

Quantifying benefits of offloading data management to storage devices

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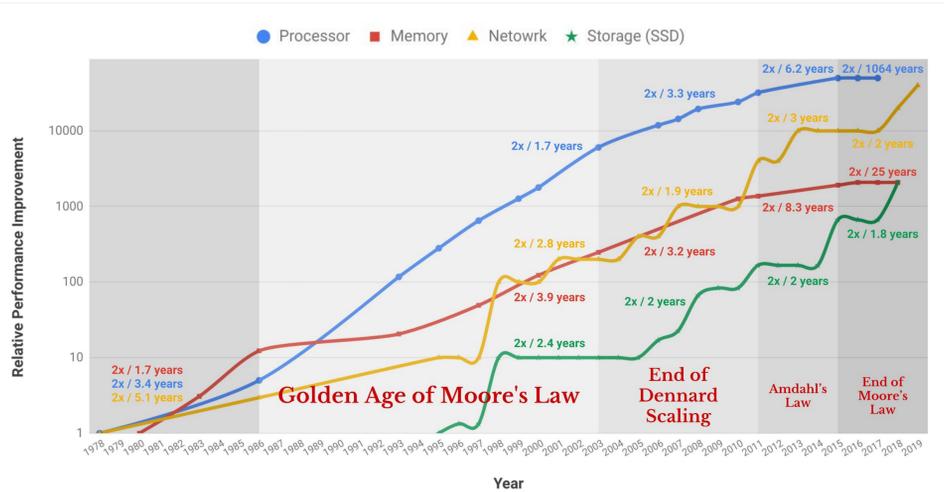
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Abstract

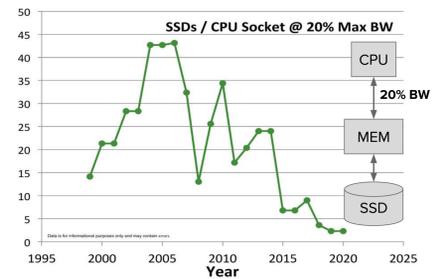
- Problem**
 It is difficult to quantify the benefits of offloading data management from hosts to storage devices.
- Approach**
 Quantify the benefits based on a reference point in order to formalize a fair comparison.
- Data Management**
 Manage data layout and placement in storage devices to ensure its durability and availability (e.g., compaction, deduplication, scrubbing, redundancy, recovery, rebalancing).
- MBWU (pronounced "MibeeWu")**
 Media-based Work Unit, as a reference point, is defined by a combination of a storage device and workload and measured in IOPS.

Trends in Technology



When performance improvement of **network** and **storage** significantly outstrips the improvement of **microprocessor** and **memory**, what can we do?

By 2020, if 20% of memory bandwidth on a host is used for storage I/O, an CPU socket can only serve less than 4 SSDs (Fig. 2).



We need to reduce the traffic between host I/O bus and storage devices.

Opportunities:

- Embedded platforms have better price/performance
- Embedded processors are getting more powerful
- Domain-specific hardware is the future! [1,2]

Data Management is ideal for offloading:

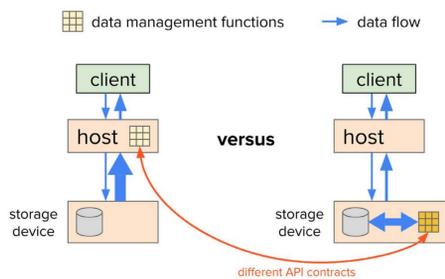
- It is data intensive
- It is responsible for a significant fraction of host-to-device traffic

Platform Comparison

Conventional servers (as host platforms) and storage devices (as embedded platforms) are hard to compare due to their significant differences in software and hardware design.

Benefit evaluations in existing research generally run into performance comparison problems because:

- Evaluation results do not isolate impacts from different implementations of data management functionality.
- Modification of the device firmware changes storage device performance



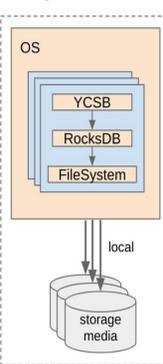
MBWU is dependent only on the **storage device** and **workload** making a reference point that is truly independent from the platform.

- The value of a single MBWU should not be throttled by any system resources on the platform. (e.g., CPU, memory).
- MBWU specifies a performance reference point for normalizing capabilities of different platforms for running a specific workload.

The capabilities of different platforms can be evaluated in terms of the number of MBWUs they can generate.

- Host platforms may be powerful enough to generate multiple MBWUs before hitting other bottlenecks.
- Embedded platforms may only be able to generate a fraction of an MBWU.

Fig. 4: Evaluate MBWUs with multiple storage devices



We can now compare the total cost of ownership of these platforms based on the MBWUs (e.g., \$/MBWU, kW·h/MBWU, m²/MBWU).

MBWU is useful for evaluating the benefits of any data management offloading to storage devices. (e.g., management of key/value data, full-memory encryption for persistent-memory technologies)

Evaluation

To demonstrate the use of this **platform-agnostic measurement methodology**, we evaluate the benefits of offloading key-value data management to storage devices. Configurations of the two platforms used in our experiment are shown in Table I.

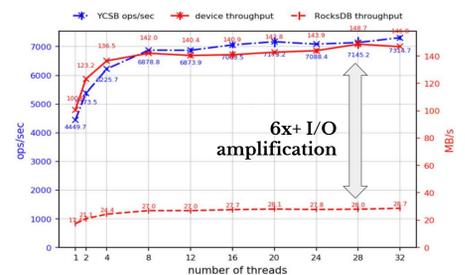
Before running the workload, we precondition the SSDs with two different preconditioning configurations (one covers the first 75% of LBAs and the other covers the entire LBA space) to illustrate the sensitivity of the evaluation results. The workload is a steady load containing key-value, read-and-write requests generated by YCSB to the underlying RocksDB that in turn uses the SSD. The detail evaluation process is shown in Fig. 6.

Platform	# of SSDs	CPU	Memory	Cost
Host	8	24-cores	64 GB	\$4,100
Embedded	1	hexa-core	4 GB	\$120

TABLE I: Configuration of Two Platforms

The key-value workload used in our experiment is:

- Key is 16 bytes, value is 4 KiB
- The read/write ratio is 50%
- The popularity of keys follows a Zipf distribution
- The size of dataset is 40 GiB.
- The number of key-value operations is 10,444,959 (~40 GiB).

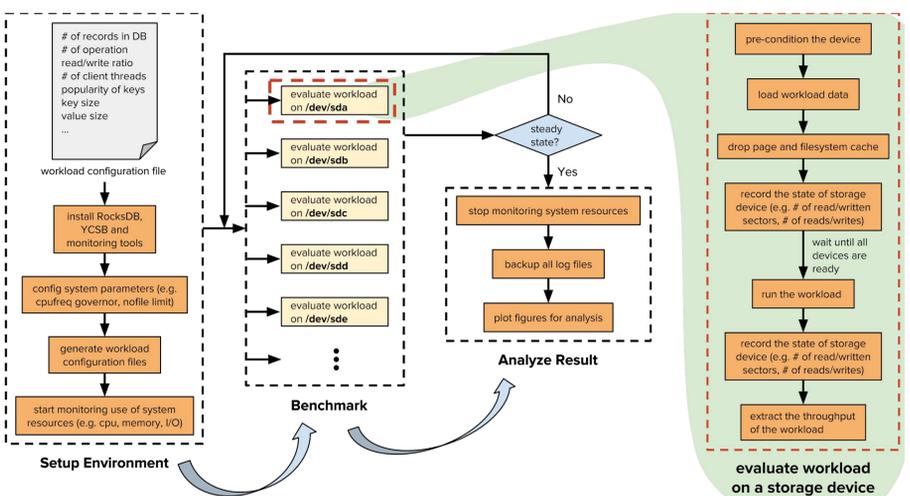


Precondition Platform	75% LBAs		100% LBAs	
	# of MBWUs	Current	# of MBWUs	Current
Host	6	1.73A	7.5	1.53A
Embedded	0.5	0.087A	0.5	0.08A

TABLE II: Evaluation Results

The evaluation results are shown in Table II. We have built a tool to automate the evaluation process. This tool allows to customize the characteristics of a workload with more options such as ratio of scan and read-modify-write operations.

With these numbers (Table II), for example, if the LBAs of the SSDs are 75% preconditioned, the performance of 12 embedded platforms is equivalent to the performance of one host platform. Therefore, the embedded platforms can save 65% of the cost per MBWU compared to running the same key-value workload on the host platform, and they can save 39.6% of energy per MBWU as well.



References

- [1] Hennessy, John, and David Patterson. "A New Golden Age for Computer Architecture: Domain-Specific Hardware/Software Co-Design, Enhanced." [2] Patterson, David. "50 Years of computer architecture: From the mainframe CPU to the domain-specific tpu and the open RISC-V instruction set." Solid-State Circuits Conference-(ISSCC), 2018 IEEE International. IEEE, 2018.

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