

Accelerating an Iterative Helmholtz Solver Using Reconfigurable Hardware

Art Petrenko

M.Sc. Defence, April 9, 2014

Seismic Laboratory for Imaging and Modelling

Department of Earth, Ocean and Atmospheric Sciences, UBC



University of British Columbia

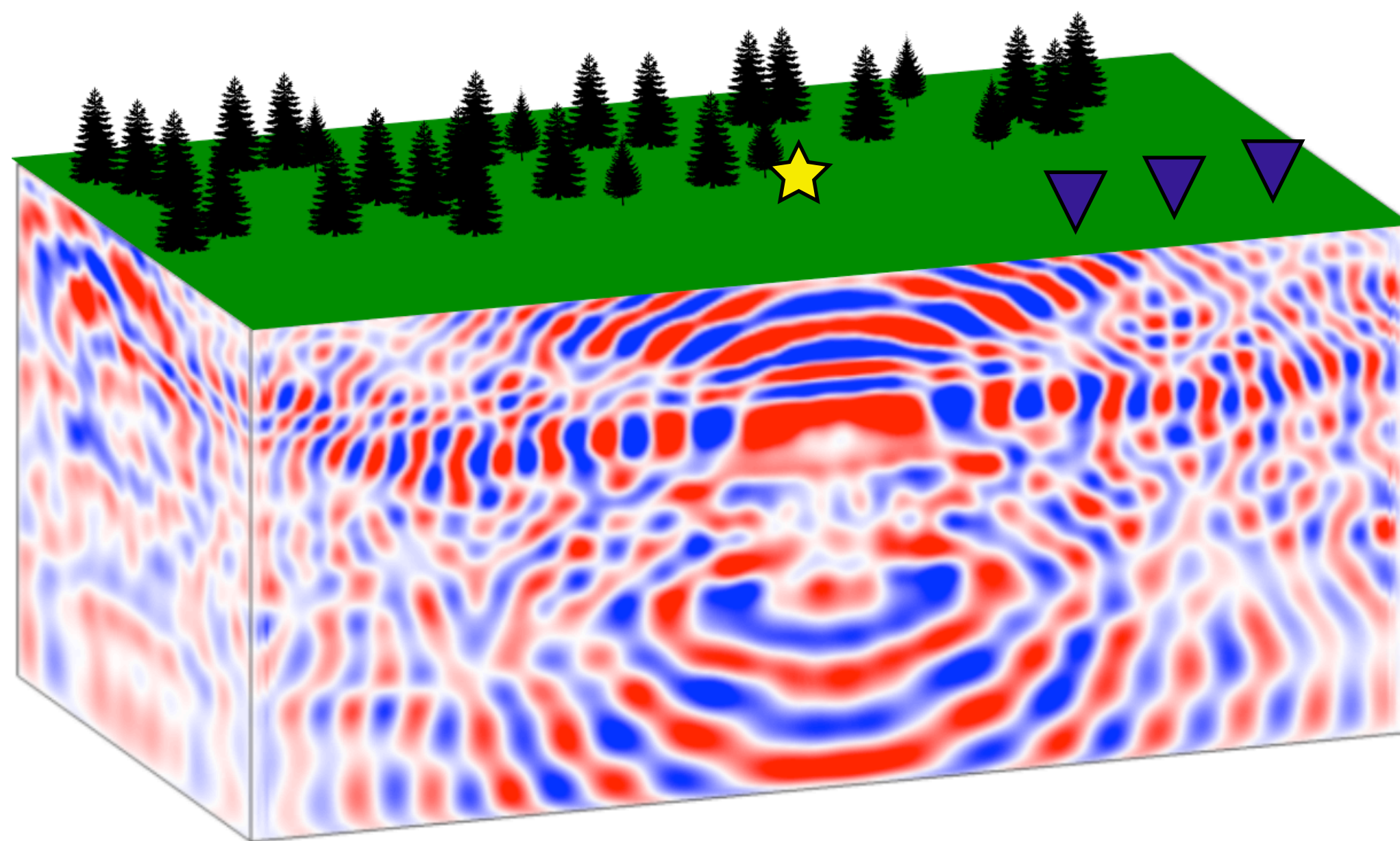
Oh by the way: I have a stutter.



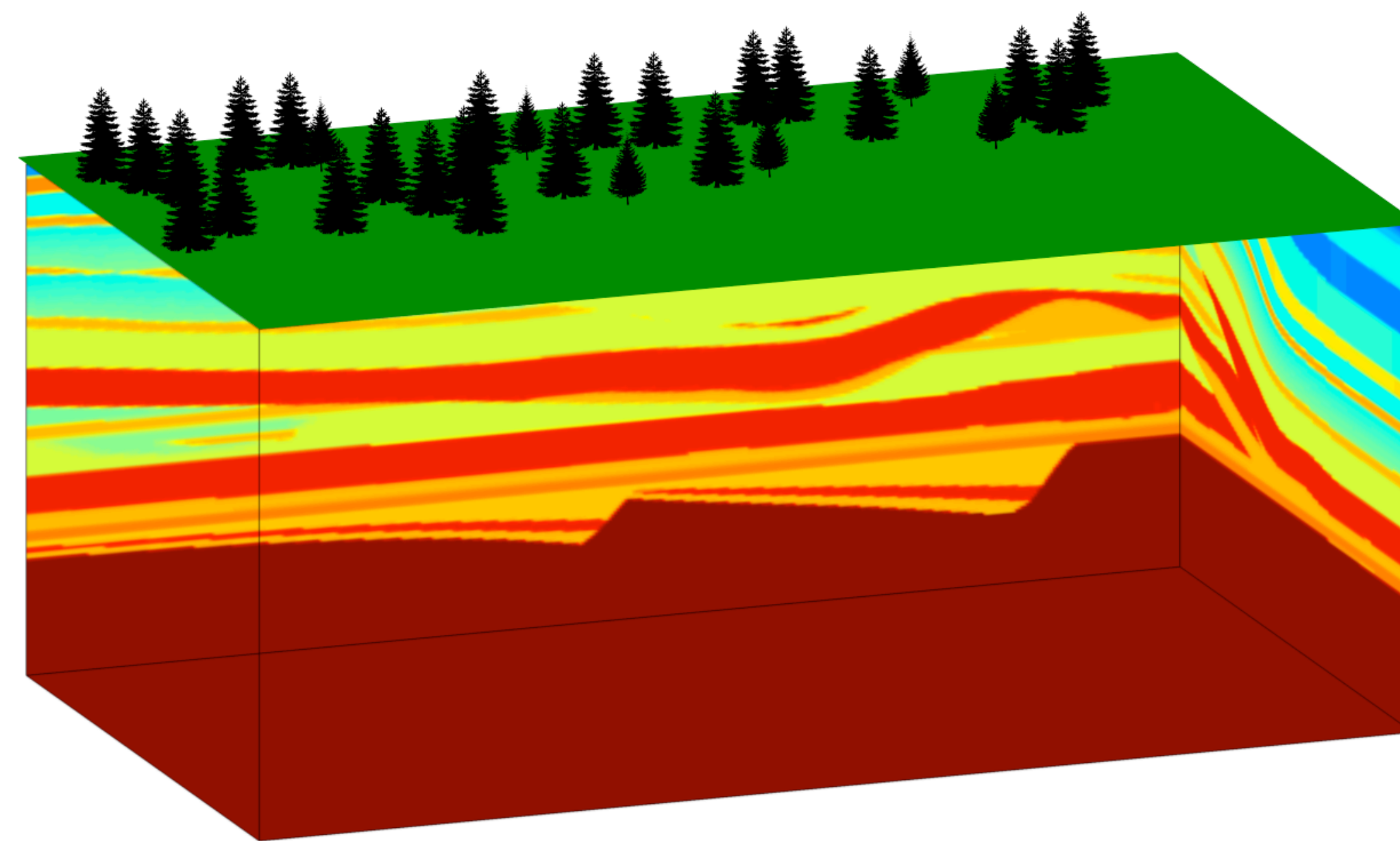
Seismic Wave Simulation

Seismic Exploration for Oil and Gas

Full-waveform Inversion



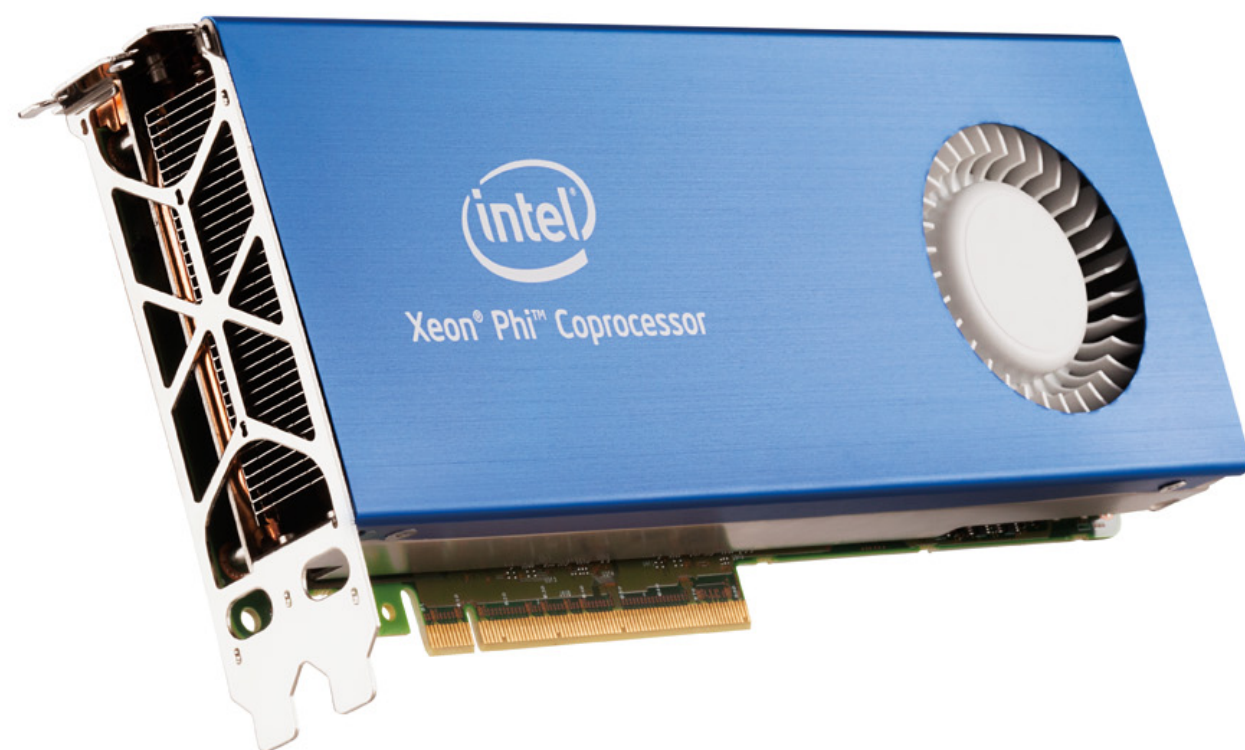
Seismic Wavefield (\mathbf{u})



Earth model (\mathbf{m})

Full-waveform inversion is SLOW

The Accelerators Have Arrived



Top 10 of "Top 500" Supercomputers

Rank	Site	System	Cores	Rmax (TFlop/s)	Rpeak (TFlop/s)	Power (kW)
1	National Super Computer Center in Guangzhou China	Tianhe-2 (MilkyWay-2) - TH-IVB-FEP Cluster, Intel Xeon E5-2692 12C 2.200GHz, TH Express-2, Intel Xeon Phi 31S1P NUDT	3,120,000	33,862.7	54,902.4	17,808
2	DOE/SC/Oak Ridge National Laboratory United States	Titan - Cray XK7 , Opteron 6274 16C 2.200GHz, Cray Gemini interconnect, NVIDIA K20x Cray Inc.	560,640	17,590.0	27,112.5	8,209
3	DOE/NNSA/LLNL United States	Sequoia - BlueGene/Q, Power BQC 16C 1.60 GHz, Custom IBM	1,572,864	17,173.2	20,132.7	7,890
4	RIKEN Advanced Institute for Computational Science (AICS) Japan	K computer , SPARC64 VIIIfx 2.0GHz, Tofu interconnect Fujitsu	705,024	10,510.0	11,280.4	12,660
5	DOE/SC/Argonne National Laboratory United States	Mira - BlueGene/Q, Power BQC 16C 1.60GHz, Custom IBM	786,432	8,586.6	10,066.3	3,945
6	Swiss National Supercomputing Centre (CSCS) Switzerland	Piz Daint - Cray XC30, Xeon E5-2670 8C 2.600GHz, Aries interconnect , NVIDIA K20x Cray Inc.	115,984	6,271.0	7,788.9	2,325
7	Texas Advanced Computing Center/Univ. of Texas United States	Stampede - PowerEdge C8220, Xeon E5-2680 8C 2.700GHz, Infiniband FDR, Intel Xeon Phi SE10P Dell	462,462	5,168.1	8,520.1	4,510
8	Forschungszentrum Juelich (FZJ) Germany	JUQUEEN - BlueGene/Q, Power BQC 16C 1.600GHz, Custom Interconnect IBM	458,752	5,008.9	5,872.0	2,301
9	DOE/NNSA/LLNL United States	Vulcan - BlueGene/Q, Power BQC 16C 1.600GHz, Custom Interconnect IBM	393,216	4,293.3	5,033.2	1,972
10	Leibniz Rechenzentrum Germany	SuperMUC - iDataPlex DX360M4, Xeon E5-2680 8C 2.70GHz, Infiniband FDR IBM	147,456	2,897.0	3,185.1	3,423

FPGAs: Reconfigurable Hardware Accelerators

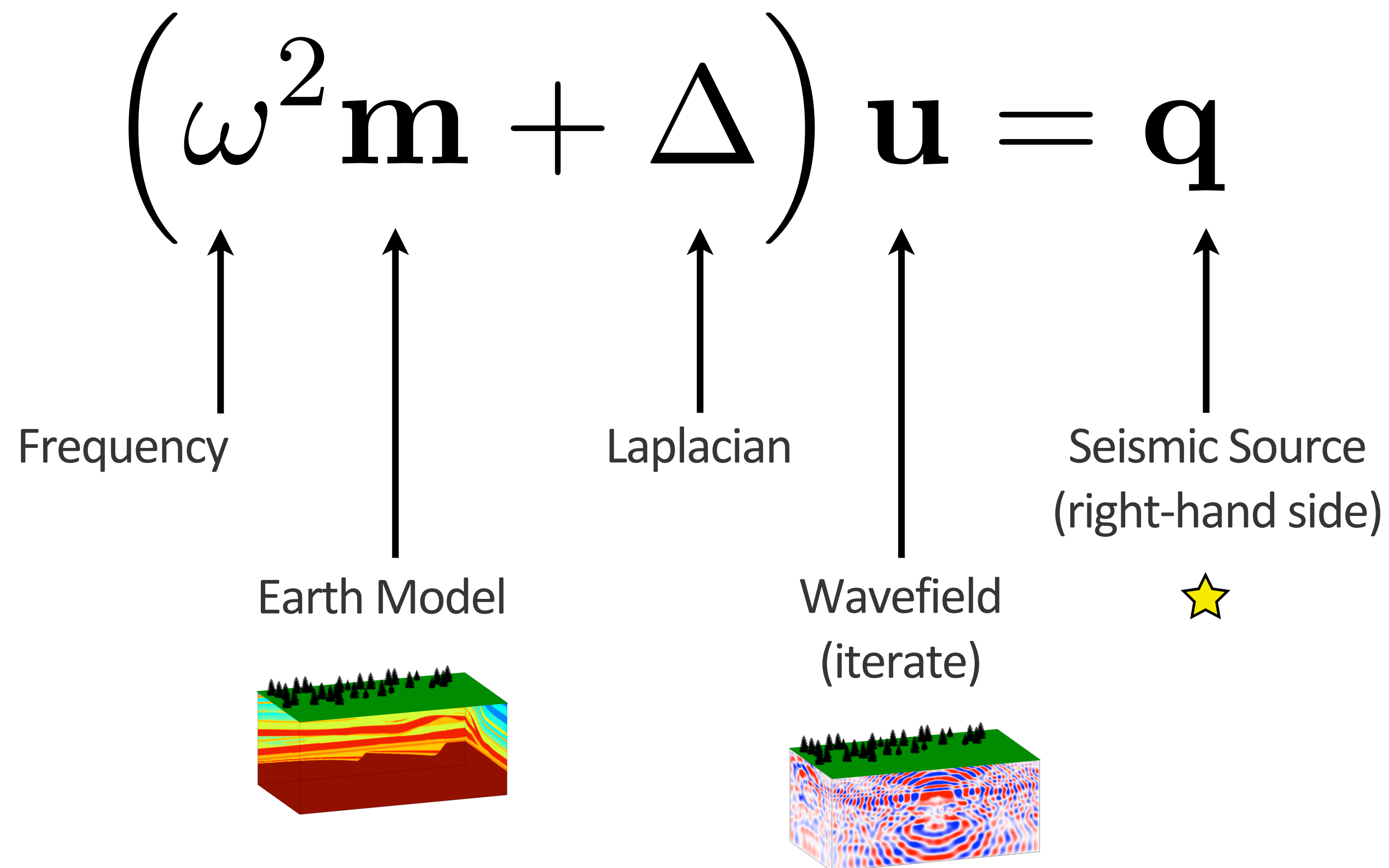


The Punchline

Modelling Seismic Waves

Mathematical Formulation

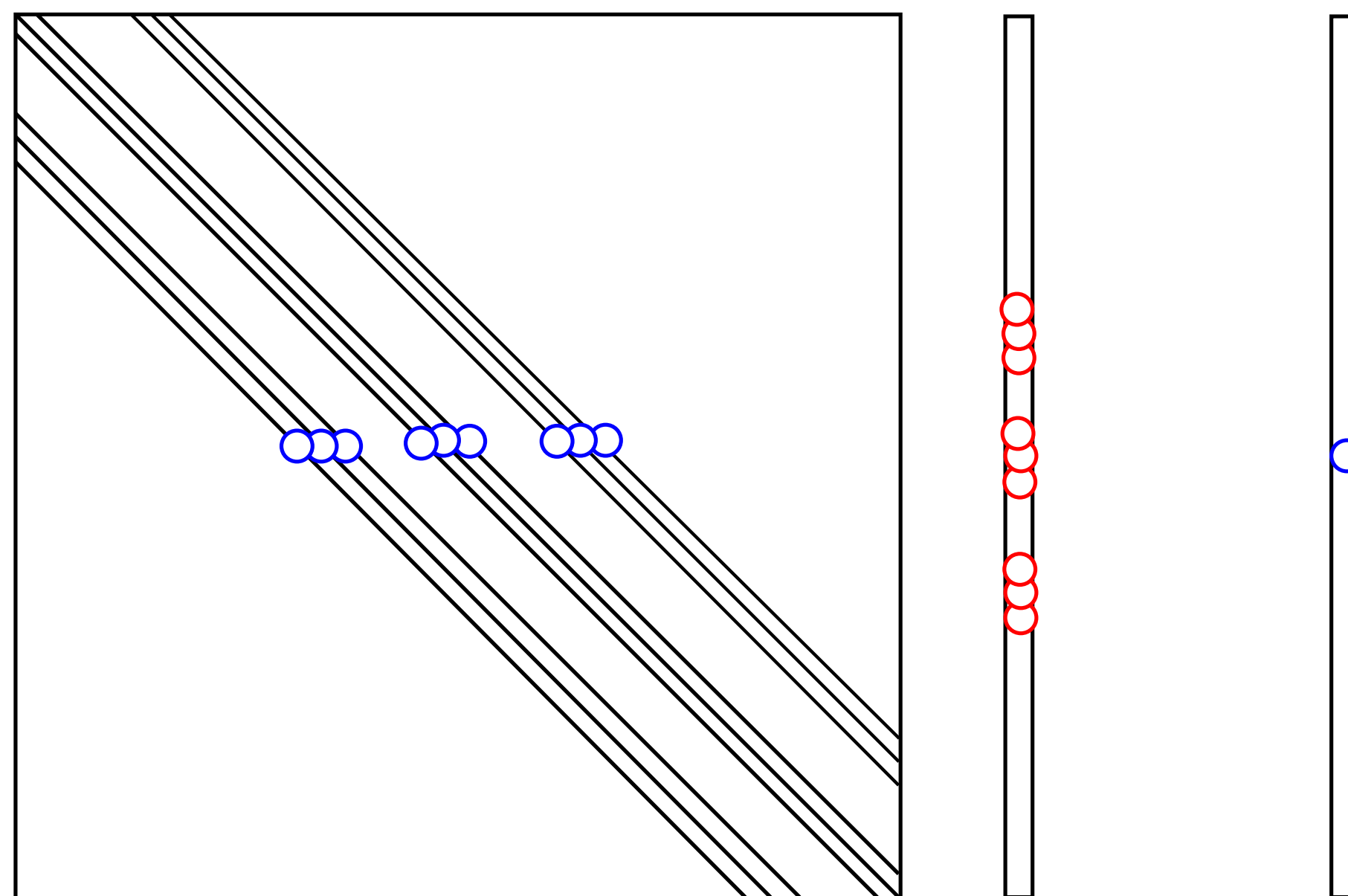
Modelling Seismic Waves: The Wave Equation



Modelling Seismic Waves: Discretization

[Operto, 2007]

$$A(\mathbf{m}, \omega) \mathbf{u} = \mathbf{q}$$

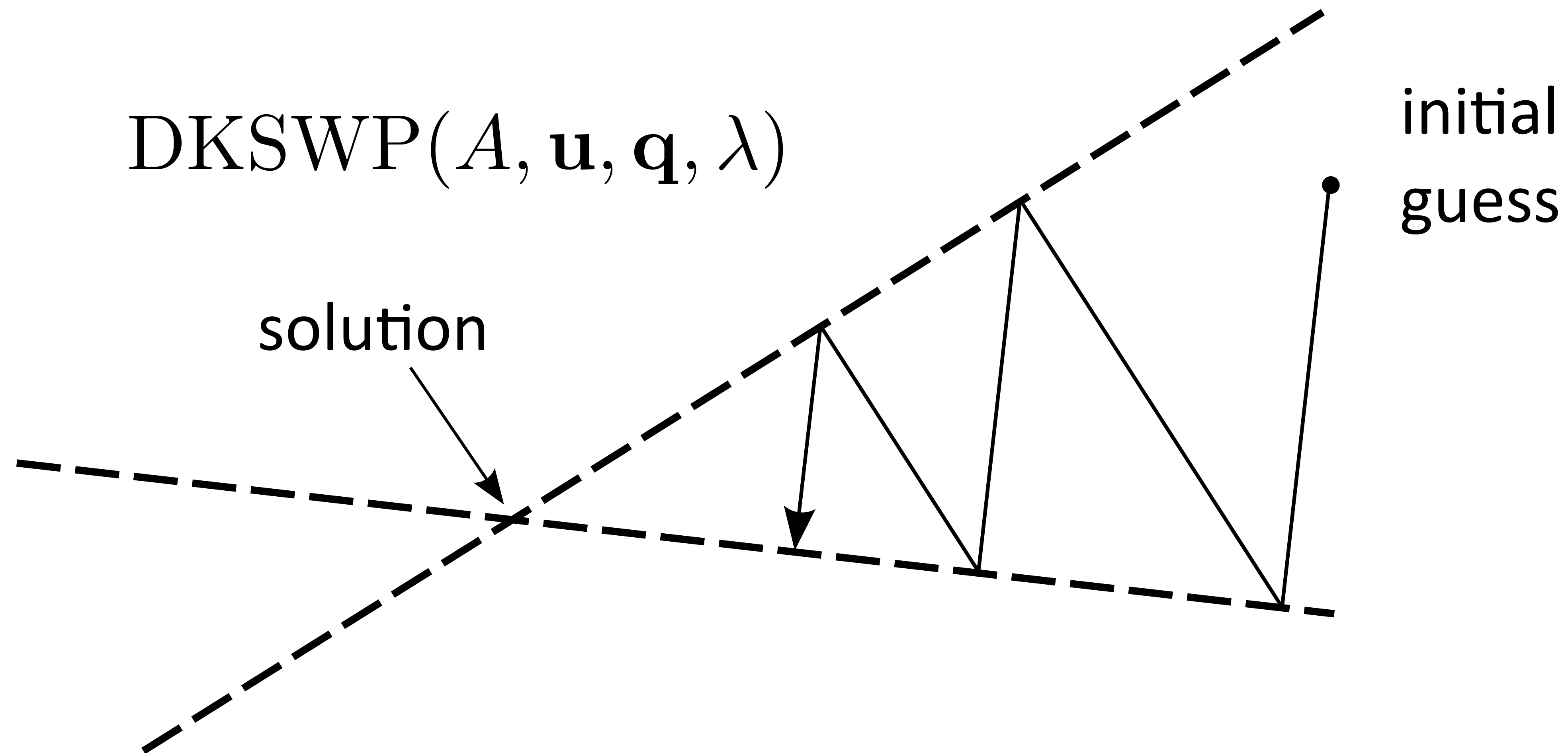


Solving the Helmholtz System

The Kaczmarz Algorithm

[Kaczmarz, 1937]

$$\text{DKSWP}(A, \mathbf{u}, \mathbf{q}, \lambda)$$



Adapted from [van Leeuwen, 2012]

The Kaczmarz Algorithm: Equivalent to SSOR-NE

[Björck and Elfving, 1979]

Double Kaczmarz sweep
on the original system:

$$A\mathbf{u} = \mathbf{q}$$



One iteration of SSOR on
the normal equations:

$$AA^*\mathbf{y} = \mathbf{q}$$

$$A^*\mathbf{y} = \mathbf{u}$$

Both are computed as:

$$\mathbf{u}_{k+1} = \mathbf{u}_k + \lambda(b_i - \langle \mathbf{a}_i, \mathbf{u}_k \rangle) \frac{\mathbf{a}_i^*}{\|\mathbf{a}_i\|^2}$$

$$k : 1 \rightarrow 2N$$

$$i : 1 \rightarrow N, N \rightarrow 1$$

Kaczmarz + CG = CGMN
[Björck & Elfving 1979]

CGMN: Solves for Fixed Point of Kaczmarz Row Projections

$$\begin{aligned} \text{DKSWP}(A, \mathbf{u}, \mathbf{q}, \lambda) &= Q_1 \cdots Q_N Q_N \cdots Q_1 \mathbf{u} + R\mathbf{q} \\ &= Q\mathbf{u} + R\mathbf{q}. \end{aligned}$$

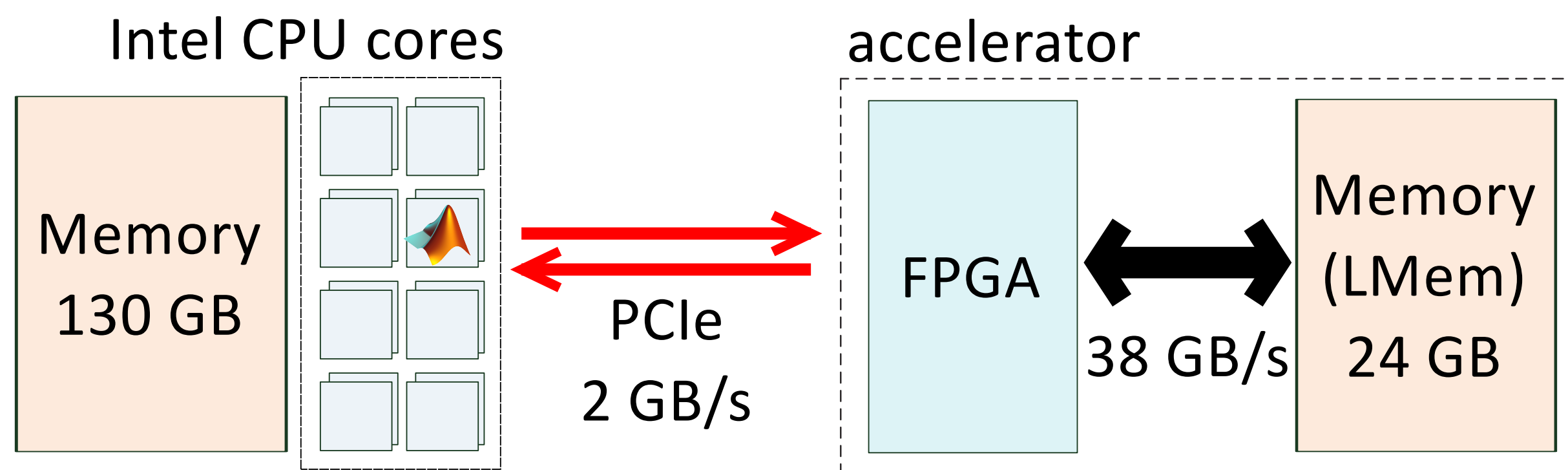
Assume \mathbf{u} is a solution and re-arrange:

$$(I - Q)\mathbf{u} = R\mathbf{q}$$

Contribution of This Work

Compute Node Overview

[Maxeler Technologies, 2011]



Adapted from [Pell, 2013]

Algorithm 1 CGMN (Björck and Elfving [4])

Input: A, u, q, λ

- 1: $Rq \leftarrow \text{DKSWP}(A, \mathbf{0}, q, \lambda)$
- 2: $\mathbf{r} \leftarrow Rq - u + \text{DKSWP}(A, u, \mathbf{0}, \lambda)$
- 3: $\mathbf{p} \leftarrow \mathbf{r}$
- 4: **while** $\|\mathbf{r}\|^2 > tol$ **do**
- 5: $\mathbf{s} \leftarrow (I - Q)\mathbf{p} = \mathbf{p} - \text{DKSWP}(A, \mathbf{p}, \mathbf{0}, \lambda)$
- 6: $\alpha \leftarrow \|\mathbf{r}\|^2 / \langle \mathbf{p}, \mathbf{s} \rangle$
- 7: $\mathbf{u} \leftarrow \mathbf{u} + \alpha \mathbf{p}$
- 8: $\mathbf{r} \leftarrow \mathbf{r} - \alpha \mathbf{s}$
- 9: $\beta \leftarrow \|\mathbf{r}\|_{\text{curr}}^2 / \|\mathbf{r}\|_{\text{prev}}^2$
- 10: $\|\mathbf{r}\|_{\text{prev}}^2 \leftarrow \|\mathbf{r}\|_{\text{curr}}^2$
- 11: $\mathbf{p} \leftarrow \mathbf{r} + \beta \mathbf{p}$
- 12: **end while**

Kernel: running on
accelerator

Output: u



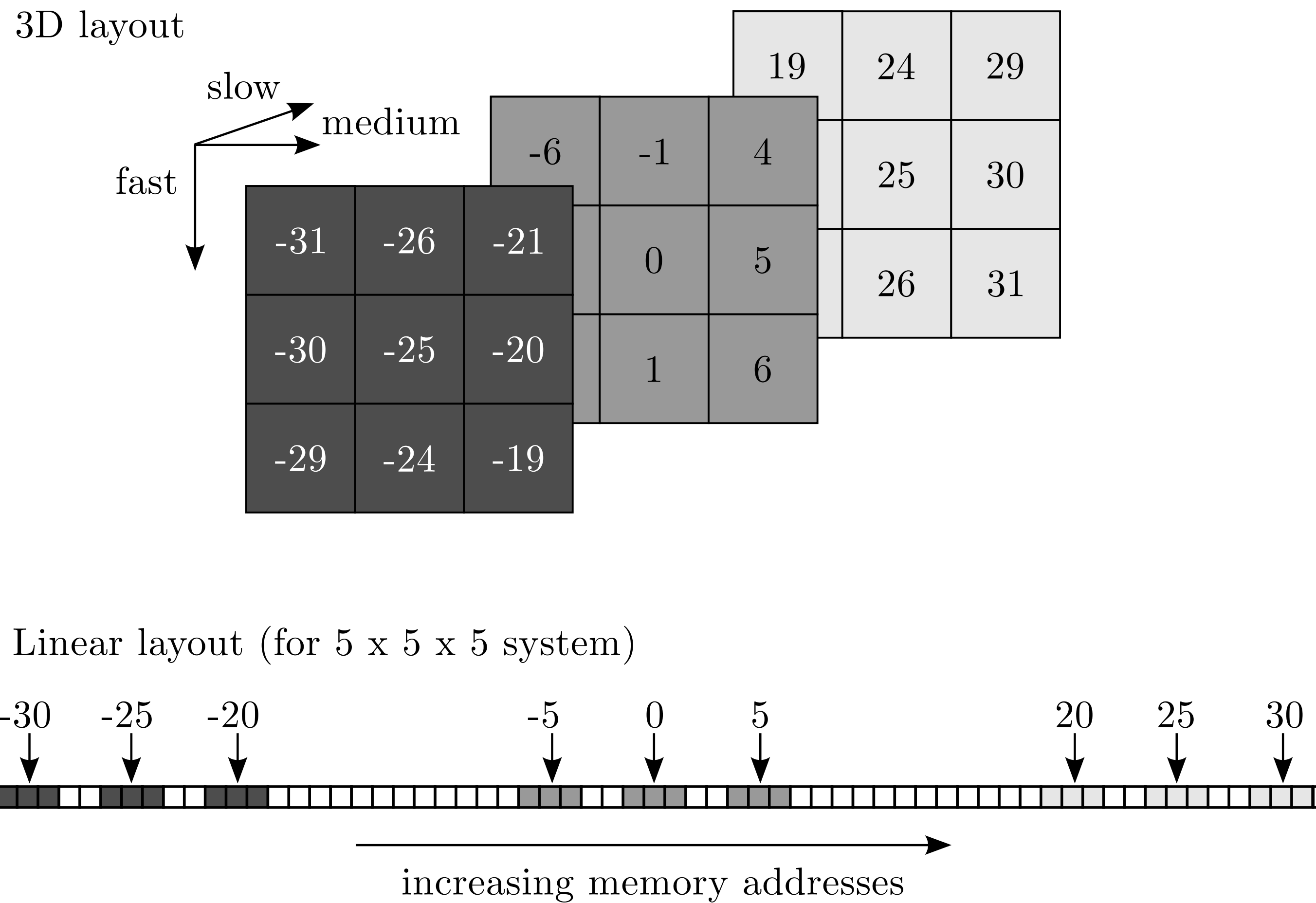
Low levels of abstraction are scary

Design at high level of abstraction

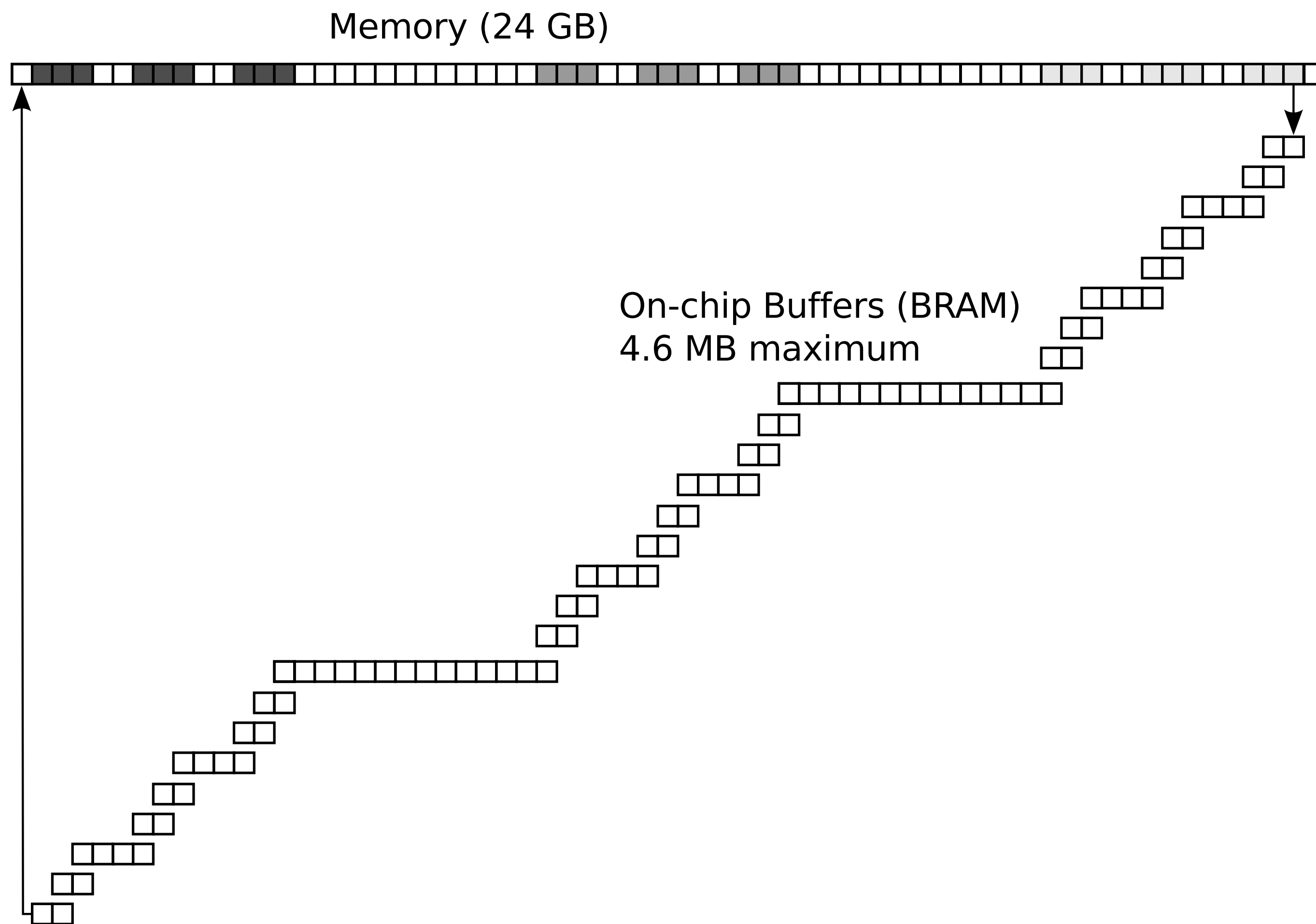
```
222         ((x[2]*R[24] + x[1]*R[25]) +
223          x[0]*R[26]));
224     DFEScalar relaxationFactor = io.scalarInput("relaxationFactor", kaczmarzEngineCode.KaczmarzWriteLMemKernel.TruncatedFloatingPoint);
225     DFEScalar kaczmarz_numerator = computation_stage ? relaxationFactor*(b - dot_product) : 0;
226     DFEScalar[] R_conj = new DFEScalar[kaczmarzEngineCode.KaczmarzManager.array_size];
227     for(int j=0; j<kaczmarzEngineCode.KaczmarzManager.array_size; j++){
228         R_conj[j] = kaczmarzEngineCode.KaczmarzWriteLMemKernel.ComplexTruncatedFloatingPoint.newInstance(this);
229         R_conj[j].setReal(R[j].getReal());
230         R_conj[j].setImaginary(-R[j].getImaginary());
231     }
232     DFEScalar[] R_scaled = new DFEScalar[kaczmarzEngineCode.KaczmarzManager.array_size];
233     for(int j=0; j<kaczmarzEngineCode.KaczmarzManager.array_size; j++){
234         R_scaled[j] = kaczmarzEngineCode.KaczmarzWriteLMemKernel.ComplexTruncatedFloatingPoint.newInstance(this);
235         R_scaled[j] = kaczmarz_numerator*R_conj[j];
236     }
237     //DFEScalar[] x_updated = new DFEScalar[kaczmarzEngineCode.KaczmarzManager.array_size];
238     for(int j=0; j<kaczmarzEngineCode.KaczmarzManager.array_size; j++){
239         x_updated[j] <= x[j] + R_scaled[j];
240     }
```

Implementation Details

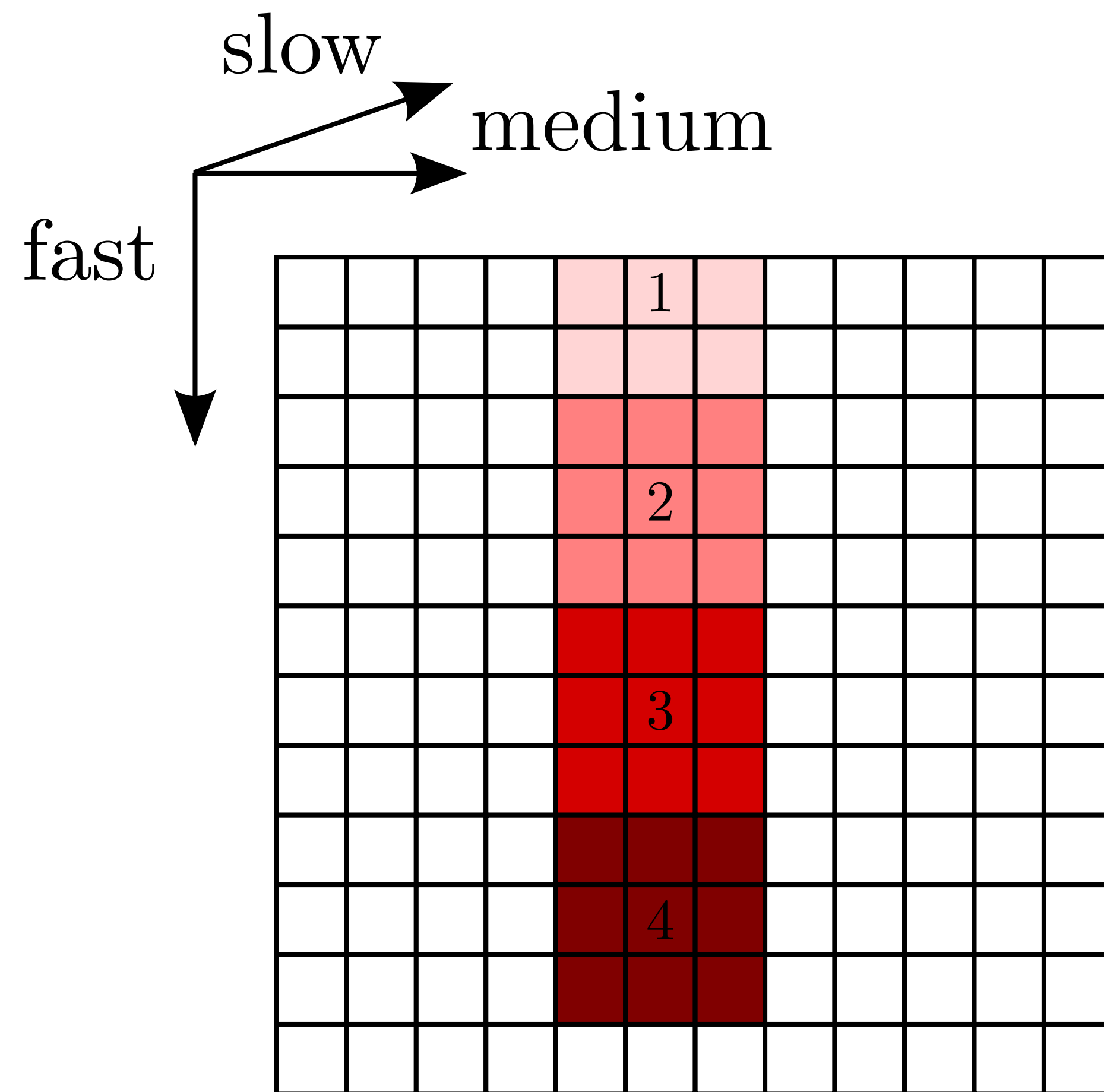
Layout of 3D Wavefields in 1D Memory



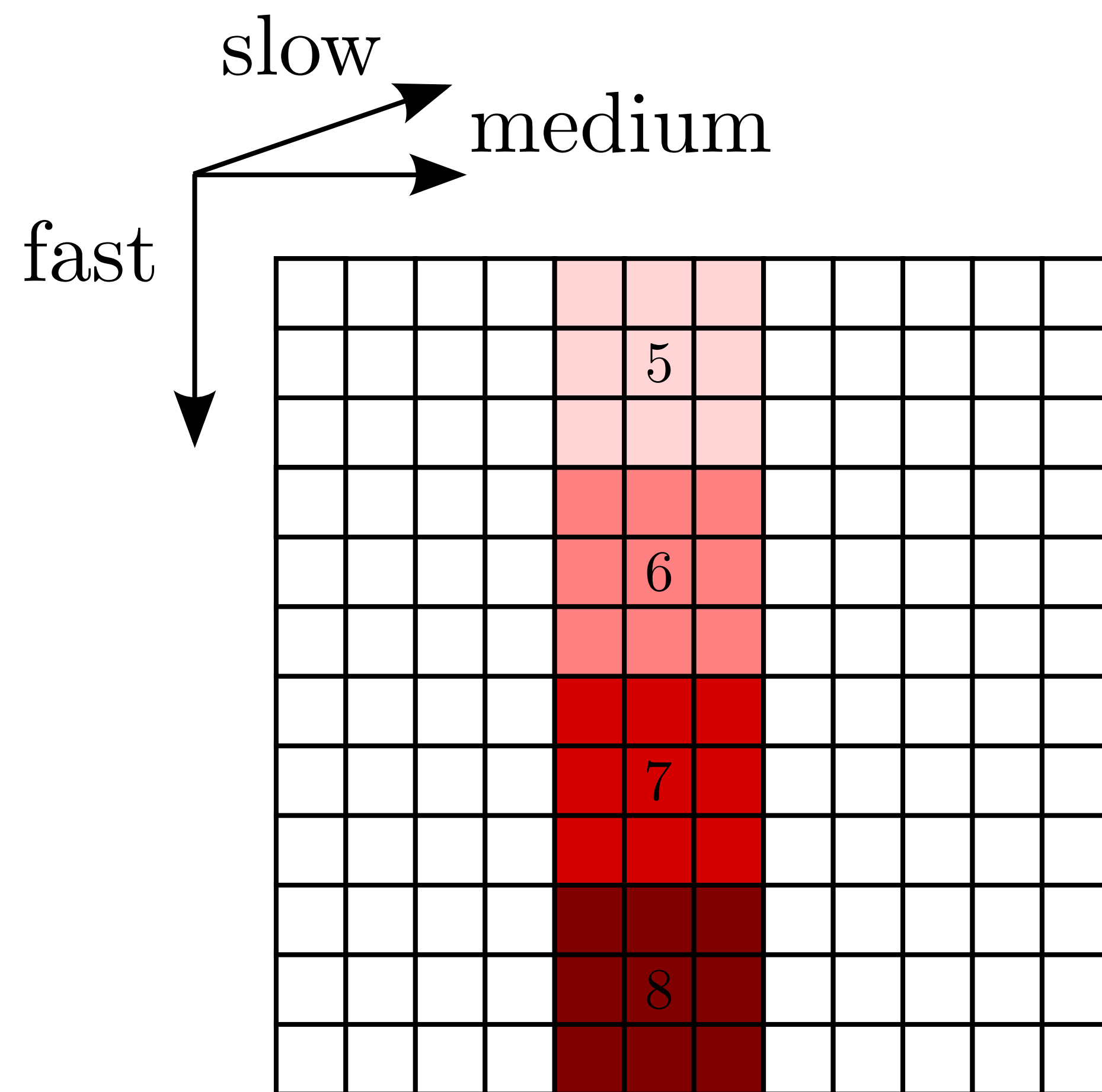
Buffering: Overcoming Latency of Memory Access



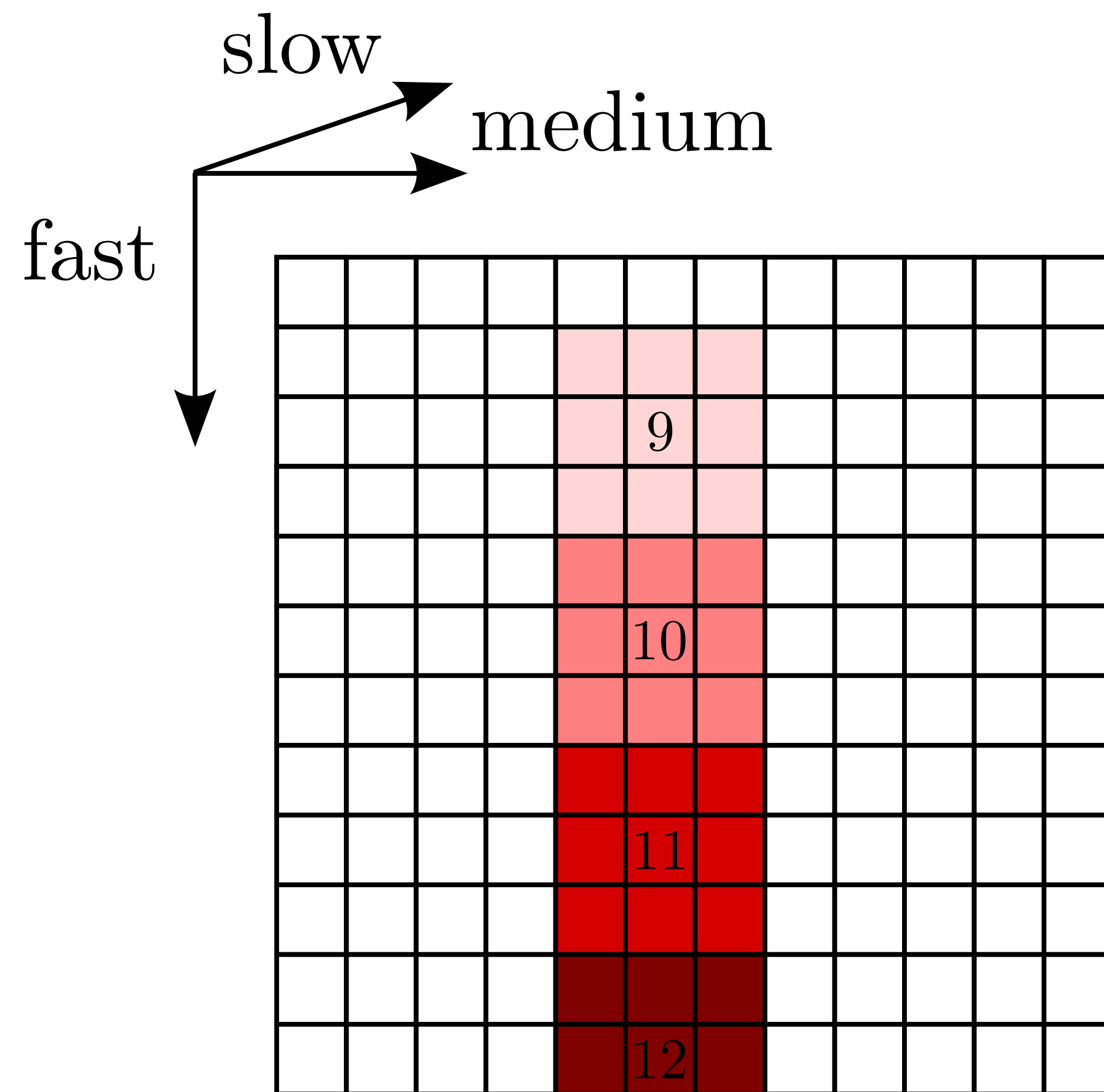
Pipelining: Overcoming Latency of Computation



Pipelining: Overcoming Latency of Computation



Pipelining: Overcoming Latency of Computation

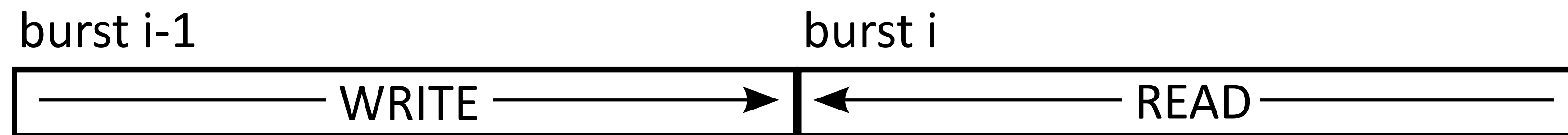


Memory Access: 384 bytes / burst

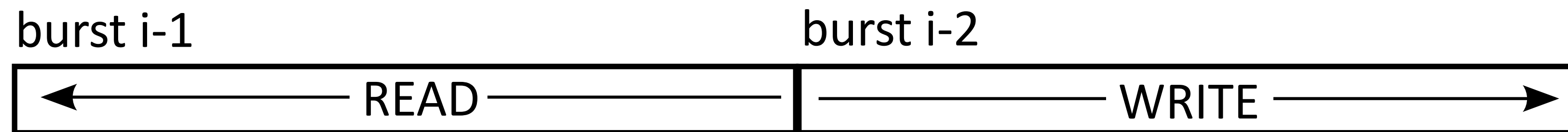
Number of bits in a real number	Number of bits in a complex number	Complex numbers per burst
24	48	64
32 (single precision)	64	48
48	96	32
64 (double precision)	128	24

Backward Sweep: Double Buffering

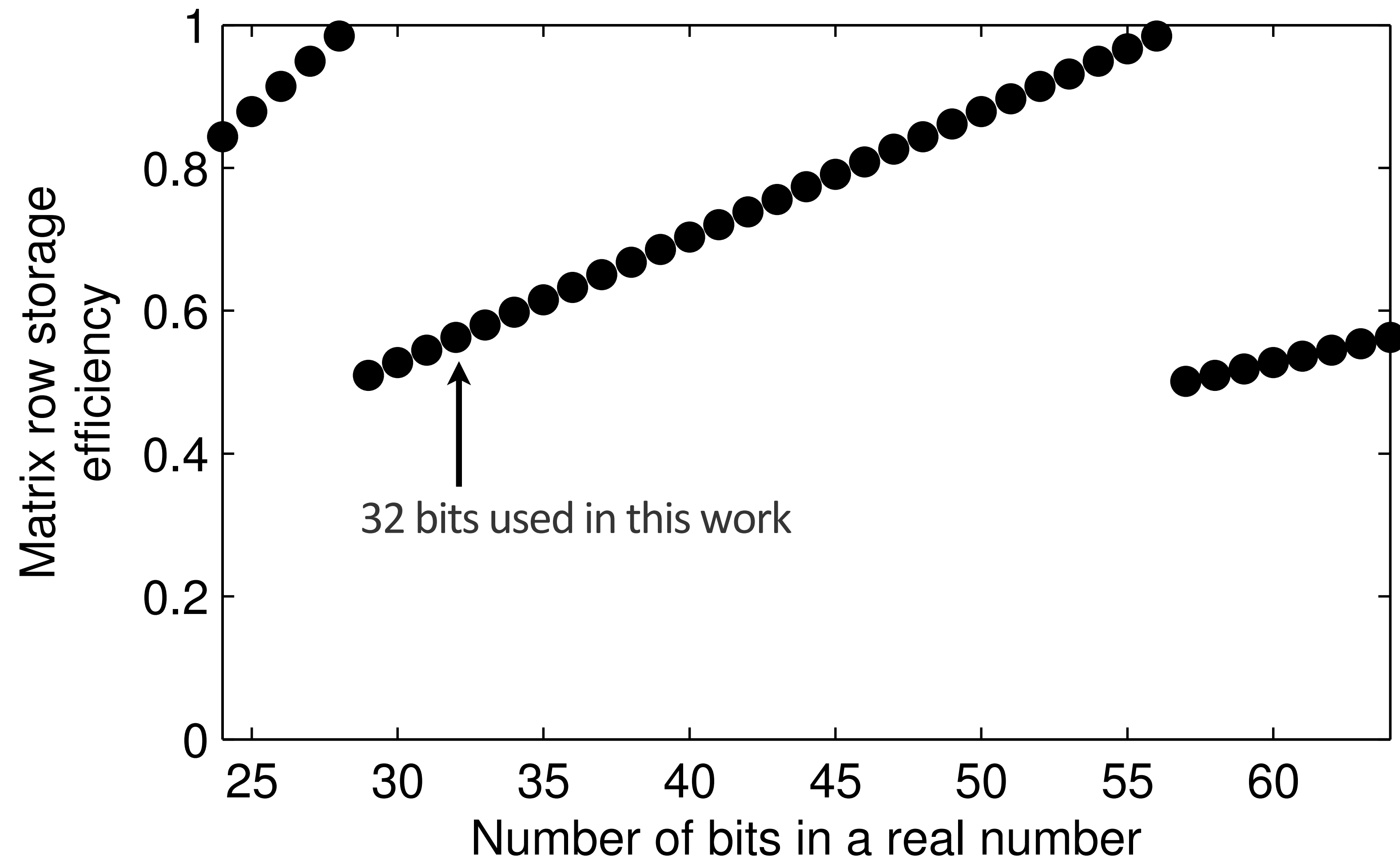
First 48 ticks



Second 48 ticks

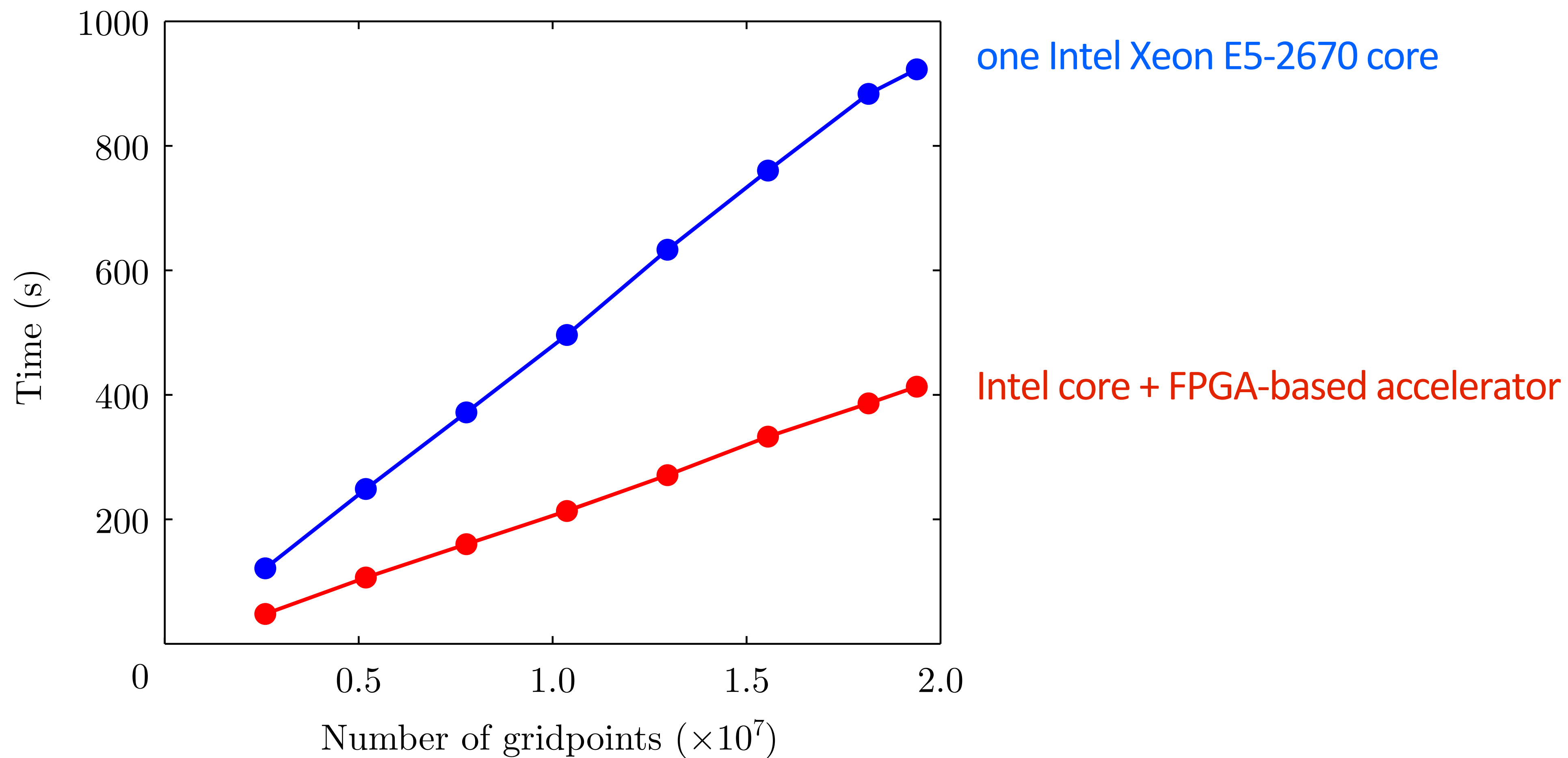


Number Representation

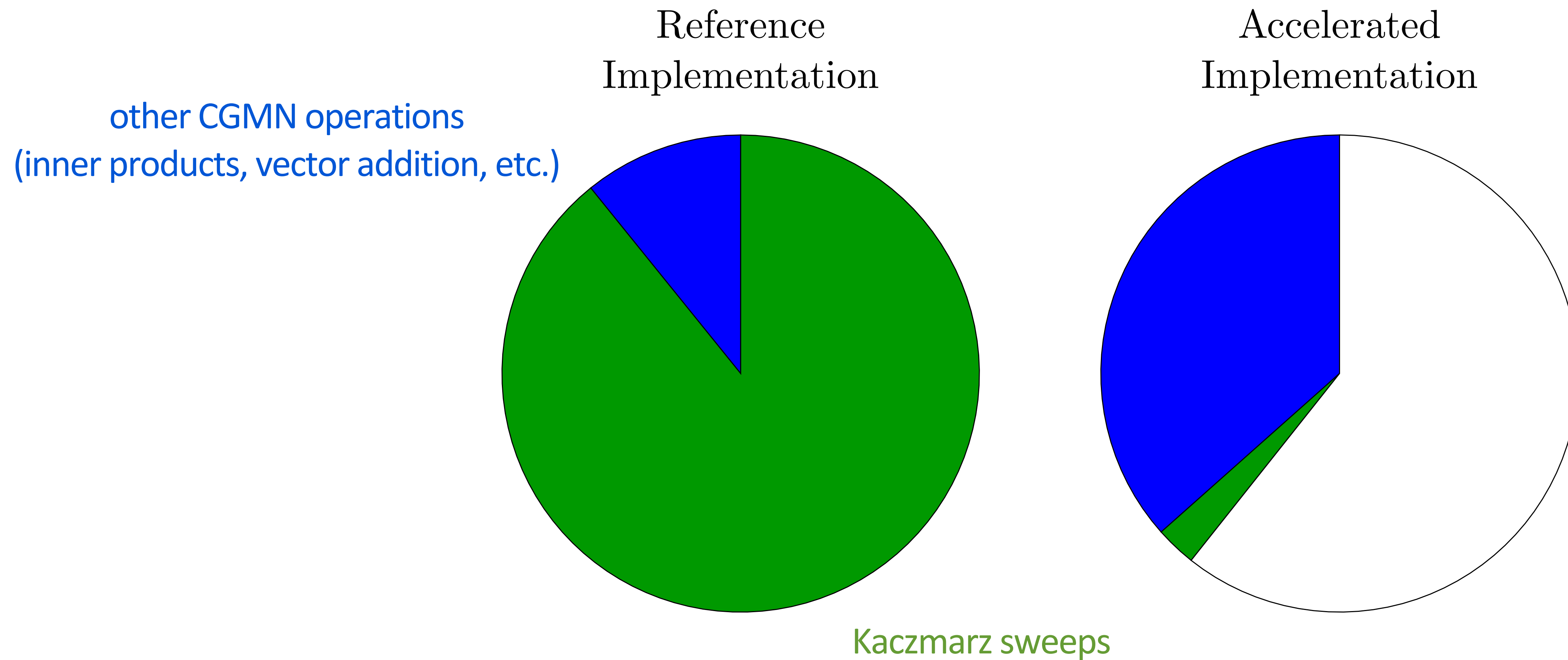


Results

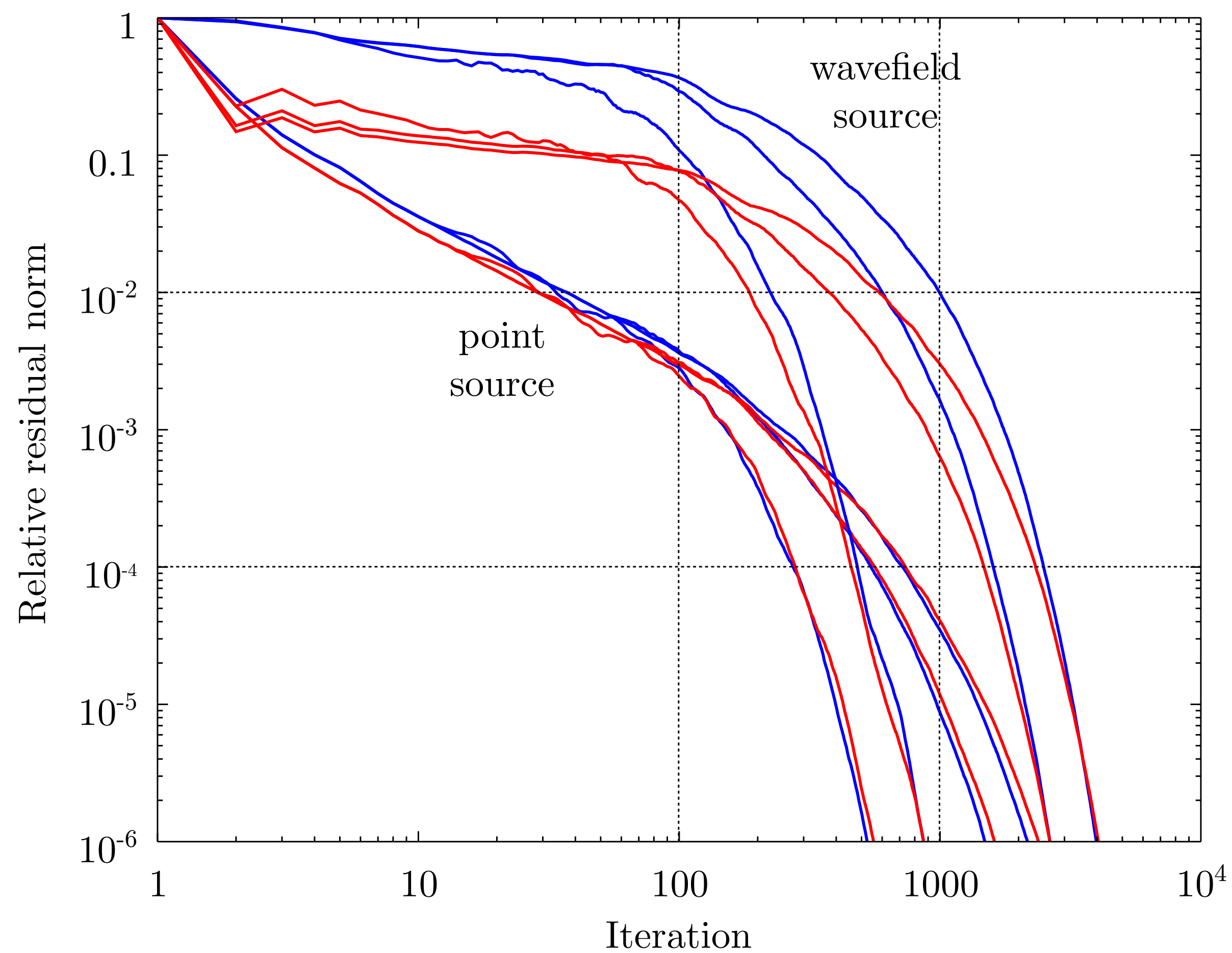
End-to-end Execution Time



Kaczmarz Sweeps: No Longer the Bottleneck



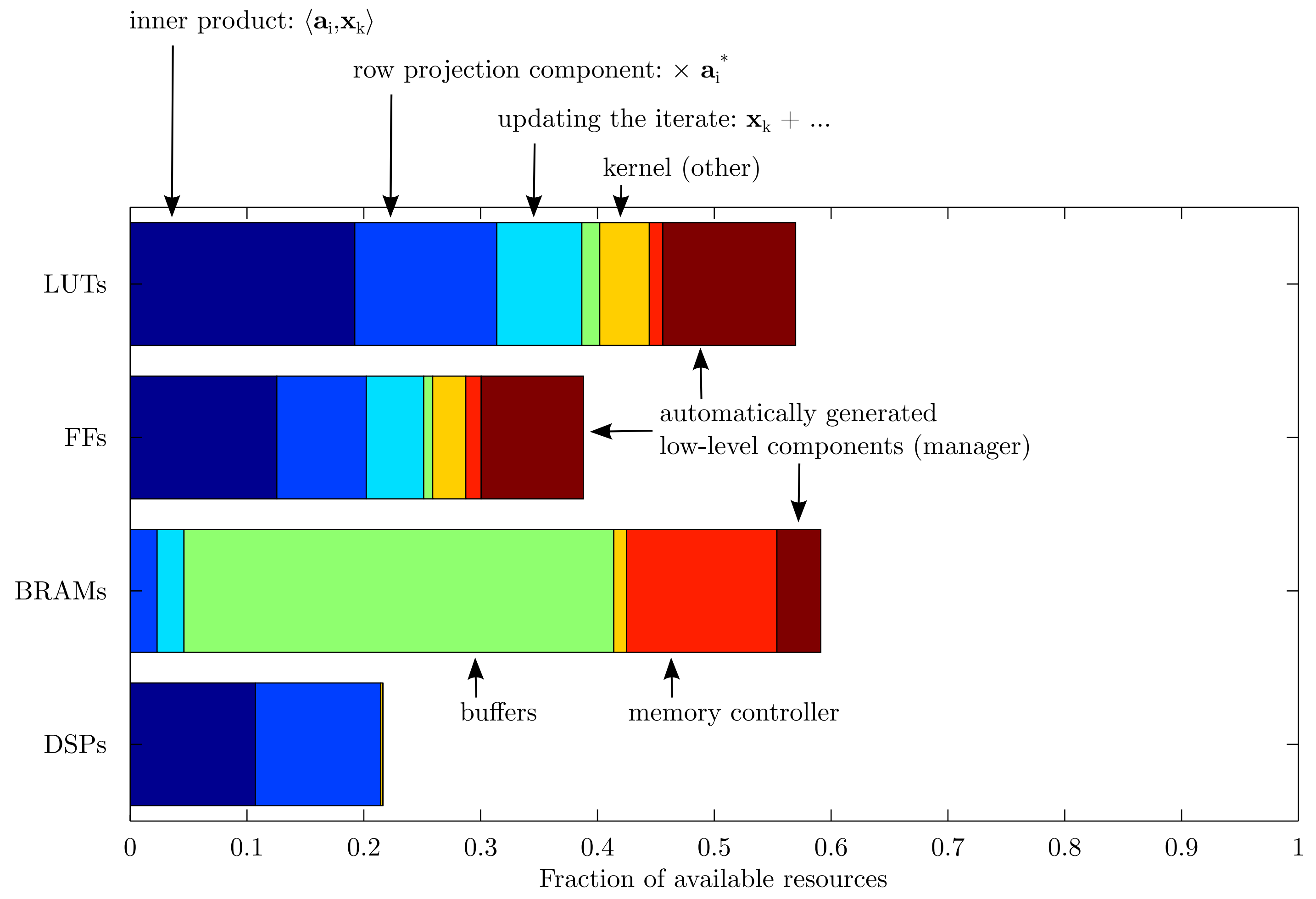
Effect of matrix row ordering on CGMN convergence



Sequential (1 to N)

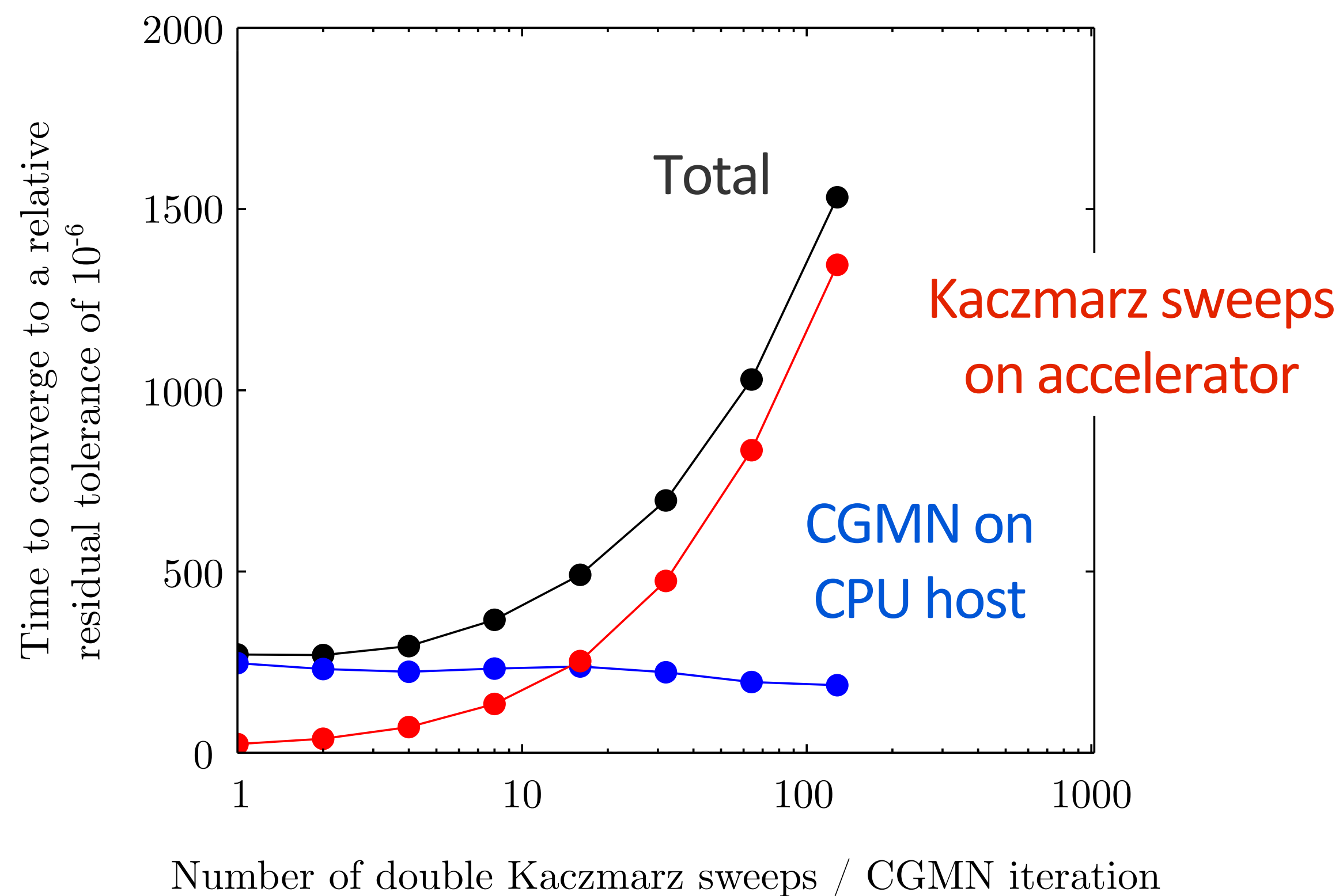
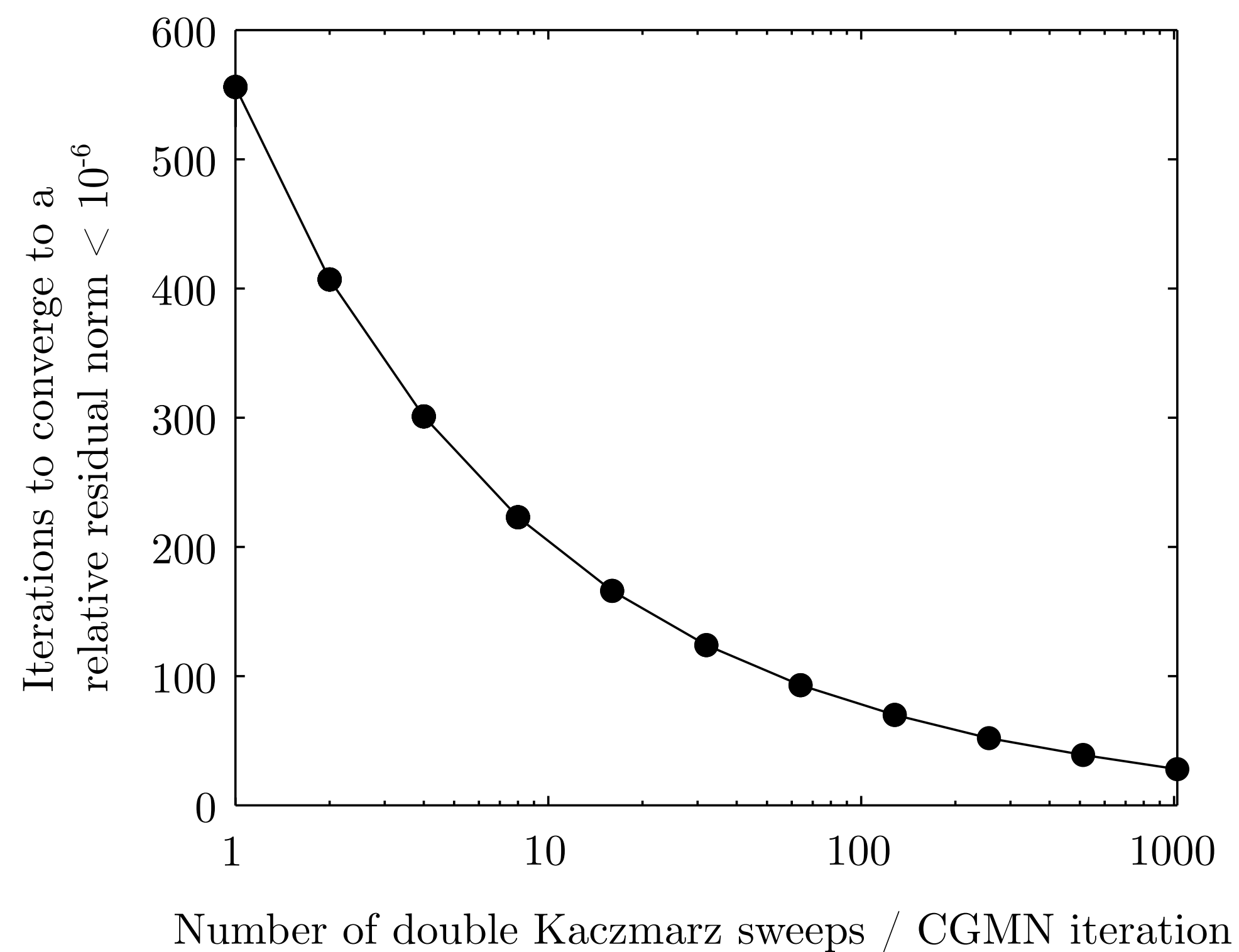
"Accelerator ordering"

FPGA Resource Usage

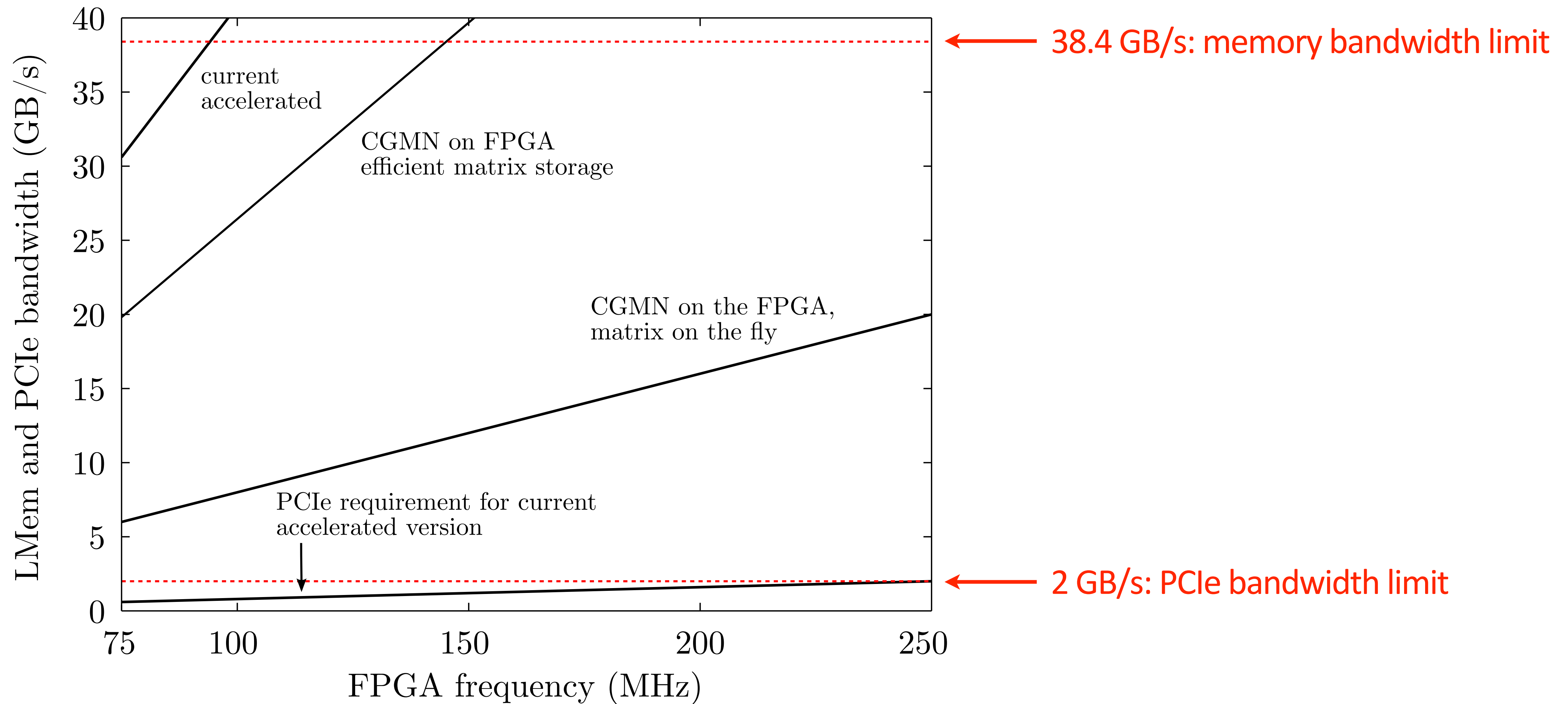


Recent Work: Multiple Kaczmarz Sweeps / CGMN Iteration

(432 x 240 x 25 system)



Avoiding Future Communication Bottlenecks



The Next Step

Problem: On-chip memory (4 MB) limits block size to **300 x 300** in the two faster dimensions.

Solution: Implement **domain decomposition** for larger systems.

Straight-forward Extension

Goal: Systematically use all 4 accelerators.

Solution: Solve **several forward problems** at once.

Future Work

Problem: Kaczmarz sweeps now account for only approximately 10% of CGMN time.

Solution: Port **all of CGMN** to the DFE.

Future Work

Fact: Reading A from memory limits optimizations like increasing FPGA frequency.

Result: Read **only earth model** m and generate A on the DFE.

Future Work

Problem: Domain size limited by memory size:
24 GB.

Solution: Parallelize CGMN to CARP-CG
[Gordon & Gordon, 2010].

Conclusion

Have **implemented** frequency-domain wave simulation using reconfigurable hardware.

A **speed-up of 2 x** 1 Intel Xeon core results from a **dataflow computing** paradigm.

Acknowledgements

Thank you to:

- Felix Herrmann, Henryk Modzelewski, Diego Oriato, Simon Tilbury, Tristan van Leeuwen, Eddie Hung, Lina Miao, Rafael Lago, my Master's committee members: Michael Friedlander, Christian Schoof, my external examiner: Steve Wilton, Maxeler Technologies, and everyone in the SLIM group!



This work was financially supported in part by the Natural Sciences and Engineering Research Council of Canada Discovery Grant (RGPIN 261641-06) and the Collaborative Research and Development Grant DNOISE II (CDRP J 375142-08). This research was carried out as part of the SINBAD II project with support from the following organizations: BG Group, BGP, BP, Chevron, ConocoPhillips, CGG, ION GXT, Petrobras, PGS, Statoil, Total SA, WesternGeco, Woodside.

References

- Å. Björck and T. Elfving. Accelerated projection methods for computing pseudoinverse solutions of systems of linear equations. *BIT Numerical Mathematics*, 19(2):145–163, 1979. ISSN 0006-3835. doi: 10.1007/BF01930845. URL <http://dx.doi.org/10.1007/BF01930845>.
- D. Gordon and R. Gordon. Component-averaged row projections: A robust, block-parallel scheme for sparse linear systems. *SIAM Journal on Scientific Computing*, 27(3):1092–1117, 2005. doi: 10.1137/040609458. URL <http://epubs.siam.org/doi/abs/10.1137/040609458>.
- D. Gordon and R. Gordon. CARP-CG: A robust and efficient parallel solver for linear systems, applied to strongly convection dominated PDEs. *Parallel Computing*, 36(9): 495–515, 2010. ISSN 0167-8191. doi: 10.1016/j.parco.2010.05.004. URL <http://www.sciencedirect.com/science/article/pii/S0167819110000827>.
- F. Grüll, M. Kunz, M. Hausmann, and U. Kobschull. An implementation of 3D electron tomography on FPGAs. In *Reconfigurable Computing and FPGAs (ReConFig)*, 2012 International Conference on, pages 1–5, 2012. doi: 10.1109/ReConFig.2012.6416732.
- S. Kaczmarz. Angenäherte auflösung von systemen linearer gleichungen. *Bulletin International de l’Academie Polonaise des Sciences et des Lettres*, 35:355–357, 1937.
- S. Kaczmarz. Approximate solution of systems of linear equations. *International Journal of Control*, 57(6):1269–1271, 1993. doi: 10.1080/00207179308934446. (translation)
- T. van Leeuwen, D. Gordon, R. Gordon, and F. J. Herrmann. Preconditioning the Helmholtz equation via row-projections. In *EAGE technical program. EAGE*, 2012. URL <https://www.slim.eos.ubc.ca/Publications/Public/Conferences/EAGE/2012/vanleeuwen2012EAGEcarpcg/vanleeuwen2012EAGEcarpcg.pdf>.
- H. Meuer, E. Strohmaier, J. Dongarra, and H. Simon. Top 500 supercomputer sites, November 2013. URL <https://www.top500.org>.
- S. Operto, J. Virieux, P. Amestoy, J.-Y. L’Excellent, L. Giraud, and H. B. H. Ali. 3D finite-difference frequency-domain modeling of visco-acoustic wave propagation using a massively parallel direct solver: A feasibility study. *Geophysics*, 72(5):SM195–SM211, 2007. doi: 10.1190/1.2759835. URL <http://geophysics.geoscienceworld.org/content/72/5/SM195.abstract>.
- O. Pell, J. Bower, R. Dimond, O. Mencer, and M. J. Flynn. Finite-difference wave propagation modeling on special-purpose dataflow machines. *Parallel and Distributed Systems, IEEE Transactions on*, 24(5):906–915, 2013. ISSN 1045-9219. doi: 10.1109/TPDS.2012.198.