

CLERECO INSTITUTIONAL REPOSITORY

[Article] Performance and Reliability Analysis of Cross-Layer Optimizations of NAND Flash Controllers

Original Citation:

Davide Bertozzi, Stefano Di Carlo, Salvatore Galfano, Marco Indaco, Piero Olivo, Paolo Prinetto, and Cristian Zambelli. 2015. Performance and Reliability Analysis of Cross-Layer Optimizations of NAND Flash Controllers. ACM Trans. Embed. Comput. Syst. 14, 1, Article 7 (January 2015)

doi: 10.1145/2629562

This version is available at: http://www.clereco.eu/images/publications/TECS.2015.2629562.pdf Since: January 2015:

Publisher: ACM

Published version: DOI: <u>http://dx.doi.org/10.1109/10.1145/2629562</u>

Terms of use: This article is made available under terms and conditions applicable to Open Access Policy Article ("Public - All rights reserved"), as described at http://www.clereco.eu/publications/item/70

Publisher copyright claim:

© 2015 ACM. Personal use of this material is permitted. Permission from ACM must be obtained for all other uses, in any current or future media, including reprinting/republishing this material for advertising or promotional purposes, creating new collective works, for resale or redistribution to servers or lists, or reuse of any copyrighted component of this work in other works

(Article begins on next page)

Performance and Reliability Analysis of Cross-Layer Optimizations of NAND Flash Controllers

DAVIDE BERTOZZI, Università di Ferrara STEFANO DI CARLO, SALVATORE GALFANO, and MARCO INDACO, Politecnico di Torino PIERO OLIVO, Università di Ferrara PAOLO PRINETTO, Politecnico di Torino CRISTIAN ZAMBELLI, Università di Ferrara

NAND flash memories are becoming the predominant technology in the implementation of mass storage systems for both embedded and high-performance applications. However, when considering data and code storage in Non-Volatile Memories (NVMs), such as NAND flash memories, reliability and performance become a serious concern for systems designers. Designing NAND flash-based systems based on worst-case scenarios leads to waste of resources in terms of performance, power consumption, and storage capacity. This is clearly in contrast with the request for runtime reconfigurability, adaptivity, and resource optimization in modern computing systems. There is a clear trend toward supporting differentiated access modes in flash memory controllers, each one setting a differentiated tradeoff point in the performance-reliability optimization space. This is supported by the possibility of tuning the NAND flash memory performance, reliability, and power consumption through several tuning knobs such as the flash programming algorithm and the flash error correcting code. However, to successfully exploit these degrees of freedom, it is mandatory to clearly understand the effect that the combined tuning of these parameters has on the full NVM subsystem. This article performs a comprehensive quantitative analysis of the benefits provided by the runtime reconfigurability of an MLC NAND flash controller through the combined effect of an adaptable memory programming circuitry coupled with runtime adaptation of the ECC correction capability. The full NVM subsystem is taken into account, starting from a characterization of the low-level circuitry to the effect of the adaptation on a wide set of realistic benchmarks in order to provide readers a clear view of the benefit this combined adaptation may provide at the system level.

Categories and Subject Descriptors: C.4 [Performance of Systems]: Design Studies

General Terms: Reliability, Performance, Design

Additional Key Words and Phrases: Adaptable memory controllers, ECC, NAND flash memories

ACM Reference Format:

Davide Bertozzi, Stefano Di Carlo, Salvatore Galfano, Marco Indaco, Piero Olivo, Paolo Prinetto, and Cristian Zambelli. 2015. Performance and reliability analysis of cross-layer optimizations of NAND flash controllers. ACM Trans. Embedd. Comput. Syst. 14, 1, Article 7 (January 2015), 24 pages. DOI: http://dx.doi.org/10.1145/2629562

1. INTRODUCTION

The application of NAND flash memory technology [Atwood et al. 1997] has faced a surprising increment, far beyond what was expected when it was originally introduced. One popular example is Solid State Disks (SSD) that feature the advent of Multi-Level Cell (MLC) NAND flash memories to store a large amount of data in flash [Ouyang et al. 2014]. However, with flash memory storage capacity roughly doubling every 18 months, designers face challenging performance and reliability problems [Micheloni et al. 2010; Mielke et al. 2008; Lee et al. 2003; Bez et al. 2003; Irom and Nguyen 2007]. MLC flash memory is require high programming time and provide reduced endurance when compared to old Single-Level Cell (SLC) devices. The Raw Bit Error Rate (RBER) of a MLC flash memory is around 10^{-6} [Cooke 2007], at least two orders of magnitude worse than that of an SLC device [Dan and Singer 2003]. These problems are further amplified by flash file systems that are often stressed by frequent write requests of small amounts of data [Di Carlo et al. 2011].

State-of-the-art NAND flash memories are tightly cost optimized. The internal operations of the memory are mostly defined at design-time based on worst-case scenarios that comply with industry standards (i.e., ONFI [ONFI Workgroup 2012]). However, a fixed system configuration based on worst-case scenarios leads to waste of resources in terms of performance, power consumption, and storage capacity. This is clearly in contrast with the request for runtime reconfigurability, adaptivity, and resource optimization in modern computing systems [Cardoso and Hübner 2011]. New mobile usage models in today's complex embedded systems require the execution of multiple use cases on the same device and adaptation to the often nonpredictive behaviors of today's complex applications [Henkel et al. 2011]. They require seamless integration of safety/time-critical functionalities with noncritical functionalities, with each one demanding different requirements from the storage system [Sampson et al. 2013]. Moreover, NAND flash reliability is not constant; it changes over the device lifetime. This must be taken into account to enable high optimization in these devices. We clearly see a trend toward supporting differentiated access modes in flash memory controllers for MLC NAND flash devices, with each one setting a differentiated tradeoff point in the performance-reliability optimization space. This trend is confirmed by some commercial devices that already enable additional levels of flexibility selectable by the user at boot-time. Mainly, these devices enable speed/power consumption optimization by changing the memory bus interface speed (e.g., Micron MT29F16G08ABABA NAND Flash) and the storage model (i.e., choosing SLC or MLC writing schemes [Samsung 2012]). Concurrently with these solutions, several researchers focused on the optimization of the flash write algorithms [Liu et al. 2012] to guarantee a performance-reliability tradeoff and on the use of adaptive ECC schemes to trade off storage space and performance for higher error correction capability [Chen et al. 2009; Fabiano et al. 2013]. Some studies also investigated how mechanisms that enable applications to store data approximately also enable improved performance whenever high-precision storage is not required [Sampson et al. 2013]. However, a clear analysis of the benefits of combining optimized write algorithms with runtime-adaptable ECC schemes is still missing in the literature, thus preventing a clear understanding of the benefits an MLC NAND flash controller would achieve by implementing this higher level of configurability. Only a few studies attempted work in this direction. Pan et al. [2011] presented a first attempt at analyzing joined limited adaptation of the flash programming step voltage coupled with programmability of the ECC, whereas a very preliminary study presented by Zambelli et al. [2012] introduced increased adaptation at the flash physical layer.

The goal of this article is to perform a comprehensive analysis of the benefits achievable by exploiting the runtime reconfigurability of an MLC NAND flash controller through the combined effect of an adaptable memory programming circuitry coupled with runtime adaptation of the ECC correction capability. Runtime reconfigurability is implemented through the definition of a set of access modes, each one setting a specific tradeoff among read throughput, write throughput, Uncorrectable Bit Error Rate (UBER), and power consumption. Rather than focusing on the architectural implementation of the considered access modes, this article focuses on the characterization of the performance-reliability tuning range achievable with them. To the best of our knowledge, this is the first time that runtime adaptation is extended to the full Non-Volatile Memory (NVM) subsystem, and a comprehensive study to quantitatively analyze the effect of this adaptability while considering a wide set of benchmarks is carried out.

The article first analyzes the tradeoff that can be achieved by considering write algorithm adaptivity and ECC adaptivity in isolation and then analyzing the combination of the two adaptation mechanisms. For this purpose, a comprehensive modeling and simulation framework has been set up for both analog and digital parts of an MLC NAND flash memory subsystem in a homogeneous 45nm industrial technology substrate. Furthermore, the benefit of the defined access modes on the software stack has been evaluated within a simulation framework based on the YAFFS2 (Yet Another Flash File System version 2) flash file system. A set of different software benchmarks with different requirements in terms of storage system have been simulated, highlighting the benefits they can achieve in terms of performance (i.e., read throughput and write throughput on the flash memory) and power consumption in the full NAND flash memory subsystem.

The article is organized as follows: Sections 2 and 3 respectively propose the result of the characterization of different flash programming algorithms and of an adaptive ECC subsystem in isolation. Section 4 explores the tradeoffs proposed by cross-layer optimization in the NAND memory controller. Section 5 shows the performance of the proposed system on a set of real-life applications; finally, Section 6 summarizes the main contributions of this work and concludes the article.

2. CHARACTERIZATION OF ADAPTIVE FLASH PROGRAMMING ALGORITHMS

The physical management subsystem of the memory controller interacts with the High-Voltage (HV) analog circuitry of the NAND flash memory, which is the block in charge of generating the voltage waveforms required to read, program, and erase the memory cells. It issues commands to a control FSM or to an embedded microcontroller in the flash device. Without loss of generality, in this work, we target a 2-bit-per-cell NAND flash memory [Mielke et al. 2008]. It stores 2 bits of information per cell by defining a set of threshold voltage levels (V_{TH}) identified by the statistical distributions L0–L3 of Figure 1.

An erase operation places all cells of a block at level L0. The following program operations set the threshold voltages of the selected cells at one of the three levels, L1–L3, according to the data that must be programmed. The cell programming operation is achieved through a standard algorithm called the Incremental Step Pulse Programming Standard Verify (ISPP-SV) [Micheloni et al. 2010]. A voltage pulse of predefined amplitude and duration is applied to the gate of each programmed cell. Afterward, a verify operation takes place. It verifies whether the V_{TH} of the programmed cells exceeds a predefined verify level V_{VFY} , as depicted in Figure 2. Since we work with a 2-bit-per-cell MLC architecture, three verify levels (i.e., VFY1, VFY2, VFY3) are defined and must be verified. If the verify is successful, the cells have reached the desired distribution level, and they are excluded from the following pulses through the so-called program-inhibition technique [Micheloni et al. 2010]. Otherwise, another cycle of ISPP is applied after incrementing the programming voltage of Δ ISPP.

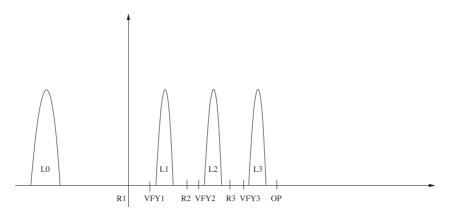


Fig. 1. Threshold voltage distributions in an MLC NAND flash. Read levels (R1, R2, and R3), Verify levels (VFY1, VFY2, VFY3), and Over-Programming level (OP) are pointed out.

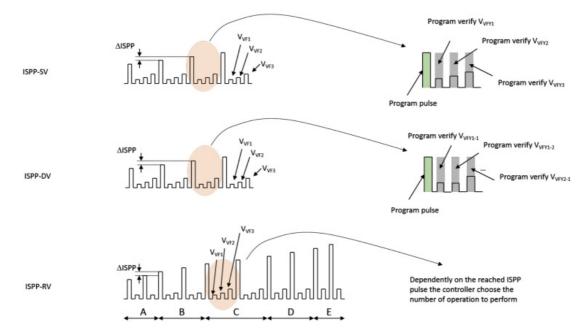


Fig. 2. Comparison of the different ISPP algorithms exploited to expose NAND flash reliability-performance tradeoffs.

Due to technological variations, V_{TH} is not perfectly related to the amplitude of the ISPP pulse. There are "fast" cells that reach the verify level with few program pulses and "slow" cells that require more pulses. Both behaviors represent a threat to the reliability of program operation. In fact, the threshold voltage distributions of the L1–L3 levels significantly deviate from an ideal Gaussian shape. They often cross the distribution read levels (R1–R3) and cause bit errors. A solution for increasing ISPP programming accuracy is the ISPP Double Verify (ISPP-DV) algorithm presented by Micheloni et al. [2010] and Miccoli et al. [2011]. The bit-line voltage of the selected cells is modulated in order to partially decrease the Δ ISPP step using a prior verify level with slightly lower voltage than the original verify level. As a result, a more compact threshold voltage distribution can be obtained (see Figure 2).

Another concern of MLC architectures is to decrease the gap in terms of writethroughput with respect to SLC memories. Both the ISPP-SV and, to a larger extent, the ISPP-DV require a large number of verify operations per single ISPP step. An interesting solution to avoid unnecessary verify operations is to use the ISPP Reduced Verify (ISPP-RV) write algorithm proposed by Micheloni et al. [2010]. In the ISPP-RV algorithm, the number of verify operations is initially small and is automatically increased based on the number of ISPP steps the algorithm performs (see Figure 2). This algorithm is able to provide increased programming speed at the cost of reduced robustness against page errors.

Currently, the flash programming algorithm is set at fabrication time in the flash controller. It is stored in a code-ROM integrated in the same memory die and is executed by an embedded microcontroller. Implementing more than one write algorithm simply requires the device to store more than one algorithm in the code-ROM by slightly increasing its capacity. Having more than one programming algorithm stored in the code-ROM also requires a mechanism to select the desired algorithm for a transaction or set of transactions. The ONFI 3.0 standard for NAND flash memories [ONFI Workgroup 2012] envisions the possibility of implementing both new vendor-specific commands and special commands in the case of the development of innovative writing methodologies. ONFI 3.0 could therefore be exploited to implement the write algorithm selection through three dedicated commands such as: 0x80 = Program with ISPP-SV, 0x81 = Program with ISPP-DV, and 0x82 = Program with ISPP-RV. The choice of the programming algorithm can be also implemented through dedicated configuration registers. This approach is consistent with the methodologies exploited in today's NAND flash memory controllers to provide reconfiguration options (e.g., changing the DDR protocol timings [Evatronix 2012] or the storage paradigm [Samsung 2012]).

It is worth mentioning here that the scaling of the flash cell geometry poses several threats to the reconfigurability of the NAND flash programming algorithms that must be carefully considered. Especially when moving to ultra-scaled devices, it is easy to incur reliability side effects. However, for technology nodes such as the 4X nm (considered in this work) and the 3X nm, it is still possible to leverage the number of verify pulses to expose tradeoffs in the reliability-performance domain. When moving to the 2X nm technology nodes and beyond, a refinement of the entire programming algorithms is required due to the increased complexity of the internal NAND flash microcontroller. However, even for these devices, it is common practice for the industry to provide on the same chip more than one programming, erasing, and even reading algorithm to be chosen by the NAND flash controller manufacturer acting upon results from test modes that are not accessible by the final users [Micheloni et al. 2010]. This suggests that programming algorithm adaptation will also be exploitable for these devices.

The next sections illustrate the models and simulations performed to characterize how different programming algorithms impact the RBER and power consumption of the memory.¹

2.1. Compact and Accurate NAND Flash Model

Our case study targets a 2-bit-per-cell NAND flash memory featuring a 45nm manufacturing process designed for low-power applications. The simulation environment includes two modules: (1) the *HV* subsystem exploited to generate the voltages required for the programming algorithms (including the verify stage), and (2) a *compact model* of the NAND flash memory with array simulation capability.

The HV module is the analog core of a NAND flash memory. Modifying or reading the number of electrons stored in the floating gate requires the generation of a set of

¹The programming of MLC NAND flash also depends on the strategy adopted for loading the data to write into the memory. Without loss of generality, we chose to investigate and explore the ISPP full-sequence strategy [Micheloni et al. 2010] instead of the two-rounds strategy because it reduces simulation time and provides faster postprocessing of experimental results.

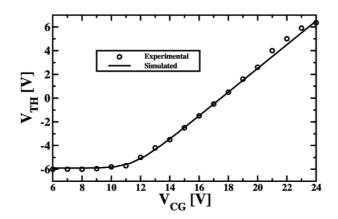


Fig. 3. Fitting results of the NAND flash compact model with experimental data during an ISPP-SV operation featuring 7μ s pulses, $1V \Delta ISPP$.

bias voltages with a desired precision, timing, and granularity. Moreover, since many voltages have a value larger than the NAND power supply, several charge pumps are required. To obtain highly accurate estimations of the energy consumption of each ISPP algorithm considered in this work, we simulated the program charge pump, the inhibit charge pump, the verify charge pump, and the regulators/limiting systems according to the guidelines proposed by Kang et al. [2008]. All blocks have been implemented in HSPICE using STM-45nm technology library [CMP 2012]. The power consumption of each pump extracted from the SPICE simulation during the various stages of the ISPP algorithms was then fed into a NAND flash power modeling framework based on the equation set provided by Mohan et al. [2010]. As input parameters of the model, we assumed a low-power NAND flash supplied with $V_{DD} = 1.8$ V using an ISPP algorithm starting from 14V to 19V, using \triangle ISPP steps of 250mV. The same settings hold for all considered programming algorithms. The simulated HV subsystem has been designed to work with all algorithms. In fact, in a NAND flash device, the timing and sequence of the analog circuitry operations are driven by the embedded microcontroller/FSM by means of a set of interface registers required to generate the enable signals for the charge pumps. Switching from one ISPP algorithm to another does not require a modification of the HV subsystem. It only implies a different sequence of enable signals notified through the same register interface.

Together with the HV module, we developed a compact model of the NAND flash cells partially based on Spessot et al. [2010]. It includes variability effects typical of nanoscaled memories, and it allows us to simulate array functionalities during a page programming operation. The considered variability effects include width and length geometrical variations of FG-MOS transistors; nonhomogeneity of tunnel oxide and substrate doping; tunneling caused by the electron injection granularity process into the cells floating gate; cell-to-cell interference caused by cross-talk between adjacent floating gates; and aging effects due to repeated program-erase (PE) cycling, which typically degrades the RBER. All these effects contribute to significantly broaden the gaussian distributions related to the programmed threshold voltage levels within the array, thus negatively impacting the RBER. A comprehensive description of the considered model is provided in the Appendix. The model has been validated by fitting it against experimental data collected from Spessot et al. [2010], as showed in Figure 3, where the cell voltage threshold is plotted during an ISPP operation for a 41nm NAND flash technology. The experimental data provided by Spessot et al. [2010] consider the study of the cell threshold voltage evolution after the application of an ISPP that ranges between 6V and 24V using Δ ISPP=1V, which is different from the specification of our

flash memory. Nevertheless, our ISPP range (14Vn19V) represents a subset of the one proposed in Spessot et al. [2010]. Moreover, modifications of \triangle ISPP do not change the physical structure of the memory cells [Cooke 2007], thus allowing us to fit our model against the data provided Spessot et al. [2010].

It is worth mentioning here that the proposed model does not take into account reliability issues due to retention errors. Retention errors are one of the major contributors to NAND flash memory reliability. However, the guidelines provided by JEDEC Solid State Technology Association [2011] about the modeling of the NAND RBER during retention clearly indicate that this value is an offset of the RBER obtained at a specific PE cycle according to the following model:

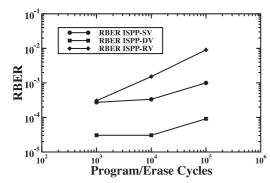
$$RBER(PE, t_{ret}) = RBER_{wr}(PE) + B_o \left(PE^n \cdot t_{ret}\right)^m$$
(1)

where t_{ret} is the page retention time measured in hours, PE is the instantaneous PE cycles count of the page, m is a coefficient whose numerical value is usually between 1 and 2, n is a power-law coefficient for PE cycles, RBER_{wr} is the error rate observed at $t_{ret} = 0$, and B_o is a scale factor, which depends on the target technological process. Therefore, to capture the impact of retention errors, a prior characterization of RBER_{wr} is required. Adding the retention contribution will introduce an offset that would not significantly change the definition of the tradeoff points while increasing the computational complexity of the model. Moreover, the simulations performed in this article consider continuous benchmarks that feature short relaxation times between read and write operations, therefore limiting the impact of the retention errors.

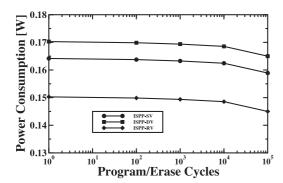
2.2. Characterization of Programming Algorithms

Power consumption, RBER, and the average page write time of the flash when using the ISPP-SV, ISPP-DV, and ISPP-RV algorithms have been characterized by means of the models presented in Section 2.1. Such parameters are derived as a function of the PE cycles of the memory and reported in Figure 4. Figures 4(a), 4(c), and 4(e) provide average results obtained by simulating a random write pattern on a 4KB memory page, whereas Figures 4(b), 4(d), and 4(f) show equivalent results considering all cells of the page programmed at one of the three threshold voltage distributions presented in Figure 1. This allows us to highlight how flash performance changes depending on the written data.

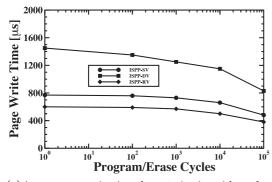
Figures 4(a) and 4(b) show the RBER estimated when programming the considered memory page. The two figures clearly show that the choice of a particular programming algorithm produces a significant modification (up to one order of magnitude) of the page's RBER. The power consumption of the memory device during a program operation with different programming algorithms is reported in Figures 4(c) and 4(d). Power measures do not include I/O pins and digital portions of the flash, which are irrelevant in the comparative analysis. The most power-demanding write strategy is the ISPP-DV. It introduces a 4% power consumption increment with respect to the standard ISPP-SV algorithm. This is due to the increased usage of the read charge pump circuitry in the HV subsystem. Nevertheless, it does not represent a major source of power drain in the overall system consumption context. The least power-demanding write strategy is instead the ISPP-RV, since the HV circuitry is enabled for a shorter lapse of time due to the reduced number of verify operations. The page write time has been calculated considering a fixed cell verify time (i.e., page read operation) of $30\mu s$. Results reported in Figures 4(e) and 4(f) show that the fastest algorithm is the ISPP-RV due to the reduced number of verify operations, which generally slow down the writing process. It is worth pointing out that the average page write time decreases as aging increases due to the faster programming behavior of the memory cells [Micheloni et al. 2010]. This effect is tightly coupled with a reduction of overall memory reliability since bit



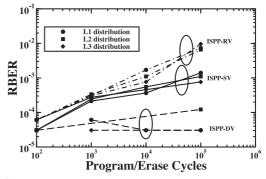
(a) Average RBER characterization with random write pattern.



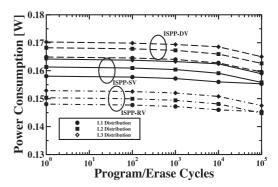
(c) Power consumption characterization with random write pattern.



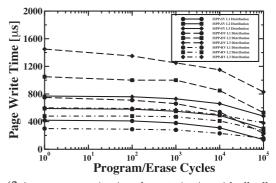
(e) Average page write time characterization with random write pattern.



(b) RBER characterization with all cells programmed with the same distribution level (i.e., L1, L2, or L3).



(d) Power consumption characterization with all cells programmed with the same distribution level (i.e., L1, L2, or L3).



(f) Average page write time characterization with all cells programmed with the same distribution level (i.e., L1, L2, or L3).

Fig. 4. RBER, power consumption, and average write time characterization for ISPP-SV, ISPP-DV, and ISPP-RV algorithms for a page of flash. For each measured parameter, two characterizations are performed based on the type of data programmed in the page. The left column reports an average characterization obtained when writing a random pattern in the memory page, whereas the right column reports a characterization obtained when programming all cells of the page at one of the three available distribution levels (L1, L2, and L3) in order to highlight the effect of the data pattern on the measured parameters.

errors tend to be more frequent [Mielke et al. 2008]. Table I summarizes the main results obtained from the characterization of the selected device.

The presented plots clearly show the potentials of an adaptive flash programming algorithm. By selecting a programming algorithm from among ISPP-SV, ISPP-DV, and ISPP-RV, we can easily trade off among RBER, power, and write throughput, with only incremental complexity added to the memory controller architecture.

Page write time (AVG) @ cycle 1	600μs (RV) 800μs (SV)
Page read time	$\frac{1,400\mu s(DV)}{75\mu s}$
Block erase time Maximum considered P/E cycles	4ms 100,000
Page Size	4KB + parity

Table I. NAND Flash Simulation Parameters

Note: Programming timings are provided at cycle 1.

3. CHARACTERIZATION OF AN ADAPTIVE ECC SUBSYSTEM

An adaptive ECC subsystem that enables us to modify the ECC correction capability in a selected range is another way of trading off NAND flash memory performance, reliability, and power consumption. In current SSDs, the ECC calculation time is often successfully hidden when the portion of errors is irrelevant compared to the effective dimension of the disk (e.g., at the beginning of the disk lifetime). However, when the number of errors due to aging of the NAND flash starts to increase, thus forcing the use of policies such as the read-retry to ease the role of the correction codes, then the ECC becomes a major system bottleneck. In this situation, multiple errors need to be corrected on multiple NAND flash devices, thus impacting on both SSD read and write throughput [Micheloni et al. 2013].

The ECC subsystem exploited in this article implements the adaptable Bose-Chaudhuri-Hocquenghem (BCH) ECC architecture presented by Fabiano et al. [2013]. BCH codes belong to the larger class of cyclic codes, which have efficient decoding algorithms due to their strict algebraic architecture [Bose and Ray-Chaudhuri 1960]. BCH codes perform corrections over single-bit symbols and perform better when bit errors are not correlated or randomly distributed. Several studies have reported that NAND flash memories manifest noncorrelated or randomly distributed bit errors over a page [Yaakobi et al. 2009]. BCH codes are therefore a perfect choice for correcting errors in these devices. The construction of a BCH code is based on Galois field $GF(2^m)$. Given a finite Galois field $GF(2^m)$ (with $m \ge 3$), a *t*-error-correcting BCH code, denoted as BCH[n, k, t], encodes a k-bit message to an n-bit codeword by adding r parity bits to the original message. The value of *m* is selected by finding the minimum value that solves the inequality $n - k < m \cdot t$, where $n = 2^m - 1$ and $r = m \cdot t$. Whenever $n = k + r < 2^m - 1$, the BCH code is called shortened or polynomial. In a shortened BCH code, the codeword includes fewer binary symbols than those the selected Galois field would allow. The missing information symbols are imagined to be at the beginning of the codeword and are considered to be 0.

The hardware BCH architecture exploited in this article enables us to encode/decode a full memory page, selecting the desired correction capability t in a given range of values. A detailed description of the employed ECC hardware architecture is outside the scope of this article and can be found in Fabiano et al. [2013]. The equation that governs the relation among t, RBER, and UBER of the flash device for a selected code is:

$$UBER = \frac{1}{n} \sum_{i=t+1}^{n} \binom{n}{i} \cdot RBER^{i} \cdot (1 - RBER)^{n-i}.$$
(2)

In our specific design, the ECC subsystem has been implemented to work on a full page of the flash (i.e., k = 4KB). We considered a target UBER equal to 1E-13, as in Mielke et al. [2008]. Based on Equation (2), Table II reports the correction capability required to achieve the target UBER considering the RBERs of the various

Table II. Correction Capability Required by the ECC to Achieve a Target UBER=1E-13

Alg/progr. cycles	1	100	1,000	10,000	100,000
ISPP-RV	1.000E-06/4	6.104 E- $05/13$	3.052E-04/31	1.526E-03/93	9.0332 E-03/399
ISPP-SV	1.000E-06/4	1.000E-06/4	2.747E-04/30	3.357E-04/33	1.000E-03/70
ISPP-DV	1.000E-06/4	1.000E-06/4	3.052E-05/10	3.052E-05/10	9.155E-05/17

Note: Every element of the table reports the memory RBERs for the different programming algorithms (random data pattern) as characterized in Section 2 and the required correction capability.

Table III. ECC Encoder and Decoder Area Footprint

	Area (μm^2)
Encoder	179586.97
Decoder	543625.58

Note: Synthesis has been performed using the STM-45nm technology library.

programming algorithms characterized in Section 2. Clearly, the correction capability required to satisfy the target UBER constraints increases over time. As expected, from the reliability standpoint, the worst performance is provided by the ISPP-RV algorithm. This algorithm requires, at the end of the life of the device, a correction capability of 399 errors per page. This value would require a considerable amount of hardware and performance resources, which leads to the conclusion that memory pages using the ISPP-RV algorithm necessarily provide reduced endurance. This result is in line with current trends for MLC memory endurance that report a typical limit of 10,000 or less PE cycles [Yaakobi et al. 2010; Grochowski and Fontana 2012]. For this reason, we selected a target maximum correction capability of 93 errors per page, corresponding to the requirement of the ISPP-RV algorithm at the end of life. Given the selected value of *k* and *t*, the resulting code is designed over GF(2¹⁶) (i.e., m = 16).

In the remainder of this section, the ECC subsystem is characterized to show the different tradeoffs offered by its programmability. It is worth mentioning here that our ECC implementation features 8-bit parallelism to meet the I/O parallelism of the target flash and 8-bit parallelism of the Chien machine to allow eight evaluations per clock cycle to speed up the decoding process. Moreover, both the encoder and the decoder are pipelined with the flash memory to optimize the performance of the block. Table III reports the area required for this block synthesized using Synopsys DesignCompiler with the STM-45nm [CMP 2012] technology library. The full design works at 100MHz clock frequency.

Let us start with the evaluation of the amount of redundancy (i.e., parity bits) introduced by the ECC. In the worst case (e.g., t = 93), the code needs to store $m \cdot t = 16 \cdot 93 = 1488$ bits = 186B. This accounts for about 83% of the spare area available on our device, which corresponds to 224B per page. ECC parity bits are not the only extra information stored in a flash memory. High-level functions such as file system management, bad blocks management, and wear-leveling need to save considerable amount of information. When the spare area is not enough, a certain amount of pages of the flash must be reserved, thus reducing overall flash capacity. As an example, YAFFS2, the file system selected for our analysis, must save 18 bytes of information for each data chunk of 2KB. Every page of our flash can store 2 chunks and requires 36 spare bytes. With UBER=1E-13 and t = 93, there are 38 spare bytes available to the file system, which is just enough for the YAFFS2 requirement. If additional functions, such as the wear-leveling, need to be managed or if increased reliability is required, the spare area may become too small. Looking at Figure 5, if reduced correction capability is required, either because the device is in the early stage of its life or because a more reliable programming algorithm is applied, the spare area occupation can be reduced

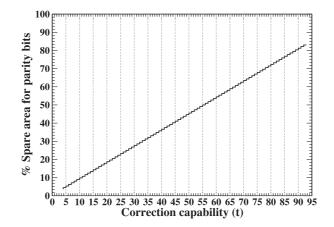


Fig. 5. Percentage of spare area dedicated to storing parity bits as a function of the selected correction capability.

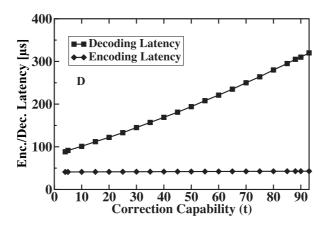


Fig. 6. Worst-case ECC encoding and decoding latency. Simulations have been performed at a clock frequency of 100MHz.

up to 78% (4.46% occupation for t = 4). This provides a high degree of freedom for the flash memory controller. It is worth noting that this does not represent the main parameter to take into account when optimizing the ECC subsystem; nevertheless, it is worth consideration in the optimization of the full flash subsystem design.

The choice of t also makes it possible to tune ECC latency. Figure 6 shows that, by carefully tuning the correction capability, the ECC subsystem can significantly save in decoding time compared to the worst case (t = 93). Simulations have been performed in the worst-case conditions (i.e., t errors injected into the last bits of the page to make sure that the full page must be checked to find the corrupted bits). The encoding latency is instead almost constant, regardless of the selected correction capability.

Similarly to the ECC latency, the ECC power consumption also can be traded-off by carefully selecting the correction capability. Figure 7 shows that, also in this case, we can save up to \sim 55% of decoding power consumption when reducing correction capability.

To conclude the characterization of the designed programmable ECC subsystem, Figure 8 reports the relation between UBER and RBER for the selected correction mode obtained by plotting Equation (2). The figure shows an additional degree of freedom that the controller can achieve in which the UBER also can be tuned together with the other parameters.

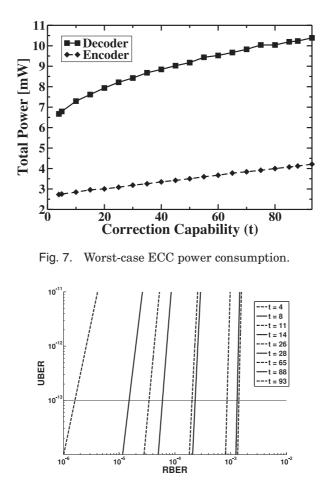


Fig. 8. RBER vs. UBER relationship for the selected code and selected correction modes.

4. CROSS-LAYER OPTIMIZED NAND FLASH ACCESS MODES

So far, we have considered the flexibility and tradeoffs that can be achieved by reconfiguring the flash programming algorithm and the ECC subsystem in isolation. However, by acting on their parameters at the same time, we show that it is possible to obtain higher optimization in terms of reliability, performance, and power consumption, thus identifying a set of differentiated access modes that can be configured in the memory controller and made available to the software stack.

Figure 9 provides an overview of how the NAND flash subsystem reacts when selecting different programming algorithms and ECC correction capabilities. Three main parameters of the flash are considered in Figure 9: (1) the UBER of the flash, (2) the read throughput (RT; i.e., the number of page read requests per second that the system is able to serve), and (3) the write throughput (WT; i.e., the number of write requests per second the system is able to serve).

If we consider the ISPP-SV programming algorithm with an ECC designed for an UBER of 1E-13 as a reference operating point, the following behaviors can be foreseen:

- —UBER worsens when lower values of t or programming algorithms with reduced verifications are used.
- -WT is mainly affected by the programming algorithm. As pointed out in Figure 6, the ECC encoding time is almost constant regardless of the selected correction capability.
- ---RT is mainly affected by the selected ECC correction capability, which directly affects the ECC decoding time (see Figure 6). It increases if a lower *t* is used.

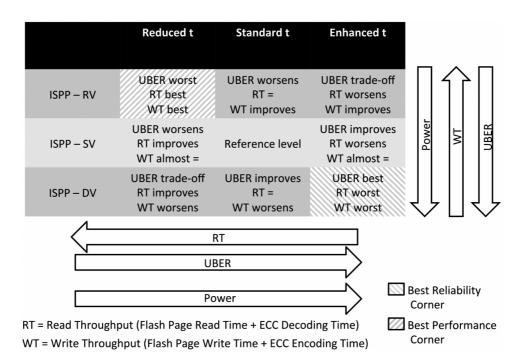


Fig. 9. Set of access modes provided when tuning the programming algorithm and the ECC correction capability in a cross-layer adaptation framework.

		PE cycles				
UBER	ISPP alg.	1	10^{2}	10^{3}	10^{4}	10^{5}
10^{-11}	RV	3	11	28	88	>93
	SV	3	3	26	29	65
	DV	3	3	8	8	14
10^{-13}	RV	4	13	31	93	399
	SV	4	4	30	33	70
	DV	4	4	10	10	17
10^{-15}	RV	4	17	34	>93	>93
	SV	4	4	34	37	74
	DV	4	4	12	12	20

Table IV. Adaptation of the ECC Correction Capability to the Flash Aging for Different Programming Algorithms and Target UBER

- —The combination of reduced *t* and ISPP-RV represents the best performance corner but offers the worst reliability.
- —The combination of increased t and ISPP-DV represents the best reliability corner but offers the worst performances.
- —In the bottom-left and upper-right access modes of the table, the UBER is adapted by acting on the correction strength and the chosen algorithm.

4.1. Access Modes Characterization

An example of the optimization that can be achieved by selecting the considered access modes is reported in Figure 10. It shows how the achievable modulation effect of the RT and WT (for a target UBER= 10^{-13}), accomplished by changing the programming algorithm and the ECC correction capability, varies over time along with memory aging. The correction capability of the ECC is adapted as aging increases (according to Table IV) to preserve the target UBER in spite of memory aging. For the sake of

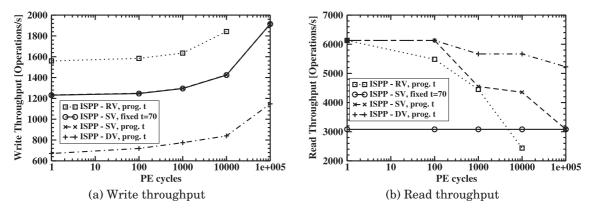


Fig. 10. WT and RT comparison among different configurations of the controller for a target UBER= 10^{-13} .

comparison, the figure shows the performance of a nonadaptive controller using the ISPP-SV programming algorithm and a fixed correction capability t = 70, which is required to meet the target UBER at the end of the memory lifetime.

Figure 10(a) clearly shows that, acting on the programming algorithm, we can modify the write performance of the flash with 30% improvement obtained with *ISPP-RV prog.* t used instead of *ISPP-SV fixed t* at the end of the flash's life. Moreover, WT modulation capability is preserved over memory cycling. The nonadaptive and the adaptive ISPP-SV solutions are almost overlapped because the encoding latency is barely affected by the ECC correction capability. Figure 10(b) instead shows that we can tune the RT of the system by using ISPP-DV as opposed to ISPP-RV. In these cases, the RT can be improved by 83% or degraded by 20%, calculated at a PE cycle of 10k, respectively, and compared to the reference-adaptive ISPP-SV solution. Of course, the RT degradation of ISPP-RV is the price to pay for its WT improvement. The comparison of the *ISPP-SV prog. t* curve with the baseline *ISPP-SV fixed t* curve shows that tuning the ECC correction capability over the life of the flash enables a significant improvement of the RT with no penalty on the WT.

Figure 10(b) also shows that, in the early stage of the memory's life, the modulation capability of the RT is marginal. The reason lies in the similar RBER figures of the programming algorithms in fresh devices. On the one hand, this means that RT improvement with respect to the reference case will be achieved only after hundreds of PE cycles. On the other hand, this also means that, in fresh devices, the WT can be broadly modulated at marginal RT penalty. Overall, Figure 10 shows a usage model of the access modes: the correction capability is used to preserve a target UBER over the flash lifetime, whereas the programming algorithm is used to trade the WT with the RT. At a given PE cycle, a higher RT can be achieved by switching the programming algorithm (i.e., from *ISPP-SV prog. t* to *ISPP-DV prog. t*) and the ECC correction capability (since ISPP-DV needs a lower t to preserve the target UBER with respect to ISPP-SV). The WT can be traded-off similarly. Regardless of the selected programming algorithm, Figure 10(b) clearly shows that for most of the memory's life, the nonadaptive approach produces a significant device underutilization from the RT standpoint.

Other usage models are clearly feasible. For instance, switching from *ISPP-SV prog.* t to *ISPP-DV prog.* t, while keeping t unchanged, minimizes the UBER beyond 10^{-13} leaving the RT unaltered at the cost of the WT. Similarly, switching to *ISPP-RV progr.* t achieves a WT improvement. If, at the same time, we decrease t, the UBER is largely degraded while the RT is improved. Otherwise, with a constant t, the UBER is degraded to the lower extent, but RT is unaltered. Finally, the upper-left access mode in Figure 9 can be used in those cases where an ultra-low-power operating mode is required while, at the same time, largely degrading UBER; therefore, application-perceived low reliability is accepted. Approximate storage of data to improve performance whenever

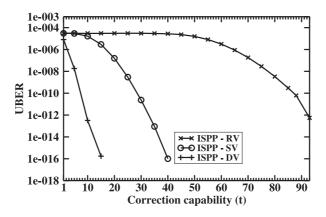


Fig. 11. Tradeoff on the storage reliability by selecting different programming algorithms and different ECC correction capabilities. UBER is computed at 10,000 PE cycles of the flash.

high-precision storage is not required has been already investigated in previous studies [Sampson et al. 2013], and the considered service represents a very efficient method for its implementation. In contrast, the lower-right access mode in Figure 9 provides the best achievable reliability at the cost of increased power consumption and largely degraded performance.

Figure 11 summarizes the way UBER can be tuned by selecting different ECC correction capabilities or programming algorithms. Values in the figure are computed considering the RBER of the flash at 10,000 PE cycles (i.e., quite late in the flash lifetime). Similar to the performance characterization, Figure 11 shows that we can achieve important tradeoffs in the reliability of the access modes, with the possibility of varying the UBER of the NVM system by several orders of magnitude.

4.2. Implementation of the Access Modes

To properly exploit the advantages provided by the combined adaptation of the flash programming algorithm and the ECC correction capability, a strategy to decide which memory access mode to use at runtime is mandatory. Although a complete discussion of this topic is out of the scope of this article, a set of preliminary insights can be provided here. There are essentially two factors that must be considered, at run-time, to properly select the optimal flash storage options: (i) application reliability/performance/power requirements and (ii) memory aging.

The first factor is static for a given application or for selected portions of data in an application. Even if not straightforward, applications can be carefully profiled in order to assign different reliability/performance/power requirements to the different sets of data they manage. The application profile can then be exploited to choose the best storage service for each type of information.

We envision in this work splitting the flash memory into different partitions to provide different storage services, according to Figure 12.

The flash file system can therefore be extended to provide dedicated API to request different classes of storage services and to properly redirect the data to the partition implementing the requested access mode. Each application can then be instrumented to request, for each flash memory, access to the best-suited storage service for the specific data. A single application can therefore benefit from data stored in different partitions with different services in order to optimize the overall reliability/performance. Moreover, considering a different scenario, the choice of the target service may be also handled by the operating system to shield the user from details of the hardware implementation and to avoid erroneous selection of the target service. The operating system may be delegated to select different access modes for an application by exploiting

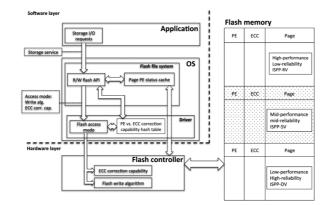


Fig. 12. Exporting storage services to the software layer.

routines that continuously analyze the behavior of the application in order to determine the optimum performance/reliability/power tradeoff configuration for the problem and to supervise program execution. Using program instrumentation gives the programmer flexibility in choosing the system configuration needed for a particular nonfunctional requirement, while the implicit approach reduces programming effort and speeds up program development.

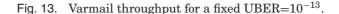
Whereas, for a given access mode, the selected programming algorithm is generally constant over the memory lifetime, the ECC correction capability must be continuously tuned at runtime to compensate for memory aging. Several models in the literature correlate the RBER of a page to the number of performed PE cycles [Sun et al. 2011], and enable us to build models fitted on experimental data to compute the best ECC correction capability to apply when a page is programmed. If the PE count is constantly tracked during flash operations, it can be exploited to adapt the ECC correction capability according to the selected aging model. In this context, one of the most efficient and easy solutions is to demand this operation to the flash file system in cooperation with the flash driver. At each programming operation, the PE count of the target page is incremented and stored together with other file system-related information. This value can be then exploited at runtime to select the best correction capability every time the page is programmed. The PE count of each page is cached in RAM using a common practice implemented by the flash file system to store management information. Every time a page must be programmed, this value is retrieved from the cache. It is used to search for the best correction capability to apply into a correction capability hash table stored in the flash drivers and containing aging information related to the specific flash technology. This value is then used to encode the target page. Similarly, whenever a page must be read, the same procedure is used to retrieve the correction capability used to encode the page, and this information is used during the ECC decoding phase.

5. STORAGE SERVICES AT WORK

To appreciate the benefits of differentiated flash access modes on the execution of a set of real applications, we constructed a simulation environment running under the Linux operating system using YAFFS2 as flash file system. The Linux Memory Technology Device (MTD) driver has been instrumented to emulate operations on a NAND flash memory with 4,096 blocks of 128 pages, with a page size of 4 kB for a total of 2GB of available storage. YAFFS2 has been also instrumented to trace the list of operations performed through the MTD. Read, write, and erase operations have been traced. The log essentially contains information about the sequence of operations, the target page

Personality	#R/#W	Avg. $#R$	Avg. $#W$		
varmail	32.5%	48,536	149,081		
webserver	webserver 153.5%		65,581		
videoserver	1077.9%	457,138	42,410		
2500 $\begin{bmatrix} \square : \square : ISPP - RV, prog. t \\ \bigcirc : \square : \square : ISPP - SV, prog. t \\ + + : ISPP - DV, prog. t \\ - \cdots = \cdots$					
E 1000		+	·*´ _		

Table V. #R/#W Ratios of Different Filebench Personalities



PE cycles

+005

address, and the timing. To obtain unbiased measurements of flash performances, the YAFFS2 caches have been disabled.

Several file system benchmarks are available on the Internet (e.g., IOzone [iozone.org 2012], Postmark [Katcher 1997], SPEC benchmarks [Standard Performance Evaluation Corporation 2013], Filebench [Wilson 2008], etc.). In this article, we selected the Filebench benchmark [Wilson 2008] that provides a large variety of behaviors that can be exploited for our analysis. They either perform simple file I/O operations or emulate complex I/O activities. Among the available benchmarks we selected three applications:

- *—varmail*: This application has different threads performing create-append-sync, readappend-sync, read, and delete operations on the files (representing emails) contained in a single directory;
- *—webserver*: Opens, reads, and closes multiple files in a directory tree while appending data in log file;
- *—videoserver*: Reads a file set containing videos that are actively served and writes another file set containing videos that are available but currently inactive.

One of the main characteristics that differentiate the three selected benchmarks is the ratio between the number of read operations (#R) and the number of write operations (#W). This is a critical parameter that influences the type of access mode required by the application to maximize its performance. Table V summarizes this information. It reports the #R/#W ratio for each benchmark, as well as the average number of actual read and write operations generated by each benchmark during the simulations. *varmail* is a typical example of a write-intensive application requiring fast programming of the flash. On the contrary, *videoserver* is a read-intensive application requiring fast read access to the data stored in flash. Finally, *webserver* is between the other two benchmarks and performs a more equalized set of read and write operations to the flash.

Figures 13, 14, and 15 show the opportunities that controller programmability provides to the three applications for a target UBER= 10^{-13} . All figures report the overall application throughput (i.e., number of operations [read or program operations]) performed on the flash per unit of time. Comparison is again performed with a non-adaptive controller using the ISPP-SV programming algorithm and fixed ECC with

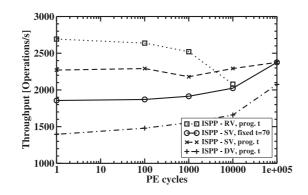


Fig. 14. Webserver throughput for a fixed UBER= 10^{-13} .

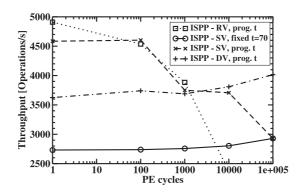


Fig. 15. Videoserver throughput for a fixed UBER= 10^{-13} .

t = 70. Simulations have been performed in order to emulate a steady state with all flash pages written at least once. This generates an average of one erase operation every 128 programmed pages, corresponding to the number of pages in a block.

Looking at Figure 13, which reports the throughput of *varmail*, it is evident that *ISPP-RV prog. t* enables a significant improvement of the overall performance of the application. This improvement comes, however, with a reduced endurance of the flash due to the high RBER introduced by this programming algorithm when the number of PE cycles exceeds 10,000.

If we move instead to the opposite application profile represented by the readintensive *videoserver* reported in Figure 15, we notice an interesting result. Looking at the overall flash lifetime, the *ISPP-SV prog. t* seems the best option for this application even if, looking at Figure 10(b), we could expect better performances from *ISPP-DV prog. t*. The main motivation for this behavior is that the flash programming time is dominant over the flash read time and therefore negatively influences overall application performance. This opens new opportunities for the proposed controller. In fact, Figure 15 suggests that not only the ECC correction capability must be adapted to compensate for page aging. In this specific application profile, the ISPP-DV can be selected when the flash reaches more than 10,000 PE cycles to sustain the overall performance and reliability level.

The last situation, represented by *webserver* (Figure 14), obviously provides an intermediate behavior. In this situation, *ISPP-DV prog. t* reduces the overall performance and is therefore not a good choice for the application. However, both *ISPP-SV prog. t* and *ISPP-RV prog. t* introduce significant performance improvements.

The analysis performed so far highlights how the proposed adaptation strategy improves the performance of selected applications when mapped to dedicated access modes. The same programmability can be also exploited to provide access modes with

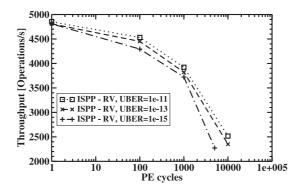


Fig. 16. Videoserver throughput with ISPP-RV program. t at different target UBER.

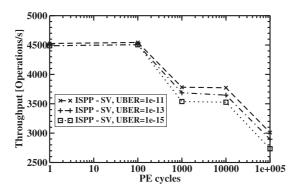


Fig. 17. Videoserver throughput with ISPP-SV program. t at different target UBER.

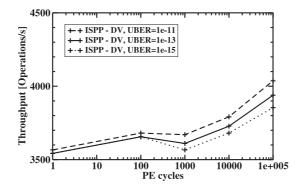


Fig. 18. Videoserver throughput with *ISPP-DV program. t* at different target UBER.

different reliability levels, as reported in Figures 16, 17, and 18 for the *videoserver* application. In this comparison, we considered a standard reliability service (UBER= 10^{-13}), an enhanced reliability service (UBER= 10^{-15}), and a reduced reliability service (UBER= 10^{-11}).

When analyzing the results reported in Figures 16, 17, and 18, it is important to take into account that the *videoserver* application is a read-intensive application. When exploiting the *ISPP-RV prog. t* (Figure 16) and the *ISPP-SV prog. t* (Figure 17) writing algorithms, which provide reduced reliability compared to the *ISPP-DV prog. t* algorithm, the ECC subsystem is particularly stressed to guarantee error-free data during the intensive read activity of the application. Since the ECC correction capability must be increased with flash aging, the throughput of the application with these two algorithms decreases over time. By contrast, when considering the *ISPP-DV prog. t*, the high reliability of this algorithm strongly relaxes the ECC requirements. This strongly

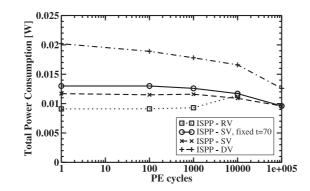


Fig. 19. Average power per operation during the execution of the videoserver benchmark.

improves the read throughput of the flash at the cost of a decreased write throughput. Write operations become therefore critical for this operation mode, and, overall, the throughput of the application decreases. Nevertheless, it is interesting to note that since the write performance of the flash increases with aging (see Figure 4(e)), we observe a slight improvement in the performance of the application at the end of the flash lifetime. Considering the increased reliability service, the target choice will be between *ISPP-SV prog. t* and *ISPP-DV prog. t*. In both cases, switching to a higher reliability level does not introduce major penalties in the performances. However, *ISPP-DV prog. t* guarantees performances that are more constant over the full flash lifetime. This could be a benefit, especially when real-time applications are considered. When moving to the reduced reliability service, instead, the choice can be between the *ISPP-RV prog. t* and *ISPP-SV prog. t*. In this case, however, the choice is a tradeoff between performance and memory endurance.

Finally, Figure 19 reports how the reliability of the memory subsystem can now be traded for reduced power consumption. In power saving scenarios, the functionalities of the system need to be preserved in order to either prolong battery life for portable and embedded systems or to reduce cooling issues in high-performance computing systems. Under such conditions, the Quality of Service (QoS) of a target application (i.e., video playback) can be degraded to a minimum acceptance level. This is the case of the *ISPP-RV prog. t* access mode, which can significantly reduce the memory energy consumption by a 10% factor at the beginning of the memory lifetime with respect to the nonadaptive ISPP-SV case.

6. CONCLUSION

In this article, we demonstrated that combining the selection of different flash programming algorithms with runtime adaptations of the ECC correction capability in an MLC NAND flash subsystem holds promise for exposing interesting tradeoffs among performance, reliability, and power for memory access. This enabled us to define differentiated access modes for ultra-high performance, for ultra-high reliability, or intermediate tradeoff requirements. When put to work for real-life workloads, the user-selectable access modes prove capable of better adapting to application requirements than do nonadaptive controllers. By modeling memory endurance effects, we pointed out that the most suitable access mode for each application is not the same through the entire memory lifetime. Based on the results of this article, the RTL coding of the runtime reconfigurable memory controller will be our future work, in order to obtain an adaptive NVM subsystem that can complement the current ongoing efforts in adaptive computing.

7. AUTHORS' CONTRIBUTIONS

D. Bertozzi, P. Olivo, and C. Zambelli mainly contributed one the characterizations of the flash programming algorithms, whereas S. Di Carlo, S. Galfano, M. Indaco, and P. Prinetto focused on the characterization of the adaptive ECC performance and on the setup of the software framework for the benchmark simulations. All authors contributed to the overall storage service analysis.

APPENDIX

The flash model developed in this work is a SPICE-based compact model devised for Monte Carlo simulation of a floating gate transistor. The model captures the threshold voltage evolution of a NAND flash cell during the ISPP algorithm within a memory array by adding to the calculated cell's threshold voltage at each time step of the writing algorithm, with the following variability sources:

- —Geometrical variability: Since the transistors within the array do not feature the same geometrical parameters, mainly due to lithographic concerns, a displacement on the channel length (L) and channel width (W) from their nominal values σ_L and σ_W is considered in each Monte Carlo run. These latter parameters feature a Gaussian distribution with mean value equal to 1nm and standard deviation of 0.2nm. Since the geometry of the transistor also affects threshold voltage evolution, these parameters are calculated before the definition of the transistor structure to be simulated and therefore affect the final cell's threshold voltage.
- --Cell-to-Cell Coupling: The SPICE compact model for the NAND array includes parasitic capacitive couplings between each cell and its first neighbors along the same word- and bit-line. The capacitances are derived from 3D-TCAD simulations and feature the typical values for a 45nm technology (i.e., roughly about 20aF). The cell's threshold voltage calculated at each ISPP step takes into account that the electron tunneling current and the channel potential of the transistor deviate from the nominal value by adding a ΔV_{TH} to the voltages exploited for the writing operation.
- --Injection statistics: The discrete nature of the electronic flow charging the floating gate represents an additional variability source to be considered when dealing with the program operation of nanoscale cells since the statistical process ruling discrete electron injection into the floating gate introduces fluctuations in cell V_{TH} after the application of a writing pulse [Spessot et al. 2010]. On this basis, we introduced this additional variability contribution into our compact model for the program operation by adding a displacement from the cell's threshold voltage having the following spread:

$$\sigma_{\Delta V_T} = \sqrt{\frac{q}{\gamma C_{PP}} \left(1 - e^{-\gamma \left(\overline{\Delta_{V_T}} \right)} \right)},\tag{3}$$

where q is the electronic charge, γ is the slope of the tunneling characteristic of the floating gate transistor, C_{pp} is the floating gate capacitance calculated with geometrical variability, and $(\Delta_{V_T})^-$ is the voltage step magnitude of the ISPP algorithm.

—Random Dopant Fluctuation (RDF): The atomistic nature of substrate doping has been clearly shown to result in a fundamental threshold voltage spread for MOS field effect transistors (MOSFETs) given by:

$$\sigma_{RDF} = 3.19 \times 10^{-8} \times \left(\frac{t_{ox} \left(N_A\right)^{0.4}}{\sqrt{WL}}\right),\tag{4}$$

where t_{ox} is the tunnel oxide thickness subjected to geometrical variability and equal to 7.5 nm + 0.1 nm, and N_A is the substrate doping of the cell, which follows a profile retrieved by TCAD simulations.

-Oxide Trap Fluctuation (OTF): Referring to traps placed at the substrate-oxide interface (where they have the strongest impact on cell V_{TH}) and assuming a Poissonian fluctuation of their number due to process variability, a spread in cell V_{TH} results, according to the following:

$$\sigma_{OTF} = K_{OX} \times t_{ox} \times \frac{\sqrt{Q_{OX}}}{\sqrt{WL}},\tag{5}$$

where Q_{ox} is the surface density of traps assumed equal to $10^{-11} cm^{-2}$, tox is the tunnel oxide thickness, and K_{ox} is a constant equal to $10^{-6}V \times cm$.

-Aging effect: The threshold voltage of a memory cell increases due to charge trapping with the number of write cycles. There are two types of traps that form in the tunnel oxide: interface traps and bulk traps, both of which contribute to the increase in the threshold voltage. It has been shown that both these traps have a power-law relation to the number of cycles on the memory cell [Spessot et al. 2010] as:

$$\Delta Nit = A \times cycle^{0.62} \tag{6}$$

$$\Delta Not = B \times cycle^{0.30},\tag{7}$$

where A and B are fitting constants, cycle is the number of write cycles on the cell, and the terms ΔN_{it} and ΔN_{ot} are the interface and bulk trap densities, respectively. In addition to providing this power-law relationship, the authors calculated the values of constants A and B to be 0.08 and 5, respectively, for the considered technology. The total threshold voltage increase due to trapping is divided into interface trap voltage shift (ΔV_{it}) and bulk trap voltage shift (ΔV_{ot}) using the following equations:

$$\Delta Vit = \frac{\Delta Nit \times q}{C_{ox}} \tag{8}$$

$$\Delta Vot = \frac{\Delta Niot \times q}{C_{ox}},\tag{9}$$

where C_{ox} is the capacitance of the tunnel oxide.

All these variability sources contribute to the final threshold voltage value with the following approximate percentile values: geometrical variability (15%); oxide trap fluctuations (15%); random dopant fluctuation (25%); and parasitic coupling capacitances, injection statistics, and aging (45%).

REFERENCES

- G. Atwood, A. Fazio, D. Mills, and B. Reaves. 1997. Intel strataflash memory technology overview. *Intel Technology Journal Quarterly* 4 (1997). Retrieved from https://noggin.intel.com/content/intel-strataflash-memory-technology-overview.
- R. Bez, E. Camerlenghi, A. Modelli, and A. Visconti. 2003. Introduction to flash memory. *Proceedings of the IEEE* 91, 4 (April 2003), 489–502.
- R. C. Bose and D. K. Ray-Chaudhuri. 1960. On a class of error correcting binary group codes. *Information* and Control 3, 1 (1960), 68–79.
- J. M. P. Cardoso and Michael Hübner. 2011. Reconfigurable Computing: From FPGAs to Hardware / Software Codesign. Springer, Berlin.
- T.-H. Chen, Y.-Y. Hsiao, Y.-T. Hsing, and C.-W. Wu. 2009. An adaptive-rate error correction scheme for NAND flash memory. In *Proceedings of the 27th IEEE VLSI Test Symposium (VTS'09)*. IEEE, 53–58.

CMP. 2012. CMP Project. Retrieved from http://cmp.imag.fr/.

- J. Cooke. 2007. The inconvenient truths of NAND flash memory. Flash Memory Summit. Retrieved from http://download.micron.com/pdf/presentations/events/flash_mem_summit_jcooke_inconvenient_truths_ nand.pdf.
- R. Dan and R. Singer. 2003. Implementing MLC NAND flash for cost-effective, high-capacity memory. M-Syst. White paper. Retrieved from http://tinyurl.com/o2443mh.
- S. Di Carlo, M. Fabiano, P. Prinetto, and M. Caramia. 2011. Design Issues and Challenges of File Systems for Flash Memories. InTech, Croatia, Chapter 1, 3–30.
- Evatronix. 2012. Evatronix NANDFLASH-CTRL NAND Flash Memory Controller. Retrieved from http://www.evatronix.pl/products/docs.html?id=10&product=TkFOREZMQVNILUN UUkw=.
- M. Fabiano, M. Indaco, S. Di Carlo, and P. Prinetto. 2013. Design and optimization of adaptable BCH codecs for NAND flash memories. *Microprocessors and Microsystems* 37, 4–5 (2013), 407–419.
- E. Grochowski and R. E. Fontana. 2012. Future technology challenges for NAND flash and HDD products. Flash Memory Summit. Retrieved from http://www.flashmemorysummit.com/English/Collaterals/ Proceedings/2012/20120821_S102A_Grochowski.pdf.
- J. Henkel, L. Bauer, M. Hübner, and A. Grudnitsky. 2011. i-Core: A run-time adaptive processor for embedded multi-core systems. In *Proceedings of the International Conference on Engineering of Reconfigurable Systems and Algorithms (ERSA'11)*. ERSA-ADN.
- iozone.org. 2012. IOzone File System Benchmark. Retrieved from http://www.iozone.org.
- F. Irom and D. N. Nguyen. 2007. Single event effect characterization of high density commercial NAND and NOR nonvolatile flash memories. *IEEE Transactions on Nuclear Science* 54, 6 (Dec. 2007), 2547–2553.
- JEDEC Solid State Technology Association. 2011. Failure mechanisms and models for semiconductor devices (JEP122G). Retrieved from http://www.jedec.org/standards-documents/docs/jep-122e.
- Y. H. Kang, J. K. Kim, S. W. Hwang, J. Y. Kwak, J. Y. Park, D. Kim, C. H. Kim, J. Y. Park, Y. T. Jeong, J. N. Baek, et al. 2008. High-voltage analog system for a mobile NAND flash. *IEEE Journal of Solid-State Circuits* 43, 2 (Feb. 2008), 507–517.
- J. Katcher. 1997. PostMark: A New File System Benchmark. Network Appliance Tech Report TR3022. Retrieved from https://communities.netapp.com/servlet/JiveServlet/download/2609-1551/Katcher97postmark-netapp-tr3022.pdf.
- J. D. Lee, J. H. Choi, D. Park, and K. Kim. 2003. Data retention characteristics of sub-100 nm NAND flash memory cells. *IEEE Electron Device Letters* 24, 12 (Dec. 2003), 748–750.
- R.-S. Liu, C.-L. Yang, and W. Wu. 2012. Optimizing NAND flash-based SSDs via retention relaxation. In Proceedings of the 10th USENIX Conference on File and Storage Technologies (FAST'12). USENIX Association. Retrieved from http://static.usenix.org/event/fast12/tech/full_papers/Liu.pdf.
- C. Miccoli, C. Monzio Compagnoni, A. S. Spinelli, and A. L. Lacaita. 2011. Investigation of the programming accuracy of a double-verify ISPP algorithm for nanoscale NAND Flash memories. In *Proceedings of the IEEE International Reliability Physics Symposium (IRPS'11)*. IEEE, MY.5.1–MY.5.6.
- R. Micheloni, L. Crippa, and A. Marelli. 2010. Inside NAND Flash Memories. Springer-Verlag, Berlin.
- R. Micheloni, A. Marelli, and K. Eshghi. 2013. Inside Solid State Drives (SSDs). Springer, Berlin.
- N. Mielke, T. Marquart, N. Wu, J. Kessenich, H. Belgal, E. Schares, F. Trivedi, E. Goodness, and L. R. Nevill. 2008. Bit error rate in NAND Flash memories. In *Proceedings of the IEEE International Reliability Physics Symposium (IRPS'08)*. IEEE, 9–19.
- V. Mohan, S. Gurumurthi, and M. R. Stan. 2010. FlashPower: A detailed power model for NAND flash memory. In Proceedings of Design, Automation & Test in Europe Conference & Exhibition (DATE). IEEE, 502–507.
- ONFI Workgroup. 2012. Open NAND Flash Interface. Retrieved from http://onfi.org.
- J. Ouyang, S. Lin, S. Jiang, Z. Hou, Y. Wang, and Y. Wang. 2014. SDF: Software-defined flash for web-scale internet storage systems. In Proceedings of the 19th International Conference on Architectural Support for Programming Languages and Operating Systems (ASPLOS'14). ACM, New York, NY, 471–484. DOI: http://dx.doi.org/10.1145/2541940.2541959
- Y. Pan, G. Dong, and T. Zhang. 2011. Exploiting memory device wear-out dynamics to improve NAND flash memory system performance. In *Proceedings of the 9th USENIX Conference on File and Storage Technologies (FAST'11)*. USENIX Association, Berkeley, CA.
- A. Sampson, J. Nelson, K. Strauss, and L. Ceze. 2013. Approximate storage in solid-state memories. In Proceedings of the 46th Annual IEEE / ACM International Symposium on Microarchitecture (MICRO-46). ACM, New York, NY, 25–36. http://www.datasheetarchive.com/dl/Datasheets-SW18/DSASW00344250. pdf.
- Samsung. 2012. Samsung KFG4GH6x4M 4Gb Flex-OneNAND M-die Datasheet.

- A. Spessot, A. Calderoni, P. Fantini, A. S. Spinelli, C. M Compagnoni, F. Farina, A. L. Lacaita, and A. Marmiroli. 2010. Variability effects on the VT distribution of nanoscale NAND flash memories. In *IEEE International Reliability Physics Symposium (IRPS'10)*. IEEE, 970–974.
- Standard Performance Evaluation Corporation. 2013. SPEC Benchmarks. Retrieved from http://www.spec.org.
- H. Sun, B. Wood, and P. Grayson. 2011. Qualifying reliability of solid-state storage from multiple aspects. In Proceedings 7th IEEE International Workshop on Storage Network Architecture and Parallel I/O (SNAPI'11). IEEE. Retrieved from http://storageconference.org/2011/Papers/SNAPI/1.Sun.pdf.
- A. Wilson. 2008. The new and improved filebench. In Proceedings of the 6th USENIX Conference on File and Storage Technologies (FAST'08). USENIX, Retrieved from https://www.usenix.org/legacy/events/ fast08/wips_posters/wilson-wip.pdf.
- E. Yaakobi, J. Ma, A. Caulfield, L. Grupp, S. Swanson, P. H. Siegel, and J. K. Wolf. 2009. Error correction coding for flash memories. Flash Memory Summit. Retrieved from http://www.bswd.com/FMS09/ FMS09-201-Yaakobi.pdf.
- E. Yaakobi, J. Ma, L. Grupp, P. H. Siegel, S. Swanson, and J. K. Wolf. 2010. Error characterization and coding schemes for flash memories. In *Proceedings of the IEEE GLOBECOM Workshops (GC Wkshps'10)*. IEEE, 1856–1860.
- C. Zambelli, M. Indaco, M. Fabiano, S. Di Carlo, P. Prinetto, P. Olivo, and D. Bertozzi. 2012. A cross-layer approach for new reliability-performance trade-offs in MLC NAND flash memories. In *Proceedings of Design, Automation Test in Europe Conference Exhibition (DATE'12)*. IEEE, 881–886.