Optimized Degree Distributions for Binary and Non-Binary LDPC Codes in Flash Memory ISITA 2014

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4 LDPC Code Design and Threshold Calculation



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5 Conclusions



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To read information, apply a specified word-line voltage to the control gate. The sense-amp comparator then provides a single bit of information (whether the transistor is "on", i.e. the drain current is above a specified threshold). The threshold voltage is the lowest voltage at which the transistor turns on. • The threshold voltage is proportional to the amount of charge in the floating gate.

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- The actual charge level written to the floating gate can vary with: Overcharge in the write operation Leakage in the retention period Interference from nearby cells
- So, there is a *distribution* associated with the threshold voltage.



One read system is equivalent to binary symmetric channel (BSC).

Two Reads



With two symmetric reads at q_1 and $-q_1$, the equivalent channel has an erasure region.

Three Reads



Additional reads result in additional outputs and more complicated equivalent communication channel.



For n reads, the channel has n+1 outputs.

Mutual Information Maximization for One Read

Find the read voltages to maximize the mutual information.

maximize
$$I(X; Y) = H(Y) - H(Y|X)$$

For the 1-read system, the optimized read is at zero.



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Mutual Information Maximization for Three Reads

Find the threshold voltages to maximize the mutual information.

maximize
$$I(X; Y) = H(Y) - H(Y|X)$$

For 2,3 reads, q_1 is the solution to $\frac{dI}{dq_1} = 0$.



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Mutual Information Maximization for Five Reads



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• Fully optimized single-read channel [x x 0 x x]: MI = 0.560

Image: A matrix and a matrix

- Fully optimized single-read channel [x x 0 x x]: MI = 0.560
- Fully optimized for 3-read channel [x -0.61 0 0.61 x]: MI = 0.652

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 $[-0.61 - 0.28 \ 0 \ 0.28 \ 0.61]$: MI = 0.663

Concluding Remarks on MMI

Optimized single read	MI = 0.560	[X	х	0	х	x]
Optimized three reads	MI = 0.652	[×	-0.61	0	0.61	x]
Optimized five reads	MI = 0.668	[-0.9	-0.4	0	0.4	0.9]
Progressive three reads	MI = 0.643	[×	-0.4	0	0.4	x]
Progressive five reads	MI = 0.663	[-0.61	-0.28	0	0.28	0.61]

The MI differences are small, but large enough to have noticeable effects on frame error rate (according to [1]).

Optimizing q_1 for three-read performance maximizes the probability of decoding successfully after three reads.

Optimizing q_1 for the five-read scenario will produce the lowest probability of losing a page.

[1] Wang, J., et al. "Enhanced Precision Through Multiple Reads for LDPC Decoding in Flash Memories," IEEE JSAC , May 2014

In order to design LDPC codes that are matched to the Flash read channel, we use EXIT functions with the Gaussian approximation and Reciprocal Channel Approximation (faster alternative to Density Evolution) to compute LDPC decoding thresholds for different number of reads.



For a fixed degree distribution and q_1 , RCA-EXIT analysis can determine the word-line voltages that minimize E_b/N_0 threshold for multiple-read Flash channels.


For a fixed degree distribution, there is a q_1 that has the absolute E_b/N_0 threshold for multiple-read Flash channels.



For each Eb/N_0 there is a q_1 that maximizes the MI.



The threshold Eb/N_0 achieved with q_1 chosen to maximize MI is only 1% away from the best possible Eb/N_0^* achieved with a density-evolution optimized q_1 .

Optimized Degree Distributions



The optimized degree distribution is found by calculating the threshold at q_{MMI} and further optimizing the q_1 .

Table: Thresholds for optimized degree distributions on SLC flash with 1, 3, and 5 reads as well as soft information.

Target	1 read	3 reads	5 reads	Soft
Threshold	Th.	Th.	Th.	Th.
Optimized for 1 read	4.752	3.728	3.542	3.398
Optimized for 3 reads	4.923	3.640	3.441	3.295
Optimized for 5 reads	4.926	3.649	3.437	3.288
Optimized for Soft	4.926	3.662	3.443	3.275
Shannon-Limit	4.400	3.495	3.328	3.198

Different codes have different thresholds for different levels of quantization. A code designed for one level of quantization is not the best for another level. Table: Gap of thresholds for optimized degree distributions on SLC flash with 1, 3, and 5 reads as well as soft information.

Target	1 read	3 reads	5 reads	Soft
threshold gap	Th. gap	Th. gap	Th. gap	Th. gap
Optimized for 1 read	0	0.088	0.105	0.123
Optimized for 3 reads	0.171	0	0.004	0.020
Optimized for 5 reads	0.174	0.009	0	0.013
Optimized for Soft	0.1740	0.022	0.006	0

The code optimized for a single read performs well across multiple reads. There is reason to consider using the code designed for a single read.

Regular LDPC Code Baselines and Bounds



The threshold of two regular binary LDPC codes vs. Shannon-limit for different number of reads.



The threshold of regular NB-LDPC code for different number of reads.

Optimized Binary LDPC Code



The threshold of an optimized binary LDPC code for different number of reads.



Huge gain can be achieved by using a rate-0.45 NB-LDPC code on MLC with the same number of reads and for the same spectral efficiency of 0.9.

Non-Binary code on MLC



Huge gain can be achieved by using a rate-0.45 NB-LDPC code on MLC with the same number of reads and for the same spectral efficiency of 0.9.

Table of Degree Distributions

# Reads	Coefficients						
SLC	λ_2	λ_3	λ_7	λ_8	λ_{27}	ρ_{61}	
1 read	0.07	0.25	0.11	0.13	0.44	1	
SLC	λ_2	λ_3	λ_6	λ_7	λ_{26}	ρ_{56}	
3 reads	0.1	0.21	0.11	0.12	0.46	1	
SLC	λ_2	λ_3	λ_5	λ_8	λ_{25}	$ ho_{56}$	
5 reads	0.11	0.21	0.09	0.14	0.45	1	
SLC	λ_2	λ_3	λ_6	λ_8	λ_{26}	$ ho_{56}$	
soft	0.11	0.21	0.21	0.14	0.47	1	
MLC	λ_2	λ_3	λ_4	λ_7	λ_8	λ_{11}	
	0.16	0.31	0.1	0.18	0.1	0.15	
3 reads	$ ho_5$	$ ho_6$	ρ_8	ρ_{22}			
	0.45	0.16	0.1	0.29			
MLC	λ_2	λ_3	λ_4	λ_7	λ_9	λ_{11}	
	0.15	0.3	0.1	0.09	0.22	0.14	
soft	$ ho_5$	$ ho_6$	$ ho_8$	$ ho_9$	ρ_{22}		
	0.44	0.07	0.1	0.11	0.27		

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- Optimized Degree Distributions can be found by calculating the threshold at q_{MMI} and further optimizing the q to get the lowest E_b/N_0 thresholds.
- Even though a code designed for one level of quantization is not the best for another level of quantization, the code optimized for a single read performs well across multiple reads.
- Using a low-rate LDPC code on MLC for the same spectral efficiency as a high-rate code on SLC can provide huge gain in terms of the thresholds.