LISA: Increasing Internal Connectivity in DRAM for Fast Data Movement and Low Latency

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This paper summarizes the idea of Low-Cost Interlinked Subarrays (LISA), which was published in HPCA 2016 [10], and examines the work’s significance and future potential. Our HPCA 2016 paper introduces a new DRAM design that enables fast and energy-efficient bulk data movement across subarrays in a DRAM chip. While bulk data movement is a key operation in many applications and operating systems, we observe that contemporary systems perform this movement inefficiently, by transferring data from DRAM to the processor, and then back to DRAM, across a narrow off-chip channel. The use of this narrow channel for bulk data movement results in high latency and energy consumption. Prior work proposes to avoid these high costs by exploiting the existing wide internal DRAM bandwidth for bulk data movement, but the limited connectivity of wires within DRAM allows fast data movement within only a single DRAM subarray. Each subarray is only a few megabytes in size, greatly restricting the range over which fast bulk data movement can happen within DRAM.

Our HPCA 2016 paper proposes a new DRAM substrate, Low-Cost Inter-Linked Subarrays (LISA), whose goal is to enable fast and efficient data movement across a large range of memory at low cost. LISA adds low-cost connections between adjacent subarrays. By using these connections to interconnect the existing internal wires (bitlines) of adjacent subarrays, LISA enables wide-bandwidth data transfer across multiple subarrays with little (only 0.8%) DRAM area overhead. As a DRAM substrate, LISA is versatile, enabling a variety of new applications. We describe and evaluate three such applications in detail: (1) fast inter-subarray bulk data copy, (2) in-DRAM caching using a DRAM architecture whose rows have heterogeneous access latencies, and (3) accelerated bitline precharging by linking multiple precharge units together. Our extensive evaluations show that each of LISA’s three applications significantly improves performance and memory energy efficiency, and their combined benefit is higher than the benefit of each alone, on a variety of workloads and system configurations.

1. Introduction

Bulk data movement, the movement of thousands or millions of bytes between two memory locations, is a common operation performed by an increasing number of real-world applications (e.g., [6, 37, 57, 58, 74, 82, 85, 88, 89, 94, 99, 110]). Therefore, it has been the target of several architectural optimizations (e.g., [4, 6, 35, 40, 58, 70, 86, 88, 103, 110]). In fact, bulk data movement is important enough that modern commercial processors are adding specialized support to improve its performance, such as the ERMSB instruction recently added to the x86 ISA [28].

In today’s systems, to perform a bulk data movement between two locations in memory, the data needs to go through the processor even though both the source and destination are within memory. To perform the movement, the data is first read out one cache line at a time from the source location in memory into the processor caches, over a pin-limited off-chip channel (typically 64 bits wide). Then, the data is written back to memory, again one cache line at a time over the pin-limited channel, into the destination location. By going through the processor, this data movement incurs a significant penalty in terms of latency and energy consumption.

To address the inefficiencies of traversing the pin-limited channel, a number of mechanisms have been proposed to accelerate bulk data movement (e.g., [35, 63, 88, 110]). The state-of-the-art mechanism, RowClone [88], performs data movement completely within a DRAM chip, avoiding costly data transfers over the pin-limited memory channel. However, its effectiveness is limited because RowClone can enable fast data movement only when the source and destination are within the same DRAM subarray. A DRAM chip is divided into multiple banks (typically 8), each of which is further split into many subarrays (16 to 64) [45], shown in Figure 1a, to ensure reasonable read and write latencies at high density [8, 32, 33, 45, 101]. Each subarray is a two-dimensional array with hundreds of rows of DRAM cells, and contains only a few megabytes of data (e.g., 4MB in a rank of eight 1Gb DDR3 DRAM chips with 32 subarrays per bank). While two DRAM rows in the same subarray are connected via a wide (e.g., 8K bits) bitline interface, rows in different subarrays are connected via only a narrow 64-bit data bus within the DRAM chip (Figure 1a). Therefore, even for previously-proposed in-DRAM data movement mechanisms such as RowClone [88], inter-subarray bulk data movement incurs long latency and high memory energy consumption even though data does not move out of the DRAM chip.

\footnote{We refer the reader to our prior works [8, 9, 10, 11, 21, 22, 39, 41, 42, 43, 44, 45, 54, 55, 56, 57, 58, 60, 61, 75, 88, 89] for a detailed background on DRAM.}
While it is clear that fast inter-subarray data movement can have several applications that improve system performance and memory energy efficiency [6, 37, 55, 74, 82, 85, 88, 89, 110], there is currently no mechanism that performs such data movement quickly and efficiently. This is because no wide datapath exists today between subarrays within the same bank (i.e., the connectivity of subarrays is low in modern DRAM). Our goal is to design a low-cost DRAM substrate that enables fast and energy-efficient data movement across subarrays.

2. Low-Cost Inter-Linked Subarrays (LISA)

We make two key observations that allow us to improve the connectivity of subarrays within each bank in modern DRAM. First, accessing data in DRAM causes the transfer of an entire row of DRAM cells to a buffer (i.e., the row buffer, where the row data temporarily resides while it is read or written) via the subarray’s bitlines. Each bitline connects a column of cells to the row buffer, interconnecting every row within the same subarray (Figure 1a). Therefore, the bitlines essentially serve as a very wide bus that transfers a row’s worth of data (e.g., 8K bits in a chip) at once. Second, subarrays within the same bank are placed in close proximity to each other. Thus, the bitlines of a subarray are very close to (but are not currently connected to) the bitlines of neighboring subarrays (as shown in Figure 1a).

**Key Idea.** Based on these two observations, we introduce a new DRAM substrate, called Low-cost Inter-linked SubArrays (LISA). LISA enables low-latency, high-bandwidth inter-subarray connectivity by linking neighboring subarrays’ bitlines together with isolation transistors, as illustrated in Figure 1b. We use the new inter-subarray connection in LISA to develop a new DRAM operation, row buffer movement (RBM), which moves data that is latched in an activated row buffer in one subarray into an inactive row buffer in another subarray, without having to send data through the narrow internal data bus in DRAM. RBM exploits the fact that the activated row buffer has enough drive strength to induce charge perturbation within the idle (i.e., precharged) bitlines of neighboring subarrays, allowing the destination row buffer to sense and latch this data when the isolation transistors are enabled. We describe the detailed operation of RBM in our HPCA 2016 paper [10].

By using a rigorous DRAM circuit model that conforms to the JEDEC standards [32] and ITRS specifications [30, 31], we show that RBM performs row buffer movement at 26x the bandwidth of a modern 64-bit DDR4-2400 memory channel (500 GB/s vs. 19.2 GB/s), even after we conservatively add a large (60%) timing margin to account for process and temperature variation.

**Die Area Overhead.** To evaluate the area overhead of adding isolation transistors, we use area values from prior work, which adds isolation transistors to disconnect bitlines from sense amplifiers [73]. That work shows that adding an isolation transistor to every bitline incurs a total of 0.8% die area overhead in a 28nm DRAM process technology. Similar to prior work that adds isolation transistors to DRAM [57, 73], our LISA substrate also requires additional control logic outside the DRAM banks to control the isolation transistors, which incurs a small amount of area and is non-intrusive to the cell arrays.

3. Applications of LISA

We exploit LISA’s fast inter-subarray movement capability to enable many applications that can improve system performance and energy efficiency. In our HPCA 2016 paper [10], we implement and evaluate three applications of LISA, which significantly improve system performance in different ways.

3.1. Rapid Inter-Subarray Bulk Data Copying (LISA-RISC)

Due to the narrow memory channel width, bulk copy operations used by applications and operating systems are performance limiters in today’s systems [35, 37, 55, 88, 110]. These operations are commonly performed due to the memcpy and memmove. Recent work reported that these two operations consume 4-5% of all of Google’s data center cycles [37], making them an important target for lightweight hardware acceleration.

Our goal is to design a new mechanism that enables low-latency and energy-efficient memory copy between rows in different subarrays within the same bank. To this end, we propose a new in-DRAM copy mechanism that uses LISA to exploit the high-bandwidth links between subarrays. The key idea, step by step, is to: (1) activate a source row in a subarray; (2) rapidly transfer the data in the activated source row buffers to the destination subarray’s row buffers, through LISA’s RBM operation; and (3) activate the destination row, which enables the contents of the destination row buffers to be latched into the destination row. We call this inter-subarray row-to-row copy mechanism LISA-Rapid Inter-Subarray Copy (LISA-RISC).

3.1.1. DRAM Latency and Energy Consumption. Figure 2 shows the DRAM latency and DRAM energy consumption of memcpy (i.e., the baseline system), RowClone [88] (state-of-the-art work), and LISA-RISC for copying a row of data (8KB). The exact latency and energy numbers are listed in...
3.2. In-DRAM Caching Using Heterogeneous Subarrays (LISA-VILLA)

Our second application aims to reduce the DRAM access latency for frequently-accessed (hot) data. We propose to introduce heterogeneity within a bank by designing heterogeneous-latency subarrays. We call this heterogeneous DRAM design Variable Latency DRAM (VILLA-DRAM). To design a low-cost fast subarray, we take an approach similar to prior work, attaching fewer cells to each bitline to reduce the parasitic capacitance and resistance. This reduces the latency of the three fundamental DRAM operations—activation, precharge, and restoration—when accessing data in the fast subarrays [37, 67, 94]. Activation “opens” a row of DRAM cells to access stored data. Precharge “closes” an activated row. Restoration restores the charge level of each DRAM cell in a row to prevent data loss. Together, these three operations predominantly define the latency of a memory request [8, 9, 10, 11, 21, 22, 39, 41, 42, 43, 44, 45, 54, 55, 56, 57, 58, 60, 61, 75, 78, 88, 89]. In this work, we focus on managing the fast subarrays in hardware, as doing so offers better adaptivity to dynamic changes in the hot data set.

In order to take advantage of VILLA-DRAM, we rely on LISA-RISC to rapidly copy rows across subarrays, which significantly reduces the caching latency. We call this synergistic design, which builds VILLA-DRAM using LISA, LISA-VILLA. Nonetheless, the cost of transferring data to a fast subarray is still non-negligible, especially if the fast subarray is far from the subarray where the data to be cached resides. Therefore, an intelligent cost-aware mechanism is required to make astute decisions on which data to cache and when.

3.2.1. Caching Policy for LISA-VILLA. We design a simple epoch-based caching policy to evaluate the benefits of caching a row in LISA-VILLA. Every epoch, we track the number of accesses to rows by using a set of 1024 saturating counters for each bank. The counter values are halved every epoch to prevent staleness. At the end of an epoch, we mark the 16 most frequently-accessed rows as hot, and cache them when they are accessed the next time. For our cache replacement policy, we use the benefit-based caching policy proposed by Lee et al. [57]. Specifically, it uses a benefit counter for each row cached in the fast subarray: whenever a cached row is accessed, its counter is incremented. The row with the least benefit is replaced when a new row needs to be inserted. Note that a large body of work proposes various caching policies (e.g., [20, 23, 26, 34, 38, 59, 66, 78, 79, 87, 91, 100, 104, 106]), each of which can potentially be used with LISA-VILLA.

3.2.2. Evaluation. Figure 3 shows the system performance improvement of LISA-VILLA over a baseline without any fast subarrays in a four-core system. It also shows the hit rate in VILLA-DRAM, i.e., the fraction of accesses that hit in the fast subarrays. We make two main observations. First, by

Table 1: Copy latency and DRAM energy. Reproduced from [10].

| Copy Commands (8KB) | Latency (ns) | Energy (pJ) |
|--------------------|--------------|-------------|
| memcpy            | 1366.25      | 6.2         |
| RC-InterSA / Bank / IntraSA | 1363.75 / 701.25 / 83.75 / 4.33 / 2.08 / 0.06 |
| LISA-RISC (1 / 7 / 15 hops) | 148.5 / 196.5 / 260.5 | 0.09 / 0.12 / 0.17 |

Table 1. For LISA-RISC, we define a hop as the number of subarrays that LISA-RISC needs to copy data across to move the data from the source subarray to the destination subarray. For example, if the source and destination subarrays are adjacent to each other, the number of hops is 1. The DRAM chips we evaluate have 16 subarrays per bank, so the maximum number of hops is 15.

We make two observations from these numbers. First, although inter-subarray RowClone (RC-InterSA) incurs similar latencies as memcpy, it consumes 1.43x less energy, as it does not transfer data over the channel and DRAM I/O for each copy operation. However, as we discuss in Section 4.1 of our HPCA 2016 paper [10], RC-InterSA incurs a higher system performance penalty because it is a blocking long-latency memory command. Second, copying between subarrays using LISA reduces the copy latency by 9x and copy energy by 48x compared to RowClone, even though the total latency of LISA-RISC grows linearly with the hop count. An additional benefit of using LISA-RISC is that its inter-subarray copy operations are performed completely inside a bank. As the internal DRAM data bus is untouched, other banks can concurrently serve memory requests, exploiting bank-level parallelism.
exploiting LISA-RISC to quickly cache data in VILLA-DRAM, LISA-VILLA improves system performance for a wide variety of workloads — by up to 16.1%, with a geometric mean of 5.1%. This is mainly due to reduced DRAM latency of accesses that hit in the fast subarrays. The performance improvement heavily correlates with the VILLA cache hit rate. Second, the VILLA-DRAM design, which consists of heterogeneous subarrays, is not practical without LISA. Figure 3 shows that using RC-InterSA (i.e., RowClone copying data across subarrays) to move data into the cache reduces performance by 52.3% due to slow data movement, which overshadows the benefits of caching. The results indicate that LISA is an important substrate to enable not only fast bulk data copy, but also a fast in-DRAM caching scheme.

![Figure 3: Performance improvement and hit rate with LISA-VILLA, and performance comparison to using RC-InterSA with VILLA-DRAM. Reproduced from [10].](image)

### 3.3. Fast Precharge Using Linked Precharge Units (LISA-LIP)

Our third application aims to accelerate the process of precharge. The precharge time for a subarray is determined by the drive strength of the precharge unit (i.e., a circuitry in a subarray’s row buffer for precharging the connected subarray). We observe that in modern DRAM, while a subarray is being precharged, the precharge units (PUs) of other subarrays remain idle.

We propose to exploit these idle PUs to accelerate a precharge operation by connecting them to the subarray that is being precharged. Our mechanism, LISA-Linked Precharge (LISA-LIP), precharges a subarray using two sets of PUs: one from the row buffer that is being precharged, and a second set from a neighboring subarray’s row buffer (which is already in the precharged state), by enabling the links between the two subarrays.

To evaluate the accelerated precharge process, we use the same DRAM circuit model described in Section 2 and simulate the linked precharge operation in SPICE. Our SPICE simulation reports that LISA-LIP significantly reduces the precharge latency by 2.6x compared to the baseline (5ns vs. 13ns). Our system evaluation shows that LISA-LIP improves performance by 10.3% on average, across 50 four-core workloads. We refer the reader to Section 6 of our HPCA 2016 paper [10] for a detailed analysis of LISA-LIP.

### 3.4. Evaluation: Putting Everything Together

As all of the three proposed applications are complementary to each other, we evaluate the effect of putting them together on a four-core system. Figure 4 shows the system performance improvement of adding LISA-VILLA to LISA-RISC, as well as combining all three optimizations, compared to our baseline using memcpy and standard DDR3-1600 memory across 50 workloads. We refer the reader to our full paper [10] for the detailed configuration and workloads. We draw several key conclusions. First, the performance benefits from each scheme are additive. On average, adding LISA-VILLA improves performance by 16.5% over LISA-RISC alone, and adding LISA-LIP further provides an 8.8% gain over LISA-(RISC+VILLA). Second, although LISA-RISC alone provides a majority of the performance improvement over the baseline (59.6% on average), the use of both LISA-VILLA and LISA-LIP further improves performance, resulting in an average performance gain of 94.8% and memory energy reduction (not plotted) of 49.0%. Taken together, these results indicate that LISA is an effective substrate that enables a wide range of high-performance and energy-efficient applications in the DRAM system.

![Figure 4: Combined weighted speedup (WS) [14, 93] improvement of LISA applications. Reproduced from [10].](image)

We conclude that LISA is an effective substrate that can greatly improve system performance and reduce system energy consumption by synergistically enabling multiple different applications. Our HPCA 2016 paper [10] provides many more experimental results and analyses confirming this finding.

### 4. Related Work

To our knowledge, this is the first work to propose a DRAM substrate that supports fast data movement between subarrays in the same bank, which enables a wide variety of applications for DRAM systems. We now discuss prior works that focus on each of the optimizations that LISA enables.

#### 4.1. Bulk Data Transfer Mechanisms

Prior works [7, 16, 17, 36, 108] propose to add scratchpad memories to reduce CPU pressure during bulk data transfers, which can also enable sophisticated data movement (e.g., scatter-gather [90]), but they still require data to first be moved on-chip. A patent proposes a DRAM design that can copy a page across memory blocks [84], but lacks concrete analysis and evaluation of the underlying copy operations. Intel I/O Acceleration Technology [27] allows for memory-to-memory DMA transfers across a network, but cannot transfer data within main memory.

Zhao et al. [110] propose to add a bulk data movement engine inside the memory controller to speed up bulk-copy
operations. Jiang et al. [35] design a different copy engine, placed within the cache controller, to alleviate pipeline and cache stalls that occur when these transfers take place. However, these works do not directly address the problem of data movement across the narrow memory channel.

A concurrent work by Lu et al. [63] proposes a heterogeneous DRAM design similar to VILLA-DRAM, called DAS-DRAM, but with a very different data movement mechanism from LISA. It introduces a row of migration cells into each subarray to move rows across subarrays. Unfortunately, the latency of DAS-DRAM is not scalable with movement distance, because it requires writing the migrating row into each intermediate subarray’s migration cells before the row reaches its destination, which prolongs data transfer latency. In contrast, LISA provides a direct path to transfer data between row buffers between adjacent subarrays without requiring intermediate data writes into any subarray.

### 4.2. Cached DRAM

Several prior works (e.g., [20, 23, 26, 38, 109]) propose to add a small SRAM cache to a DRAM chip to lower the access latency for data that is kept in the SRAM cache (e.g., frequently or recently used data). There are two main disadvantages of these works. First, adding an SRAM cache into a DRAM chip is very intrusive: it incurs a high area overhead (38.8% for 64KB in a 2Gb DRAM chip) and design complexity [45, 57]. Second, transferring data from DRAM to SRAM uses a narrow global data bus, internal to the DRAM chip, which is typically 64-bit wide. Thus, installing data into the DRAM cache incurs high latency. Compared to these works, our LISA-VILLA design enables low latency without significant area overhead or complexity.

### 4.3. Heterogeneous-Latency DRAM

Prior works propose DRAM architectures that provide heterogeneous latency either spatially (dependent on where in the memory an access targets) or temporally (dependent on when an access occurs).

**Spatial Heterogeneity.** Prior work introduces spatial heterogeneity into DRAM, where one region has a fast access latency but fewer DRAM rows, while the other has a slower access latency but many more rows [57, 94]. Recent works show that latency heterogeneity inherent in DRAM chips due to process or design-induced variation can also naturally enable such heterogeneous-latency substrates [9, 54]. The fast region in DRAM can be utilized as a caching area, for the frequently or recently accessed data. We briefly describe two state-of-the-art works that offer different heterogeneous-latency DRAM designs.

CHARM [94] introduces heterogeneity within a rank by designing a few fast banks with (1) shorter bitlines for faster data sensing, and (2) closer placement to the chip I/O for faster data transfers. To exploit these low-latency banks, CHARM uses an OS-managed mechanism to statically map hot data to these banks, based on profiled information from the compiler or programmers. Unfortunately, this approach cannot adapt to program phase changes, limiting its performance gains. If it were to adopt dynamic hot data management, CHARM would incur high migration costs over the narrow 64-bit bus that internally connects the fast and slow banks.

TL-DRAM [57] provides heterogeneity within a subarray by dividing it into fast (near) and slow (far) segments that have short and long bitlines, respectively, using isolation transistors. The fast segment can be managed as an OS-transparent hardware cache. The main disadvantage is that it needs to cache each hot row in two near segments as each subarray uses two row buffers on opposite ends to sense data in the open-bitline architecture (as discussed in our HPCA 2016 paper [10]). This prevents TL-DRAM from using the full near segment capacity. As we can see, neither CHARM nor TL-DRAM strike a good design balance for heterogeneous-latency DRAM. Our proposal, LISA-VILLA, is a new heterogeneous DRAM design that offers fast data movement with a low-cost and easy-to-implement design.

**Temporal Heterogeneity.** Prior work observes that DRAM latency can vary depending on when an access occurs. The key observation is that a recently-accessed or refreshed row has nearly full electrical charge in the cells, and thus the following access to the same row can be performed faster [21, 22, 92]. We briefly describe two state-of-the-art works that focus on providing heterogeneous latency temporally.

ChargeCache [22] enables faster access to recently-accessed rows in DRAM by tracking the addresses of recently-accessed rows. NUAT [92] enables accesses to recently-refreshed rows at low latency because these rows are already highly-charged. In contrast to ChargeCache and NUAT, LISA does not require data to be recently-accessed/refreshed in order to reduce DRAM latency. Adaptive-Latency DRAM (AL-DRAM) [56] adapts the DRAM latency of each DRAM module to temperature, observing that each module can be operated faster at lower temperatures. LISA is orthogonal to AL-DRAM. The ideas of LISA can be employed in conjunction with works that exploit the temporal heterogeneity of DRAM latency.

### 4.4. Other Latency Reduction Mechanisms

Many prior works propose memory scheduling techniques, which generally reduce latency to access DRAM [3, 13, 15, 29, 43, 44, 51, 52, 53, 68, 69, 71, 72, 96, 97, 98, 102]. Other works propose mechanisms to perform in-memory computation to reduce data movement and access latency [1, 2, 5, 6, 18, 24, 25, 40, 46, 62, 76, 77, 88, 89, 95, 107]. LISA is complementary to these works, and it can work synergistically with in-memory computation mechanisms by enabling fast aggregation of data.

### 5. Significance

Our HPCA 2016 paper [10] proposes a new DRAM substrate that significantly improves the performance and effi-
ciency of bulk data movement in modern systems. In this section, we briefly discuss the expected future impact of our work, and discuss several research directions that our work motivates.

5.1. Potential Industry Impact

We believe that our LISA substrate can have a large impact on mobile systems as well as data centers that consume a significant amount of cycle time performing bulk data movement. A recent study [37] by Google reports that memcpy() and memmove() library functions alone represent 4-5% of their data center cycles even though Google has a significant workload diversity running within their data centers. Another recent study shows that 62.7% of system energy is spent on data movement on consumer devices (e.g., smartphones, wearable devices, web-based computers such as Chromebooks) [6]. In this work, we demonstrate that one potential application of using the LISA substrate is to accelerate memcpy() and memmove(), as discussed in Section 3.1. Our detailed DRAM circuit model reports that LISA reduces the latency and DRAM energy of these functions by 9x and 69x compared to today’s systems, respectively. Hence, we expect LISA can improve the efficiency and performance of both mobile and data center systems.

5.2. Future Research Directions

This work opens up several avenues of future research directions. In this section, we describe several directions that can enable researchers to tackle other problems related to memory systems based on the LISA substrate.

Reducing Subarray Conflicts via Remapping. When two memory requests access two different rows in the same bank, they have to be served serially, even if they are to different subarrays. To mitigate such bank conflicts, Kim et al. [45] propose subarray-level parallelism (SALP), which enables multiple subarrays to remain activated at the same time. However, if two accesses are to the same subarray, they still have to be served serially. This problem is exacerbated when frequently-accessed rows reside in the same subarray. To help alleviate such subarray conflicts, LISA can enable a simple mechanism that efficiently remaps or moves the conflicting rows to different subarrays by exploiting fast RBM operations.

Enabling LISA to Perform 1-to-N Memory Copy or Move Operations. A typical memcpy or memmove call only allows the data to be copied from one source location to one destination location. To copy or move data from one source location to multiple different destinations, repeated calls are required. The problem is that such repeated calls incur long latency and high bandwidth consumption. One potential application that can be enabled by LISA is performing memcpy or memmove from one source location to multiple destinations completely in DRAM without requiring multiple calls of these operations.

By using LISA, we observe that moving data from the source subarray to the destination subarray latches the source row’s data in all the intermediate subarrays’ row buffer. As a result, activating these intermediate subarrays would copy their row buffers’ data into the specified row within these subarrays. By extending LISA to perform multi-point (1-to-N) copy or move operations, we can significantly increase system performance of several commonly-used system operations. For example, forking multiple child processes can utilize 1-to-N copy operations to efficiently copy memory pages from the parent’s address space to all the children. As another example, LISA can extend the range of in-DRAM bulk bitwise operations [85, 89]. Thus, LISA can efficiently enable architectural support to a new, useful system and programming primitive: 1-to-N bulk memory copy/movement.

In-Memory Computation with LISA. One important requirement of efficient in-memory computation is being able to move data from its stored location to the computation units with very low latency and energy. We believe using the LISA substrate can enable a new in-memory computation framework. The idea is to add a small computation unit inside each or a subset of banks, and connect these computation units to the neighboring subarrays which store the data. Doing so allows the system to utilize LISA to move bulk data from the subarrays to the computation units with low latency and low area overhead.

Extending LISA to Non-Volatile Memory. In this work, we only focus on the DRAM technology. A class of emerging memory technology is non-volatile memory (NVM), which has the capability of retaining data without power supply. We believe that the LISA substrate can be extended to NVM (e.g., PCM [48, 49, 50, 80, 104, 105] and STT-MRAM [12, 19, 47]) since the memory organization of NVM mostly resembles that of DRAM. A potential application of LISA in NVM is an efficient file copy operation that does not incur costly I/O data transfer. We believe LISA can provide further benefits when main memory becomes persistent [65].

6. Conclusion

We present a new DRAM substrate, low-cost inter-linked subarrays (LISA), that expedites bulk data movement across subarrays in DRAM. LISA achieves this by creating a new high-bandwidth datapath at low cost between subarrays, via the insertion of a small number of isolation transistors. We describe and evaluate three applications that are enabled by LISA. First, LISA significantly reduces the latency and memory energy consumption of bulk copy operations between subarrays over state-of-the-art mechanisms [88]. Second, LISA enables an effective in-DRAM caching scheme on a new heterogeneous DRAM organization, which uses fast subarrays for caching hot data in every bank. Third, we reduce precharge latency by connecting two precharge units of adjacent subarrays together using LISA. We experimentally show that the three applications of LISA greatly improve system
performance and memory energy efficiency when used individually or together, across a variety of workloads and system configurations.

We conclude that LISA is an effective substrate that enables several effective applications. We believe that this substrate, which enables low-cost interconnections between DRAM subarrays, can pave the way for other applications that can further improve system performance and energy efficiency through fast data movement in DRAM. We greatly encourage future work to 1) investigate new applications and benefits of LISA, and 2) develop new low-cost interconnection substrates within a DRAM chip to improve internal connectivity and data transfer ability.

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