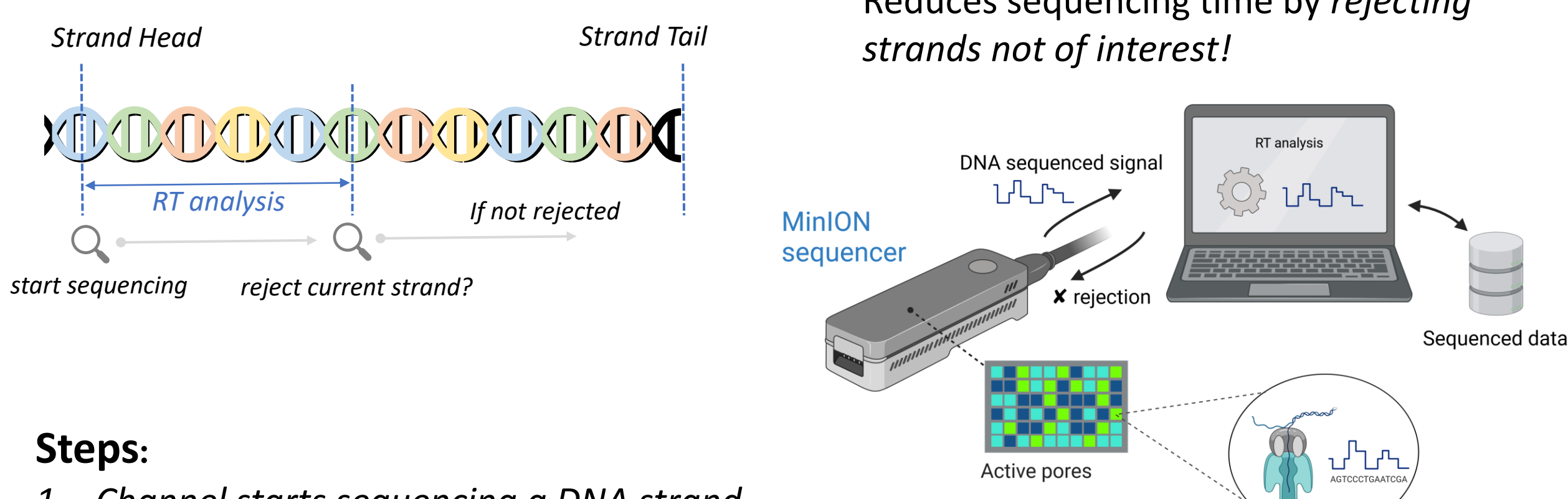


Introduction

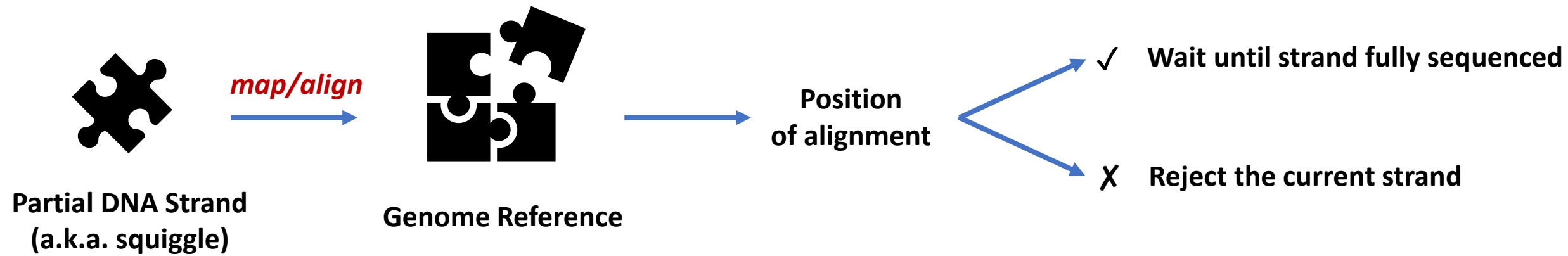
Nanopore sequencers provide **portable** long-read sequencing and the ability to **access, analyse, and filter reads in real-time** → **Read Until**



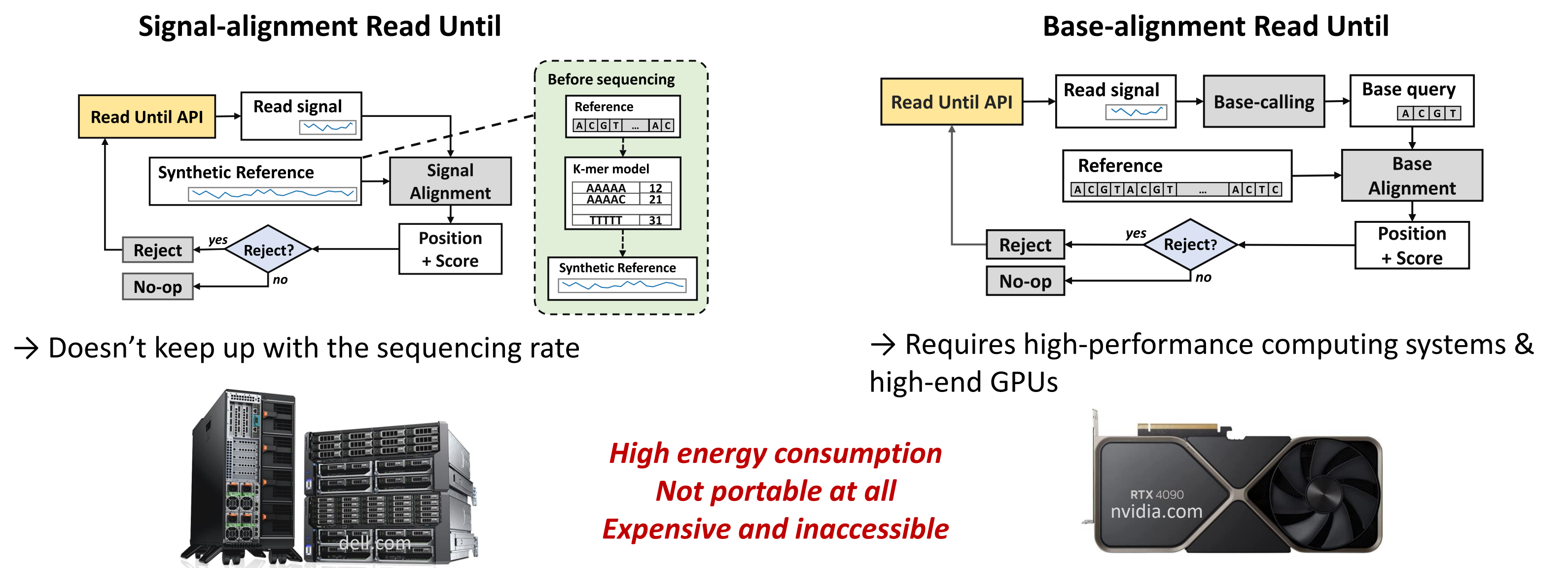
Steps:

1. Channel starts sequencing a DNA strand
2. Host machine performs **real-time analysis** as we sequence
3. If analysis results is to "skip", reject the strand
4. If analysis results is to "continue", let the pore finish sequencing

Typical analysis flow



Problem Existing analysis pipelines are compute-intensive → *costly compute hardware requirements!*



Our goal Real-time analysis on an SoC for Read Until on Nanopore Sequencers!

- Low energy
- Ultra portable and accessible
- Scalable & adaptive to future sequencing advancements
- Fully working from end-to-end
- HIGH THROUGHPUT!

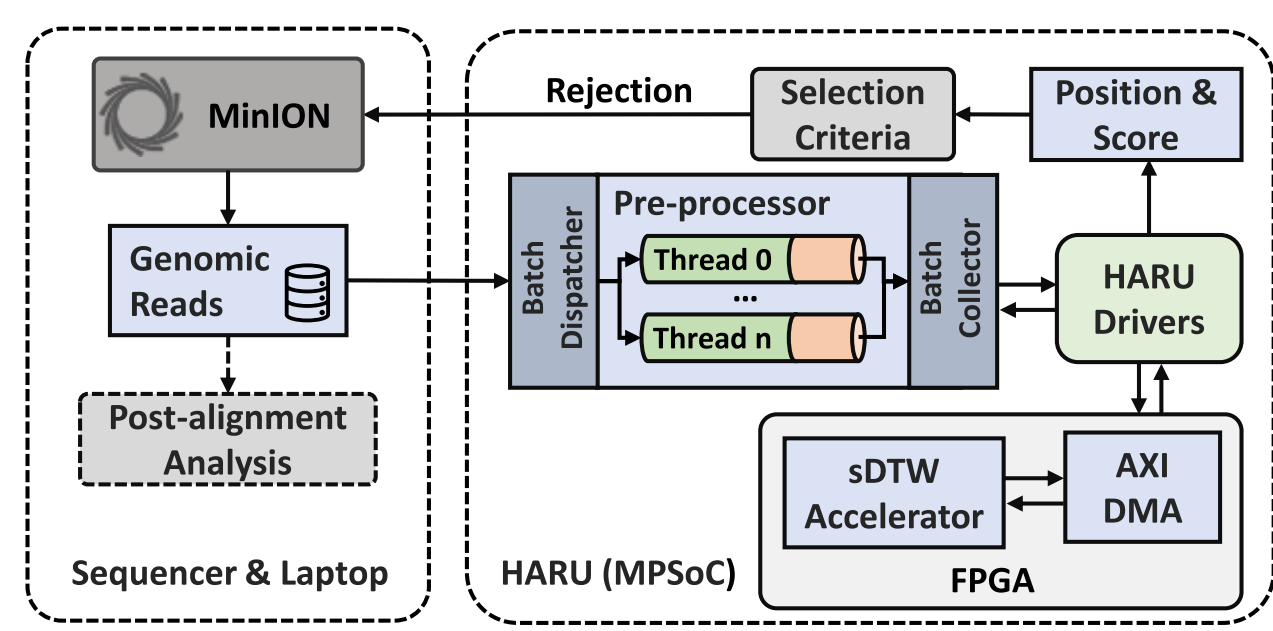
Hardware Accelerated Read Until



System Design and Methodologies

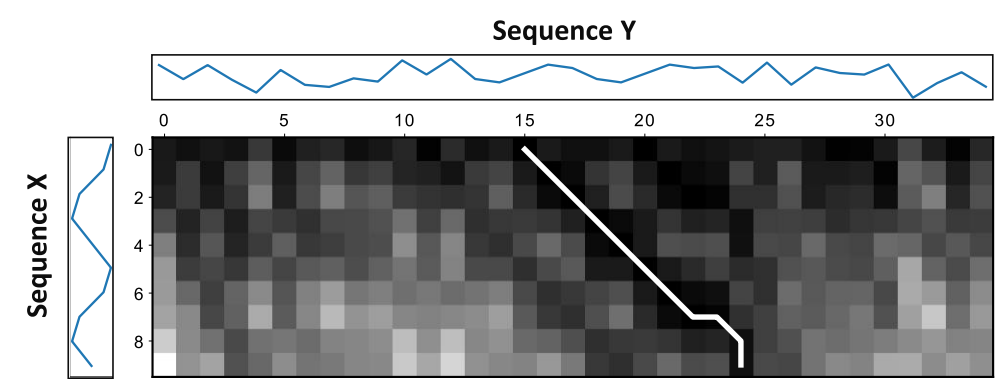
HARU Overview

HARU is a **signal-alignment** hardware-software co-design pipeline for Read Until. It features an efficient **subsequence dynamic time warping (sDTW) hardware accelerator** running in the FPGA of an AMD Zynq MPSoC (an SoC with an ARM processor and FPGA).



Subsequence DTW

Finds the optimal (warped) alignment of a subsequence to reference → sequences are represented in raw signal values (as opposed to bases) → **computationally intensive!** Naively quadratic for both space and time complexity → takes up 98% of processing time for signal-alignment Read Until!

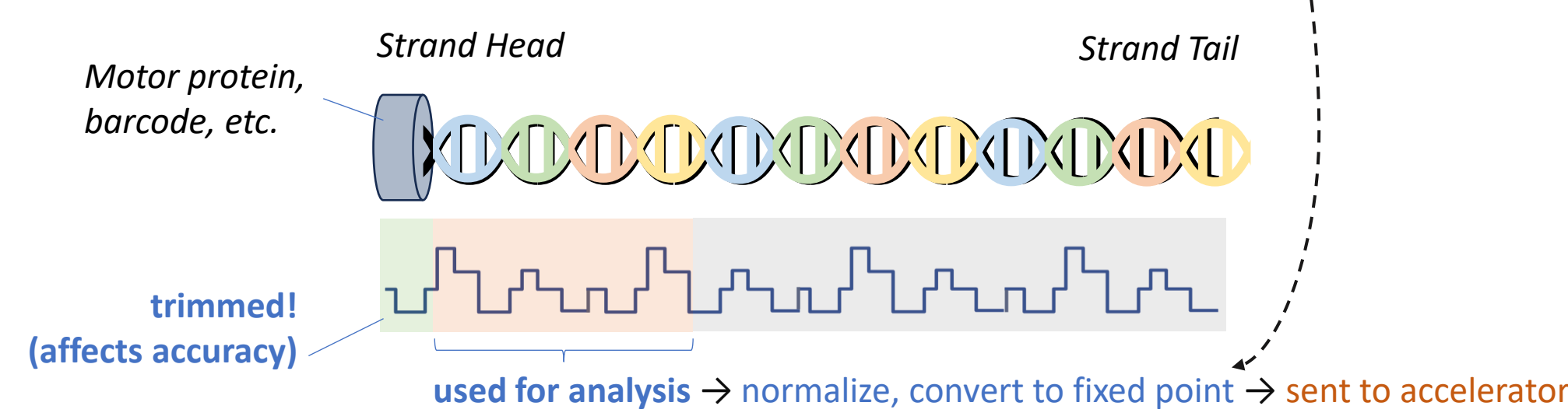
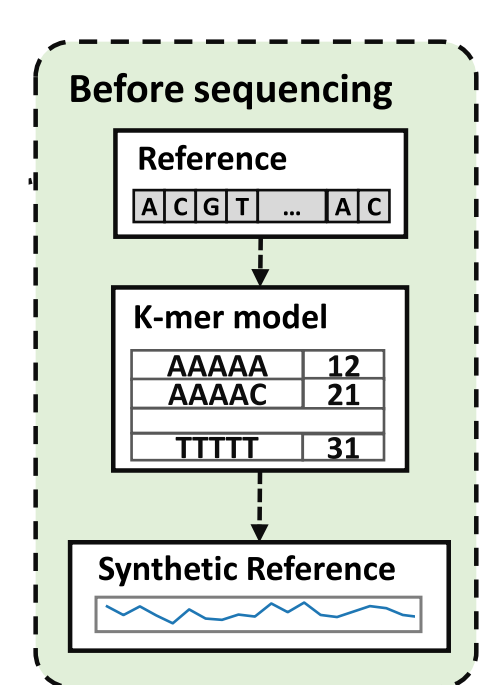
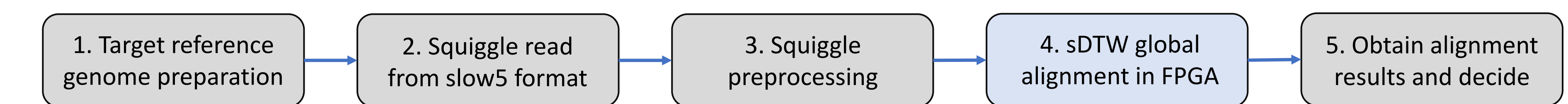


```

Algorithm 1: Subsequence DTW
Input: X[1..M], Y[1..N], M, N
Output: position, score
1. cost ← matrix of size M+1, N+1
2. position ← -1
3. score ← -∞
4. for j in range 1 to N do
5.   c[1, j] ← abs(X[1] - Y[j])
6. end
7. for i in range 2 to M do
8.   for j in range 1 to N do
9.     c[i, j] ← min(c[i-1, j-1], c[i-1, j], c[i, j-1]) + c[i, j]
10.  end
11. end
12. min_score ← c[M, N]
13. position ← j
14. score ← c[M, N]
15. end
  
```

Base-alignment (e.g. SmithWatermann) execution is similar to signal-alignment but requires basecalling (expensive). With enough optimization and acceleration effort, signal-alignment is more direct!

HARU Co-Design Execution Breakdown

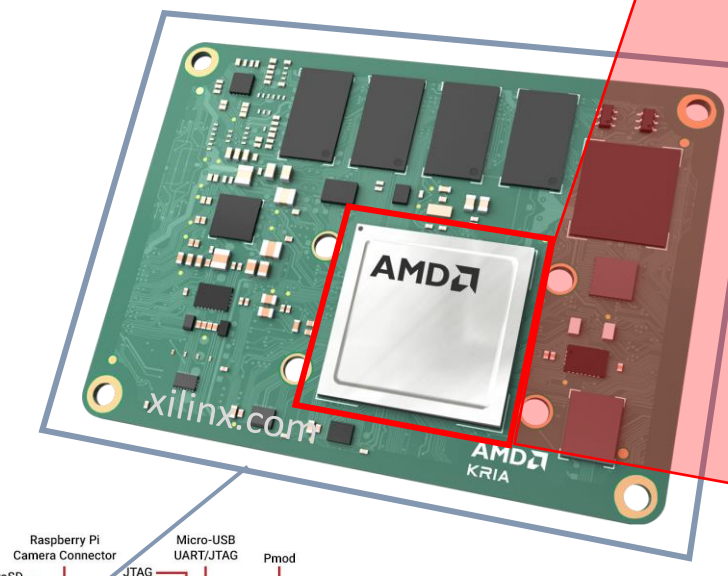
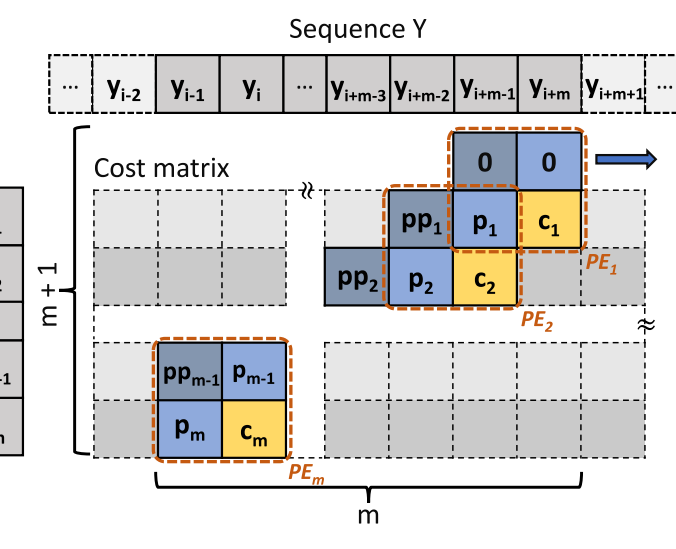


HARU's sDTW accelerator

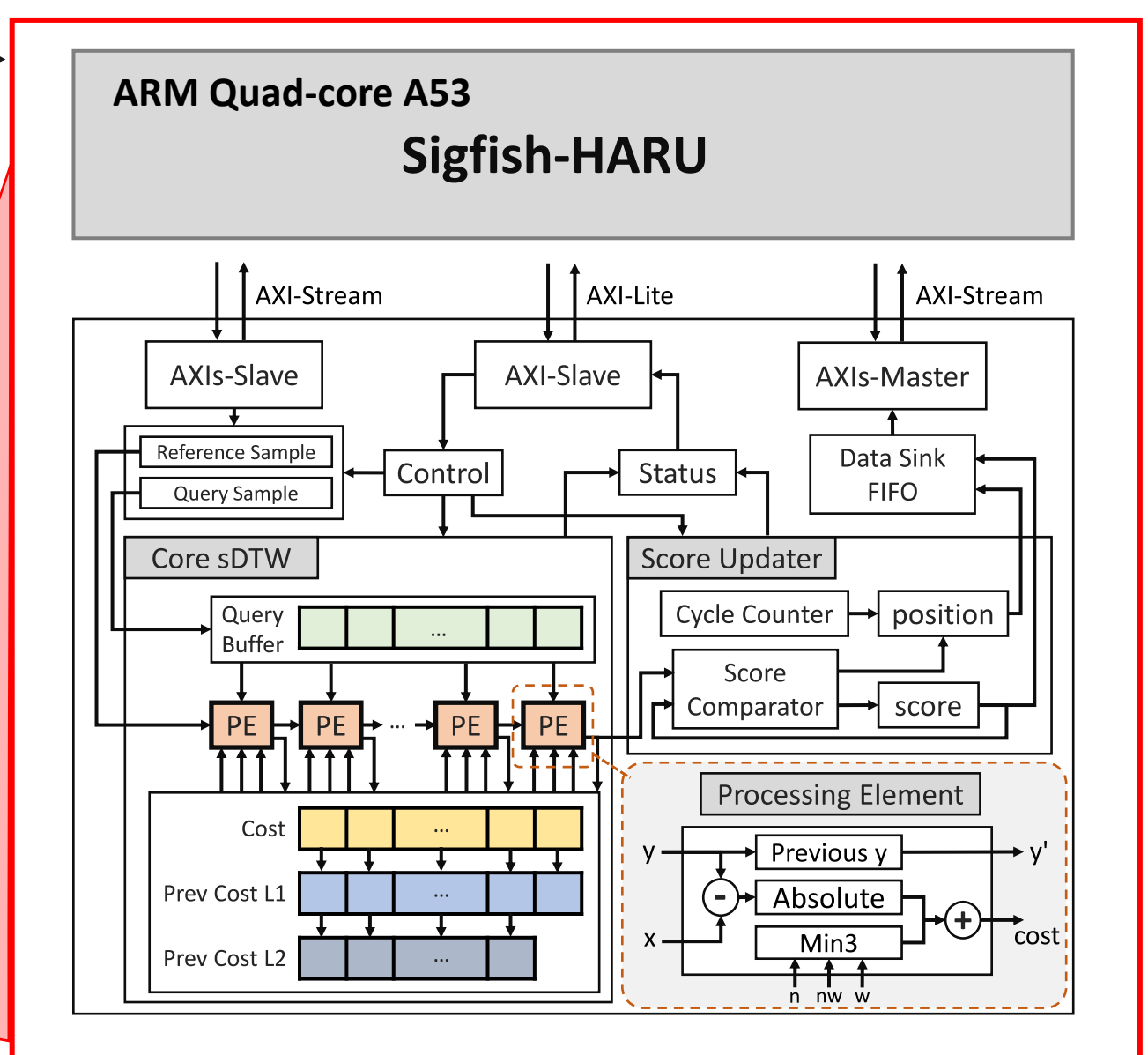
- Optimizations**
- Column cells computed in parallel (via pipelining)
 - fixed-point data & tuned scaling factor (reduce computation cost)
 - data reusing, no backtracing (keeping only necessary data of the DP memoization)
- Becomes a chain of Processing Elements (PEs) sliding through the reference (linear)

```

Algorithm 2: Memory-efficient subsequence DTW
Input: X[1..M], Y[1..N], M, N
Output: position, score
1. C: array of size M+1 initialised to ∞;
2. score ← -∞
3. position ← -1
4. for j in range 1 to N do
5.   w ← c[j, 1]
6.   u ← c[1, j]
7.   m ← min(w, u)
8.   for i in range 2 to M do
9.     c[i, j] ← min(c[i-1, j-1], c[i-1, j], c[i, j-1]) + m
10.  end
11. end
12. min_score ← c[M, N]
13. position ← j
14. score ← c[M, N]
15. end
  
```



AMD Kria AI Starter Kit (249 USD)



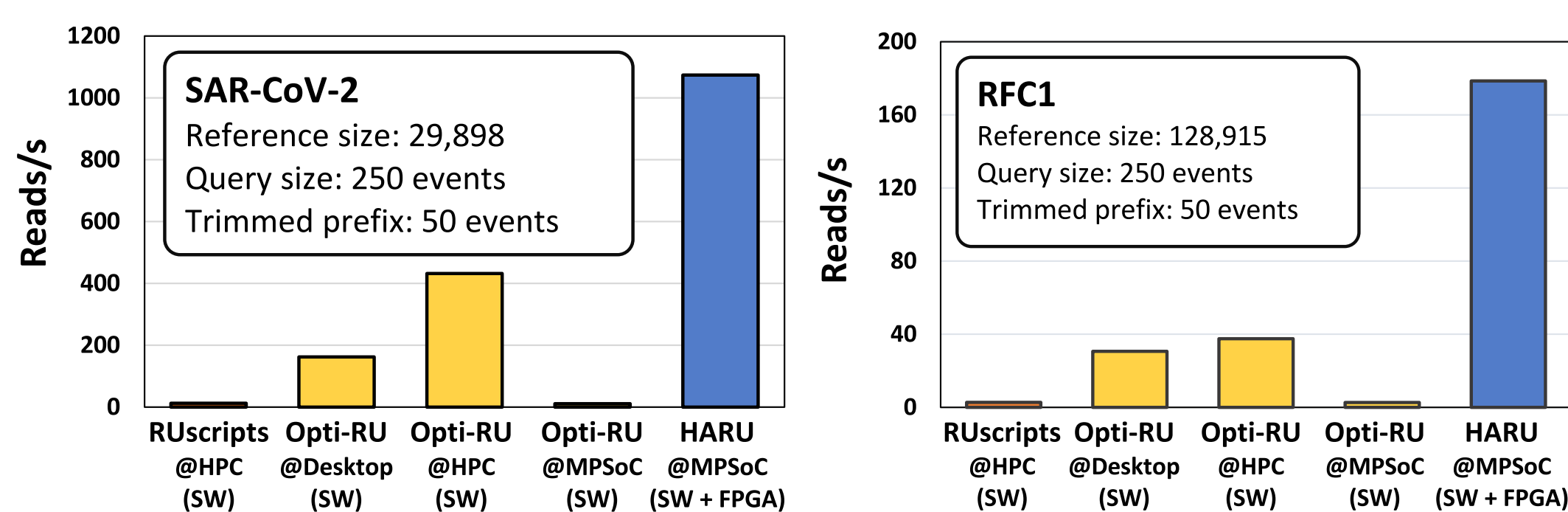
Scan to access paper and codebase



<https://doi.org/10.1093/gigascience/giad046>

Results and Outcomes

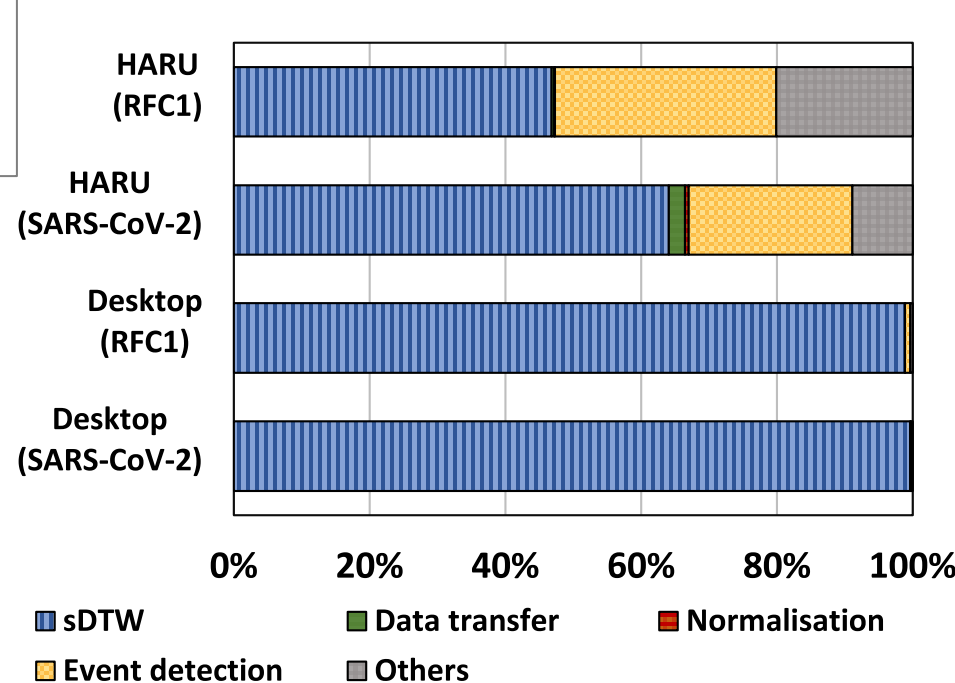
Processing capability (HARU vs sDTW implementations)



Desktop - Intel Core i9-10850K (10 cores, 32GB RAM)
HPC - Intel Xeon Gold 6154 (36 cores, 377GB RAM)
MPSoC - ARM Cortex A53 (4 cores, 4GB RAM)

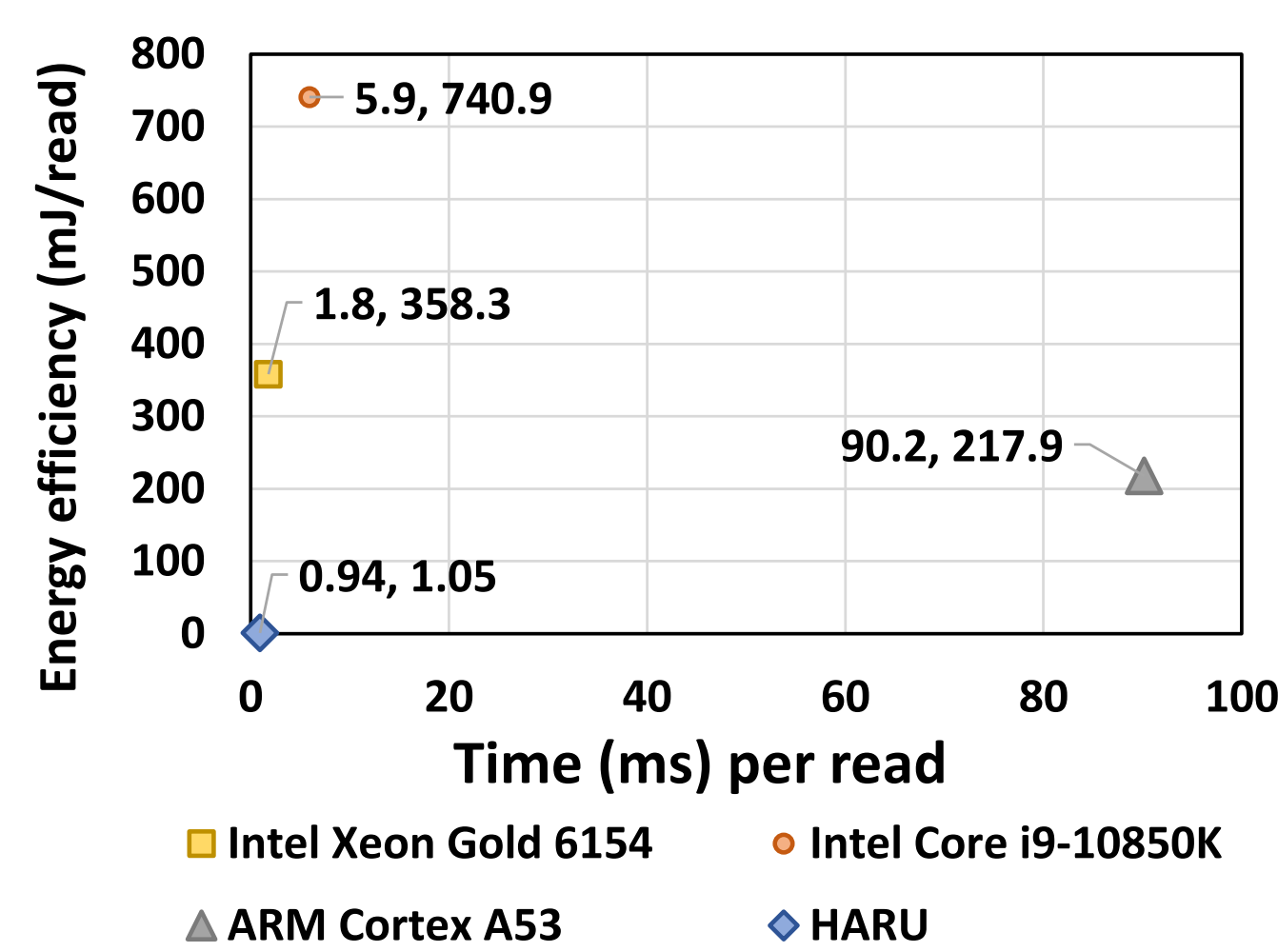
Speedups (HARU vs):

- RUScripts [1] @HPC → 85.8 x
- Optimized RUScripts @Desktop → 6.6 x
- Optimized RUScripts @HPC → 2.5 x



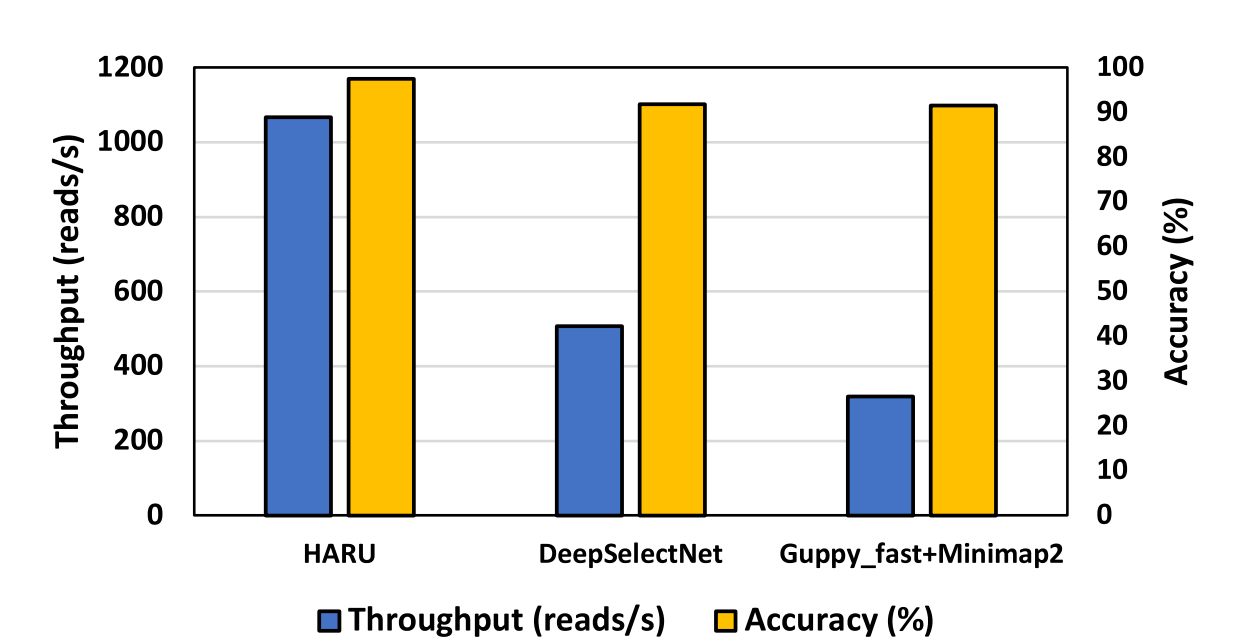
Smaller, cheaper, but higher throughput!

Energy-delay product (HARU vs sDTW implementations)



Significant improvement in energy efficiency!

Processing capability & Accuracy (HARU vs other SOTA implementations)



DeepSelectNet [2] running on HPC + Tesla V100 GPU
Guppy_fast + Minimap2 [3] running on Nvidia Jetson Xavier

The reduction of processing requirements enables in-field selective sequencing! + Fully working system!



[1] Loose, M., Malla, S., & Stout, M. (2016). Real-time selective sequencing using nanopore technology. *Nature methods*, 13(9), 751-754.

[2] Senanayake, A., Gamaarachchi, H., Herath, D., & Ragel, R. (2023). DeepSelectNet: deep neural network based selective sequencing for oxford nanopore sequencing. *BMC bioinformatics*, 24(1), 31.

[3] Li, H. (2018). Minimap2: pairwise alignment for nucleotide sequences. *Bioinformatics*, 34(18), 3094-3100.

