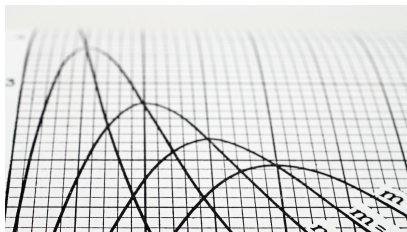


Moments of permutation statistics and central limit theorems



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(joint work with Niraj Khare)

Open University Discrete Mathematics Seminar Series
November 17, 2021

Consider the set, Π_n , of all partitions of $[n] := \{1, 2, \dots, n\}$.

Their number is B_n - the n -th Bell number. For example, $B_3 = 5$:

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Let $cr_2(\lambda) :=$ the number of 2-crossings in $\lambda \in \Pi_n$,

i.e., numbers $i_1 < i_2 < j_1 < j_2$, such that i_1, j_1 and i_2, j_2 are in two different blocks. For instance, $cr_2(\{\{1, 3\}, \{2, 4, 5\}\}) = 2$.

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Can we find $M(cr_2; n)$?

$M(cr_2; n) = \frac{1}{4}(-5B_{n+2} + (2n + 9)B_{n+1} + (2n + 1)B_n)$ (Kasraoui, 2013 [6]).

Chern, Diaconis, Kane and Rhoades [4] found that

$$M(X_1^2; n) = nB_{n-1} + (n^2 - n)B_{n-2}.$$

and

$$M(cr_2^2; n) = \frac{1}{144} (225B_{n+4} - (180n + 814)B_{n+3} + (36n^2 + 156n + 489)B_{n+2} + (72n^2 + 72n - 260)B_{n+1} + (36n^2 + 24n - 23)B_n).$$

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Theorem 1 (CDKR)

For a family of set partition statistics, the moments can be written as linear combinations of shifted Bell numbers, where the coefficients are polynomials in n .

Khare, Lorentz and Yan [7] developed the same approach on the set of perfect matchings (set partitions with blocks of size 2) on $[2m]$.

Theorem 2 (KLY)

For a family of statistics on perfect matchings, the moments can be written as linear combinations of double factorials with constant coefficients.

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For example, they found:

$$\sum_{M \in \mathcal{M}_{2m}} cr_2^2(M) = \binom{2m}{4} T_{2m-4} + 12 \binom{2m}{6} T_{2m-6} + 70 \binom{2m}{8} T_{2m-8},$$

where $T_{2m} = |\mathcal{M}_{2m}| = (2m-1)(2m-3) \cdots 3 \cdot 1 = (2m-1)!!$.

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Goal: Develop the same approach for permutations!

- *permutation* - ordering of the numbers in $[n]$.
- S_n - the set of permutations of size n .
Example: $4172365 \in S_7$.
- Let $A(\pi) := \{(u, v) \mid u = \pi_i, v = \pi_j, i < j\}$ be the arc set of π .
Example: $A(312) = \{(3, 1), (3, 2), (1, 2)\}$.
- $\text{red}(s_1 s_2 \cdots s_k) := p_1 \cdots p_k \in S_k$, where $p_i < p_j$ iff $s_i < s_j$.
Example: $\text{red}(6253) = 4132$.

Definition

- (i) A *permutation pattern* \underline{P} of size k is a tuple $\underline{P} = (P, \mathbf{C}(\underline{P}), \mathbf{D}(\underline{P}))$, where $P = p_1 \cdots p_k \in S_k$ and $\mathbf{C}(\underline{P}) \subseteq [k-1]$, $\mathbf{D}(\underline{P}) \subseteq [k-1]$.
- (ii) An *occurrence* of the pattern $\underline{P} = (p_1 p_2 \cdots p_k, \mathbf{C}(\underline{P}), \mathbf{D}(\underline{P}))$ of size k in $\sigma \in S_n$ is a tuple $t = (t_1, t_2, \dots, t_k)$ with $t_i \in [n]$, such that:
 - a) $t_1 < t_2 < \cdots < t_k$.
 - b) $(t_i, t_j) \in A(\sigma)$, if and only if $(i, j) \in A(P)$.
 - c) if $i \in \mathbf{C}(\underline{P})$, then the positions of t_{p_i} and $t_{p_{i+1}}$ in σ are consecutive.
 - d) if $i \in \mathbf{D}(\underline{P})$, then $t_{i+1} = t_i + 1$.

Write $t \in_{\underline{P}} \sigma$, if t is an occurrence of \underline{P} in σ .

Examples:

1. $\underline{P} = 132 = (132, \emptyset, \emptyset)$ [*classical patterns*].
 $t = (3, 4, 5) \in_{\underline{P}} 31524$, since $\text{red}(354) = 132$.

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Examples:

2. $\underline{P} = \underline{3214} = (3214, \{1\}, \emptyset)$ [*vincular patterns*].

$t = (2, 3, 5, 7) \in_{\underline{P}} 4536217$, since $\text{red}(5327) = 3214$
and the positions of $t_3 = 5$ and $t_2 = 3$ are consecutive.

Definition

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Examples:

$$3. \underline{P} = \frac{12\overline{34}}{\underline{43}12} = (4312, \{2\}, \{3\}) \text{ [bivincular patterns].}$$

$t = (1, 3, 5, 6) \in_{\underline{P}} 625143$, since $\text{red}(6513) = 4312$,
the positions of $t_3 = 5$ and $t_1 = 1$ are consecutive, and $t_4 = 6 = t_3 + 1$.

simple statistic: a pattern \underline{P} of size k and a valuation function $Q(t, w) = Q_1(t)Q_2(w)$, where $Q_1, Q_2 \in \mathbb{Z}[y_1, \dots, y_k, m]$.

$$f(\sigma) = f_{\underline{P}, Q}(\sigma) := \sum_{t \in \underline{P}\sigma} Q(t, \sigma^{-1}(t)) = \sum_{t \in \underline{P}\sigma} Q_1(t)Q_2(\sigma^{-1}(t)).$$

f is of degree $d(f) := 2k + \deg(Q)$.

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Examples (simple statistic):

1. $\text{cnt}_{\underline{P}}(\sigma) := f_{\underline{P}, 1}(\sigma) = \sum_{t \in \underline{P}\sigma} 1$, for any pattern \underline{P} ,
e.g., 21, 1324, 123, $\overline{123}$, $\underline{312}$.

simple statistic: a pattern \underline{P} of size k and a valuation function

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e.g., $21, 1324, \underline{123}, \overset{123}{312}$.

2. $\text{drops}(\sigma) := \sum_{\sigma_i > \sigma_{i+1}} \sigma_i - \sigma_{i+1} = \sum_{(t_1, t_2) \in \underline{21}\sigma} t_2 - t_1$.

$\underline{P} = \underline{21}$, $Q(t, w) = Q_1(t)Q_2(w)$, where

$Q_1(t) = Q_1(t_1, t_2) = t_2 - t_1$ and $Q_2(w) = 1$.

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Example (statistic):

$$\text{peakSqSum}(\sigma) := \sum_{\sigma(i-1) < \sigma(i) > \sigma(i+1)} \sigma(i)^2 = \sum_{(t_1, t_2, t_3) \in \underline{132}\sigma} t_3^2 + \sum_{(t_1, t_2, t_3) \in \underline{231}\sigma} t_3^2.$$

This is a sum of the simple statistics $f_{\underline{132}, t_3^2}$ and $f_{\underline{231}, t_3^2}$.

Theorem 3

Let $f_{\underline{P}, \underline{Q}}$ be a simple statistic of degree m , where $|P| = k$, $|\mathbf{C}(\underline{P})| = c$ and $|\mathbf{D}(\underline{P})| = d$. Then

$$M(f_{\underline{P}, \underline{Q}}, n) = R(n)(n - k)!,$$

where $R(x)$ is a polynomial of degree no more than $m - c - d$.
Equivalently for $n \geq k$,

$$M(f_{\underline{P}, \underline{Q}}, n) = \begin{cases} 0 & n < k \\ \sum_{i=0}^{m-c-d} c_i (n - k + i)! & n \geq k \end{cases},$$

for some constants $c_i \in \mathbb{Q}$.

Simple statistics:

1. cnt_{1324} .

$$M(\text{cnt}_{1324}, n) = \frac{1}{24}n! - \frac{1}{6}(n+1)! + \frac{1}{8}(n+2)! - \frac{1}{36}(n+3)! + \frac{1}{576}(n+4)!.$$

In fact, $M(\text{cnt}_P, n) = \frac{1}{k!} \binom{n}{k} n!$ for any classical pattern P of size k .

Express rising factorials in terms of falling factorials to get

$$M(\text{cnt}_P, n) = \frac{1}{k!} \binom{n}{k} n! = \frac{(-1)^k}{k!} n! + \sum_{j=1}^{k-1} \frac{(-1)^{k-j}}{(j!)^2 (k-j)!} (n+j)! + \frac{1}{(k!)^2} (n+k)!.$$

2. *Descent drop*.

$$M(\text{drops}, n) = -\frac{1}{2}(n+1)! + \frac{1}{6}(n+2)!.$$

Theorem 4

For any statistic f of degree m , there is a positive integer $L \leq \frac{m}{2}$, such that for all $n \geq L$,

$$M(f, n) = U(n)(n - L)!,$$

where $U(n)$ is a polynomial of degree no more than $m + L$. Equivalently, if $n \geq L$,

$$M(f, n) = \sum_{-L \leq i \leq m} \alpha_i (n + i)!,$$

for some constants $\alpha_i \in \mathbb{Q}$.

Example: *Sum of peak squares.*

$$M(\text{peakSqSum}, n) = (n + 1)! - \frac{5}{4}(n + 2)! + \frac{1}{5}(n + 3)!.$$

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Key observation: The union of two (or more) occurrences of a pattern σ in π is an occurrence of another pattern in π .

Example:

$$\pi = 516243$$

$$\sigma = \underline{132}$$

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Example:

$$\pi = 516243$$

$$\sigma = \underline{132}$$

$$(1, 4, 6) \in_{\underline{132}} \mathbf{516243},$$

$$(2, 3, 4) \in_{\underline{132}} \mathbf{516243}$$

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$$(1, 2, 3, 4, 6) \in_{\underline{15} \underline{243}} \mathbf{516243}$$

Let \underline{P}_1 , \underline{P}_2 and \underline{P}_3 be patterns of sizes k_1 , k_2 and k_3 , respectively.

A *merge* of \underline{P}_1 and \underline{P}_2 onto \underline{P}_3 is a pair of increasing functions $m_1 : [k_1] \rightarrow [k_3]$ and $m_2 : [k_2] \rightarrow [k_3]$ with certain properties.

Denote a merge by $m_1, m_2 : \underline{P}_1, \underline{P}_2 \rightarrow \underline{P}_3$.

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Denote a merge by $m_1, m_2 : \underline{P}_1, \underline{P}_2 \rightarrow \underline{P}_3$.

Example:

$$m_1(1) = 1, m_2(1) = 2,$$

$$m_1(2) = 4, m_2(2) = 3,$$

$$m_1(3) = 5, m_2(3) = 4.$$

Then $m_1, m_2 : \underline{132}, \underline{132} \rightarrow \underline{15\ 243}$.

Lemma 1

Let \underline{P}_1 and \underline{P}_2 be two patterns. For any $\sigma \in S_n$, there is a one-to-one correspondence between the following sets.

$$\{(s_1, s_2) : s_1 \in_{\underline{P}_1} \sigma, s_2 \in_{\underline{P}_2} \sigma\} \leftrightarrow \{s_3 \in_{\underline{P}_3} \sigma \mid m_1, m_2 : \underline{P}_1, \underline{P}_2 \rightarrow \underline{P}_3\}$$

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Using Lemma 1, we prove that the product of two simple statistics is a statistic:

$$\begin{aligned} f_{\underline{P}_1, \mathcal{Q}_1}(\sigma) g_{\underline{P}_2, \mathcal{Q}_2}(\sigma) &= \sum_{s_1 \in \underline{P}_1 \sigma} \mathcal{Q}_1(s_1) \mathcal{Q}'_1(\sigma^{-1}(s_1)) \sum_{s_2 \in \underline{P}_2 \sigma} \mathcal{Q}_2(s_2) \mathcal{Q}'_2(\sigma^{-1}(s_2)) \\ &\stackrel{(\text{by Lemma 1})}{=} \sum_{\underline{P}_3} \left(\sum_{s_3 \in \underline{P}_3 \sigma} \left(\sum_{m_1, m_2 : \underline{P}_1, \underline{P}_2 \rightarrow \underline{P}_3} \mathcal{Q}_{m_1, m_2, \mathcal{Q}_1, \mathcal{Q}_2}(s_3) \mathcal{Q}'_{m_1, m_2, \mathcal{Q}_1, \mathcal{Q}_2}(\sigma^{-1}(s_3)) \right) \right) = \sum_{\underline{P}_3} f_{\underline{P}_3, \tilde{\mathcal{Q}}}. \end{aligned}$$

Theorem 5

Let f be any statistic of degree m . Then, for any positive integer r , the r -th moment of f is given by

$$M(f^r, n) = \sum_{-I \leq i \leq J} \alpha_i (n + i)!,$$

where I and J are constants that satisfy $-I \geq \frac{-rm}{2}$, $J \leq mr$ and $n \geq I$, and the α_i 's are rational constants.

Corollary 1

If \underline{P} is a vincular pattern of size k , such that $|\mathbf{C}(\underline{P})| = c$, then

$$M(\text{cnt}_{\underline{P}}^r, n) = \sum_{0 \leq i \leq r(k-c)} \alpha_i (n+i)!,$$

for $n \geq rk$.

Zeilberger [8] showed that if \underline{P} is a classical pattern of size k , then $\mathbb{E}(\text{cnt}_{\underline{P}}^r)$ for a random permutation of size n , is a polynomial in n of degree rk .

Corollary 1 is a generalization.

1. Second moment of the number of double ascents.

$$M(\text{cnt}_{\underline{123}}^2, n) = -\frac{1}{12}n! - \frac{1}{15}(n+1)! + \frac{1}{36}(n+2)!.$$

2. Second moment of $\text{cnt}_{\underline{123}}^2$.

$$M(\text{cnt}_{\underline{123}}^2, n) = \frac{1}{2}n! - \frac{9}{28}(n+1)! + \frac{29}{672}(n+2)! + \frac{11}{10080}(n+3)! - \frac{1}{45360}(n+4)!.$$

Corollary 2

Let \underline{P} be a pattern of size k with $|\mathbf{C}(\underline{P})| = c$, $|\mathbf{D}(\underline{P})| = d$. Then,

$$M(\text{cnt}_{\underline{P}}^r, n) = \sum_{\tilde{k}, \tilde{c}, \tilde{d}} w_{\tilde{k}, \tilde{c}, \tilde{d}}^{(r)} \binom{n - \tilde{c}}{\tilde{k} - \tilde{c}} \binom{n - \tilde{d}}{\tilde{k} - \tilde{d}} (n - k)!,$$

where $w_{\tilde{k}, \tilde{c}, \tilde{d}}^{(r)}$ is the number of ways to merge r copies of \underline{P} and get a pattern \underline{P}^r of size \tilde{k} , with $|\mathbf{C}(\underline{P}^r)| = \tilde{c}$, $|\mathbf{D}(\underline{P}^r)| = \tilde{d}$ and where $k \leq \tilde{k} \leq rk$, $c \leq \tilde{c} \leq rc$ and $d \leq \tilde{d} \leq rd$.

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Next goal: Apply Corollary 2 to some simple patterns.

Theorem 6

Let $\text{des} := \text{cnt}_{\underline{21}}$. Consider a random permutation in S_n . Then, for any $r \geq 2$,

$$\mathbb{E}(\text{des}^r) = \sum_{m=2}^{\min(n, 2r)} \sum_{u=1}^{\lfloor \frac{m}{2} \rfloor} \left(\sum_{w=0}^{m-u} (-1)^w \binom{m-u}{w} (m-u-w)^r \right) \left(\sum_{\substack{q_1 + \dots + q_u = m \\ q_i \geq 2}} \binom{m}{q_1, \dots, q_u} \right) \frac{\binom{n-(m-u)}{u}}{m!}.$$

Theorem 6

Let $\text{des} := \text{cnt}_{\underline{21}}$. Consider a random permutation in S_n . Then, for any $r \geq 2$,

$$\mathbb{E}(\text{des}^r) = \sum_{m=2}^{\min(n, 2r)} \sum_{u=1}^{\lfloor \frac{m}{2} \rfloor} \left(\sum_{w=0}^{m-u} (-1)^w \binom{m-u}{w} (m-u-w)^r \right) \left(\sum_{\substack{q_1 + \dots + q_u = m \\ q_i \geq 2}} \binom{m}{q_1, \dots, q_u} \right) \frac{\binom{n-(m-u)}{u}}{m!}.$$

Sketch of proof:

Let \underline{P}^r be a pattern of size m , obtained after a merge of r copies of $\underline{21}$.

Every such \underline{P}^r with u segments has $|\mathcal{C}(\underline{P}^r)| = m - u$.

Thus $M(\text{cnt}_{\underline{P}^r}, n) = \frac{\binom{n-(m-u)}{u}}{m!} n!$ and

$$w_{\underline{k}, \underline{c}}^{(r)} = w_{m, m-u}^{(r)} = \sum_{u=1}^{\lfloor \frac{m}{2} \rfloor} \left(\sum_{w=0}^{m-u} (-1)^w \binom{m-u}{w} (m-u-w)^r \right) \sum_{\substack{q_1 + \dots + q_u = m \\ q_i \geq 2}} \binom{m}{q_1, \dots, q_u}.$$

Theorem 7

Let $\text{adj} := \text{cnt}_{\overline{12}}$. Consider a random permutation in S_n . Then, for any $r \geq 1$,

$$\mathbb{E}(\text{adj}^r) = \sum_{m=2}^{\min(n, 2r)} \sum_{u=1}^{\lfloor \frac{m}{2} \rfloor} \left(\left(\sum_{w=0}^{m-u} (-1)^w \binom{m-u}{w} (m-u-w)^r \right) \binom{m-u-1}{u-1} u! \frac{\binom{n-(m-u)}{u}^2}{n^{(m)}} \right).$$

We will use this result to prove a limit theorem for adj .

The asymptotic distribution of $\text{cnt}_{\underline{P}}(\sigma)$, when σ is chosen uniformly at random from S_n , is Normal:

- i.** True, when \underline{P} is a classical pattern (Bóna, 2007 [1]) .
- ii.** True, when \underline{P} is a vincular pattern (Hofer, 2017 [5])
- iii.** Not true for an arbitrary bivincular pattern
(last proof by Corteel, Louchard and Pemantle, 2004 [3])

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We will reprove **i.** and **iii.** and give a lemma that would imply **ii.**

Bóna [1] uses the method of *dependency graphs* to obtain that the following theorem implies the CLT for an arbitrary classical pattern.

Theorem 8

Let $X_n := \text{cnt}_\sigma$ be the number of occurrences of a classical pattern $\sigma \in S_k$ in a random permutation of size n . Then, there exists $c > 0$, such that for all n ,

$$\text{Var}(X_n) \geq cn^{2k-1}.$$

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Sketch of proof:

Use Corollary 2 to obtain that

$$\text{Var}(X_n) = \mathbb{E}(X_n^2) - \mathbb{E}^2(X_n) = [a_\sigma(2k) \frac{\binom{n}{2k}}{(2k)!} + a_\sigma(2k-1) \frac{\binom{n}{2k-1}}{(2k-1)!} + \mathcal{O}(n^{2k-2})] - \frac{\binom{n}{k}^2}{(k!)^2},$$

where $a_\sigma(r)$ is the number of ways to merge two copies of σ and get a pattern of size r .

Sketch of proof (cont.)

Note that $a_\sigma(2k) = \binom{2k}{k}^2$ and simplify to get:

$$\text{Var}(X_n) \geq cn^{2k-1} \iff a_\sigma(2k-1) > \binom{2k-1}{k}^2.$$

Sketch of proof (cont.)

Note that $a_\sigma(2k) = \binom{2k}{k}^2$ and simplify to get:

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Lemma 2 (Burstein and Hästö, [2, Lemma 4.3])

For any classical pattern $\sigma \in S_k$,

$$a_\sigma(2k-1) > \binom{2k-1}{k}^2.$$

$$A_{\sigma,\sigma'}(2k-1) := \{(\pi, x, y) \mid \pi \in \mathcal{S}_{2k-1}, x, y \in \text{subs}(\pi), \text{red}(x)=\sigma, \text{red}(y)=\sigma', |x \cap y| = 1\},$$

where $\text{subs}(\pi)$ denotes the set of the subsequences of the permutation π .

$a_{\sigma}(2k-1)$, is the number of triples in the set $A_{\sigma,\sigma}(2k-1)$.

Example: $A_{312,312}(5)$ contains $(54213, 523, 413)$.

5	4	2	1	3
5		2		3
	4		1	3

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Theorem 9

If $a_{\sigma,\sigma'}(2k-1) := |A_{\sigma,\sigma'}(2k-1)|$, then Lemma 2 is equivalent to

$$a_{\sigma}(2k-1) > \mathbb{E}(a_{\sigma,\sigma'}(2k-1)),$$

for any fixed $\sigma \in S_k$ and σ' chosen uniformly at random in S_k .

Theorem 10 (Hofer, [5])

Let $X_n = \text{cnt}_{\underline{\sigma}}$ be the number of occurrences of a vincular pattern $\underline{\sigma}$ with j blocks, in a random permutation of size n . Then, there exists $c > 0$, such that for all n ,

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Let $b_{\underline{\sigma}}(m, j')$ be the number of merges of two copies of $\underline{\sigma}$, where the resulting pattern is of size m and has j' blocks.

Example: $\underline{\sigma} = \underline{431} \underline{52}$. Below is a merge of two copies of σ .

The resulting pattern, $\underline{6531} \underline{84} \underline{72}$, has size $m = 8$ and $j' = 3$ blocks:

6	5	3	1	8	4	7	2
6	5	3		8	4		
	5	3	1			7	2

Merge of two copies of the pattern $\underline{431} \underline{52}$.

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$$\text{Var}(X_n) \geq cn^{2j-1}.$$

Theorem 11

Theorem 10 is equivalent to

$$\sum_{l=1}^{M_{\underline{\sigma}}} (2k)_l b_{\underline{\sigma}}(2k-l, 2j-1) > \binom{2k}{k} \binom{2j-1}{j} j,$$

for any vincular pattern $\underline{\sigma}$ with j blocks, where $M_{\underline{\sigma}}$ is the maximal size of a block of $\underline{\sigma}$.

Recall that adj denotes $\text{cnt}_{\overline{12}}$.

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We use Theorem 7 and the method of moments to prove the following.

Theorem 12

adj converges in distribution to $\text{Po}(1)$.

Sketch of proof:

Show that $\lim_{n \rightarrow \infty} \mathbb{E}(\text{adj}^r) = B_r$, where B_r is the r -th Bell number.

It is well-known that the r -th moment of $\text{Po}(1)$ variable is B_r .

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It is well-known that the r -th moment of $\text{Po}(1)$ variable is B_r .

Therefore, $\text{cnt}_{\underline{P}}$ does not necessarily converge to a Normal distribution, when \underline{P} is a bivincular pattern.

- 1) Can we find a combinatorial proof of Lemma 2 and the corresponding fact for vincular patterns?

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“... This would be hopefully followed by sequels applied to other combinatorial objects like graph-colorings, Boolean functions, and Random Walks...”

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Thanks for the attention!

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