

# Counting Contingency Tables

Igor Pak, UCLA

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# Contingency tables

Fix  $\mathbf{a} = (a_1, \dots, a_m)$ ,  $\mathbf{b} = (b_1, \dots, b_n)$ ,  $a_i, b_j > 0$ , s.t.

$$\sum_{i=1}^m a_i = \sum_{j=1}^n b_j = N.$$

A *contingency table* with *margins*  $(\mathbf{a}, \mathbf{b})$  is an  $m \times n$  matrix  $X = (x_{ij})$ , s.t.

$$\sum_{j=1}^n x_{ij} = a_i, \quad \sum_{i=1}^m x_{ij} = b_j, \quad x_{ij} \geq 0 \quad \forall i, j.$$

We denote by  $\mathcal{T}(\mathbf{a}, \mathbf{b})$  the set of all such matrices, and  $T(\mathbf{a}, \mathbf{b}) := |\mathcal{T}(\mathbf{a}, \mathbf{b})|$ .

**Main problem:** Compute  $T(\mathbf{a}, \mathbf{b})$ .

**That means:** *formula, algorithm, asymptotics, bounds, etc.*

**More precisely:** Do your best!

## Examples:

$$\mathbf{a} = \mathbf{b} = (1, 1, 1) \longrightarrow T(\mathbf{a}, \mathbf{b}) = 6$$

$$\mathbf{a} = \mathbf{b} = (100, 100, 100) \longrightarrow T(\mathbf{a}, \mathbf{b}) = 13268976 \approx 1.3 \times 10^7$$

$$m = n = 10, \mathbf{a} = \mathbf{b} = (20, \dots, 20) \longrightarrow T(\mathbf{a}, \mathbf{b}) \approx 1.1 \times 10^{59} \text{ [Canfield–McKay, 2010]}$$

$$m = n = 30, \mathbf{a} = \mathbf{b} = (3, \dots, 3) \longrightarrow T(\mathbf{a}, \mathbf{b}) \approx 2.2 \times 10^{92}$$

$$m = n = 9, \mathbf{a} = \mathbf{b} = (10^5, \dots, 10^5) \longrightarrow T(\mathbf{a}, \mathbf{b}) \approx 6.1 \times 10^{279} \text{ [Beck–Pixton, 2003]}$$

$$m = n = 9, \mathbf{a} = (220, 215, 93, 64), \mathbf{b} = (108, 286, 71, 127) \longrightarrow T(\mathbf{a}, \mathbf{b}) = 1225914276768514 \approx 1.2 \times 10^{15} \text{ [Des Jardins, 1994]}$$

$$\mathbf{a} = (13070380, 18156451, 13365203, 20567424), \mathbf{b} = (12268303, 20733257, 17743591, 14414307) \longrightarrow T(\mathbf{a}, \mathbf{b}) \approx 4.3 \times 10^{61} \text{ [De Loera, 2009]}$$

$$m = n = 15, \mathbf{a} = \mathbf{b} = (10^5, \dots, 10^5) \longrightarrow T(\mathbf{a}, \mathbf{b}) \approx 1.7 \times 10^{819} \text{ [good estimate]}$$

$$m = n = 100, \mathbf{a} = \mathbf{b} = (10^3, \dots, 10^3) \longrightarrow T(\mathbf{a}, \mathbf{b}) \approx 6.3 \times 10^{33470} \text{ [good estimate]}$$

$$m = n = 100, \text{ nonuniform margins average } 10 \longrightarrow ??? \text{ [can be done via SHM in under 200h CPU time]}$$

$$m = n = 1000, \text{ nonuniform margins average } 100 \longrightarrow ??? \text{ [currently cannot be done in our lifetime]}$$

## More Examples:

**Permutations:**  $m = n$ ,  $\mathbf{a} = \mathbf{b} = (1, \dots, 1) \longrightarrow T(\mathbf{a}, \mathbf{b}) = n!$

**Magic squares:**  $m = n$ ,  $\mathbf{a} = \mathbf{b} = (k, \dots, k)$  [when  $k$  fixed,  $T(\mathbf{a}, \mathbf{b})$  is P-recursive]

$k = 2 \longrightarrow T(\mathbf{a}, \mathbf{b}) = c(n)$ , where  $c(n) = n^2 c(n-1) - \frac{1}{2} n(n-1)^2 c(n-2)$ , so

$$c(n) \sim \frac{\sqrt{e} (n!)^2}{\sqrt{\pi n}}$$

$k = 3 \longrightarrow T(\mathbf{a}, \mathbf{b}) = n! v(n)$ , where

$$\begin{aligned} 576n \cdot v(n) &= (2880n^2 - 5760n + 3456) v(n-1) + (324n^5 - 3564n^4 + 14148n^3 - 26028n^2 + 21312n - 6192) v(n-2) \\ &+ (81n^6 - 1377n^5 + 7209n^4 - 13203n^3 - 3402n^2 + 32076n - 21384) v(n-3) \\ &+ (-81n^7 + 1944n^6 - 20232n^5 + 115578n^4 - 383283n^3 + 724230n^2 - 708372n + 270216) v(n-4) \\ &+ (-72n^6 + 1440n^5 - 10890n^4 + 40500n^3 - 78678n^2 + 75780n - 28080) v(n-5) \\ &+ (81n^9 - 3321n^8 + 59004n^7 - 594054n^6 + 3718687n^5 - 14927199n^4 + 38152096n^3 - 59311746n^2 + 50236612n - 17330160) v(n-6) \\ &+ (72n^8 - 2520n^7 + 37347n^6 - 304479n^5 + 1484133n^4 - 4394565n^3 + 7642248n^2 - 7039116n + 2576880) v(n-7) \\ &+ (-198n^9 + 8712n^8 - 165175n^7 + 1764196n^6 - 11643772n^5 + 48965728n^4 - 130257475n^3 + 209370724n^2 - 182126340n + 64083600) v(n-8) \\ &+ (36n^{10} - 1944n^9 + 45884n^8 - 621504n^7 + 5330892n^6 - 30123576n^5 + 112954596n^4 - 275612976n^3 + 415021552n^2 - 343920960n + 116928000) v(n-9) \\ &+ (-9n^{11} + 585n^{10} - 16800n^9 + 280800n^8 - 3027357n^7 + 22034565n^6 - 110039130n^5 + 375129450n^4 - 849926784n^3 + 1208298600n^2 - 958439520n + 315705600) v(n-10) \\ &+ (-7n^{10} + 385n^9 - 9240n^8 + 127050n^7 - 1104411n^6 + 6314385n^5 - 23918510n^4 + 58866500n^3 - 89275032n^2 + 74400480n - 25401600) v(n-11) \\ &+ (n^{11} - 66n^{10} + 1925n^9 - 32670n^8 + 357423n^7 - 2637558n^6 + 13339535n^5 - 45995730n^4 + 105258076n^3 - 150917976n^2 + 120543840n - 39916800) v(n-12), \end{aligned}$$

so

$$v(n) \sim e^2 \sqrt{\frac{3\pi n}{2}} \left( \frac{3n^3}{4e^3} \right)^n$$

# Complexity aspects: bad news all around

**Theorem** [Narayanan, 2006]

Computing  $T(\mathbf{a}, \mathbf{b})$  is  $\#\mathbf{P}$ -complete.

**Theorem** [P.–Panova, 2020+, former *folklore conjecture*]

Computing  $T(\mathbf{a}, \mathbf{b})$  is *strongly*  $\#\mathbf{P}$ -complete (i.e. for the input  $a_i, b_j$  in unary).

**Corollary** [P.–Panova, 2020+] Computing:

- *Kostka numbers*  $K_{\lambda\mu}$  and *Littlewood–Richardson coefficients*  $c_{\mu\nu}^\lambda$  is *strongly*  $\#\mathbf{P}$ -complete
- *Schubert coefficients* is  $\#\mathbf{P}$ -complete
- *Kronecker coefficients*  $g(\lambda, \mu, \nu)$  and *reduced Kronecker coefficients*  $\bar{g}(\lambda, \mu, \nu)$  is  $\#\mathbf{P}$ -hard

**Note:** The last part is known [Ikenmeyer–Mulmuley–Walter, 2017] and [P.–Panova, 2020], resp.

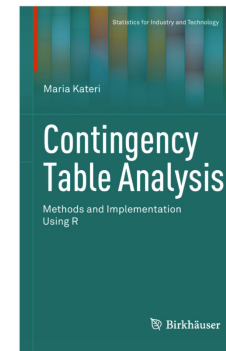
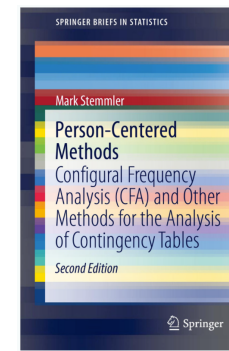
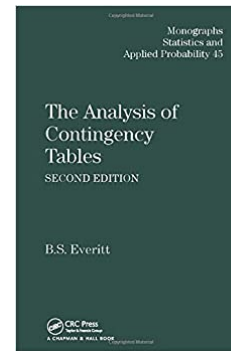
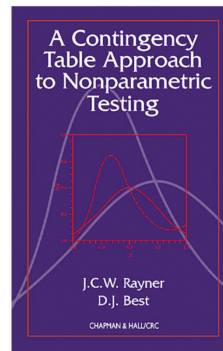
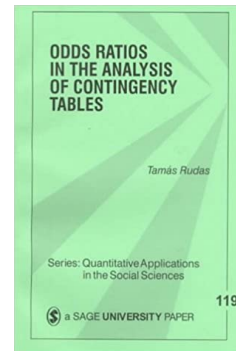
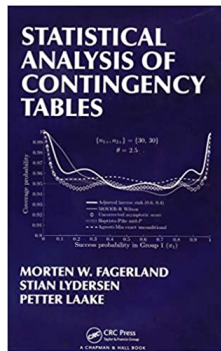
**Moral:** Asymptotic formulas and approximate counting is the best one can hope for.

# Connections and Applications

- **Random networks:** contingency tables  $\leftrightarrow$  bipartite graphs with fixed degrees

**Note:** graphs with fixed degrees  $\leftrightarrow$  symmetric binary (0-1) CTs with 0 diagonal,  
 numerous papers on all aspects of these, see e.g. [Wormald, 2018 ICM survey]

- **Statistics**



**Key observation:** Random sampling  $\longleftrightarrow$  approximate counting

*Self-reduction:*

$$\mathbb{P}(x_{11} \geq t) = \frac{T(a_1 - t, a_2, \dots; b_1 - t, b_2, \dots)}{T(a_1, a_2, \dots; b_1, b_2, \dots)}$$

# Descendants of Queen Victoria (1819 – 1901)



Month of birth	Month of death												Total
	Jan	Feb	March	April	May	June	July	Aug	Sept	Oct	Nov	Dec	
Jan	1	0	0	0	1	2	0	0	1	0	1	0	6
Feb	1	0	0	1	0	0	0	0	0	1	0	2	5
March	1	0	0	0	2	1	0	0	0	0	0	1	5
April	3	0	2	0	0	0	1	0	1	3	1	1	12
May	2	1	1	1	1	1	1	1	1	1	1	0	12
June	2	0	0	0	1	0	0	0	0	0	0	0	3
July	2	0	2	1	0	0	0	0	1	1	1	2	10
Aug	0	0	0	3	0	0	1	0	0	1	0	2	7
Sept	0	0	0	1	1	0	0	0	0	0	1	0	3
Oct	1	1	0	2	0	0	1	0	0	1	1	0	7
Nov	0	1	1	1	2	0	0	2	0	1	1	0	9
Dec	0	1	1	0	0	0	1	0	0	0	0	0	3
Total	13	4	7	10	8	4	5	3	4	9	7	8	82

FIGURE 1. Month of birth and death for descendants of Queen Victoria (as of 1990).

**Question:** Is there a dependence between **Birthday** and **Deathday** of the 82 (dead) descendants?

**Testing correlation for**  $X = (x_{ij})$  (after Diaconis–Efron, 1985):

- Sample large number  $N$  of random samples, compute their  $\chi^2$ ,
- Output fraction  $a/N$ , where  $a =$  number of samples with  $\chi^2 \leq \chi(X)$ .

# Birthday–Deathday example analysis:

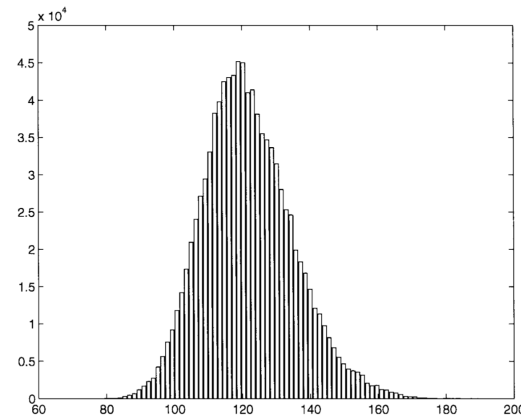


FIG. 2. Histogram of the chi-square statistic for Table 1.

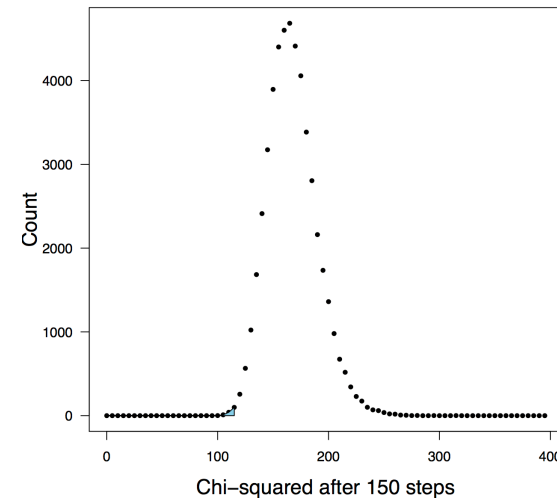


FIGURE 1. Plot of  $\chi^2$  from [Diaconis–Sturmfels] and [Dittmer–Pak]

**Setup:**  $\chi^2(X) \approx 115.56$ , so p-value = % of tables have  $\chi^2 \leq 115.56$

**Hypothesis:** There is NO dependence between Birthday and Deathday.

[Diaconis–Sturmfels, 1998]: From the  $10^6$  trials of *Diaconis–Gangolli MC*, they get  $p \approx 37.75\%$   $\longrightarrow$  Accept!

[Dittmer–P., 2019+]: From the  $5 \times 10^4$  trials using our new *SHM MC*, we get  $p \approx 0.10\%$   $\longrightarrow$  Reject!

**First Moral:** It's important to get good uniform samples from  $\mathcal{T}(\mathbf{a}, \mathbf{b})$ . Otherwise, you *might* actually start to believe that there is NO dependence.

**Second Moral:** Dependence, really??? Ah, well, the model was faulty...



# Exact and approximate counting results

Below:  $m \leq n$ ,  $a_1 \geq \dots \geq a_m$ ,  $b_1 \geq \dots \geq b_n$ .

- Exact counting in poly-time for  $m, n = O(1)$  [Barvinok'93]
- Exact counting in poly-time for  $a_1, b_1 = O(1)$  via dynamic programming.
- Quasi-poly time approx counting for  $a_1/a_m, b_1/b_n < 1.6$  and  $m = \Theta(n)$  [Barvinok et al, 2010].
- Poly-time approx counting for  $m = O(1)$  [Cryan, Dyer 2003]
- Poly-time approx counting for  $a_m = \Omega(n^{3/2}m \log m)$  and  $b_n = \Omega(m^{3/2}n \log n)$  [Dyer–Kannan–Mount, 1997], [Morris, 2002]
- Poly-time approx counting for  $a_1, b_1 = \Omega(n^{1/4-\varepsilon})$ ,  $\varepsilon > 0$  and  $m = \Theta(n)$  [Dittmer–P., 2019+]
- Poly-time approx counting for all  $a_i, b_j = \Theta(n^{1-\varepsilon})$ ,  $\varepsilon > 0$  and  $m = \Theta(n)$  [Dittmer–P., 2019+]

**Note:** These four are all MCMC based FPFAS.

## Diaconis–Gangolli Markov chain (1995)

STEP: choose a random  $2 \times 2$  submatrix, and make either of the following changes:

$$\begin{array}{cc} +1 & -1 \\ -1 & +1 \end{array} \quad \text{or} \quad \begin{array}{cc} -1 & +1 \\ +1 & -1 \end{array}$$

(stay put if this is impossible). **Note:** Use *hit-and-run* for large  $a_1, b_1$ .

**Note:** Early theoretical results in [Diaconis – Saloff-Coste, 1995], [Chung–Graham–Yau, 1996]

## Split–Hyper–Merge (SHM) Markov chain [Dittmer–P., 2019+]

**Idea:** Use *Burnside processes* [Jerrum, 1993]  $\leftarrow$  probabilistic version of the *Burnside Lemma*.

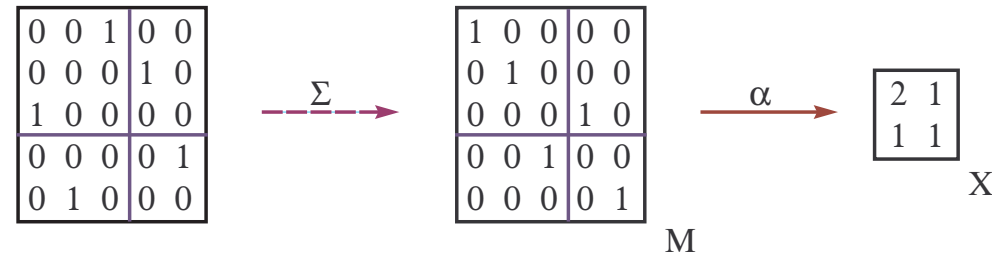
**Lemma:**  $\mathcal{T}(\mathbf{a}, \mathbf{b})$  is in bijection with the set of orbits of group

$$\Sigma := \text{Sym}(a_1) \times \dots \times \text{Sym}(a_m) \times \text{Sym}(b_1) \times \dots \times \text{Sym}(b_n)$$

acting on  $S_N = N \times N$  permutation matrices.

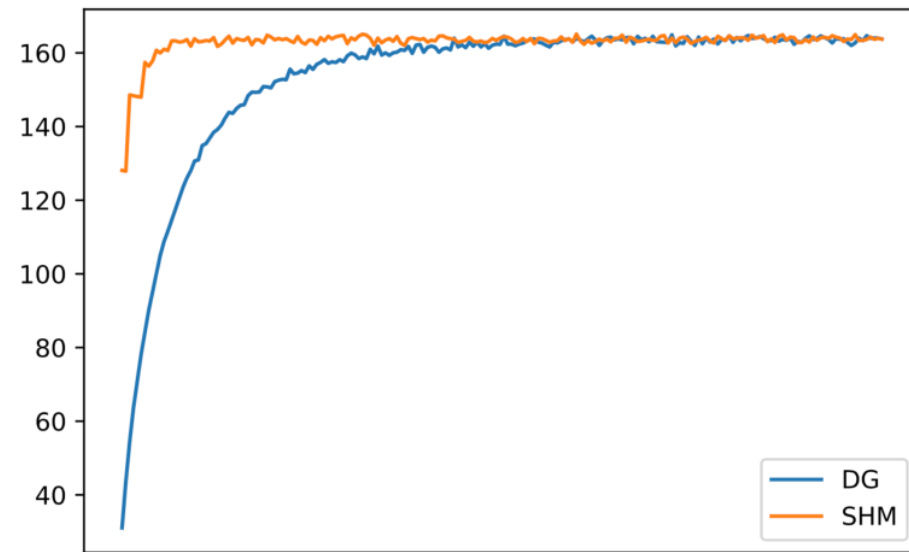
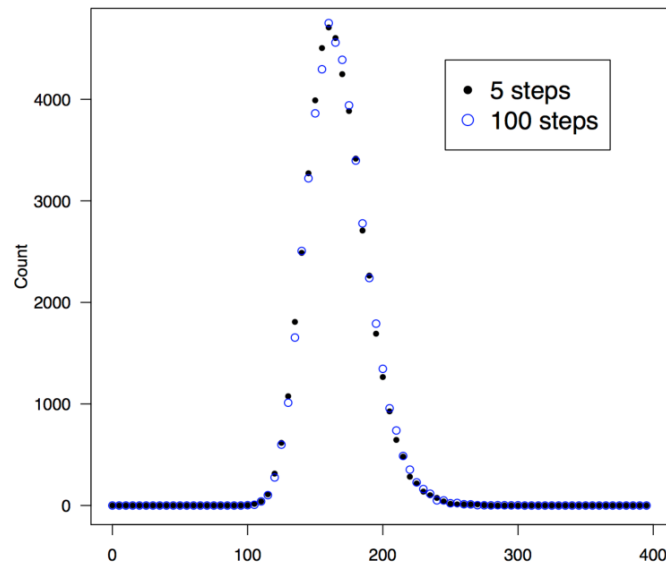
**Conjecture:** For  $a_1 b_1 \leq \text{poly}(mn)$ , both DG and SHM Markov chains mix in polynomial time.

# Why contingency tables are orbits:



Here  $X \in \mathcal{T}(3, 2; 3, 2)$  corresponds to orbit representative  $M \in S_5$  under the action of  $\Sigma = S_3 \times S_2 \times S_3 \times S_2$ .

# Testing SHM chain on the Birthday–Deathday example (plot of $\chi^2$ )



**Independence heuristic** [Good, 1950]:  $T(\mathbf{a}, \mathbf{b}) \approx G(\mathbf{a}, \mathbf{b})$ , where

$$G(\mathbf{a}, \mathbf{b}) := \binom{N + mn - 1}{mn - 1}^{-1} \prod_{i=1}^m \binom{a_i + n - 1}{n - 1} \prod_{j=1}^n \binom{b_j + m - 1}{m - 1}.$$

**Good's reasoning** [Good, 1976]: Let  $\mathcal{S}(N, m, n)$  be the set of  $m \times n$  tables with total sum  $N$ , so

$$|\mathcal{S}(N, m, n)| = \binom{N + mn - 1}{mn - 1}$$

Observe:

$$\begin{aligned} \mathbb{P}(X \text{ has row sums } \mathbf{a}) &= \frac{1}{|\mathcal{S}(N, m, n)|} \prod_{i=1}^m \binom{a_i + n - 1}{n - 1}, \\ \mathbb{P}(X \text{ has column sums } \mathbf{b}) &= \frac{1}{|\mathcal{S}(N, m, n)|} \prod_{j=1}^n \binom{b_j + m - 1}{m - 1}. \end{aligned}$$

If these events are asymptotically independent:

$$\begin{aligned} \frac{T(\mathbf{a}, \mathbf{b})}{|\mathcal{S}(N, m, n)|} &= \mathbb{P}(X \text{ has row sums } \mathbf{a}, \text{ column sums } \mathbf{b}) \\ &\approx \frac{1}{|\mathcal{S}(N, m, n)|} \prod_{i=1}^m \binom{a_i + n - 1}{n - 1} \times \frac{1}{|\mathcal{S}(N, m, n)|} \prod_{j=1}^n \binom{b_j + m - 1}{m - 1}. \end{aligned}$$

“the conjecture appears to be confirmed” [...] “leaving aside finer points of rigor”.  $\square$

# Does the independence heuristic work?

For the Birthday–Deathday example with  $N = 592$ :  $T(\mathbf{a}, \mathbf{b}) = 1.226 \times 10^{15}$  vs.  $G(\mathbf{a}, \mathbf{b}) = 1.211 \times 10^{15}$

For the large  $4 \times 4$  case with  $N = 65159458$  [De Loera]:  $T(\mathbf{a}, \mathbf{b}) = 4.3 \times 10^{61}$  vs.  $G(\mathbf{a}, \mathbf{b}) = 3.7 \times 10^{61}$

**Theorem** [Canfield–McKay, 2010] For  $m = n$ ,  $\mathbf{a} = \mathbf{b} = (k, \dots, k)$ ,  $k = \omega(1)$ ,  $k = O(\log n)$ :

$$T(\mathbf{a}, \mathbf{b}) \sim \sqrt{e} \cdot G(\mathbf{a}, \mathbf{b}) \quad \text{as } n \rightarrow \infty.$$

**Theorem** [Greenhill–McKay, 2008] For  $m = n$ ,  $a_1 b_1 = o(N^{2/3})$ :

$$T(\mathbf{a}, \mathbf{b}) \sim \sqrt{e} \cdot G(\mathbf{a}, \mathbf{b}) \quad \text{as } n \rightarrow \infty.$$

**Theorem** [Barvinok, 2009] For  $m = n$ ,  $\mathbf{a} = \mathbf{b} = (Bn, \dots, Bn, n, \dots, n)$ , with  $\theta n$  sums  $Bn$

$$\lim_{n \rightarrow \infty} \frac{1}{n^2} \log T(\mathbf{a}, \mathbf{b}) > \lim_{n \rightarrow \infty} \frac{1}{n^2} \log G(\mathbf{a}, \mathbf{b}) \quad \text{for all } B > 1.$$

# Two valued margins: second order phase transition

**Theorem** [Lyu-P., 2020+]

Let  $m = n$ ,  $\mathbf{a} = \mathbf{b} = (Bn, \dots, Bn, n, \dots, n)$ , with  $n^\delta$  sums  $Bn$ ,  $0 < \delta < 1$  fixed.

Let  $B_c = 1 + \sqrt{2}$ . Then:

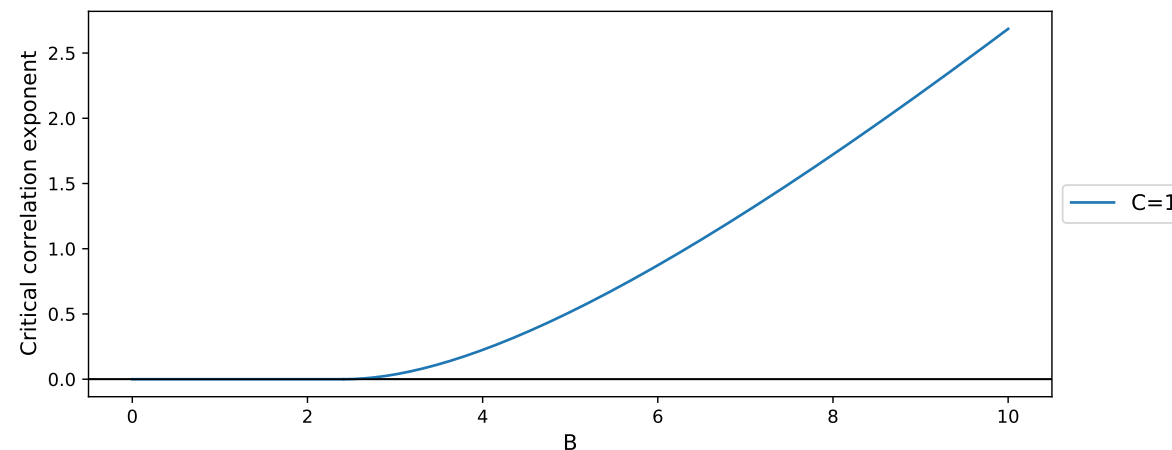
$$\lim_{n \rightarrow \infty} \frac{1}{n^2} \log T(\mathbf{a}, \mathbf{b}) = \lim_{n \rightarrow \infty} \frac{1}{n^2} \log G(\mathbf{a}, \mathbf{b}) = 2 \log 2.$$

On the other hand:

$$\lim_{n \rightarrow \infty} \frac{1}{n^{1+\delta}} \log \frac{T(\mathbf{a}, \mathbf{b})}{G(\mathbf{a}, \mathbf{b})} = \begin{cases} 0 & \text{for } 1 \leq B < B_c \\ (B - B_c) \log B_c - 2f(B) + 2f(B_c) & \text{for } B > B_c \end{cases}$$

where  $f(x) := (x + 1) \log(x + 1) - x \log x$ .

The proof is based on [Barvinok, 2009] and [Dittmer-Lyu-P., 2020].



# Combinatorial optimization approach

**Theorem** [Barvinok'09, Barvinok–Hartigan'12]

$$N^{-\gamma(m+n)} g(\mathbf{a}, \mathbf{b}) \lesssim T(\mathbf{a}, \mathbf{b}) \leq g(\mathbf{a}, \mathbf{b}),$$

for some  $\gamma > 0$ , where

$$g(\mathbf{a}, \mathbf{b}) := \inf_{\substack{x_i \in (0,1) \\ 1 \leq i \leq m}} \inf_{\substack{y_j \in (0,1) \\ 1 \leq j \leq n}} \left[ \prod_{i=1}^m x_i^{a_i} \prod_{j=1}^n y_j^{b_j} \right]^{-1} \prod_{i=1}^m \prod_{j=1}^n \frac{1}{1 - x_i y_j}.$$

The lower bound is hard, but made explicit. The upper bound is immediate from the GF:

$$\prod_{i=1}^m \prod_{j=1}^n \frac{1}{1 - x_i y_j} = \sum_{\mathbf{a} \in \mathbb{N}^m, \mathbf{b} \in \mathbb{N}^n} T(\mathbf{a}, \mathbf{b}) \prod_{i=1}^m x_i^{a_i} \prod_{j=1}^n y_j^{b_j}$$

**Theorem** [Brändén–Leake–P., 2020+] For all margins  $(\mathbf{a}, \mathbf{b})$  we have:

$$\left[ \frac{1}{e^{m+n-1}} \prod_{i=2}^m \frac{1}{a_i + 1} \prod_{j=1}^n \frac{1}{b_j + 1} \right] g(\mathbf{a}, \mathbf{b}) \leq T(\mathbf{a}, \mathbf{b}) \leq g(\mathbf{a}, \mathbf{b}),$$

The proof involves the technology of (denormalized) Lorentzian polynomials [Brändén–Huh, 2019], and the approach in [Gurvits '08, '09, '15].

## Applications of the New LB

For the Birthday–Deathday example with  $N = 592$ :  $T(\mathbf{a}, \mathbf{b}) = 1.2 \times 10^{15}$ , New LB =  $9.5 \times 10^{12}$ , Old LB =  $4.6 \times 10^8$

For the large  $4 \times 4$  case with  $N = 65159458$ :  $T(\mathbf{a}, \mathbf{b}) = 4.3 \times 10^{61}$ , New LB =  $5.8 \times 10^{58}$ , Old LB  $\leftarrow$  hard to compute.

## Volumes of transportation polytopes:

Observe that  $\mathcal{T}(\mathbf{a}, \mathbf{b})$  are integer points in  $Q(\mathbf{a}, \mathbf{b}) := \mathcal{T}_{\mathbb{R}}(\mathbf{a}, \mathbf{b}) \subset \mathbb{R}_+^{mn}$ . Then:

$$\text{vol}Q(\mathbf{a}, \mathbf{b}) = \sqrt{m^{n-1}n^{m-1}} \cdot \lim_{M \rightarrow \infty} \frac{T(M\mathbf{a}, M\mathbf{b})}{M^{(m-1)(n-1)}}$$

[Canfield–McKay, 2009]  $\longrightarrow$  asymptotics for the volume of the *Birkhoff polytope*  $Q(\mathbf{1}, \mathbf{1})$ .

[Brändén–Leake–P., 2020+]  $\longrightarrow$  new lower bounds for the volume of general *transportation polytopes*  $Q(\mathbf{a}, \mathbf{b})$ .

**Note:** Barvinok and [BLP] results generalize to all subsets of zeros in  $[m \times n]$ . These give lower bounds for all *bipartite flow polytopes*. Using [Baldoni et al., 2004] these give lower bounds for all *flow polytopes*.



**Thank you!**

