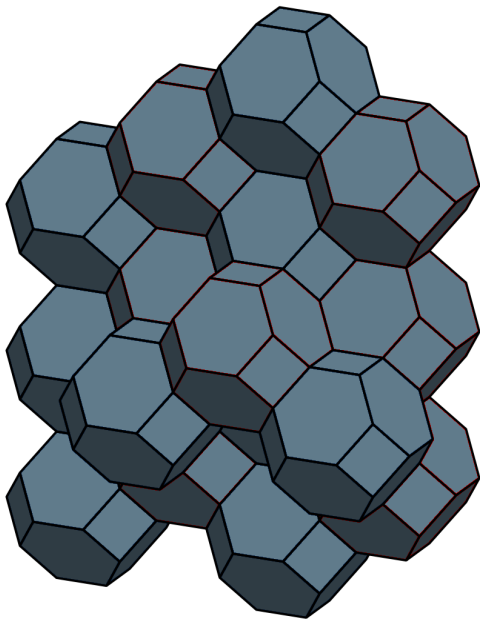


Balls of Permutations in the l_∞ -Metric

Moshe Schwartz

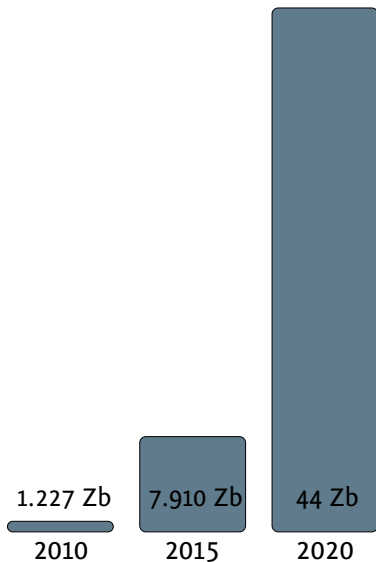
Electrical and Computer Engineering
Ben-Gurion University of the Negev



An age of data

Data storage is growing exponentially.

Storage technologies are crucial to enabling **big data** and **cloud** paradigms.



IDC Digital Universe Study, 2011 & 2014

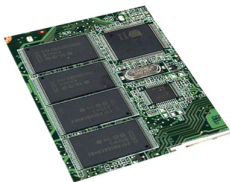
What storage options do we have?

What storage options do we have?

Magnetic HDD



Flash SSD



Cheap

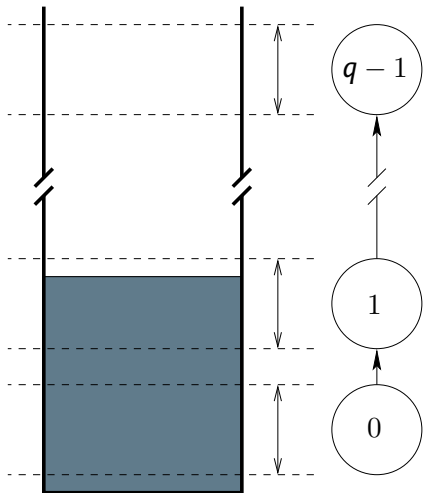
Many R/W operations

Fast

Energy Efficient

Resilient (no mechanical parts)

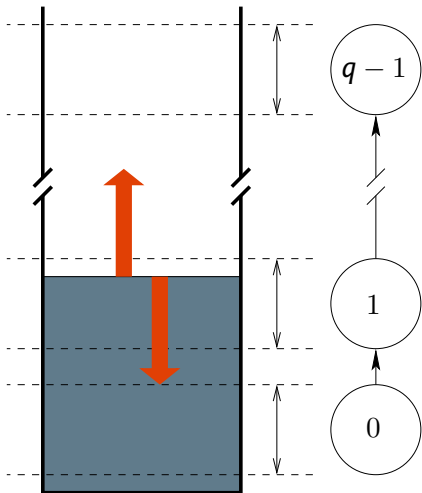
A flash memory cell is a bucket



Flash memory is an electronic non-volatile memory.

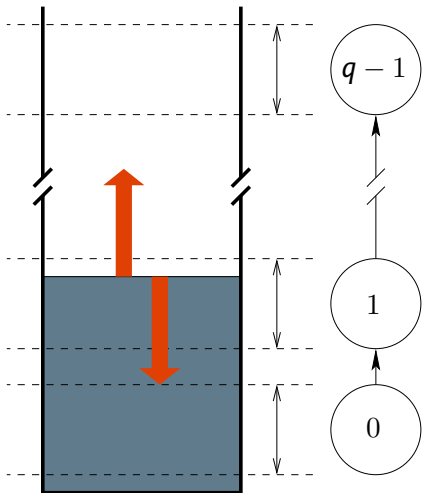
Conventional multi-level flash technology allows each cell to store q discrete values. These are represented by appropriate charge-level intervals.

Flash is an asymmetric noisy bucket



- 1 Adding charge is much easier than removing it. (asymmetric)
- 2 Errors commonly change charge level by a limited magnitude. (noisy)

Flash is an asymmetric noisy bucket

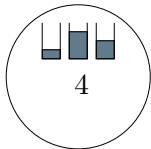
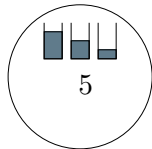
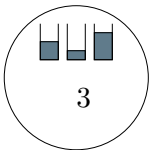
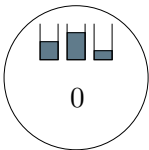
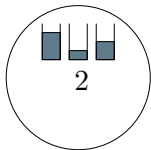
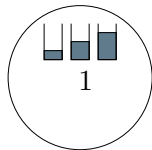


- 1 Adding charge is much easier than removing it. (asymmetric)
- 2 Errors commonly change charge level by a limited magnitude. (noisy)



- 1 Writing is slow.
- 2 Data loss occurs with time.

Rank Modulation - Rethinking flash

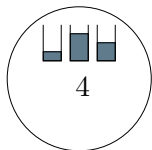
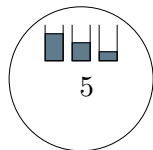
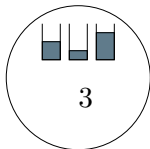
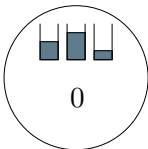
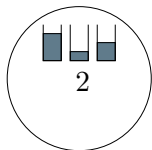
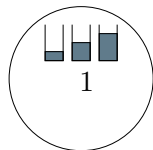


New Paradigm

Cells are compared against each other.

Information is stored in the **permutation** induced by their ranking.

Rank Modulation - Rethinking flash



New Paradigm

Cells are compared against each other.

Information is stored in the **permutation** induced by their ranking.

- 1 Faster writing
- 2 Less errors.

Jiang, Mateescu, Schwartz, and Bruck, IEEE Trans. Inform. Th., 2009

The goal

Construct Error-Correcting Codes for Rank Modulation

What do errors look like?

Limited-magnitude change in charge level



Limited-magnitude change in ranks of cells



Limited ℓ_∞ -distance between stored permutations

Definition

For all $f, g \in S_n$, $d_\infty(f, g) = \max_i |f(i) - g(i)|$.

A simple code

(S_n, d_∞) is a metric space, so we can define error-correcting codes.

Definition

An (n, M, d) -LMRM code (Limited-Magnitude Rank-Modulation code) is a subset $C \subseteq S_n$ of size M such that $d_\infty(f, g) \geq d$ for all $f, g \in C, f \neq g$.

Tamo and Schwartz, IEEE Trans. Inform. Th., 2010

Kløve, Lin, Tsai, and Tzeng, IEEE Trans. Inform. Th., 2010

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Construction

Given $n, d \in \mathbb{N}$ we construct

$$C = \{f \in S_n \mid \forall i \in [n], f(i) \equiv i \pmod{d}\}.$$

How good is the code?

We are interested in the **asymptotics** of the code. For an (n, M, d) -LMRM code we define the **normalized distance** and **rate**:

$$\delta = \frac{d}{n-1}, \quad R = \frac{\log_2 M}{n}.$$

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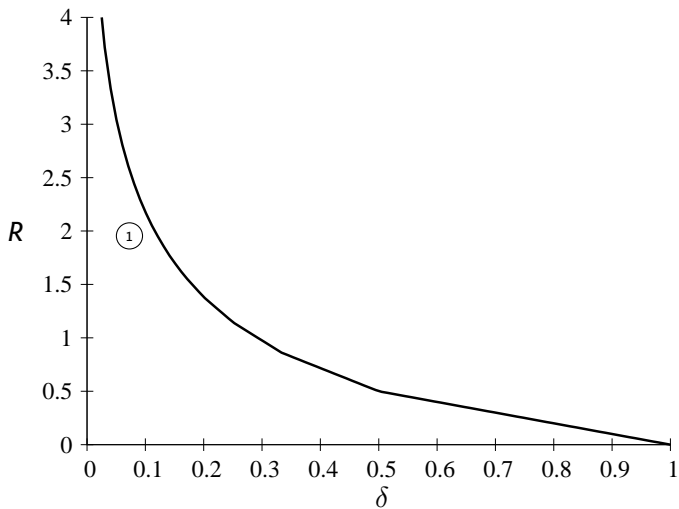
$$\delta = \frac{d}{n-1}, \quad R = \frac{\log_2 M}{n}.$$

Theorem

For $0 < \delta \leq 1$, the construction produces codes of rate

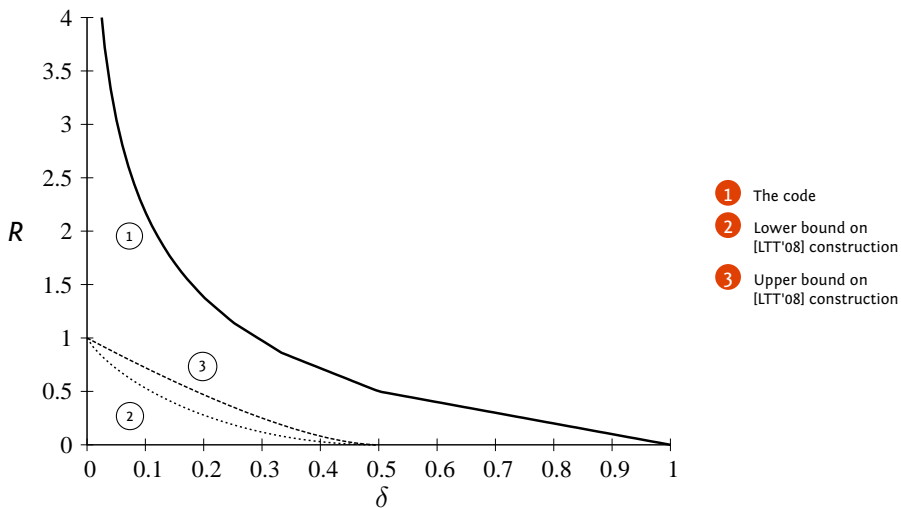
$$R = \left(1 - \delta \left\lfloor \frac{1}{\delta} \right\rfloor\right) \log_2 \left(\left\lceil \frac{1}{\delta} \right\rceil!\right) + \left(\delta + \delta \left\lfloor \frac{1}{\delta} \right\rfloor - 1\right) \log_2 \left(\left\lfloor \frac{1}{\delta} \right\rfloor!\right) + o(1).$$

How good is the code?



1 The code

How good is the code? (Cont.)



Lin, Tsai, and Tzeng, *Int. Symp. on Inform. Th.*, 2008

We can obtain better bounds

- A **ball** of radius r centered at f is $\mathcal{B}_{r,n}(f) = \{g \in \mathcal{S}_n \mid d_\infty(f, g) \leq r\}$.

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Theorem (Gilbert-Varshamov-Like Bound)

Let n, M , and d , be positive integers such that $|\mathcal{B}_{d-1,n}| M \leq n!$. Then there exists an (n, M, d) -LMRM code.

Theorem (Ball-Packing Bound)

Let C be an (n, M, d) -LMRM code. Then $|\mathcal{B}_{\lfloor (d-1)/2 \rfloor, n}| M \leq n!$.

But wait...

What is the size of a ball?

Permanents count certain permutation sets

Definition

For an $n \times n$ binary matrix $A = (a_{i,j})$, the **permanent** of A is defined as

$$\text{per}(A) = \sum_{f \in \mathcal{S}_n} \prod_{i=1}^n a_{i,f(i)}.$$

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Example

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

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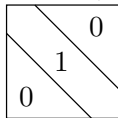
Example

$$A = \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \end{pmatrix}, f = [1, 4, 5, 3, 2] \text{ does not contribute to } \text{per}(A).$$

The size of a ball is a permanent

Let us define the following family of $n \times n$ matrices, $A_{r,n} = (a_{i,j}^{r,n})$, where

$$a_{i,j}^{r,n} = \begin{cases} 1 & |i - j| \leq r, \\ 0 & \text{otherwise.} \end{cases}$$



Theorem

$$|\mathcal{B}_{r,n}| = \text{per}(A_{r,n}).$$

Proof.

$f \in \mathcal{S}_n$ contributes to $\text{per}(A_{r,n})$ iff $|f(i) - i| \leq r$ for all i , which holds iff $d_\infty(f, \text{Id}) \leq r$. □

Off-the-shelf bounds on permanents exist

Theorem (Brégman's Theorem)

Let A be an $n \times n$ binary matrix, with r_i 1's in the i th row. Then

$$\text{per}(A) \leq \prod_{i=1}^n (r_i!)^{\frac{1}{r_i}}.$$

Theorem (Van der Waerden's Theorem)

Let A be an $n \times n$ doubly-stochastic matrix, then

$$\text{per}(A) \geq \frac{n!}{n^n}.$$

Brégman, *Soviet Math. Dokl.*, 1973

Falikman, *Math. Zametki*, & Egorychev, *Adv. in Math.*, 1981

Bound application

Brégraman's Theorem gives an immediate upper bound:

$$|\mathcal{B}_{r,n}| = \text{per}(A_{r,n}) \leq \begin{cases} ((2r+1)!)^{\frac{n-2r}{2r+1}} \prod_{i=r+1}^{2r} (i!)^{\frac{2}{i}} & 0 \leq r \leq \frac{n-1}{2}, \\ (n!)^{\frac{2r+2-n}{n}} \prod_{i=r+1}^{n-1} (i!)^{\frac{2}{i}} & \frac{n-1}{2} \leq r \leq n-1. \end{cases}$$

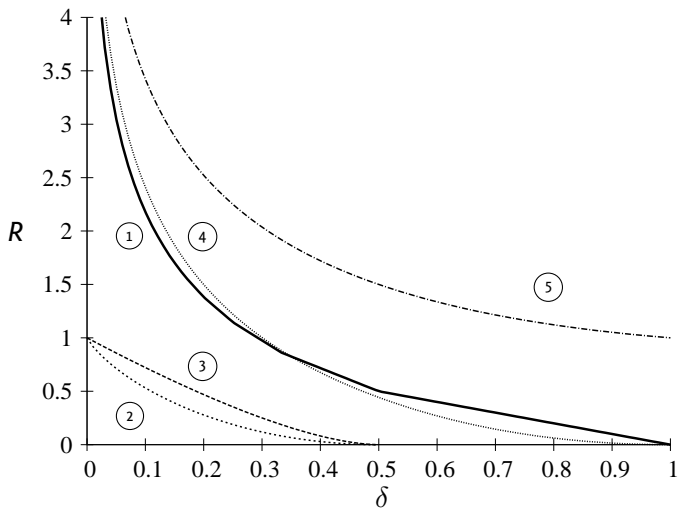
Van der Waerden's Theorem gives a lower bound:

$$|\mathcal{B}_{r,n}| = \text{per}(A_{r,n}) \geq \begin{cases} \frac{n!(2r+1)^n}{2^{2r} n^n} & 0 \leq r \leq \frac{n-1}{2}, \\ \frac{n!}{2^{2(n-r-1)}} & \frac{n-1}{2} \leq r \leq n-1. \end{cases}$$

Proof Sketch

$$\text{per} \begin{array}{|c|c|c|} \hline & & 0 \\ \hline & \diagdown & \\ \hline & 1 & \\ \hline 0 & & \\ \hline \end{array} = \frac{1}{2^{2r}} \text{per} \begin{array}{|c|c|c|} \hline & 2 & 0 \\ \hline & \diagdown & \\ \hline & 1 & \\ \hline 0 & & 2 \\ \hline \end{array} \geq \frac{1}{2^{2r}} \text{per} \begin{array}{|c|c|c|} \hline & 2 & 0 \\ \hline & \diagdown & \\ \hline & 1 & \\ \hline 0 & & 2 \\ \hline \end{array}$$

Adding the bounds to the picture



- 1 The code
- 2 Lower bound on [LTT'08] construction
- 3 Upper bound on [LTT'08] construction
- 4 GV lower bound on optimum code rate
- 5 Ball-packing upper bound on optimum code rate

A wide gap is the cause (in part)

Rewriting the bounds in asymptotic forms, we get

$$\begin{aligned} |\mathcal{B}_{\rho(n-1),n}| &\leq \begin{cases} \left(\frac{2\rho n}{e}\right)^n \cdot \left(\frac{2}{e}\right)^{2\rho n} \cdot 2^{o(n)} & 0 \leq \rho \leq \frac{1}{2} \\ \frac{n^n}{e^{n(3-2\rho)} \rho^{2\rho n}} \cdot 2^{o(n)} & \frac{1}{2} \leq \rho \leq 1 \end{cases} &:= \Phi(\rho, n) \\ |\mathcal{B}_{\rho(n-1),n}| &\geq \begin{cases} \frac{(2\rho n)^n}{2^{2\rho n} e^n} \cdot 2^{o(n)} & 0 \leq \rho \leq \frac{1}{2} \\ \frac{n^n}{e^n 2^{2n(1-\rho)}} \cdot 2^{o(n)} & \frac{1}{2} \leq \rho \leq 1 \end{cases} &:= \phi(\rho, n) \end{aligned}$$

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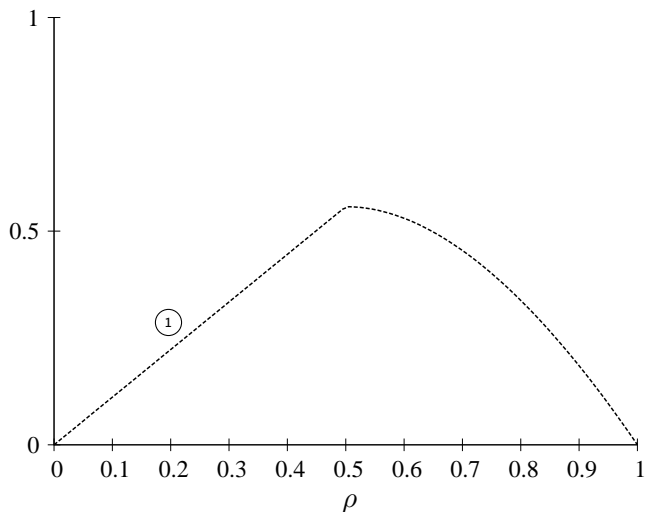
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The asymptotic gap is therefore,

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log_2 \frac{\Phi(\rho, n)}{\phi(\rho, n)} = \begin{cases} (4 - 2 \log_2 e) \rho & 0 \leq \rho \leq \frac{1}{2}, \\ 2(1 - \rho)(1 - \log_2 e) - 2\rho \log_2 \rho & \frac{1}{2} \leq \rho \leq 1. \end{cases}$$

A wide gap is the cause (in part)



1 Asymptotic gap for off-the-shelf ball-size bounds

We can have zero gap for constant radii

The strategy

- 1 Show that permutations in a ball centered at the identity form a regular language.
- 2 Use Perron-Frobenius Theory to find the asymptotic size of the regular language.

We can have zero gap for constant radii

The strategy

- 1 Show that permutations in a ball centered at the identity form a regular language.
 - 2 Use Perron-Frobenius Theory to find the asymptotic size of the regular language.
- The method is more general.
 - For ease of presentation, only $(0, 1)$ -matrices will be shown.

Lehmer, Comb. Th. and its Applications, 1970
Schwartz, Linear Algebra and its Applications, 2009

Iteratively building permutations in $\mathcal{B}_{r,n}$

Example: For $n = 9$, $r = 2$, $i = 5$, the current status is

$$\begin{array}{ccccccccc} f(1) & f(2) & f(3) & f(4) & f(5) & f(6) & f(7) & f(8) & f(9) \\ \hline 3 & 2 & 1 & 6 & ? & & & & \end{array}$$

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- At step i , $b_j = 1$ iff $i + j$ was already assigned.
- Moving from step i to step $i + 1$ we drop the left bit, and append a 0 rightmost bit.

More notation is needed

For a binary $b = b_1 b_2 \dots b_m \in \{0, 1\}^m$:

$b0$	Appending a 0 to the end of b
$\text{wt}(b)$	The number of non-zero entries in b
$L(b)$	Erasing the leftmost bit, i.e., $b_2 \dots, b_m$
e_k	The k th standard unit vector

We now build a finite state machine

Construction

We construct a digraph $\mathcal{G}_{[-r,r]} = (V, E)$ in the following way:

$$V = \left\{ b = b_{-r}b_{-r+1} \dots b_{r-1} \in \{0, 1\}^{2r} \mid \text{wt}(b) = r \right\},$$

$$E = \left\{ b \xrightarrow{k} L(b0 + e_k) \mid b, L(b0 + e_k) \in V, \quad k \in T, \quad (b0)_k = 0 \right\}.$$

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Theorem

There is a bijection between the set of permutations $\mathcal{B}_{r,n}$, and the paths of length n in $\mathcal{G}_{[-r,r]}$ starting and ending in the vertex $1^r 0^r$.

Exact asymptotic size

Theorem

For a constant r ,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \log_2 |\mathcal{B}_{r,n}| = \log_2 \lambda(\mathcal{A}(\mathcal{G}_{[-r,r]})),$$

where $\mathcal{A}(\mathcal{G}_{[-r,r]})$ is the adjacency matrix of $\mathcal{G}_{[-r,r]}$, and $\lambda(\mathcal{A}(\mathcal{G}_{[-r,r]}))$ is the spectral radius of $\mathcal{A}(\mathcal{G}_{[-r,r]})$.

Exact asymptotic size

Theorem

For a constant r ,

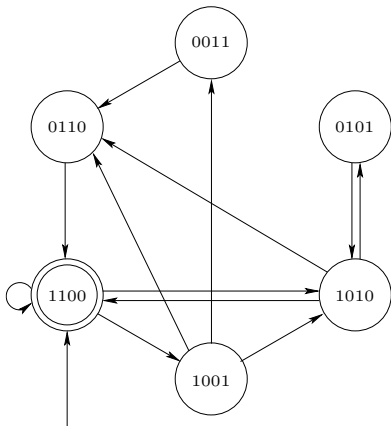
$$\lim_{n \rightarrow \infty} \frac{1}{n} \log_2 |\mathcal{B}_{r,n}| = \log_2 \lambda(\mathcal{A}(\mathcal{G}_{[-r,r]})),$$

where $\mathcal{A}(\mathcal{G}_{[-r,r]})$ is the adjacency matrix of $\mathcal{G}_{[-r,r]}$, and $\lambda(\mathcal{A}(\mathcal{G}_{[-r,r]}))$ is the spectral radius of $\mathcal{A}(\mathcal{G}_{[-r,r]})$.

Proof.

By the previous theorem, $|\mathcal{B}_{r,n}| = (\mathcal{A}(\mathcal{G}_{[-r,r]})^n)_{1,1}$. The matrix $\mathcal{A}(\mathcal{G}_{[-r,r]})$ has non-negative entries, and can be shown to be primitive. The claim follows immediately by Perron-Frobenius. \square

Taking radius 2 as an example



Taking λ to be the largest (real) root of

$$\chi_{\mathcal{A}(\mathcal{G}_{[-2,2]})}(\lambda) = \lambda^6 - \lambda^5 - 2\lambda^4 - 2\lambda^3 - 2\lambda^2 + \lambda + 1,$$

we get

$$|\mathcal{B}_{2,n}| = \lambda^{n(1+o(1))} \approx 2.34148^n.$$

What about $r = \Theta(n)$?

Theorem

Let $M = (m_{i,j})$ be a $n \times n$ matrix with non-negative entries and $\text{per}(M) > 0$. Additionally, we require that there are two diagonal matrices D and D' with positive diagonal elements such that $D \cdot M \cdot D'$ is a doubly-stochastic matrix. Let $Q = (q_{i,j})$ be an $n \times n$ doubly-stochastic matrix. Then

$$\log_2 \text{per}(M) \geq \log_2 \frac{n!}{n^n} + \sum_{i,j \in [n]} \left(-q_{i,j} \log_2 \frac{q_{i,j}}{m_{i,j}} \right).$$

Using the Sinkhorn balancing

Proof.

$$\begin{aligned}\log_2 \text{per}(M) &= \log_2 \text{per}(D \cdot M \cdot D') - \sum_{i \in [n]} \log_2(d_{i,i}) - \sum_{j \in [n]} \log_2(d'_{j,j}) \\ &\geq \log_2 \frac{n!}{n^n} - \sum_{i \in [n]} \log_2(d_{i,i}) - \sum_{j \in [n]} \log_2(d'_{j,j}) \\ &\geq \log_2 \frac{n!}{n^n} - \sum_{i \in [n]} \log_2(d_{i,i}) - \sum_{j \in [n]} \log_2(d'_{j,j}) \\ &\quad - \sum_{i,j \in [n]} q_{i,j} \log_2 \frac{q_{i,j}}{d_{i,i} \cdot m_{i,j} \cdot d'_{j,j}} \\ &= \log_2 \frac{n!}{n^n} - \sum_{i,j \in [n]} q_{i,j} \log_2 \frac{q_{i,j}}{m_{i,j}} \quad \square\end{aligned}$$

We can find the optimal matrix

Lemma

Fix $\frac{n-1}{2} \leq r \leq n-2$. The optimal $Q = (q_{i,j})$ for $A_{r,n}$ is given by

$$q_{i,j} = a_{i,j} \cdot C \cdot 2^{\lambda_i} \cdot 2^{\lambda_j}, \quad i, j \in [n],$$

$$\lambda_i = \begin{cases} ((n-r) - i) \cdot \log_2(\alpha_{r,n}) & 1 \leq i \leq n-r \\ 0 & n-r \leq i \leq r+1, \\ (i - (r+1)) \cdot \log_2(\alpha_{r,n}) & r+1 \leq i \leq n \end{cases}$$

$$C = (\alpha_{r,n} - 1) \cdot \alpha_{r,n}^{-(n-r)} = \frac{\alpha_{r,n} - 1}{(2r-n+2) - (2r-n) \cdot \alpha_{r,n}},$$

and $\alpha_{r,n}$ is the unique positive solution of the equation

$$\alpha_{r,n}^{n-r} + (2r-n) \cdot \alpha_{r,n} - (2r-n+2) = 0.$$

Finding $\alpha_{r,n}$ is more tricky

Lemma

Let $r = \rho \cdot (n-1)$, with $\frac{1}{2} < \rho < 1$ a constant. Then $\alpha_{r,n}$ is

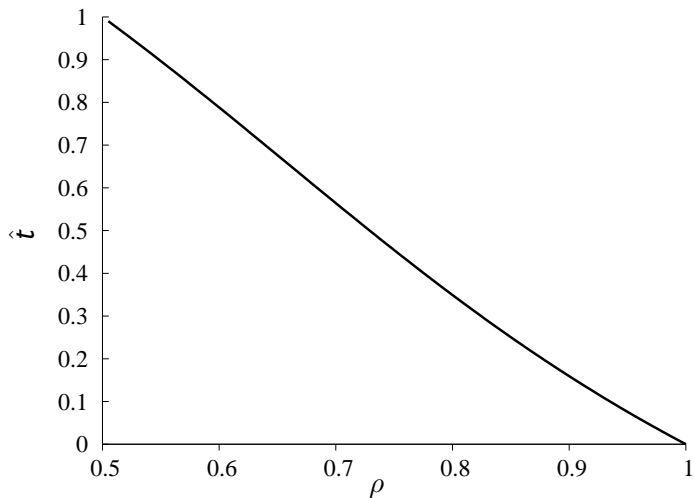
$$\alpha_{r,n} = 1 + \left(\hat{t} + \Theta \left(\frac{1}{n} \right) \right) \left(2^{\frac{1}{(n-1)(1-\rho)+1}} - 1 \right),$$

where

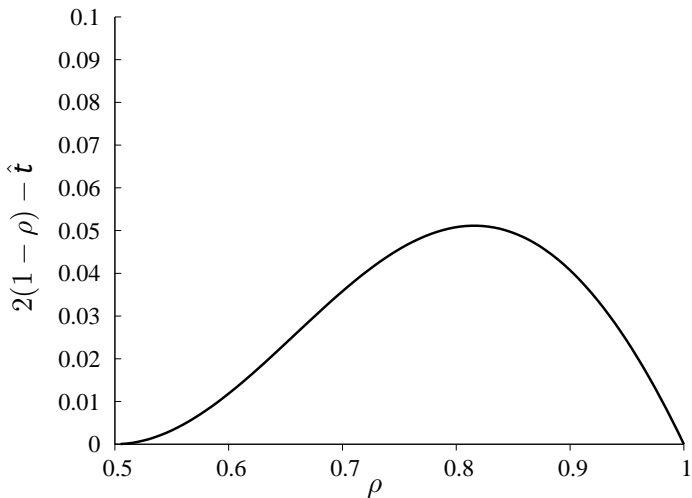
$$\hat{t} = \frac{1}{\ln(2)} \left(\frac{2(1-\rho)}{2\rho-1} - W \left(\frac{(1-\rho) \exp \left(\frac{2(1-\rho)}{2\rho-1} \right)}{2\rho-1} \right) \right),$$

and where $W(\cdot)$ denotes Lambert's function, i.e., $W(\cdot)$ is defined by $z = W(z) \exp(W(z))$.

\hat{t} is linear?



\hat{t} is linear? Almost...



And the lower bound is...

Theorem

Let $r = \rho \cdot (n-1)$, with $\frac{1}{2} < \rho < 1$ a constant, then

$$|B_{r,n}| \geq \frac{n^n \cdot 2^{\hat{t}(2\rho-1)n} \cdot (1-\rho)^n}{(e\hat{t}\ln(2))^n} \cdot 2^{o(n)}.$$

And the lower bound is...

Theorem

Let $r = \rho \cdot (n-1)$, with $\frac{1}{2} < \rho < 1$ a constant, then

$$|B_{r,n}| \geq \frac{n^n \cdot 2^{\hat{t}(2\rho-1)n} \cdot (1-\rho)^n}{(e\hat{t}\ln(2))^n} \cdot 2^{o(n)}.$$

Proof.

$$\begin{aligned} \ln |B_{r,n}| &\geq \ln(\text{per}(A_{r,n})) \geq \ln(n!/n^n) + \sum_{i,j \in [n]} \left(-q_{i,j}^* \ln \frac{q_{i,j}^*}{a_{i,j}} \right) \\ &= -n - n \ln(\alpha_{r,n} - 1) + (n-r)(2r-n+2) \ln(\alpha_{r,n}) + o(n). \end{aligned}$$

And the lower bound is...(Cont.)

Proof.

We have:

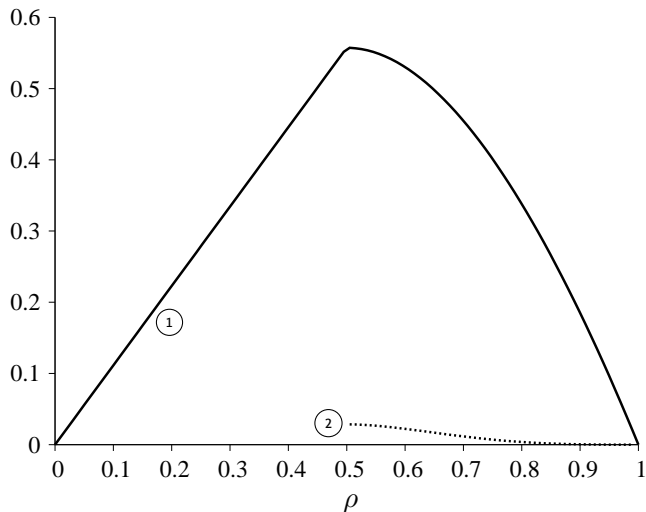
$$\begin{aligned} -n \ln(\alpha_{r,n} - 1) &= -n \ln \left((\hat{t} + \Theta(n^{-1})) \left(2^{\frac{1}{n-r}} - 1 \right) \right) \\ &= -n \ln(\hat{t} + \Theta(n^{-1})) - n \ln \left(\frac{\ln(2)}{(1-\rho)n} + O(n^{-2}) \right) \\ &= -n \ln(\hat{t}) - n \ln \ln(2) + n \ln(1-\rho) + n \ln(n) + o(n). \end{aligned}$$

Similarly,

$$(n-r)(2r-n+2) \ln(\alpha_{r,n}) = n(2\rho-1)\hat{t} \ln(2) + o(n).$$

Combining both we get the result. □

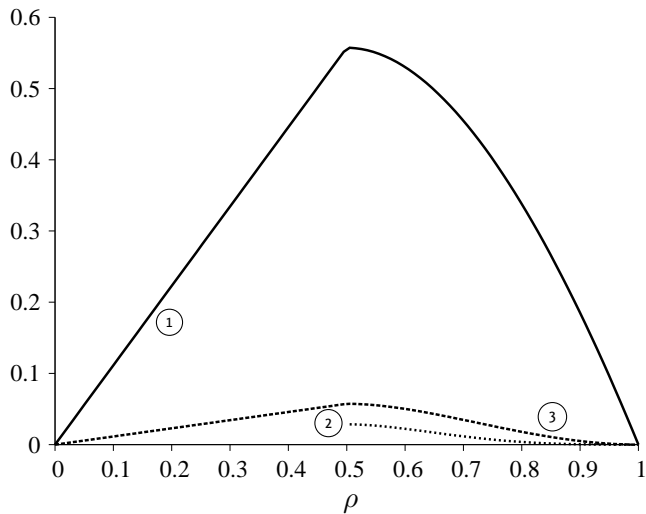
The new bounds reduces the gap



- 1 Asymptotic gap for off-the-shelf ball-size bounds
- 2 Sinkhorn-balancing bounds

Schwartz and Vontobel, to appear

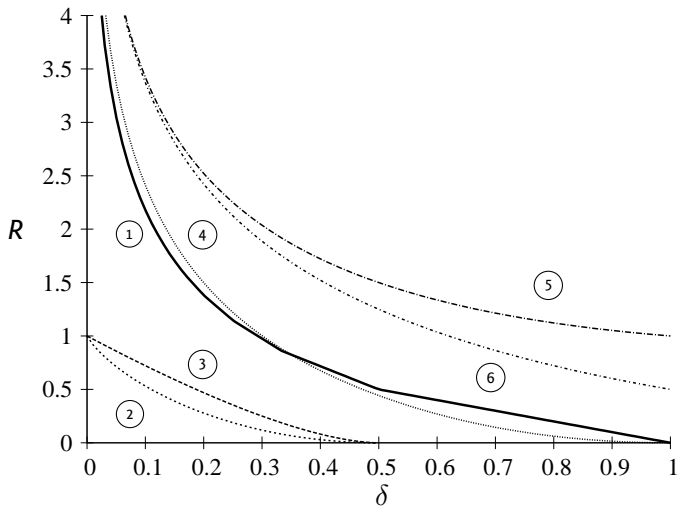
Wait, there's more...



- 1 Asymptotic gap for off-the-shelf ball-size bounds
- 2 Sinkhorn-balancing bounds
- 3 Bethe-permanent bounds

Schwartz and Vontobel, to appear

The final picture (for now)



- 1 The code
- 2 Lower bound on [LTT'08] construction
- 3 Upper bound on [LTT'08] construction
- 4 GV lower bound on optimum code rate
- 5 Ball-packing upper bound on optimum code rate
- 6 Ball-packing upper bound with improved ball-size bound