A cross-layer approach for new reliability-performance trade-offs in MLC NAND flash memories

Original

Availability:
This version is available at: 11583/2462376 since: 2020-12-12T17:44:02Z

Publisher:
IEEE Computer Society

Published
DOI:10.1109/DATE.2012.6176622

Terms of use:
openAccess

This article is made available under terms and conditions as specified in the corresponding bibliographic description in the repository

(Article begins on next page)
A cross-layer approach for new reliability-performance trade-offs in MLC NAND flash memories

Authors: Zambelli C., Indaco M., Fabiano M., Di Carlo S., Prinetto P., Olivo P., Bertozzi D.,

Published in the Proceedings of the Design, Automation and Test in Europe, Conference and Exhibition (DATE), 12-16 March 2012, Dresden, DE.

N.B. This is a copy of the ACCEPTED version of the manuscript. The final PUBLISHED manuscript is available on IEEE Xplore®:

URL: http://ieeexplore.ieee.org/stamp/stamp.jsp?tp=&arnumber=6176622

© 2011 IEEE. Personal use of this material is permitted. Permission from IEEE 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.
A Cross-Layer Approach for New Reliability-Performance Trade-Offs in MLC NAND Flash Memories

Abstract

Multi-level cell (MLC) NAND Flash memories are very popular storage media because of their power efficiency and big storage density. In spite of the mature cell structure, the memory controller architecture is evolving fast in an attempt to improve the uncorrected/miscorrected bit error rate (UBER) and to provide a more flexible usage model where the performance-reliability trade-off point can be adjusted at runtime. However, optimization techniques in the memory controller architecture cannot avoid a strict trade-off between UBER and read throughput. In this paper, we show that co-optimizing ECC architecture configuration in the memory controller with program algorithm selection at the technology layer, a more flexible memory sub-system arises, which is capable of new trade-offs between performance and reliability. In particular, the cross-layer configuration framework enables to improve UBER without sacrificing read throughput, and vice versa to improve read throughput without impacting UBER, but at the cost of write throughput, a scenario which is of great interest to read-intensive applications.

1. Introduction

Flash memory is an important building block for modern embedded systems due to its high data transfer rate, low power consumption and long mechanical durability. In addition, the advent of multi-level cell (MLC) NAND flash memories [1] has opened unprecedented opportunities for embedded systems and laptops to store larger amounts of data in flash, thus replacing power-hungry, relatively unreliable hard drives.

Unfortunately, flash memory devices come with their unique reliability concerns, making them a highly vulnerable portable storage. MLC technology has further exacerbated the problem with respect to traditional SLC one. The primary failure mechanisms for MLC flash memories include threshold voltage distribution shifting [2], program/read disturb [3], data retention [4], endurance [5] and single event upset[6]. The raw bit error rate (RBER) of MLC flash memory is around $10^{-6}$ [7] and is at least two orders of magnitude worse than that of the SLC device [8]. Since page errors are expected to be the primary failure pattern in flash-based systems.

Design techniques at the level of the memory controller, such as page-level error correcting codes (ECC), are systematically used to achieve an acceptable uncorrectable bit error rate (UBER). The choice of the most suitable error correcting scheme is tightly application dependent. Reliability requirements of relatively small flash memories that hold non-critical, error-tolerant data such as photographs and digitized music are radically different from those of multi-gigabyte file systems based on MLC flash devices [9]. For this reason, practical ECC solutions are typically market segment-specific and range from derivatives of the Hamming code [2] to the BCH [10, 11, 12, 13] or the Reed-Solomon code [14].

However, statically setting memory controller features at design-time is a practice that is rapidly running out of steam. On one hand, the new mobile usage models that are coming about require the execution of multiple use cases on the same device while optimizing resource consumption for each of them. On the other hand, the hardware and software design convergence in today’s complex embedded systems call for an upgrade of architecture building blocks in the direction of runtime re-configurability and adaptivity. The main idea of this paper is that a better and more flexible performance-reliability trade-off for runtime adaptivity could be achieved by combining ECC architecture configuration with settings in the physical layer, thus giving rise to a cross-layer optimization framework. In fact, the program algorithm and voltage waveforms that are applied for memory writing are typically defined at fabrication time by the memory vendor and hardwired in memory operation. Accurate programming of NAND Flash memories is usually obtained by the incremental step pulse programming (ISPP) algorithm [3]. However, a number of variants does exist to counter the dispersion of programmed cell thresholds distributions in nano-scaled flash devices, such as the double verify algorithm in [19, 20, 21], which can potentially improve the RBER on the same device or sustain the RBER on scaled devices.

In this paper we propose to dynamically tune the programming algorithm for MLC NAND flash memories so to provide differentiated reliability provisions for memory storage. Above all, this paper for the first time explores and quantifies the synergies between architecture-level (configuration of the ECC decoding structure) and physical-level (selection of the program algorithm and waveforms) settings, proving that a cross-layer optimization framework can materialize new trade-offs between read, write throughput, reliability and power consumption.

An extensive modeling, simulation and implementation framework has been set up for both the analog and the digital parts of an MLC NAND flash memory sub-system in an homogeneous 45nm technology substrate, thus coming up with the accurate quantification of these trade-offs.

2. Related Work

In traditional NAND product a single algorithm for writing data into its addressable space is used, not allowing to devise a reliability/performance degree of freedom which is intrinsically available in non-volatile memory technology. There exists the possibility of having a segmented memory with the possibility to dynamically configure the segment sizes during the memory boot stage, but once again only one write/erase algorithm will be used for all the segments, thus losing the intrinsic degree of freedom of having a reliability/speed trade-off (as envisioned but not still implemented in [20]). The only available degree of freedom presented in nowadays NAND Flash resides on the choice of having a segmented memory with mixed SLC/MLC structure [21], but this feature is something that can be changed only at boot time and directly does not affect the write algorithms, which are unique for every segment.

Error correction is also an integral part of using flash memories that ensure data integrity.

The literature is rich of publications proposing ECC based solutions for flash memories with fixed correction capability [35]. However, to the best of our knowledge, only Chen et al. proposes an solution allowing limited adaptation [28]. It introduces an adaptable-rate Bose-Chaudhuri-Hocquenghem (BCH) codec with a controllable parallel Linear Feedback Shift Register (LFSR) and a pipelined decoder. The main drawback of this solution is the ap-
plication of the code to small blocks of data (512B compared to the typical 2/4KB page of a NAND flash memory). Small blocks make it difficult to handle high concentrations of errors and require a high number of parity bits and higher hardware complexity. These extra bits tend to saturate the spare area usually available on a flash and may also consume extra pages. This represents a major problem since the spare area is usually dedicated to system management and not only to ECC. Moreover, adaptability is limited to a reduced set of values.

This paper advances state-of-the-art with innovations on the usage model both of NAND Flash devices and ECC systems by providing a cross-layer optimization framework that changes the traditional performance-reliability trade-off and enables increasingly flexible usage models of mobile devices.

3. Advanced Controller Architecture for a NAND Flash Memory

The NAND Flash sub-system consists not only of the Flash memory device but also of the memory controller, a key component for determining the quality metrics of the whole non-volatile memory sub-system. State-of-the-art controllers for NAND Flash memories typically consist of one or more bus interfaces toward the rest of the MPSoC, of the Flash device interface and of sophisticated error correctors. In some cases, queues are implemented for data exchange (typically, a single page) between the ECC and the flash device and to decouple blocks clocked by different signals.

Figure 1. Memory controller architecture for a NAND Flash device in the short-to-medium run.

The focus of this paper is on memory controllers that can be envisioned for the largely integrated MPSoCs in the short-to-medium run. In this context, system components are typically connected via an on-chip interconnection network. At network boundary, components can be attached to the network via standard socket interfaces like the AXI or the OCP protocol. We projected a realistic controller architecture for the near future and illustrate it in Fig.1.

The OCP interface connects the controller to the on-chip network, which routes read and write access requests or configuration commands. The network is typically much faster than the Flash device and to decouple blocks clocked by different signals. The network is typically much faster than the Flash device and to decouple blocks clocked by different signals. The network is typically much faster than the Flash device and to decouple blocks clocked by different signals. The network is typically much faster than the Flash device and to decouple blocks clocked by different signals.

4. Adaptive ECC Architecture

The adaptive ECC architecture analysed in this work implements a Bose-Chaudhuri-Hocquenghem (BCH) ECC with programmable correction capability. Bose-Chaudhuri-Hocquenghem (BCH) codes are a family of ECCs largely applied to NAND flash memories [28]. BCH codes are less complex than other ECCs and provide high code efficiency. Moreover, errors in flash memories are in general non-correlated and BCH codes are particularly efficient in this situation. The construction of a BCH code is based on Galois field $GF(2^m)$.

A binary BCH code, denoted as $BCH[r,k]$, encodes a $k$-bit message (in our case consisting of a full 4KB page of the flash) to an $n$-bit codeword ($n > k$) by adding $r$ parity bits to the original message. Parity bits are stored in the spare area of the flash. The number $r$ of parity bits required to correct $t$ errors in the $k$-bit message is computed by finding the minimum $m$ that solves the inequality $2^m - 1 > t + 1$ where $r = m \cdot t$ [29].

The correction capability of the code must provide depends on the RBER. The Flash device is able to provide and on the UBER the target service requires, according to the following relation:

$$UBER = \frac{(n+1)RBER^{r+1}(1 - RBER)^{n-(r+1)}}{r}$$

(1)

The adaptable ECC block employed in this work makes it possible to dynamically change its correction capability $t$ in a range between $l$ and a maximum value denoted here with $t_{\text{max}}$ by using a dedicated input port.

The BCH encoder computes the $r$ parity bits for a $k$-bit block of data by computing the reminder of the division between the message and the code generator polynomial. This computation can be efficiently implemented using a $r$-bit linear feedback shift register (LFSR) with characteristic polynomial being equal to the code generator polynomial. To obtain adaptability to different correction capabilities a parallel programmable LFSR able to support different characteristic polynomials is employed [28]. The sequences of bits representing each polynomial are stored in a ROM which is used to control a set of multiplexers able to dynamically insert XOR gates in the LFSR network according to the requested polynomial. Given the parallelism $p$ of the parallel LFSR (i.e., the block receives the message to encode in words of $p$ bits starting from the most significant word), $\frac{r}{p}$ clock cycles are required to compute the parity bits. The encoding latency is therefore not influenced by the selected correction capability.

The BCH decoding process aims at identifying the position of the erroneous bits of the codeword. This operation is more
complex than the encoding and requires three main computational steps highlighted in Fig. 2.

![Figure 2. BCH decoding flow.](image)

Given the selected correction capability $t$ the Syndrome block computes $2t$ syndromes of the codeword to decode. The syndrome computation requires calculating the reminder of the division between the codeword and each of the $2t$ polynomials $\psi_i(x)$ generating the generator polynomial of the code. Similarly to the encoding, this step requires a set of parallel LFSR (one for each polynomial $\psi_i(x)$). If all reminders are null the codeword is error-free and the decoding process ends. If not, the syndromes are computed by evaluating each reminder in the corresponding elements of $\text{GF}(2^m)$ using a set of dedicated combinational networks (see [29] for additional details). In an adaptable decoder with maximum correction capability $2t$ LFSR compose the syndrome block. However, depending on the selected correction capability only $2t$ of them will be actually enabled. Moreover, depending on $t$, the number $r$ of parity bits included in the codeword changes. If this number does not perfectly fit the parallelism of the decoder, which in our case is the same selected for the encoder, a preliminary alignment phase is required.

When all syndromes have been computed, the Berlekamp Massey block computes the error locator polynomial $\lambda$ whose roots represent the inverse of the error positions in the codeword. To implement an adaptable Berlekamp Massey Machine we considered the hardware implementation of the Berlekamp-Massey (IBM) algorithm proposed in [29]. It iteratively computes the coefficients of the error locator polynomial without requiring complex matrix inversions. The number of iterations required to compute the coefficients is equal to the selected correction capability $t$ thus easily allowing adaptability.

Finally, the Chien Search block searches for the roots of the error locator polynomial computed by the Berlekamp Massey block. This is the most complex and time intensive process of the decoder since it basically requires evaluating the error locator polynomial $\lambda$ into each element of $\text{GF}(2^m)$. In fact, according to the BCH theory, given the chosen correction capability $t$ not all elements of $\text{GF}(2^m)$ must be considered. The adaptable decoder stores in a small ROM, for each possible correction capability, a set of coefficients indicating the first element of $\text{GF}(2^m)$ from which the Chien search must initiate.

The performance of the Chien search strongly depends on its parallelism denoted here with $h$, i.e., on the number of parallel evaluations the block is able to perform. A high parallelism allows for fast search but requires a considerable set of hardware resources ($t \times h$ constant Galois multipliers), while a low parallelism reduces the amount of requested resources but, at the same time, penalizes the decoding latency.

5. Runtime-Selection of Program Algorithm

The other sub-system affected by the new trade-offs targeted by this work is the high-voltage sub-system of the Flash memory device, which is in charge of generating the voltage waveforms for flash cell read, program and erase operations and for address decoding. Its operation is regulated by the commands from a control FSM or from an embedded microcontroller in the flash device itself.

The physical layer considered in this work refers to state of the art 2-MLC memories [3], which store two bits per cell by accurately placing four different $V_{TH}$ levels, identified by the statistical distributions L0-L3 shown in Fig. 3. An erase operation places all the cells within a block on the L0 level (the threshold voltage distribution typically below zero), representing the starting point for each subsequent Program operation which will end up on placing threshold voltages of the selected cells on L1-L3 levels. In order to accurately fulfill the aforementioned operation, a standard algorithm is exploited in NAND Flash memories: the Incremental Step Pulse Programming (ISPP) [22]. A voltage step (whose amplitude and duration are predefined) is applied to the gate of the cells. Afterwards, a Verify operation (i.e. threshold voltage Read) takes place in order to check if the cells $V_{TH}$ have exceeded a predefined voltage value $V_{TH}$ (in MLC architectures more than one Verify level is present). 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 [2]. Otherwise, another cycle of ISPP is applied to the cells, where the programming voltage is incremented by $\Delta V_{ISPP}$.

Because of the technological variations, $V_{TH}$ is not perfectly related to the amplitude of the ISPP pulse: there are “fast” cells which reach the Verify level with few Program pulses, while other “slow” cells require more pulses. Both kinds of behaviors represent a threat for the reliability of the Program operation, since the threshold voltage distributions of the L1-L3 levels significantly deviate from the ideal gaussian-like shape, thus crossing the distribution read levels (R1-R3) and causing bit errors.

Different technological approaches for achieving distribution compactness are commonly pursued, although they share the same underlying principle: acting directly on the ISPP pulse characteristics by decreasing the $\Delta V_{ISPP}$ per step or by increasing the total number of pulses per Program operation. However, although these methodologies could effectively increase the accuracy of the ISPP algorithm in terms of threshold voltage placement, a substantial penalty both in power consumption and write throughput is paid.

An alternative solution for increasing ISPP Programming accuracy with minimal burden on the programming time and complexity has been recently presented in [2-19]. This algorithm exploits a Double Verify (DV) approach, where the bitline voltage of the selected cells is modulated in order to partially decrease the $\Delta V_{ISPP}$ step using a prior Verify level with slightly lower voltage than the original Verify level, hence compacting the final desired threshold voltage distribution.

In current flash device controllers, the program algorithm is set at fabrication time and hardwired in memory operation, thus preventing runtime trade-offs. We find that making the algorithm selectable at runtime is a feasible option that comes with minor implementation cost (see section 6.4) and opens up interesting implications in memory usage. Next section illustrates the modeling effort of the high-voltage memory sub-system with the capability to execute both the ISPP-SV (Standard ISPP Single Verify) and ISPP-DV (Double Verify ISPP) algorithms. The objective is to capture how different program algorithms impact the raw bit error
rate RBER and the energy consumption of the memory.¹

5.1 Compact NAND Flash Model

We targeted as a case study a 2-bit per cell (4LC) NAND Flash Memory featuring a 45 nm manufacturing process designed for low-power applications. The entire framework has been implemented on a SPICE-like environment using the STM-45nm technology library [23]. The simulation environment is constituted by two distinct modules: the high-voltage (HV) subsystem of the memory, including the charge pumps, and the voltage regulators exploited for the generation of the voltages required for the programming algorithm (including the verify stage), and a compact model for NAND Flash memories with array simulation capability.

The HV module represents the analog core of a NAND Flash memory. Modifying or reading the number of electrons stored into the floating gate requires a set of bias voltages with a desired precision, timing and granularity. Moreover, many voltages have a value larger than the NAND power supply, requiring the use of several charge pumps. In order to achieve a highly accurate estimation of the energy consumption of each ISPP algorithm considered in this work we have simulated the following blocks of the NAND HV subsystem:

- **Program Charge pump**: this is a conventional 12-stages Dickson modified charge pump [24]. It supplies the high voltage necessary for the ISPP pulse ranging from 14V to 19V.
- **Inhibit Charge pump**: during the Program operation a single page can be addressed in a memory block whereas the other ones need to be inhibited through channel self-boosting techniques [2]. This pump shares the same architecture of the Program Charge pump except for the number of pumping stages which has been reduced to 8 in order to obtain 8V.
- **Verify Charge pump**: the Verify operation consists of a read of the actually programmed page within a block. As for the Program operation, only the selected cells within a block should be read while bypassing the unselected ones with an intermediate voltage of 4.5V. This pump is a 4-stage high-speed Dickson modified pump.
- **Regulators and limiting systems**: each pump generates a growing voltage ramp till the regulation system shuts it down. This is done by comparing a voltage divider in feedback between the output of a charge pump and one input of a differential amplifier, biased with a reference voltage on the other input. The charge pump is then shut down when a target voltage is reached and possibly restarted when the target voltage drops below a reference level. This is the only viable solution for an accurate control of the threshold voltages in a MLC NAND Flash device.

The power consumption of each pump during the various stages of the ISPP algorithms, as measured from the SPICE simulation, is then fed into a NAND Flash power modeling framework based on the equation set provided in [25]. As input parameters of the model, we assumed a low-power NAND Flash supplied with $V_{DD} = 1.8V$ using an ISPP algorithm starting from 14V to 19V, using ∆ISPP steps of 250mV.

The same settings hold both for the ISPP-SV and the ISPP-DV. Similarly, the HV sub-system functionality simulated in this work is designed to work with both algorithms. In fact, in a NAND Flash device the timing and sequence of analog circuitry operations are driven by the embedded microcontroller/FSM by means of a set of interface registers, generating the enable signals for the charge pumps. Switching from ISPP-SV to ISPP-DV does not require a modification of the HV subsystem but rather implies a different sequence of enable signals notified through the same register interface. Implications on the embedded microcontroller/FSM will be discussed in section 6.4.

An additional modeling effort was devoted to NAND Flash cells. We developed a compact model partially based on [26], which includes variability effects typical of nanoscaled memories. This allowed to simulate array functionalities during a page-wide programming operation. Variability effects included the following: width and length geometrical variations of FG-MOS transistors; non-homogeneity 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; aging effects due to repeated Program/Erase 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, negatively impacting the RBER. For the sake of model validation, we were able to fit experimental data collected from [26] as showed in Fig. 4, where cell voltage threshold is plotted during an ISPP operation for a 41nm NAND Flash technology.

![Figure 4. Fitting results of the NAND Flash compact model with experimental data during an ISPP operation featuring 1μs pulses, 1V ∆ISPP.](image)

6. Experimental results

The different sections of the memory sub-system are at first characterized in isolation, and then combined in section 6.3 to quantify the trade-offs of the cross-layer approach to memory configuration.

6.1 Characterization of programming algorithms

Power consumption of the NAND Flash device and its RBER when using the ISPP-SV and the ISPP-DV algorithms have been characterized by means of the developed simulation framework. Such parameters are derived as a function of the Program/Erase cycles of the memory, thus enabling lifetime-wide assessment of memory features.

Fig. 5 shows RBER results for a simulated 4KB page Program of a NAND Flash. Acting only upon Program algorithm selection to improve memory reliability allows to significantly improve RBER figures up to one order of magnitude.

However, this reliability improvement comes with two major penalties: an increase of the power consumption due to the additional verify operations required by the ISPP-DV algorithm, and a reduction of the write throughput due to the increased algorithm run time.

The increased power consumption of the memory device during a program operation with ISPP-DV instead of ISPP-SV has been measured and reported in Fig. 6. Power numbers do not include I/O pins and the digital part, which are irrelevant in the comparative analysis. A shift of just 7.5mW between the two algorithms is measured, which is a marginal 4 to 5% increment with respect to the baseline power with ISPP-SV. Such a consumption mismatch is ascribed mainly to the increased usage of the read charge pump circuitry in the memory HV-subsystem, which anyway does not represent a major source of power drain in the overall

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 it in a way that enforces the ISPP full sequence strategy [2] instead of the two-seconds one since it allows reduced simulation time and faster post-processing of the experimental results.
different RBERs, which even degrade over time, as illustrated in section 6.1. Those RBER ranges then become the x-axis values in Fig. ?? and Fig. 7.

Both Figures show that in the best case (left-hand side), \( t_{\text{MIN}} = 3 \) is sufficient to meet the required reliability constraints. Fig. ?? shows that, in the worst case, the correction capability required by the code is \( t_{\text{MAX}} = 65 \) errors for the ISPP-DV algorithm. Fig. 7 shows that this value grows to \( t_{\text{MAX}} = 65 \) errors for ISPP-SV.

We therefore instantiated a BCH codec architecture for the worst case with correction capability in the range \( t = 3 \div 65 \), which is able to accommodate also the relaxed requirements of ISPP-DV and of both program algorithms over time.

We then characterized the encoding and decoding latency of this ECC architecture when required to guarantee a constant UBER of \( 10^{-11} \) over time and in the presence of the ISPP-SV and ISPP-DV in the flash device. Results are reported in Fig.8. We can see that with ISPP-SV the adaptive ECC is reconfigured over time in an attempt to meet the reliability requirement, thus clearly resulting in longer (de-)coding latencies. Since ISPP-DV can contain the RBER with memory aging, ECC requirements can be relaxed accordingly hence almost keeping a constant latency. As will be pointed out in next section, this latency deviations translates into different write and read throughput figures for the two program algorithms.

### 6.3 The new performance-reliability trade-offs

The cross-domain approach proposed in this work for performance-reliability trade-off leverages the possibility to act both upon physical and architectural parameters, thus combining their settings into unprecedented combinations.
We assume that ISPP-SV and the ECC settings meeting the $10^{-11}$ requirement with that algorithm over time are the baseline configurations of the memory sub-system. With respect to that average case, we then ask for improved read throughput or for improved UBER through our cross-layer memory configuration.

![Figure 9. Write throughput penalty.](image)

**Figure 9.** Write throughput penalty.

![Figure 10. UBER improvement obtained with our approach with respect to nominal NAND Flash operating mode.](image)

**Figure 10.** UBER improvement obtained with our approach with respect to nominal NAND Flash operating mode.

![Figure 11. Read throughput gain as an effect of the cross-layer optimization.](image)

**Figure 11.** Read throughput gain as an effect of the cross-layer optimization.

### 6.3.1 Minimizing UBER

Some mission-critical applications may require an UBER value lower than the typical $10^{-11}$ one, which was originally projected for the SSD and HDD markets [3]. The improvement of the UBER can be obtained by keeping constant the ECC correction capability, while tuning only on the physical layer. Indeed, by switching from the ISPP-SV algorithm to the more reliable ISPP-DV one it is possible to reduce the overall RBER by an order of magnitude, which directly maps into a reduction of the UBER.

Fig. 5 shows that by exploiting this methodology during the whole memory lifetime, it is possible to achieve an UBER boost which on average amounts to two orders of magnitude but peaks at four orders of magnitude at the end of memory lifetime with respect to the nominal setting. The key benefit of this approach is that the UBER boost does not come at the cost of read throughput penalty, since the decoding time is unaffected. There is a power penalty for this strategy, approximately consisting of the 7mW documented in Fig. 6.

This cross-layer configuration provides a new operating point for the memory sub-system which better meets the requirements of a range of applications in the embedded computing domain. For instance, secure transactions for web-based payments, critical data transfers such as operating system (OS) upgrades or internal data backup would all benefit from UBER minimization with respect to the average case.

### 6.3.2 Maximizing Read Throughput

Read intensive scenarios require the maximum read throughput achievable by the memory, eventually penalizing write throughput performance. By acting only upon the memory controller parameters, the ECC strength would have to be reduced, thus degrading the UBER. In contrast, with a cross-layer approach, it is possible to select the ISPP-DV algorithm to provide the best RBER feature during the whole memory lifetime (see Fig. 5), and to continuously relax the ECC configuration to provide a constant UBER of $10^{-11}$.

This solution bears two benefits: the encoding/decoding latency can be reduced since a less aggressive correction capability is required; the power consumption of the ECC can be reduced for the same reason (from 7mW to 1mW in this work). The decoding latency savings enable to improve the memory read throughput of up to 30% at the end of memory lifetime, as shown in Fig. 11. This depends on two factors. On one hand, read throughput is dominated by decoding latency and not by page read time (which takes up to 75µs against the 150µs of the decoding operation [27]). On the other hand, as previously shown in Fig. 8, the improvement is a strict function of memory aging.

The key novelty of this approach lies in the fact that read throughput can be maximized with respect to the average case without impacting UBER. Moreover, the relaxation of ECC performance allows to keep the memory power budget constant since the increased power needs of the physical layer are compensated by the lower power of the ECC sub-system. Of course, this result is strictly ECC architecture dependent. This way, a new operating point is made available to embedded systems where applications such as music playback, video streaming and digitized pictures would greatly benefit from it. All those multimedia-oriented applications dictate as a main requirement a high quality of service (QoS) in terms of decoding/encoding capabilities, occasionally tolerating glitches in video/audio playback or image viewing. Finally, such enhanced application perceived performance comes at no increased power cost.

### 6.3.3 Trading Write Throughput for Adaptivity

In both adaptivity cases discussed above, when switching away from the baseline memory setting, a loss in write throughput has to be expected for improved read throughput or UBER.

In fact, in both cases the program algorithm is switched to ISPP-DV, which takes a longer time to run than the ISPP-SV. Since this time dominates with respect to the encoding latency (1.5ms against the ECC encoder latency which is about two orders of magnitude lower), then the longer program time of the memory can be directly referred to the longer ISPP-DV algorithm. As shown in Fig. 9, the write throughput loss with respect to the baseline setting on average amounts to 40%.

Although the overhead is relevant, it can be mitigated by using a two-round data load strategy on the page buffer of the NAND flash, and is however acceptable for read-intensive applications, where it is the price to pay for improved UBER (without sacrificing read throughput) or for improved read throughput (without sacrificing UBER). In the extreme case, an ideal target for our methodology is the NAND flash sector typically reserved as One-Time-Programmable (for instance for execute-in-place).
6.4 Implementation Complexity

What is worth to be pointed out is that the programming algorithms for NAND Flash memories are generally hardened in a code-ROM memory integrated in the same memory die and accessed by an embedded microcontroller, which executes all the necessary control operations for the algorithm to take place. The physical layer optimization moves from the assumption that more than one algorithm can be stored on the code-ROM, being also runtime user-selectable. This feature anyway does not lead to a complexity overhead of the microcontroller structure as the only change to be devised resides in the small increase of the code-ROM capacity.

Another possible solution, which would increase the reconfigurability of the entire NAND memory, may include the possibility of substituting the code-ROM with a SRAM memory to be written by the memory controller with the most suitable algorithm for the memory transaction at hand. The approach of integrating more than one RAM devices on the memory controller is not unusual [2], especially for debug and test purposes.

7. Conclusions

In this paper, we demonstrate that combining settings at the physical and architectural level in a memory sub-system holds promise of exposing unprecedented trade-off points between performance and reliability. In particular, read throughput can be improved upon demand at runtime without sacrificing UBER and vice versa. This comes with a loss in write throughput and with a marginal power penalty in specific use cases. When combining these trade-offs with those traditionally provided by design techniques for the memory controller architecture, we broach the available trade-off points for memory operation, thus paving the way for a more fine-grained optimization of applications and for a higher degree of memory self-adaptivity. In future work, we intend to implement the memory controller taking advantage of the new trade-offs, thus exposing differentiated storage services to applications.

8. REFERENCES