LaLDPC: Latency-aware LDPC for Read Performance Improvement of Solid State Drives

Yajuan Du†‡, Deqing Zou*, Qiao Li‡, Liang Shi‡, Hai Jin* and Chun Jason Xue‡

*Service Computing Technology and System Lab, Cluster and Grid Computing Lab, Big Data Technology and System Lab, School of Computer Science and Technology, Huazhong University of Science and Technology, Wuhan, 430074, China
Email: dyjcitu2013@gmail.com, deqingzou@hust.edu.cn, hjin@hust.edu.cn

†Department of Computer Science, City University of Hong Kong, Kowloon, Hong Kong
Email: jasonxue@cityu.edu.hk

‡College of Computer Science, Chongqing University, Chongqing, China
Email: qiaoli045@gmail.com, shiliang@cqu.edu.cn

Abstract—High-density Solid State Drives (SSDs) have to use Low-Density Parity-Check (LDPC) codes to store data reliably. Current LDPC implementations apply multiple read-retry steps to find the appropriate set of multiple Reference Voltage (RVs) for successful decoding. These appropriate RVs, number of which is represented by read level, cannot be obtained in advance, instead, an iterative read-retry LDPC process happens that starts with the first read level with the smallest number of RVs and then increases the level until that the data have been successfully decoded. Only read-retry with the last read level, called the appropriate read level, is necessary. Read latency can be improved by avoiding read-retries. This becomes particularly important in the late stage of SSD life when high read levels are needed.

To avoid unnecessary read-retries, this paper proposes Latency-aware LDPC (LaLDPC). We study read level characteristics along with data retention and observe that a single read level stays appropriate for a long time. During this time, all reads in the same page have probably the same appropriate read level, which we call the temporal read level locality. Using this value, LaLDPC estimates the starting read voltage level, which we store in the flash translation layer of the controller. As an example, we show how to integrate LaLDPC into the Demand-based Flash Translation Layer (DFTL). A new cache eviction algorithm is also proposed to leave entries with high read levels in the cache as long as possible. Experimental results show that LaLDPC saves 56% of read retries and improves SSD read performance by 18% on average compared with the existing LDPC method.

I. INTRODUCTION

Flash memory based Solid State Drives (SSDs) have been gradually replacing Hard Disk Drives (HDD) as the main storage medium in some applications, such as mobile phones and personal computers, with its characteristics of non-volatility, shock resistance, high speed, and low energy consumption [1]. In order to reduce the production cost, high-density flash memories have been developed, such as Multi-Level per Cell (MLC) flash [2], Triple-Level per Cell (TLC) flash [3] as well as 3D NAND flash [4], [5]. Although with larger storage space and lower price, these flash memories have worse endurance [6], [7]. One way to improve the endurance is to strengthen data reliability by Error Correction Codes (ECCs) [8], [9].

Traditional ECCs, such as BCH codes [10] cannot satisfy the reliability need of the exacerbated endurance problem in high-density flash memory. Advanced codes, Low-Density Parity-Check (LDPC) codes [11]–[13], have been studied and implemented to provide higher error correction capability. However, LDPC codes often lead to long read latency, especially when SSDs are in the late stage of life [14]. This paper strives to improve read performance of flash-based SSDs that use LDPC codes.

LDPC reads on flash memory involve three main processes: flash sensing, data transferring, and decoding. Flash sensing is first performed on flash cells. The sensed information is then transferred to flash controller and taken as the input of decoding. The long latency of LDPC reads comes from two aspects. On one hand, flash sensing is time-consuming because a set of multiple Reference Voltages (RVs) are applied [15], [16], the number of which is represented as a read level. As one sensing applies one RV, multiple sensing times with a fixed read level containing multiple RVs have to be performed to determine the threshold voltage states of flash cells. Furthermore, for high Raw Bit Error Rates (RBERs) caused by increased number of Program/Erase (P/E) cycles, a larger read level with more RVs, providing higher error correction capability, has to be applied to successfully decode older flash with LDPC codes [14], [17], [18]. This, of course, further increases the latency of flash sensing.

From the latency specifications of MLC flash [15], flash sensing takes about 50μs with a low read level but more than 100μs with a high read level, which is much longer than the other two processes in LDPC reads. In a word, LDPC with a low read level is faster but can only correct errors in the early stages of flash life, whereas in the latter stages of life, a higher read level must be used at the cost of longer sensing latency. One read level, that happens to successfully decode the data and does not induce excessive sensing latency, is called the appropriate read level.

On the other hand, in order to adapt error correction to the RBERs over the entire lifetime of flash, that require varied appropriate levels of RVs, current LDPC implementations...
apply an iterative read-retry process, which starts with a low level of RVs and increases until the appropriate read level is found. This process accumulates read latencies of all read-retries, which degrades read performance of SSDs with LDPC codes. If one could know the appropriate read level from the start, unnecessary read-retries can be avoided and SSD read performance would be largely optimized.

This paper proposes Latency-aware LDPC (LaLDPC) to estimate an appropriate level of RVs. In a preliminary experiment, we observe that one read level can be appropriately used to read a page for a long time. During this time, all reads on this page have the same appropriate read level, which we call the temporal read level locality. LaLDPC implements two functions with marginal storage overhead. On one hand, LaLDPC applies the level of RVs used in the last read on one page as the start value of the read-retry iteration for the next reads on the same page. On the other hand, LaLDPC stores these read levels into Flash Translation Layer (FTL) and timely updates the level information once a higher read level happens. In this manner, unnecessary read-retries can be significantly reduced. As an example, we show how to integrate LaLDPC into the Demand-based Flash Translation Layer (DFTL) [19]. By noticing that pages with high read levels can achieve more performance benefits from LaLDPC, a new mapping cache eviction algorithm is proposed to leave these pages in the cache as long as possible, which can further strengthen the performance improvements of LaLDPC.

In order to evaluate effectiveness of LaLDPC, we perform a series of experiments on the simulation platform, Disksim with SSD extensions [20], [21]. The read performance results of LaLDPC as well as the existing LDPC method have been collected from fifteen real workloads. The sensitivity of LaLDPC as well as the existing LDPC method have been shown in Section V. Related work and conclusion of this paper are presented in Section VI and Section VII, respectively.

II. BACKGROUND AND MOTIVATION

This section first reviews the architecture of SSDs. Then, the read process with LDPC is analyzed. At last, the problem of read latency accumulation is studied.

A. Basics of Solid State Drives

Solid state drives consist of two main components: controller and NAND flash, as shown in Figure 1. The controller receives requests from host and decides how to organize data in flash. Several components are implemented in controller for better utilization of flash, such as FTL [22], [23], garbage collection, wear leveling, and LDPC encoder/decoder. FTL translates logical addresses requested by host into physical addresses with a mapping table. The mapping cache proposed in DFTL [19] keeps part of mapping entries with recently accessed pages while the whole mapping table is stored in translation blocks of flash. When the mapping cache runs out of space, one entry in the table has to be selected to evict and flush to translation blocks in NAND flash. Due to the out-of-place update characteristics of flash, data can not be directly overwritten but have to be programmed into other blocks. Garbage collection and wear leveling are responsible to reclaim blocks with invalid data and balance wears of blocks [24], [25]. LDPC encoder and decoder [26] ensure data reliability of flash. The raw data are first encoded into codewords and then stored in flash during write operations while codewords are decoded when they are read. Details of the encoder and decoder are presented in Section II-B.

Fig. 1. The architecture of solid state drives (SSDs). Several components are integrated into the controller for better utilization of flash.
B. Read Process with LDPC

This section introduces the process of LDPC read: flash sensing, data transferring, and LDPC decoding. After FTL finds the physical addresses requested by host, flash sensing happens with multiple RVs. In this paper, the read level of LDPC is defined as the number of RVs between a pair of adjacent threshold voltage states of flash memory. An example of these RVs is shown in Figure 2, in which $V_{th}$ in the $x$-axis represents threshold voltages while $\#\text{cells}$ in the $y$-axis represents the number of cells. Based on threshold voltages in programming shown in Figure 2(a), the $3^{rd}$ read level is applied with three RVs between two adjacent flash states, as shown in Figure 2(b). There are overall nine RVs applied in flash sensing. After flash sensing, the sensed data are transferred to controller by flash-to-controller bus. As the bus can only transfer a fixed amount of data, the transfer latency is related to size of flash data. LDPC decoding happens after data are transferred to controller.

LDPC decoding has a large difference with traditional ECCs and involves a complex process. The sensed data are first transferred into log-likelihood ratio (LLR) information [27], [28], the input of LDPC decoder. The decoding iteratively passes messages through the Tanner graph [29] generated from a parity-check matrix [30], as shown in Figure 3. Once a codeword is found, the decoding process finishes. The iteration counts are related to RBERs of data. As LDPC codes can be implemented by hardware, the decoding often takes small time cost.

In current LDPC implementations, three stages can happen multiple times when LDPC decoding fails, as shown in the left part of Figure 4. For the read-retry in the $i^{th}$ read level with $i$ ranging from 1 to the maximum iteration $N$, sensed data are transferred to LDPC decoder to perform the message passing decoding. When the decoding fails, $i$ is incremented. With more reference voltages in a higher read level, the LDPC capability to correct errors is increased. The read-retry phase ends when LDPC decoding succeeds or the maximum read level is reached.

C. The Problem of Read Latency Accumulation

This section analyzes the latency involved in each part of current LDPC implementation. The decoding process and latency details are presented in Figure 4. As the first step of LDPC read, flash sensing detects flash cell states with given RVs of the first read level. Then, the sensed data are transferred to flash controller by flash-to-controller bus. Subsequently, LDPC decoder performs the decoding algorithm on the data. The latencies of flash sensing, data transferring, and LDPC decoding with level $i$ are denoted as $S_i$, $R_i$, and $D_i$, respectively. Thus, the read latency $T_i$ with level $i$ is computed as $T_i = S_i + R_i + D_i$. When decoding fails with the $i^{th}$ read level, the sensing with read level $i + 1$ is retried with incremented RVs. The read-retry process does not finish until the appropriate RVs are found. Assuming the read level with appropriate RVs on one flash page is $L$, overall read latency of current LDPC method [15] is computed as $\sum_{i=1}^{L} T_i$, which is the accumulation of read latencies in all read-reties.

From above analysis, if the appropriate RVs are known in advance, only the latency $T_L$ happens and the read performance is optimal. For the case with $L > 1$, latencies with read levels from 1 to $L - 1$ are redundant. In order to investigate the read latency accumulation, quantitative analysis is performed.
to show latency differences between the current LDPC and the optimal case with known RVs. According to the basic latencies listed in Table I [31], the latency results are shown in Figure 5(a). It can be found that there exists a large latency gap between current LDPC and the optimal case, especially for large read levels. The proposed technique in this paper aims to narrow this gap and solve the latency accumulation problem.

**TABLE I**

<table>
<thead>
<tr>
<th>Read latency</th>
<th>Steps in LDPC read</th>
</tr>
</thead>
<tbody>
<tr>
<td>The first level sensing</td>
<td>50μs</td>
</tr>
<tr>
<td>Sensing one extra level</td>
<td>14μs</td>
</tr>
<tr>
<td>Data transfer of the first level</td>
<td>20μs</td>
</tr>
<tr>
<td>Data transfer of one extra level</td>
<td>10μs</td>
</tr>
<tr>
<td>Decoding latency on average</td>
<td>15μs</td>
</tr>
</tbody>
</table>

![Fig. 5. The illustration of read latency accumulation problem and read level characteristics.](a) Read latency comparison (b) Read level characteristics

D. Exploiting Temporal Read Level Locality

In order to propose an accurate estimation of the appropriate read level, we investigate read level characteristics for different P/E cycles and retention time. The appropriate read level of one page with given P/E cycles and retention is obtained by two steps. In the first step, RBERs are computed according to the error model of Gaussian noises [32], described in Equation 1, considering random telegraph noises [33] and retention errors [8], [34]. In the model, \( N \) and \( t \) represent P/E cycle and retention time, respectively, while \( x \) and \( x_0 \) represent the threshold voltage of the programmed states and that of the erased state, respectively. Other parameters are referenced from the settings in [32]: \( K_0 = 0.333, K_1 = 4 \times 10^{-4}, K_2 = 2 \times 10^{-6} \) and \( x_0 = 1.4 \).

\[
\begin{align*}
\mu &= K_0 K_1 (x - x_0) \ln(1 + t/t_0) N^{0.5} \\
\sigma^2 &= K_0 K_2 (x - x_0) \ln(1 + t/t_0) N^{0.6}
\end{align*}
\]

(1)

In the second step, appropriate read levels for RBERs computed in the first step are estimated from the result of decoding failure probability presented in [15]. For example, data with RBERs between 0.006 and 0.008 can be corrected by LDPC codes with the 3rd read level with a decoding failure probability smaller than \( 10^{-15} \). Thus, the appropriate read level of RBERs in this range is 3. The correspondence between RBERs and the appropriate read levels is shown in Table II.

Based on above simulation method, appropriate read levels of LDPC codes along with retention for different P/E cycles, are shown in Figure 5(b). It can be found that read level changes with retention and keeps appropriate for a long time ranging from several days to several months. The step size of a read level jumping to the next level is related to P/E cycles. For all P/E cycles, the read level changes infrequently in the studied period. From this result, it can be concluded that one read level applied in the last read is probably appropriate for successive reads, which we call the temporal read level locality. By exploiting this characteristic, the proposed method, LaLDPC, takes the read level applied in the last read as the start read level of the next read. We present details of LaLDPC in the next section.

III. LATENCY-AWARE LDPC

This section presents the design of our proposed Latency-aware LDPC (LaLDPC), as shown in Figure 6. One storage module and two functional modules are developed in LaLDPC: read level storage, mapping cache management, and LDPC assistant. The read level storage module stores the level information of each page while cache management module manages the value of read levels. The LDPC assistant module applies the information in read level storage module to assist LDPC decoding process. LaLDPC can be applied in all kinds of flash translation layers. This paper implements LaLDPC into DFTL as an example. In order to further strengthen the advantage of LaLDPC, a new mapping cache eviction algorithm is proposed and implemented in the cache management module.

![Fig. 6. System architecture of LaLDPC. There are three components in LaLDPC: one storage component to store read levels in mapping cache and two functional components to manage read levels and assist LDPC decoding.](A. The Storage of Read Levels

LaLDPC stores the last read levels applied in LDPC for all pages into the mapping cache entries in the controller. As one entry in the cache represents one page mapping, the read level is added into the end of each entry, as shown in Figure 6. For alignment consideration, four bits are used to represent...
the level. In our implementation, the read level ranges from 1 to 7. In LaLDPC, the information of read levels is only stored in the mapping cache but not in translation blocks based on DFTL. There are several reasons for this design. First, for timeliness consideration on read requests, the levels used by recent reads has high reference value, and mapping cache keeps these recent reads. Second, as translation blocks store the mapping entries for all addresses, large storage overhead has to be paid on the read levels. Third, read levels have to be updated when read failure happens with lower read levels. If these updates happen in translation blocks, extra writes will be induced and impact read performance.

B. Read Level Management and Utilization

This section introduces the functional modules in LaLDPC, read level management and utilization. The mapping cache management module of LaLDPC, as shown in Figure 6, estimates the read level of each page by leveraging the read level of the last read to this page. There are four management scenarios to initialize and update the read levels. The first scenario is that when one entry is added to the mapping cache because of a first read, the read level is stored into level bits of the corresponding page as the initial read level. The second scenario is the update of read levels. When the new read level of the page is different from the stored read level, level bits are updated to the new one. The third scenario is that when one entry is added to the mapping cache because of a recent write, level bits of the corresponding page are initialized to be the first read level as data have not been kept. The fourth scenario involves page migration in garbage collection (GC) and wear leveling, the read levels of pages in new addresses are reset to be the first read level. We further introduce GC in Section III-D.

The LDPC assistant module in LaLDPC assists the decoding process of LDPC decoder, as shown in Figure 6. It contains two sub-modules, read level determination and level difference detection, as shown in Figure 7. Read levels applied in LDPC decoder are determined and level difference detection assists the management of mapping cache in LaLDPC. As read level has to jump to the next level when LDPC decoding failure happens, the stored level bits should be updated to a newer one. The detailed read process with LaLDPC is listed as follows:

Step 1: Read request to mapping entries enters read level determination module;
Step 2: Read levels are determined to be different values in three cases. The first case is that when mapping cache hit happens, the starting level of read-retry steps in decoding is decided as the value of level bits. The second case is that when mapping cache miss happens, the starting level is set as the first read level. The third case is that when decoding failure happens in LDPC decoder, one extra level is added to perform memory sensing again;
Step 3: LDPC decoding failure happens and be back to the read level determination module to obtain the next read level. When the maximum read level is reached, read failure is returned to the host;
Step 4: Level difference detection module collects the read level from LDPC decoder and compares it with the stored read level to detect difference;
Step 5: No read level difference is detected and read result is directly returned to the host;
Step 6: Read level difference is detected and level bits of corresponding mapping entries are updated.

C. Latency-aware Mapping Cache Eviction

As traditional least recently used (LRU) cache eviction policy in DFTL can not be aware of the LDPC read latency of pages, this section presents a new cache eviction algorithm, as shown in Algorithm 1, to strengthen the performance advantage of LaLDPC. There are two considerations when choosing the item to evict in the proposed algorithm. From the analysis in Section II-C, the pages with high read levels induce more redundant read latency, which has larger space to be improved by LaLDPC. Thus, the first consideration prefers to evict pages with lower read levels, which ensures that high-read-level pages can exist in the cache as long as possible. This method can improve the hit rate of pages with high read levels but may sacrifice the overall access locality. In order to reduce this side effect, the second consideration is to make quite recently accessed entries fixed and not be evicted. Then, the cache is partitioned into two parts: the part with fixed entries and the part with unfixed entries. In this way, a part of locality can be reserved.

There are two possible cases in the proposed cache eviction algorithm, as shown in Figure 8. The first case is that there are entries that exist in the unfixed part and with the smallest read level of all stored levels. The last appearance of the entry of this kind is chosen to evict, which makes the choice by the first consideration suboptimal, as shown in Case 1 of Figure 8. The second case is that entries with the smallest read level all exist in the fixed part. Only entries in the unfixed part and with a larger read level can be chosen to evict. In this case, the algorithm chooses the last appeared entry with a read level as
small level as possible, as shown in Case 2 of Figure 8. This case integrates the partial locality in the second consideration. The chosen entry is not optimal to be aware of read latency but can benefit cache hit ratio.

In summary, the proposed algorithm searches the last cache entry with the lowest read levels within unfixed entries for eviction, as shown in Lines 6-12 of Algorithm 1. With the proposed algorithm, mapping cache can be aware of the LDPC latency of pages and keeps a part of locality in some degree, which ensures that the read levels can be utilized more efficiently in LDPC.

D. Garbage Collection

The connection between the proposed LaLDPC and garbage collection is two-fold. On one hand, LaLDPC can improve the performance of garbage collection by reducing read latency of valid pages in victim blocks. The valid pages in garbage collection are often cold data kept for long and have high error rates. Thus, valid pages often require high read latency. The situation can be relieved when the migrated pages hit the mapping cache in LaLDPC. Redundant read retries can be avoided and garbage collection performance can be improved. On the other hand, garbage collection changes the read levels of valid pages. As valid pages in victim blocks are decoded by LDPC in garbage collection, the error-free data can be obtained. Then, the read levels stored for these pages have to be reset to the initial read level when they are written to new blocks. Garbage collection decreases the read latency of future reads on valid pages.

E. Analysis of Performance and Overhead

This section analyzes read performance and overhead of LaLDPC with DFTL. Suppose the appropriate read level for current read is \( L_1 \) and that applied in the last read and stored in mapping cache is \( L_0 \) with \( L_0 \leq L_1 \). Read latency of the two reads are \( T_1 \) and \( T_0 \), respectively. The cache hit rate is denoted as \( r \). Read latency involved in LaLDPC is computed as \( T_{LaLDPC} = r \cdot \sum_{j=L_0}^{L_1} T_j + (1 - r) \cdot \sum_{j=1}^{L_1} T_j \) while that of existing LDPC is \( T_{LDPC} = \sum_{j=1}^{L_1} T_j \). Then, the latency difference between LaLDPC and the existing LDPC method is computed as \( T_{diff} = T_{LDPC} - T_{LaLDPC} = r \cdot (\sum_{j=1}^{L_1} T_j - \sum_{j=L_0}^{L_1} T_j) = r \cdot \sum_{j=1}^{L_0} T_j \). Read performance benefits of LaLDPC are decided by two factors: the cache hit rate and the difference between the stored level \( L_0 \) and the appropriate level \( L_1 \). For reads with higher cache hit rate and smaller difference, more redundant read-retries of LDPC can be reduced and read performance can be improved.

Overhead of LaLDPC comes from the storage component and the functional components. The storage overhead in LaLDPC is taken by the read levels in mapping cache entries. Assuming the size of one mapping cache entry is 8 Bytes, the portion of space taken by the four level bits is \( \frac{4}{8} \times 100\% = 6.25\% \). For mapping cache with the size of 256MB, level bits take 16MB storage space. As operations in controller are much faster than flash device, extra computational overhead caused by LaLDPC functional components can be ignored.

IV. Experiment

In order to evaluate effectiveness of the proposed technique, this paper simulates flash storage system using Disksim [21] with SSD extensions [20]. Experimental setup and parameters are presented in this section. The SSD system with the total size of 32GB is configured with eight packages, each of which has eight planes. Each plane contains 1024 blocks and each block has 64 pages. The page size is set to be 4KB. The
write latency for one page and block erase latency are set to be 900\(\mu s\) and 3.5ms for MLC flash, respectively [31]. Read latency for seven read levels in LDPC decoding are computed according to analysis in Section II-C and presented in Table II based on the latency of Table I. Statistics and performance results of fifteen real workloads from MSR [35], [36] and UMass [37] are collected. Specifications of the workloads including read/write ratio and duration time are listed in Table III.

<table>
<thead>
<tr>
<th>read level</th>
<th>RBER</th>
<th>read latency ((\mu s))</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>&lt; 0.005</td>
<td>85</td>
</tr>
<tr>
<td>2</td>
<td>(0.005, 0.006)</td>
<td>109</td>
</tr>
<tr>
<td>3</td>
<td>(0.006, 0.008)</td>
<td>133</td>
</tr>
<tr>
<td>4</td>
<td>(0.008, 0.009)</td>
<td>157</td>
</tr>
<tr>
<td>5</td>
<td>0.009, 0.01</td>
<td>181</td>
</tr>
<tr>
<td>6</td>
<td>0.01, 0.012</td>
<td>205</td>
</tr>
<tr>
<td>7</td>
<td>0.012, 0.013</td>
<td>229</td>
</tr>
</tbody>
</table>

The performance results for four methods are collected, i.e., LDPC-in-SSD, Ideal, LaLDPC\(_{LRU}\), and LaLDPC\(_{new}\), the implementations of which are listed as follows:

- LDPC-in-SSD is the current fine-grained progressive LDPC method [15];
- Ideal assumes that the appropriate RVs are known in advance and directly applied for LDPC read;
- LaLDPC\(_{LRU}\) is the proposed LaLDPC method implemented into DFTL with LRU cache eviction algorithm;
- LaLDPC\(_{new}\) is the proposed LaLDPC method implemented into DFTL with the proposed cache eviction algorithm.

The evaluation contains two aspects: the basic experiment and sensitivity study. In the basic experiment, parameters are set as follows. The cache size in LaLDPC\(_{LRU}\) and LaLDPC\(_{new}\) is set to be 64MB. The fixed entry length of mapping cache in the eviction algorithm is set to be 2000. SSD is assumed to be in the middle stage of flash life with P/E cycle of 10000 and data retention of one year, which can reach RBERs that require LDPC read levels around 3 and 4 based on the model in Section II-D. In order to generate high RBERs that require LDPC soft sensing, the technique of programming speedup [38] is used to accelerate wears of flash cells. The programming step size is set to be 0.38V.

In order to comprehensively investigate read performance of these methods, sensitivity study is performed on more settings of three parameters: mapping cache sizes, fixed entry lengths in the new cache eviction algorithm, and stages of SSD life. Besides the parameters in the basic experiment, the cache size is set to be two extra values of 32MB and 128MB. The fixed entry length is set to be 500 and 5000 as additional values. The SSD lifetime is set to be in two extra stages of early and late, with P/E cycles of 8000 and 12000 and the same retention as the basic experiment.

V. Results and Analysis

This section presents experimental results for the four methods: LDPC-in-SSD, Ideal, LaLDPC\(_{LRU}\), and LaLDPC\(_{new}\). As different workloads have varied SSD read patterns and cache characteristics, which closely relate to the performance benefit of LaLDPC\(_{LRU}\) and LaLDPC\(_{new}\), we first present statistical results of the fifteen workloads to provide references for performance results. Then, read performance results of the basic experiment are presented. The read performance metrics include flash read latency and SSD read response time. Flash read latency computes the overall latency taken by flash sensing, data transferring, and LDPC decoding. SSD read response time is the time from that the reads are issued in host to that read results are returned to the host. Then, results of system response time are presented to show the overall impact of LaLDPC\(_{LRU}\) and LaLDPC\(_{new}\) on SSD performance. System response time computes the average response time of both reads and writes. Lastly, results of sensitivity study are presented on the three parameters mentioned in Section IV.

A. Statistics of Workloads

This section presents three statistics of workloads. The first statistic is the distribution of the appropriate read levels, as shown in Table IV. We define hard read as the read with read level equal to one and soft read as that with read level larger than one. The difference between hard and soft reads is decided by the data retention patterns in workloads. As we assume one-year initial retention, data need soft reads at the beginning of workloads. When data are updated, read levels are reset to be one and become hard reads. This statistic reflects the potential performance improvement of workloads because only performance of soft reads can be improved.

The second statistic is the ratio of soft-start reads, which indicates the LDPC reads that begin from soft read levels. Only with the assistant of LaLDPC\(_{new}\) and LaLDPC\(_{LRU}\), LDPC read can start from soft read levels. The ratio of soft-start reads reflects that how many reads in the workloads benefit from the two LaLDPC methods. From the result in
The read performance is evaluated in two aspects, flash read latency and SSD read response time, and the results are shown in Figure 11. We first analyze the performance benefits brought by LaLDPC<sub>new</sub> and LaLDPC<sub>LRU</sub> compared with LDPC-in-SSD and then present the comparison of read performance results with LaLDPC<sub>new</sub> and LaLDPC<sub>LRU</sub>.

Read performance comparison between LDPC-in-SSD and the proposed LaLDPC methods. Read latency benefits from LaLDPC<sub>new</sub> and LaLDPC<sub>LRU</sub>, shown in Figure 11(a), are affected by two factors: the ratio of soft appropriate read levels and cache hit ratio, statistics of which are shown in Section V-A. The first factor decides how many latencies can be improved and decided by access patterns of workloads. For the workloads with high ratios, such as rsrch1, F1, and F2, there are a lot of redundant read latencies that can be improved and decided by access patterns of workloads.

For the workloads with high ratios, such as rsrch1, F1, and F2, there are a lot of redundant read latencies that can be improved and decided by access patterns of workloads.

SSD read response time, as shown in Figure 11(b), is also affected by read ratios of workloads besides the two factors mentioned above. Among workloads with similar values in above two factors, such as wdev0, ts0, and hm0, workloads wdev0 and hm0 with higher read ratios achieve more read performance benefits from LaLDPC<sub>new</sub> and LaLDPC<sub>LRU</sub> compared with LDPC-in-SSD.

Read performance comparison between LaLDPC<sub>LRU</sub> and LaLDPC<sub>new</sub>. Based on LaLDPC, the main difference between LaLDPC<sub>LRU</sub> and LaLDPC<sub>new</sub> exists in cache hit ratios because of different cache eviction algorithms. From the results in Figure 10 and Figure 11, it can be concluded that LaLDPC<sub>new</sub> leads to lower cache hit ratios but can improve flash read performance compared with LaLDPC<sub>LRU</sub>. However, for the workload F2 with little writes, the read performance of LaLDPC<sub>new</sub> is no better than LaLDPC<sub>LRU</sub>. This may be because that LaLDPC<sub>new</sub> decreases the read locality, which affects performance improvement because of the special access pattern of this workload. This performance degradation can be relieved by keeping more fixed entries in mapping cache.

In summary, the proposed two methods can achieve more performance benefits for workloads with higher read ratios, higher ratios of soft reads, and higher cache hit ratios. On average, LaLDPC<sub>new</sub> can remove 56% of redundant read latencies.
performance by comparing with the Ideal results and can improve read performance of LDPC-in-SSD by 18%.

C. System Response Time

This section presents and analyzes the results of SSD response time, as shown in Figure 12. As read latency to flash affects on the latency of other parts in SSD, such as the request queuing time in controller and the write latency because of utilization of DFTL and the new cache eviction algorithm, the overall SSD response time may also be affected. From Figure 12, it can be observed that improvements of SSD system response time by $LaLDPC_{LRU}$ and $LaLDPC_{new}$ are similar with flash read performance but have small differences for some workloads, such as $ts0$ and $hm1$, which is decided by the access patterns of workloads. $LaLDPC_{new}$ shows improved system performance of these workloads than $LaLDPC_{LRU}$. On average, about 24% of system response time in LDPC-in-SSD can be improved. Above results verify that LaLDPC is effective on improving not only read performance but also system performance of SSDs.

D. Sensitivity Study

This section analyzes the sensitivity of read performance improvement by $LaLDPC_{new}$ on three parameters: mapping cache size, fixed entry length of mapping cache and stages of flash life. The experimental results are shown in Figure 13. From the results of different cache sizes, it can be found that when a larger mapping cache is used, more performance benefits of $LaLDPC_{new}$ and $LaLDPC_{LRU}$ can be achieved. This is because that read levels of more mapping entries can be stored in the larger cache, which brings higher cache hit rates for soft reads and more read latency reductions. From the results of different fixed cache lengths, it can be found that when more entries are fixed in mapping cache, advantages of the new cache eviction algorithm $LaLDPC_{new}$ are decreased. This is because that with larger fixed entry length, long-latency reads stay in the mapping cache for a shorter time, which decreases benefits of $LaLDPC_{new}$. However, the differences among three fixed entry lengths are not obvious, which means that read performance are not sensitive to this parameter. From the results of different stages of flash life, it can be found that for SSDs in the late life stage, high performance improvements can be achieved. This is because that when SSDs are in this stage, latency accumulation becomes more severe. And $LaLDPC_{new}$ and $LaLDPC_{LRU}$ have the capability to remove more redundant latencies. Compared with the other two parameters, LaLDPC is more sensitive to this parameter.

Besides, there is a tradeoff between performance and overhead in LaLDPC methods. For example, larger cache size takes more space overhead although with better performance.
These parameters should be comprehensively considered in real SSDs.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Read Performance</th>
<th>LaLDPC new</th>
<th>LaLDPC LRU</th>
<th>LDPC-in-SSD</th>
</tr>
</thead>
<tbody>
<tr>
<td>Cache size</td>
<td>0.2</td>
<td>0.4</td>
<td>0.6</td>
<td>0.8</td>
</tr>
<tr>
<td>Fixed entry length</td>
<td>0.2</td>
<td>0.4</td>
<td>0.6</td>
<td>0.8</td>
</tr>
<tr>
<td>Flash lifetime</td>
<td>0.2</td>
<td>0.4</td>
<td>0.6</td>
<td>0.8</td>
</tr>
</tbody>
</table>

Fig. 13. Read performance results of sensitivity study on three parameters normalized to LDPC-in-SSD.

VI. RELATED WORK

Existing work to optimize read performance with LDPC in flash-based SSDs can be mainly categorized into three groups. The first group decreases the RBERs of data, as lower RBERs require less read latency. EC-Cache [39] proposes to store errors detected in last reads into an error correction cache. In the next read, these errors can be corrected before LDPC decoding. Li et al. [38] analyzed the cost models in flash reads and writes and proposed to obtain low RBERs of pages by applying fine-grained programming steps in writes. FlexLevel [40] optimizes read performance by strengthening error-tolerant capability of flash cells with less states. Only three states rather than traditional four states are used to store raw data in MLC flash, which can lower RBERs of flash pages. The refresh methods [41]–[43] are proposed to periodically correct data with long retention and reprogram them into new blocks, which can reduce retention-induced errors.

The second group optimizes the flash sensing process by searching the appropriate RVs for decoding. LDPC-in-SSD [15] proposed to progressively apply fine-grained levels in LDPC reads. When the read with a lower level fails, the next level with several extra RVs is applied for flash sensing. This technique improves the read performance in the early stage of SSD life. In order to read out correct data with BCH codes, Cai et al. [34] proposed to record the optimal threshold voltages (OPTs) of the last programmed page in each block into the DRAM buffer of flash controller. The stored OPTs are applied on all page reads in the block, which largely reduces the response time for successful reads. Peleato et al. [44] presented a mathematical model based on the RVs in last read to adaptively estimate the appropriate RVs of current read. This paper belongs to this ground. Different from above works, the proposed LaLDPC method exploited the read level locality characteristics for LDPC read performance improvement.

The third group optimizes LDPC decoding algorithm. Sun et al. [45] exploited intra-cell error characteristics to speed up LDPC decoding by reducing overall error probability and decoding latency. REAL [46] incorporates numeric-correlation characteristic of different error patterns in most significant bit and least significant bit of MLC flash into the message passing process of the decoding. LDPC decoder can achieve successful decoding faster with the extra information provided by this characteristics. Aslam et al. [47] proposed a two-round LDPC decoding process by reusing the read-back voltages and the decoded results for flash cells from retention-induced failure, which can improve the decoding success probability and further read performance. In this paper, we did not change the design of LDPC decoder but optimized the communication mechanism between LDPC decoder and NAND flash with little overhead to improve read performance.

VII. CONCLUSION

In order to solve the latency accumulation problem caused by existing LDPC methods, this paper proposes LaLDPC aiming at reducing redundant read-retries that cause latency accumulation. LaLDPC stores RVs in last read into FTL and apply them in future reads. LaLDPC has been implemented into DFTL as an example. In order to further strengthen the capability of LaLDPC, a new cache eviction algorithm is proposed. Experimental results show that the proposed method significantly improves read performance of SSDs compared with the existing LDPC method.

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