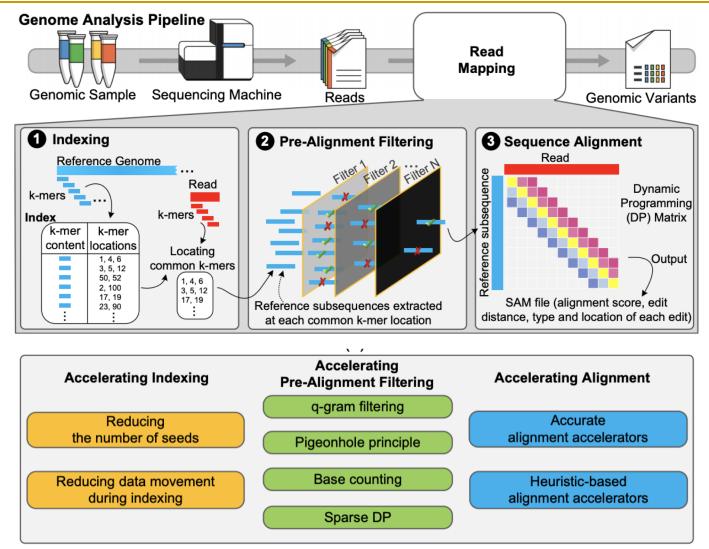
REVIEW Open Access

Technology dictates algorithms: recent developments in read alignment



Mohammed Alser^{1,2,3†}, Jeremy Rotman^{4†}, Dhrithi Deshpande⁵, Kodi Taraszka⁴, Huwenbo Shi^{6,7}, Pelin Icer Baykal⁸, Harry Taegyun Yang^{4,9}, Victor Xue⁴, Sergey Knyazev⁸, Benjamin D. Singer^{10,11,12}, Brunilda Balliu¹³, David Koslicki^{14,15,16}, Pavel Skums⁸, Alex Zelikovsky^{8,17}, Can Alkan^{2,18}, Onur Mutlu^{1,2,3†} and Serghei Mangul^{5*†}

Accelerating Read Mapping

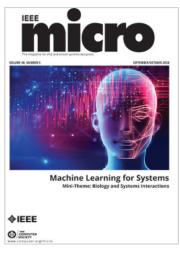


Alser+, "Accelerating Genome Analysis: A Primer on an Ongoing Journey", IEEE Micro, 2020.

Detailed Analysis of Tackling the Bottleneck

Mohammed Alser, Zülal Bingöl, Damla Senol Cali, Jeremie Kim, Saugata Ghose, Can Alkan, Onur Mutlu

"Accelerating Genome Analysis: A Primer on an Ongoing Journey" IEEE Micro, August 2020.





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IEEE Micro

Accelerating Genome Analysis: A Primer on an Ongoing Journey

Sept.-Oct. 2020, pp. 65-75, vol. 40

DOI Bookmark: 10.1109/MM.2020.3013728

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More on Fast Genome Analysis ...

Onur Mutlu,

"Accelerating Genome Analysis: A Primer on an Ongoing Journey"

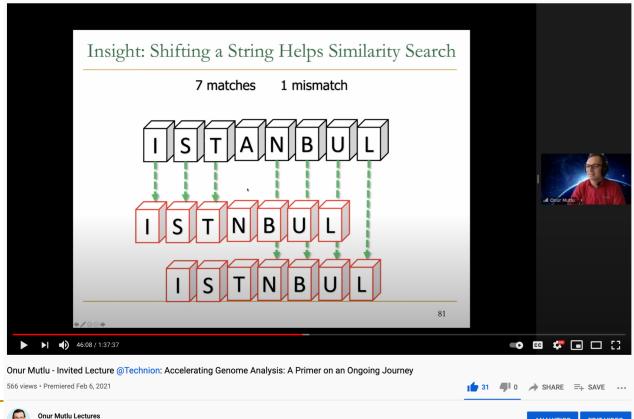
Invited Lecture at Technion, Virtual, 26 January 2021.

[Slides (pptx) (pdf)]

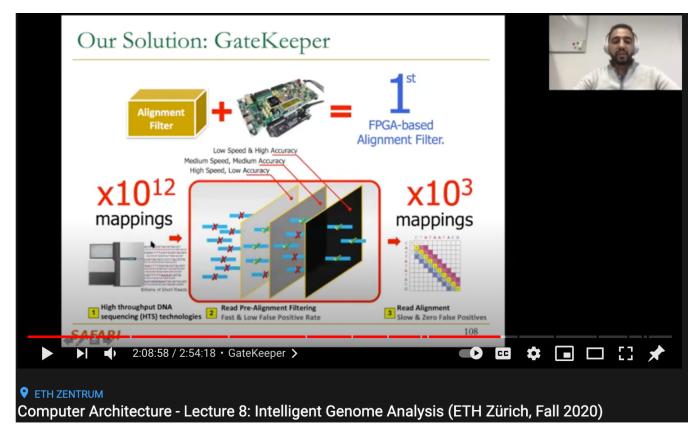
[Talk Video (1 hour 37 minutes, including Q&A)]

[Related Invited Paper (at IEEE Micro, 2020)]

13.9K subscribers



More on Intelligent Genome Analysis ...





https://www.youtube.com/watch?v=ygmQpdDTL7o

Detailed Lectures on Genome Analysis

- Computer Architecture, Fall 2020, Lecture 3a
 - Introduction to Genome Sequence Analysis (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=CrRb32v7SJc&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=5
- Computer Architecture, Fall 2020, Lecture 8
 - □ **Intelligent Genome Analysis** (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=ygmQpdDTL7o&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=14
- Computer Architecture, Fall 2020, Lecture 9a
 - □ **GenASM: Approx. String Matching Accelerator** (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=XoLpzmN Pas&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=15
- Accelerating Genomics Project Course, Fall 2020, Lecture 1
 - Accelerating Genomics (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=rgjl8ZyLsAg&list=PL5Q2soXY2Zi9E2bBVAgCqL gwiDRQDTyId

Prior Research on Genome Analysis (1/2)

- Alser + "SneakySnake: A Fast and Accurate Universal Genome Pre-Alignment Filter for CPUs, GPUs, and FPGAs.", Bioinformatics, 2020.
- Senol Cali+, "GenASM: A High-Performance, Low-Power Approximate
 String Matching Acceleration Framework for Genome Sequence Analysis",
 MICRO 2020.
- Alser+, "<u>Technology dictates algorithms: Recent developments in read alignment</u>", arXiv, 2020.
- Kim+, "<u>AirLift: A Fast and Comprehensive Technique for Translating</u>
 <u>Alignments between Reference Genomes</u>", *arXiv*, 2020
- Alser+, "<u>Accelerating Genome Analysis: A Primer on an Ongoing Journey</u>", IEEE Micro, 2020.

Prior Research on Genome Analysis (2/2)

- Firtina+, "Apollo: a sequencing-technology-independent, scalable and accurate assembly polishing algorithm", Bioinformatics, 2019.
- Alser+, "Shouji: a fast and efficient pre-alignment filter for sequence alignment", Bioinformatics 2019.
- Kim+, "GRIM-Filter: Fast Seed Location Filtering in DNA Read Mapping Using Processing-in-Memory Technologies", BMC Genomics, 2018.
- Alser+, "GateKeeper: A New Hardware Architecture for Accelerating Pre-Alignment in DNA Short Read Mapping", Bioinformatics, 2017.
- Alser+, "MAGNET: understanding and improving the accuracy of genome pre-alignment filtering", IPSI Transaction, 2017.

GateKeeper: A New Hardware Architecture for Accelerating Pre-Alignment in DNA Short Read Mapping

Mohammed Alser, Hasan Hassan, Hongyi Xin, Oğuz Ergin, Onur Mutlu, Can Alkan Bioinformatics, 2017

Presented by: Mohammed Alser







Thank you. Questions?

Seminar in Computer Architecture

Meeting 2: GateKeeper

Dr. Mohammed Alser



ETH Zurich
Fall 2021
30 September 2021



