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LDPC Codes for Quantitative Group Testing with a Non-Binary Alphabet

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Joint work with Mgeni Makambi Mashauri[†] and Alexandre Graell i Amat[‡]

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Munich Workshop on Shannon Coding Techniques

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Background: Group Testing

- ▶ We have a large population of items



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- ▶ Very few of them are "defective" (probability of being defective, γ is very small)



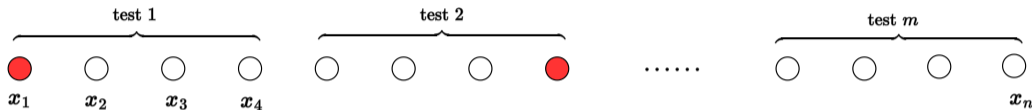
- ▶ **Goal:** Identify \mathbf{x} : defective ($x_i = 1$), non-defective ($x_i = 0$)

[Dorfman1943] Robert Dorfman, "The Detection of Defective Members of Large Populations," *The Annals of Mathematical Statistics*, vol. 14, no. 4, pp. 436–440, 1943.



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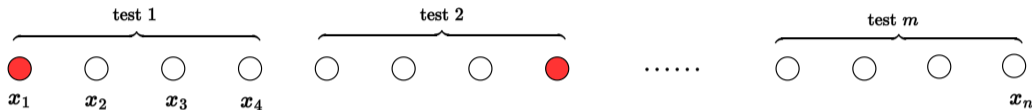
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- ▶ Rate, $\Omega = \frac{m}{n}$ (smaller is better)

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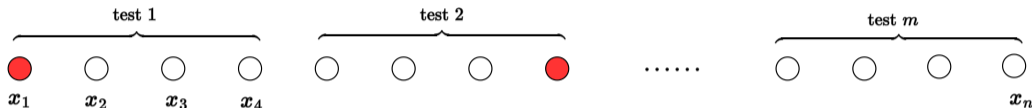
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Background: Graphical Representation

- ▶ For non-adaptive group testing the pooling can be represented by a test matrix A

$$A = \begin{pmatrix} 1 & 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 & 1 \end{pmatrix}$$

x_1 x_2 x_3 x_4 x_5 x_6



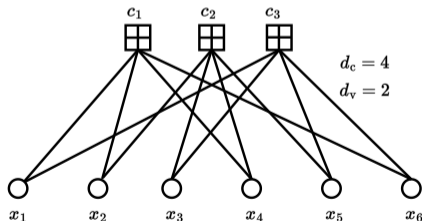
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$x_1 \quad x_2 \quad x_3 \quad x_4 \quad x_5 \quad x_6$

- ▶ The matrix can be represented by a bipartite graph G



- ▶ We consider the scenario where the graph is sparse



Non-quantitative vs Quantitative

- **Non-quantitative:** test result, $s_i = 1$ if at least one item is defective otherwise $s_i = 0$ (logical OR)

	●	●	○	○	○	●		Non-Quantitative
	0	0	1	1	1	0	s_1	0
	1	1	0	1	0	0	s_2	1
	0	1	1	0	1	0	s_3	1
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)								3

- For **quantitative** group testing, a test result shows the number of defective items

$$s_i = \sum_{j=1}^n x_j a_{ij} \rightarrow s = Ax$$



Quantitative Group Testing with Sparse Graphs: Prior work

- ▶ The test results show the number of defectives
- ▶ Best known scheme with sparse graph uses GLDPC [KAR2019]

$$U = \begin{pmatrix} \overbrace{\begin{matrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 \end{matrix}}^{d_c} \\ t = 1, m_U = 3 \end{pmatrix}$$

- ▶ A t -error-correcting BCH code is used as a component code
- ▶ An additional row of ones to identify # of defective items

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- ▶ $d_c = 2^{m_u} - 1 \rightarrow m_u = \log_2(d_c + 1)$
- ▶ Tests per subcode = $t \log_2(d_c + 1) + 1$

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► Density Evolution

For each iteration ℓ

$q^{(\ell)}$: probability a test sends **resolved** to item

$p^{(\ell)}$: probability a defective item is **unresolved**

Test to item:

$$q^{(\ell)} = \sum_{i=0}^{t-1} \binom{d_c - 1}{i} \left(p^{(\ell-1)}\right)^i \left(1 - p^{(\ell-1)}\right)^{d_c - 1 - i}$$

Item to test:

$$p^{(\ell)} = \gamma (1 - q^{(\ell-1)})^{d_v - 1}$$



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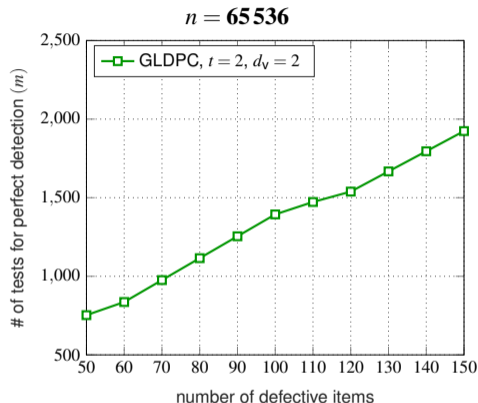
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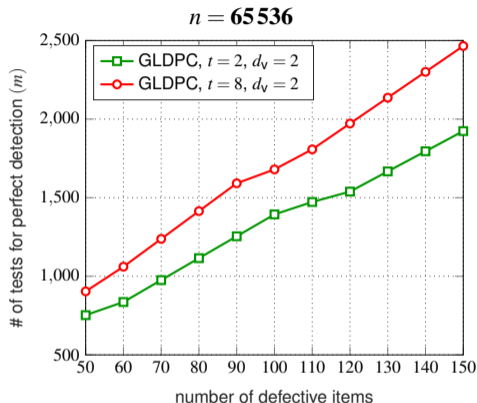
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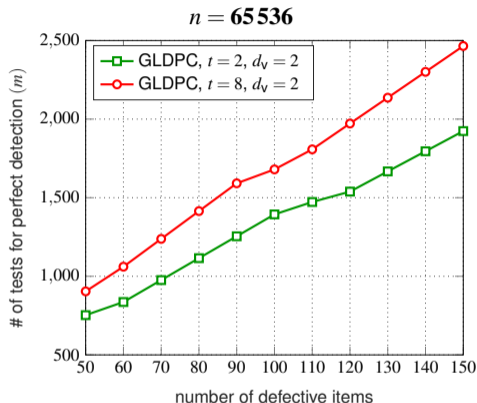
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■ What about using $t = 0$?



Proposed scheme: Group Testing with LDPC

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?	?	?	?	?	?		
●	○	●	●	●	○	$s^{(1)}$	$d_c^{(1)}$
0	0	1	1	1	0	3	3
1	1	0	1	0	0	2	3
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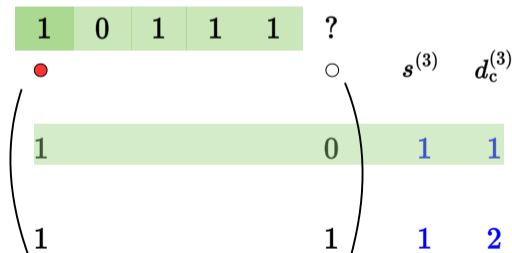
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?	0	1	1	1	?		$s^{(2)}$	$d_c^{(2)}$
●	○				○			
(1	1			0		1	2
	0	1			0		0	1
	1	0			1		1	2
)								



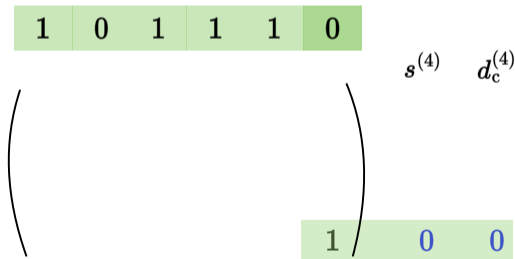
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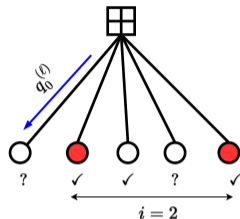
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Density Evolution

- ▶ $p_1^{(\ell)}$: probability that a message from a defective is *unresolved*
- ▶ $q_0^{(\ell)}$: probability that a message to a non-defective is *resolved*
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From test to item

$$q_0^{(\ell)} = \sum_{i=0}^{d_c-1} \binom{d_c-1}{i} \gamma^i (1-\gamma)^{d_c-1-i} (1-p_1^{(\ell-1)})^i$$

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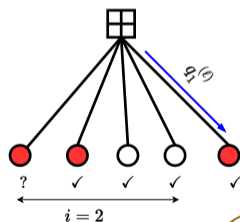
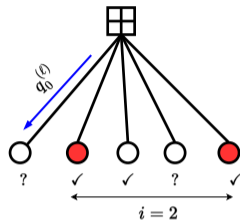
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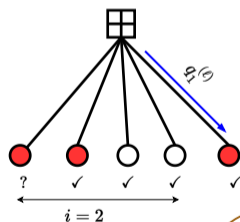
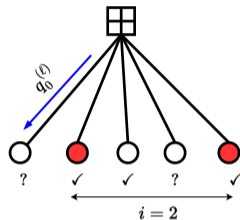
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From item to test

$$p_0^{(\ell)} = (1 - q_0^{(\ell-1)})^{d_v-1}$$

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Performance Comparison

- ▶ We consider two scenarios
 - **Fixing the proportion of defective items γ and changing the rate $\Omega = \frac{m}{n}$**



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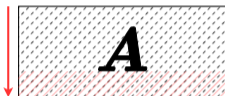
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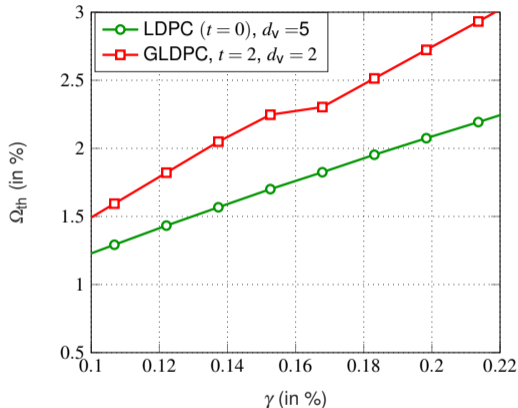
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Minimum rate required for a fixed γ



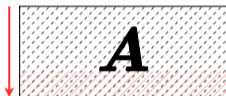
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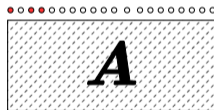
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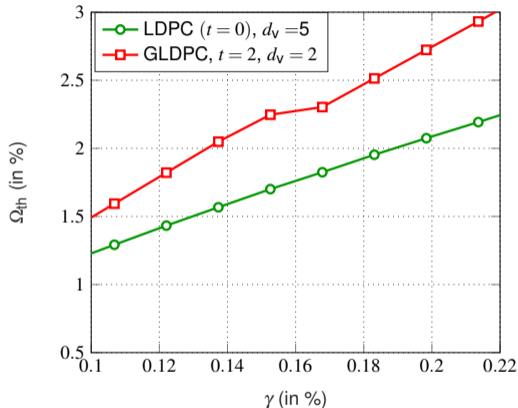


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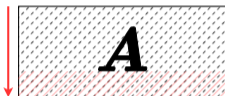
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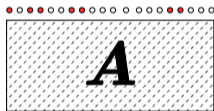
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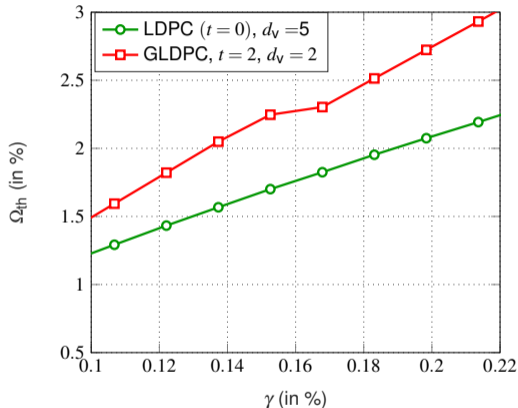
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- **Fixing the rate Ω and changing γ**



- ▶ A new perspective considering A (code) as fixed

Minimum rate required for a fixed γ



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Performance Comparison: Fixed Rate, $\Omega = 5\%$

Table: GLDPC Based

t	d_v	γ_{th}
1	2	0.2487
	3	0.3708
	4	0.3510
2	2	0.3983
	3	0.3372
	4	0.2884
3	2	0.3784
	3	0.3189
	4	0.2441
5	2	0.3418
	3	0.2686
	4	0.2014



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Table: LDPC Based

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5	0.6416
6	0.6464
7	0.6353
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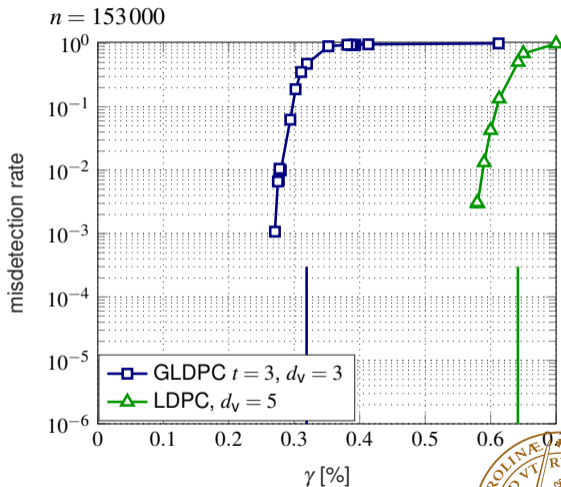
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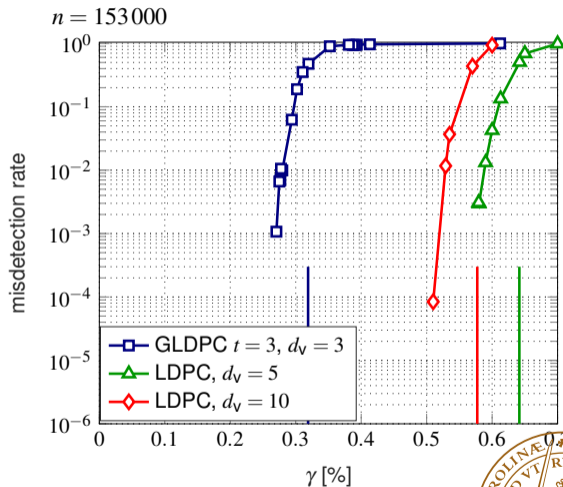
Performance Comparison: Fixed Rate, $\Omega = 5\%$

Table: GLDPC Based

t	d_v	γ_{th}
1	2	0.2487
	3	0.3708
	4	0.3510
2	2	0.3983
	3	0.3372
2	4	0.2884
	2	0.3784
3	3	0.3189
	4	0.2441
5	2	0.3418
	3	0.2686
4	0.2014	

Table: LDPC Based

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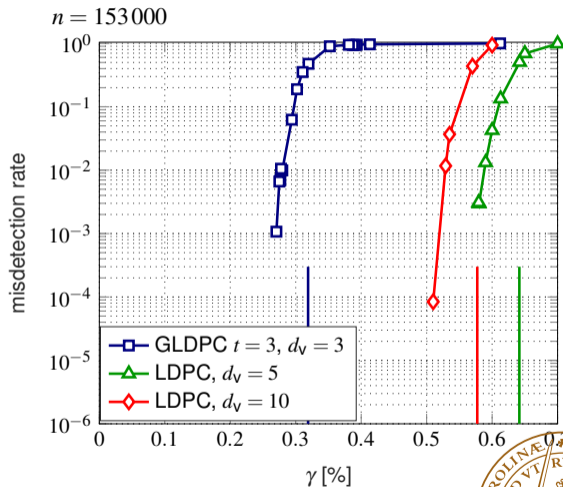
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■ Can we get consistently better with increasing d_v ? What about spatial coupling?



Improved Decoding with a Non-Binary Alphabet

- ▶ **Idea:** for a subset of tests, items occur only in bundles of size q



Improved Decoding with a Non-Binary Alphabet

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Example: $d_v = 3$, $d_c = 4$

$$\mathbf{A} = \begin{array}{c} \begin{array}{cccc} z_1 & z_2 & z_3 & z_4 \end{array} \\ \left[\begin{array}{cccccc} \boxed{1} & \boxed{1} & 0 & 0 & 0 & 0 & \boxed{1} & \boxed{1} \\ 0 & 0 & \boxed{1} & \boxed{1} & \boxed{1} & \boxed{1} & 0 & 0 \\ \boxed{1} & \boxed{1} & 0 & 0 & \boxed{1} & \boxed{1} & 0 & 0 \\ 0 & 0 & \boxed{1} & \boxed{1} & 0 & 0 & \boxed{1} & \boxed{1} \\ \hline 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 \end{array} \right] \end{array} \quad q = 2$$

$x_1 \ x_2 \ x_3 \ x_4 \ x_5 \ x_6 \ x_7 \ x_8$



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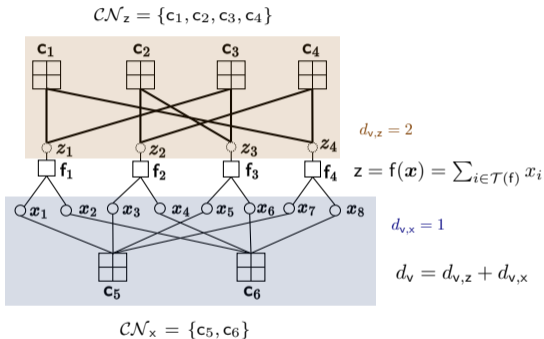
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	z_1	z_2	z_3	z_4				
$A =$	1	1	0	0	0	0	1	1
	0	0	1	1	1	1	0	0
	1	1	0	0	1	1	0	0
	0	0	1	1	0	0	1	1
	1	0	1	0	1	0	1	0
	0	1	0	1	0	1	0	1
	x_1	x_2	x_3	x_4	x_5	x_6	x_7	x_8

$q = 2$

Factor graph representation:



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Local Decoding with q -ary Variables

Extension of the erasure decoder to $q > 1$:

- ▶ messages: $\mu \in \{0, q, ?\}$
- ▶ **Problem:** can still only resolve $s = 0$ and $s = d_C$, no gain with q



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APP decoding (SISO):

- ▶ messages are probability vectors $\mu = [P(z = 0), P(z = 1), \dots, P(z = q)]$, computed in a trellis
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Proposed decoder: motivated by works on counter braids [LM+2008][RG2018]

- ▶ messages $\mu = [L, U]$ consist of **lower bound** L and **upper bound** U on $z \in \{0, \dots, q\}$
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[LM+2008] Y. Lu, A. Montanari, B. Prabhakar, S. Dharmapurikar, and A. Kabbani, "Counter braids: A novel counter architecture for per-flow measurement," *Int. Conf. Meas. Modeling Comput. Syst. (SIGMETRICS)*, Annapolis, June 2008.

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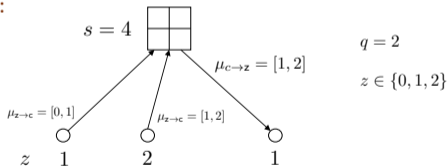
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Example:



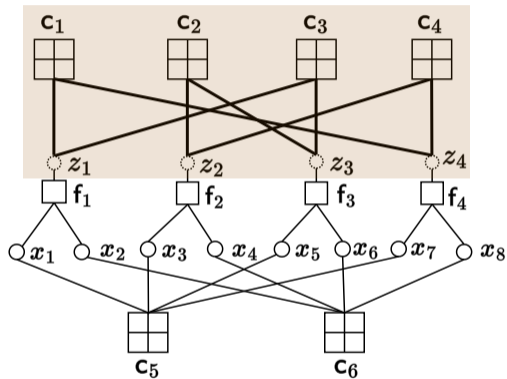
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Message Passing between Bundles and Tests

$$\mathcal{CN}_z = \{c_1, c_2, c_3, c_4\}$$

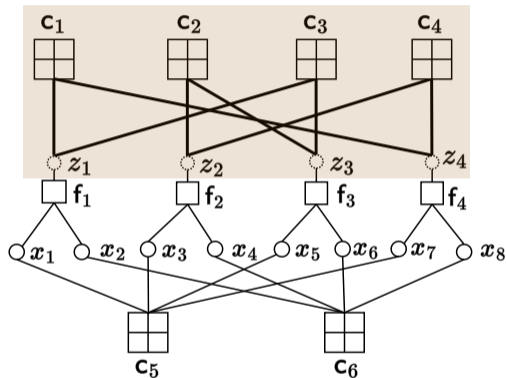


$$\mathcal{CN}_x = \{c_5, c_6\}$$



Message Passing between Bundles and Tests

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$$\mathcal{CN}_x = \{c_5, c_6\}$$

Test to bundle:

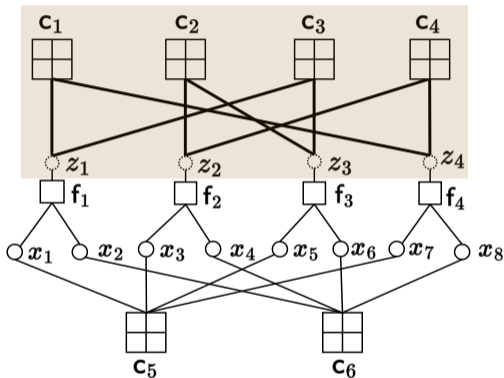
$$L_{c \rightarrow z}^{(\ell)} = \max \left\{ s(c) - \sum_{z' \in \mathcal{T}(c) \setminus z} U_{z' \rightarrow c}^{(\ell-1)}, 0 \right\}$$

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Bundle to test:

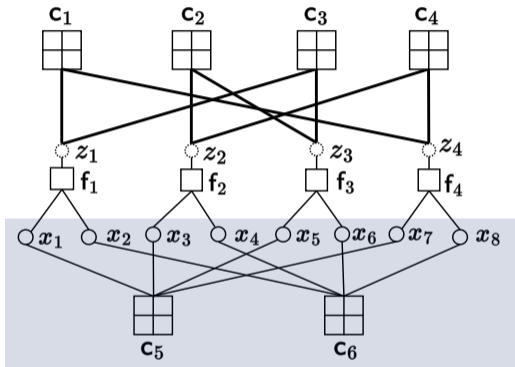
$$L_{z \rightarrow c}^{(\ell)} = \max \left\{ \max_{c' \in \mathcal{T}(z) \setminus c} L_{c' \rightarrow z}^{(\ell-1)}, L_{f \rightarrow z}^{(\ell)} \right\}$$

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Message Passing between Items and Tests

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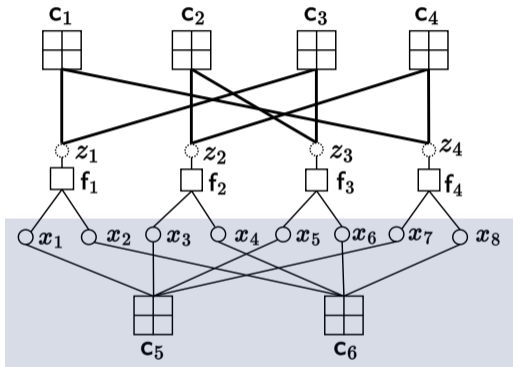


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Message Passing between Items and Tests

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Test to item:

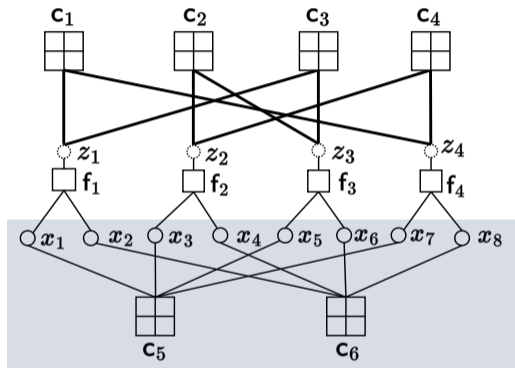
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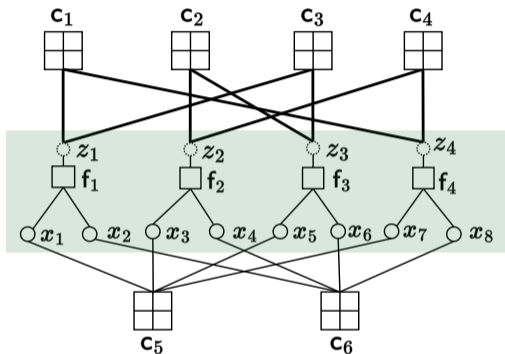
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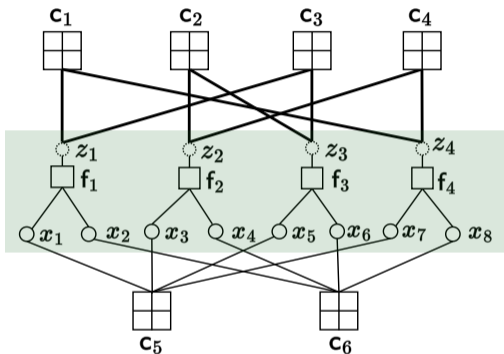


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Bundle to item:

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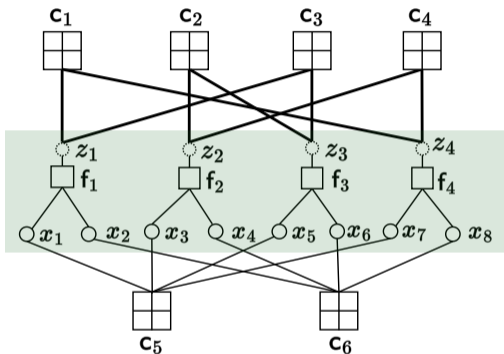
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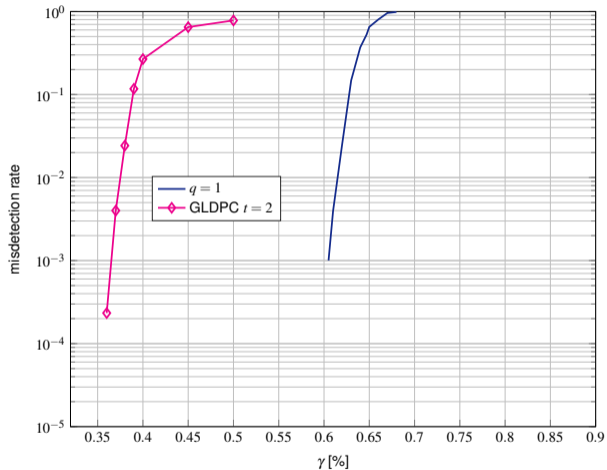
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Performance Evaluation



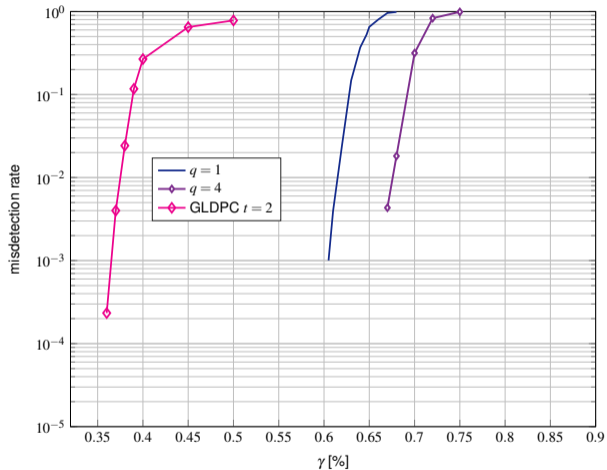
Simulation results:

$n = 210000$ items

Fixed rate $\Omega = 5\%$, i.e., $m = 10500$ tests



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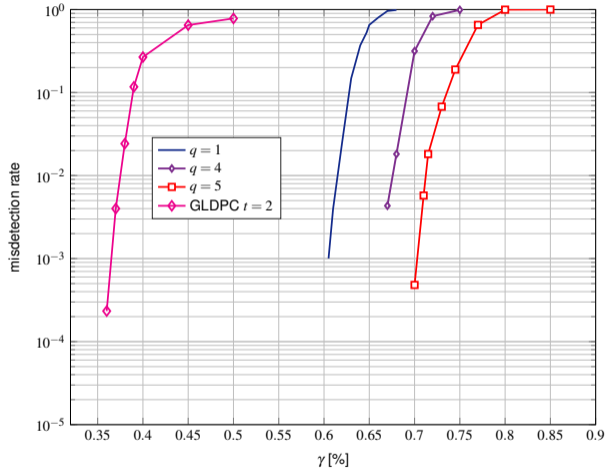


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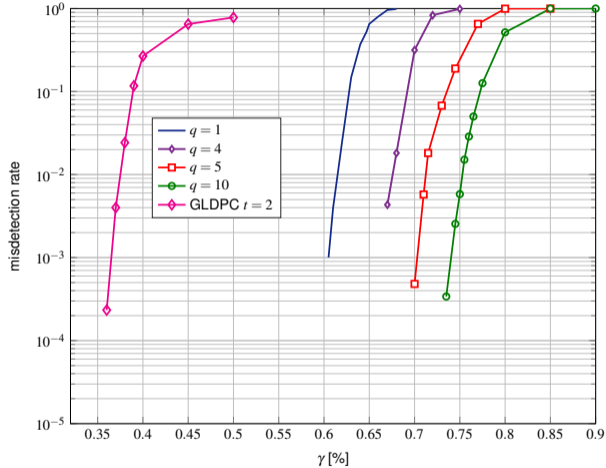


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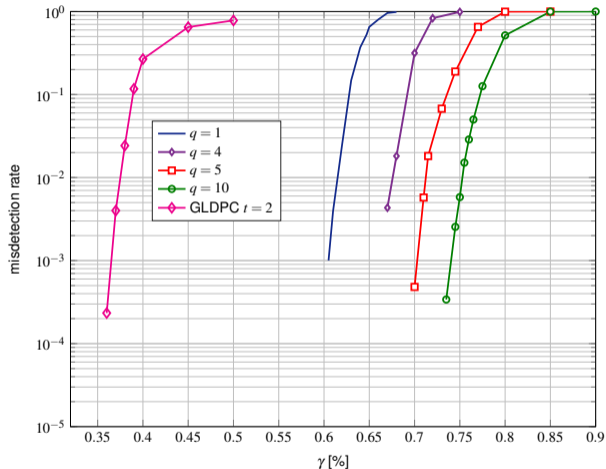
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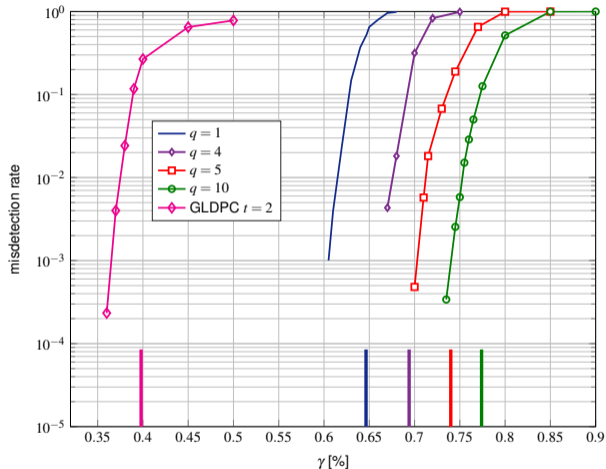
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Density evolution thresholds: γ_{th}

q	$d_{v,x}$	$d_v = 4$	$d_v = 5$	$d_v = 6$	$d_v = 7$	$d_v = 8$
1		0.598	0.641	0.646	0.635	0.618
4	2	0.590	0.660	0.694	0.706	0.702
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10	3	0.549	0.636	0.693	0.774	0.694



Performance Evaluation



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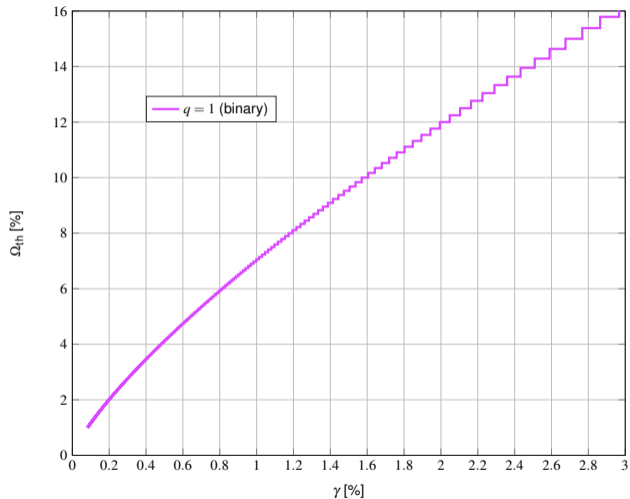
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Minimum Rate Ω_{th} for a Fixed γ

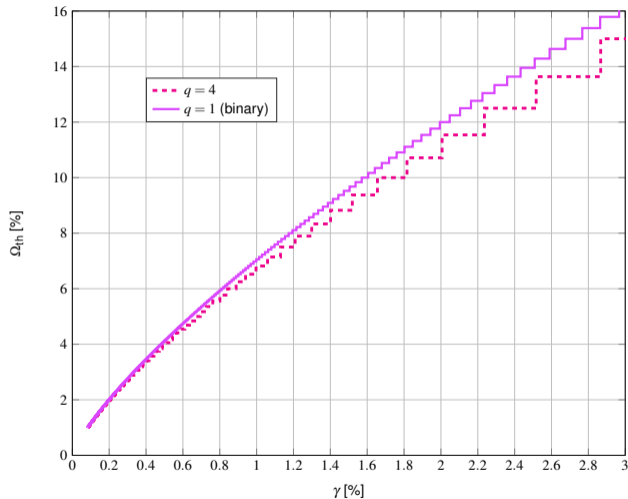


$$\Omega_{\text{th}}^* = \frac{d_V}{d_C} = \frac{m}{n}$$

(smaller is better)



Minimum Rate Ω_{th} for a Fixed γ

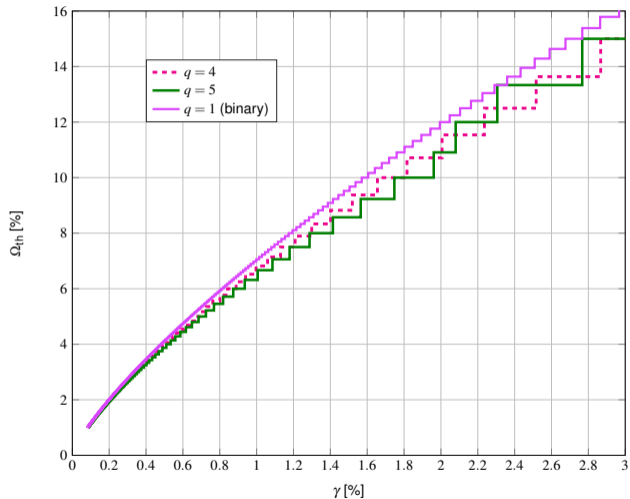


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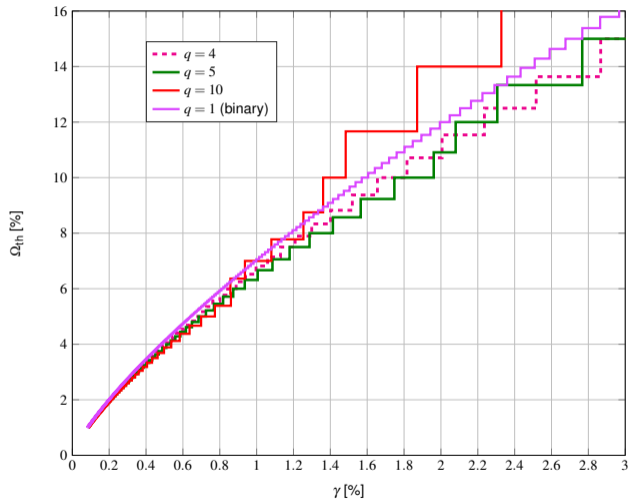


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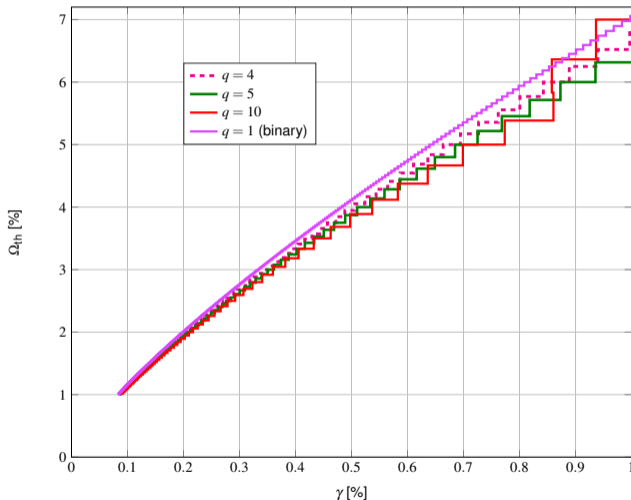
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(smaller is better)



Minimum Rate Ω_{th} for a Fixed γ

Consider a smaller range:



$$\Omega_{\text{th}}^* = \frac{d_V}{d_C} = \frac{m}{n}$$

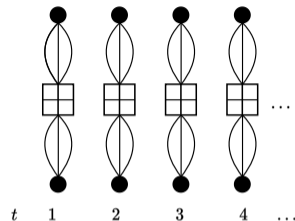
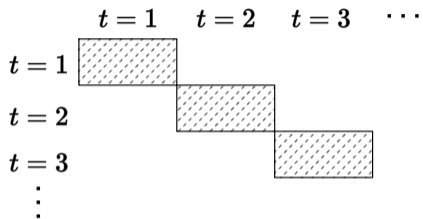
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Observe:
best q depends on range of γ



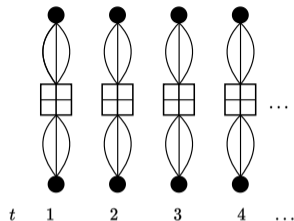
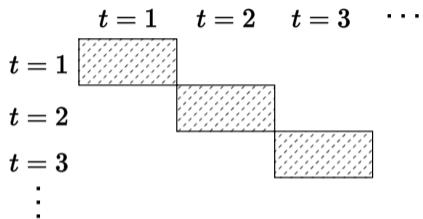
Group Testing with Spatial Coupling

► **Classical approach:** test each block of items separately

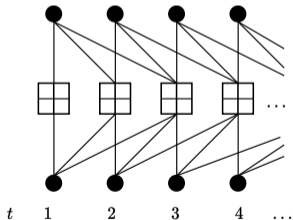
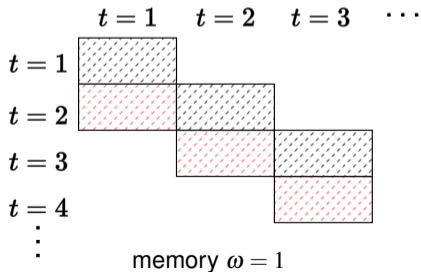


Group Testing with Spatial Coupling

- **Classical approach:** test each block of items separately



- **Spatial Coupling:** Interconnect blocks (motivated by results in coding theory)



Spatial coupling: Performance for Fixed Rate ($q = 1$)

Table: γ_{th} for $\Omega = 5\%$ with GLDPC Code-Based

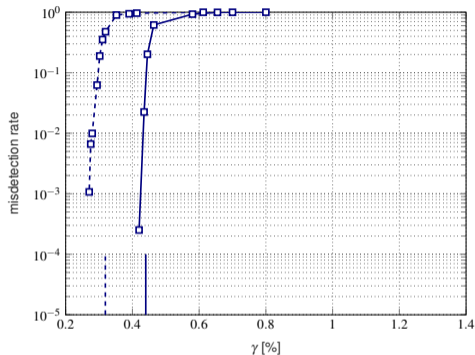
t	d_V	$\omega = 0$	$\omega = 1$	$\omega = 5$	$\omega = 10$
1	3	0.3708	0.4166	0.4166	0.4166
	4	0.3510	0.4395	0.4425	0.4425
3	3	0.3189	0.4257	0.4379	0.4395
	4	0.2441	0.3662	0.4028	0.4028
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—□— GLDPC $t = 3, d_V = 3$

■ $n = 153\,000, L = 200, \omega = 5$

■ solid(coupled) dashed(uncoupled)



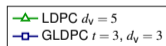
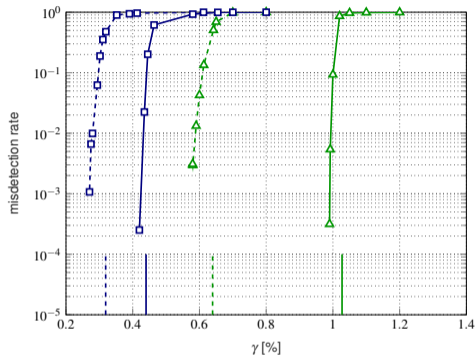
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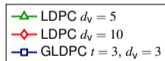
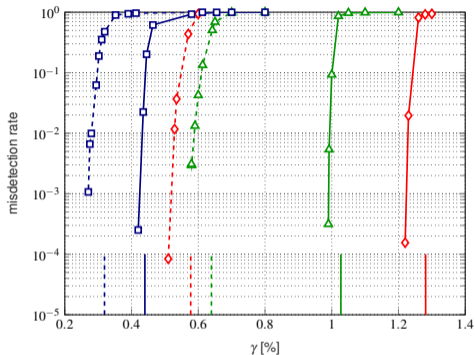
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Proving threshold saturation: minimum Ω for a fixed γ ($q = 1$)

- **Vector admissible system:** [YED2012] a recursion (\mathbf{f}, \mathbf{g}) with

$$\mathbf{x}^{(\ell)} = \mathbf{f}\left(\mathbf{g}(\mathbf{x}^{(\ell-1)}); \varepsilon\right), \quad \mathbf{x}^{(0)} = \mathbf{1}, \quad \varepsilon \in [0, 1]$$

where $\mathbf{f}(\mathbf{x}) = [f_1(\mathbf{x}), \dots, f_d(\mathbf{x})]$ and $\mathbf{g}(\mathbf{x}) = [g_1(\mathbf{x}), \dots, g_d(\mathbf{x})]$ are twice continuously differentiable and strictly increasing in all arguments.

[YED2012] A. Yedla, Y.-Y. Jian, P. S. Nguyen, and H. D. Pfister, "A simple proof of threshold saturation for coupled vector recursions," in *Proc. IEEE Inf. Theory Workshop (ITW)*, 2012.



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- ▶ Setting $\varepsilon = 1 - \frac{1}{d_c}$ we get from density evolution equations:

$$\mathbf{f}(y_0, y_1; \varepsilon) = \left[1 - (1 - y_1)^{\frac{\varepsilon}{1-\varepsilon}}, \quad 1 - (1 - y_0)^{\frac{\varepsilon}{1-\varepsilon}} \right]$$
$$\mathbf{g}(x_0, x_1) = \left[(1 - \gamma) \cdot x_0^{d_v - 1}, \quad \gamma \cdot x_1^{d_v - 1} \right]$$

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- ▶ The potential function is then given as

$$U(\mathbf{x}; \varepsilon) = \int_0^1 \left((\mathbf{z}(\lambda) - \mathbf{f}(\mathbf{g}(\mathbf{z}(\lambda)); \varepsilon)) \mathbf{D}\mathbf{g}'(\mathbf{z}(\lambda)) \right) \cdot \mathbf{z}'(\lambda) d\lambda$$

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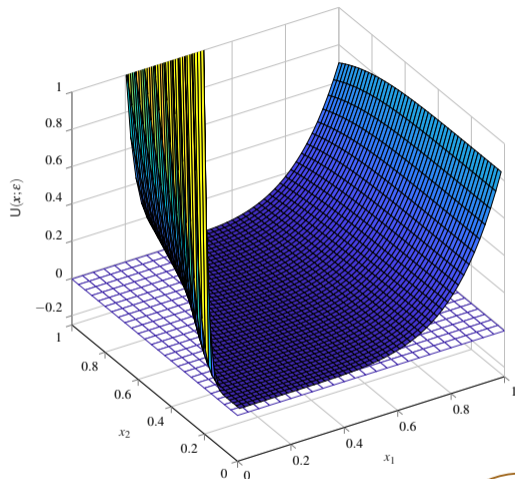


Potential function ($q = 1$)

$$U(\mathbf{x}; \varepsilon) = (1-p)x_1^{d_v-1} \left((1-\varepsilon) \frac{1 - (1-px_2^{d_v-1})^{\frac{1}{1-\varepsilon}}}{px_2^{d_v-1}} + \frac{(d_v-1)}{d_v}x_1 - 1 \right) \\ + px_2^{d_v-1} \left((1-\varepsilon) \frac{1 - (1-(1-p)x_2^{d_v-1})^{\frac{1}{1-\varepsilon}}}{(1-p)x_1^{d_v-1}} + \frac{(d_v-1)}{d_v}x_2 - 1 \right)$$

Potential threshold:

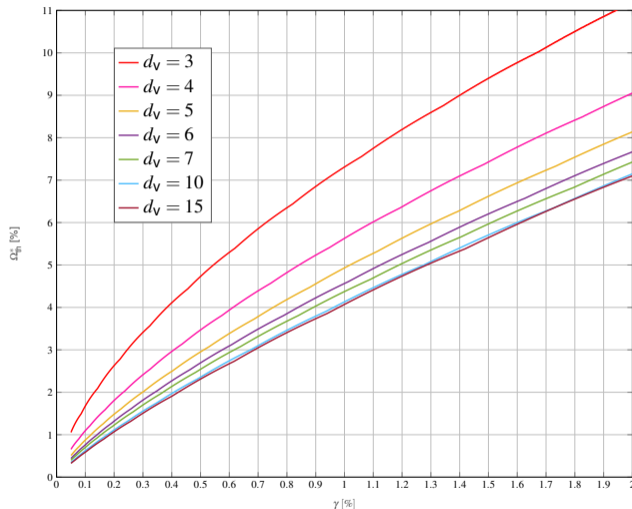
$$\varepsilon^* = \sup\{\varepsilon \in [0, 1] \mid \min_{\mathbf{x}} U(\mathbf{x}; \varepsilon) \geq 0\}.$$



$d_v = 6$, $\gamma = 1\%$ with $\varepsilon^* = 0.9924$. $U(\mathbf{x}; \varepsilon)$ is above the $z = 0$ plane since $\varepsilon = 0.9667 < \varepsilon^*$.



Potential thresholds ($q = 1$)



$$\Omega_{\text{th}}^* = \frac{d_v}{d_c} = d_v(1 - \varepsilon^*).$$

$$\varepsilon^* = \sup\{\varepsilon \in [0, 1] \mid \min_{\mathbf{x}} U(\mathbf{x}; \varepsilon) \geq 0\}.$$

The minimum rate Ω_{th}^* for a fixed γ computed from the potential threshold ε^* .



Conclusions and Outlook

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- ▶ We can measure the performance by two different approaches
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Outlook

- ▶ Spatial coupling with q -bundles
- ▶ Looking at soft message passing

