

Generalized carry process and riffle shuffle

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Carries in addition

Adding 2 numbers with randomly chosen digits,

Introduction

Amazing
Matrix

(b, n, p) -
process

Riffle Shuffle

$(-b, n, p)$ -
process

Miscellaneous

Application

Summary

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Adding 3 numbers,

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Adding 3 numbers,

10111	10210	11102	11122	01011	11210	2112
43443	07082	04401	15299	64642	73497	38426
00171	55077	11440	95932	91116	17255	19649
49339	70267	68885	98147	70311	43856	37376
92954	32426	84728	09380	26070	34608	95451

then 1 seems to appear frequently. ($\#0 : \#1 : \#2 = 7 : 20 : 7$).

Transition Probability 1

$$P_{ij} := \mathbf{P}(C_{k+1} = j \mid C_k = i), \quad i, j \in \{0, 1, \dots, n-1\}$$

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Example 1 ($b = 2, n = 2$)

0	0	0	0	0	0	1	0
0	1	0	0	1	1	1	1
0	1	1	1	0	0	0	0

$$\implies (P_{0,0}, P_{0,1}) = \frac{1}{2^2} (3, 1)$$

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 $\implies (P_{0,0}, P_{0,1}) = \frac{1}{2^2} (3, 1)$

For $b = 2, n = 2$

$$P = \frac{1}{2^2} \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix} \implies \text{Stationary dist. } \pi = \left(\frac{1}{2}, \frac{1}{2} \right)$$

Transition Probability 2

Example 2 ($b = 2, n = 3$)

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$$\begin{array}{cccc}
 \begin{array}{c} 0 \ 0 \\ \hline 0 \end{array} & \begin{array}{c} 0 \ 0 \\ \hline 1 \end{array} & \begin{array}{c} 1 \ 0 \\ \hline 1 \end{array} & \begin{array}{c} 1 \ 0 \\ \hline 1 \end{array} \\
 \begin{array}{c} 0 \\ \hline 0 \end{array} & \begin{array}{c} 0 \\ \hline 1 \end{array} & \begin{array}{c} 1 \\ \hline 0 \end{array} & \begin{array}{c} 1 \\ \hline 1 \end{array} \\
 \begin{array}{c} 0 \\ \hline 0 \end{array} & \begin{array}{c} 0 \\ \hline 1 \end{array} & \begin{array}{c} 0 \\ \hline 0 \end{array} & \begin{array}{c} 1 \\ \hline 1 \end{array}
 \end{array}$$

$$\implies (P_{0,0}, P_{0,1}, P_{0,2}) = \frac{1}{2^3} \cdot (4, 4, 0)$$

Transition Probability 2

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$$\begin{array}{c}
 \begin{array}{cc|cc|cc|cc}
 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 \\
 \hline
 & & & & & & & \\
 0 & & 1 & & 1 & & 1 & \\
 \hline
 0 & & 0 & & 1 & & 1 & \\
 \hline
 0 & & 0 & & 0 & & 1 & \\
 \hline
 0 & & 1 & & 0 & & 1 &
 \end{array}
 \end{array}$$

$$\Rightarrow (P_{0,0}, P_{0,1}, P_{0,2}) = \frac{1}{2^3} \cdot (4, 4, 0)$$

For $b = 2, n = 3$

$$P = \frac{1}{2^3} \begin{pmatrix} 4 & 4 & 0 \\ 1 & 6 & 1 \\ 0 & 4 & 4 \end{pmatrix} \Rightarrow \pi = \frac{1}{3!} \cdot (1, 4, 1)$$

Carries Process

Add n base- b numbers ($b, n \in \mathbf{N}, b, n \geq 2$)

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C_k : the carry coming out in the k -th digit.

Carry	C_{k+1}	C_k	\cdots	C_1	$C_0 = 0$
Addends		$X_{1,k}$	\cdots	$X_{1,1}$	$X_{1,0}$
		\vdots		\vdots	\vdots
		$X_{n,k}$	\cdots	$X_{n,1}$	$X_{n,0}$
Sum		S_k	\cdots	S_1	S_0

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$$C_k + X_{1,k} + \cdots + X_{n,k} = C_{k+1}b + S_k, \quad S_k \in D_b$$

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$\{C_k\}_{k=0}^{\infty}$ ($C_k \in \{0, \dots, n-1\}$) is called the **carries process**.

Amazing Matrix : Holte(1997)

$$P_{ij} := \mathbf{P}(C_{k+1} = j \mid C_k = i), \quad i, j = \underline{0}, 1, \dots, n-1$$

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E-values and left E-vectors of Amazing Matrix

E-values/ E-vectors depends only on b / n .

$$P = L^{-1}DL, \quad D = \text{diag} \left(1, \frac{1}{b}, \frac{1}{b^2}, \dots, \frac{1}{b^{n-1}} \right)$$

$$L_{ij} = v_{ij}(n) = [x^j](A_m(x)).$$

$A_m(x) := (1-x)^{n+1} \sum_{j \geq 0} (j+1)^m x^j$: Eulerian polynomial.

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Remark. $P(b_1)P(b_2) = P(b_1 \cdot b_2)$.

Property of Left Eigenvectors

$$[1] \quad L = \begin{pmatrix} (n\text{-th Eulerian num.}) \\ \vdots \\ (-1)^j((n-1)\text{-th Pascal num.}) \end{pmatrix}$$

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$$\begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}, \begin{pmatrix} 1 & 4 & 1 \\ 1 & 0 & -1 \\ 1 & -2 & 1 \end{pmatrix}, \begin{pmatrix} 1 & 11 & 11 & 1 \\ 1 & 3 & -3 & -1 \\ 1 & -1 & -1 & 1 \\ 1 & -3 & 3 & -1 \end{pmatrix}$$

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$E(n, k) := \#\{ \sigma \in S_n \text{ with } k\text{-descents} \} : n\text{-th Eulerian num.}$

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$$E(3, 0) = \#\{(123)\} = 1,$$

$$E(3, 1) = \#\{(13\underline{2}), (3\underline{1}2), (23\underline{1}), (2\underline{1}3)\} = 4,$$

$$E(3, 2) = \#\{(3\underline{2}1)\} = 1.$$

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[2] L is equal to the Foulkes character table of S_n
(Diaconis-Fulman, 2012).

Foulkes character

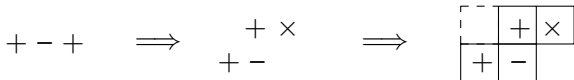
Example

$$\begin{aligned} & \#\{\sigma \in S_4 \mid \sigma(1) < \sigma(2) > \sigma(3) < \sigma(4)\} \\ &= \{(1324), (1423), (2314), (2413), (3412)\} = 5 \end{aligned}$$

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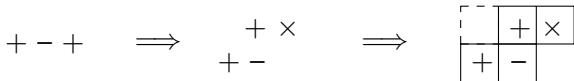
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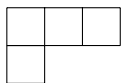
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LR
 \approx



$\dim = 3$

\oplus



$\dim = 2$

Property of Right Eigenvectors

Right Eigenvector of P

$$P = RDR^{-1}$$

$$R_{ij} = \sum_{r=n-j}^n (-1)^{n-r} \begin{bmatrix} n \\ r \end{bmatrix} \binom{r}{n-j} (n-1-i)^{r-(n-j)}$$

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$$\begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \quad \begin{pmatrix} 1 & 3 & 2 \\ 1 & 0 & -1 \\ 1 & -3 & 2 \end{pmatrix} \quad \begin{pmatrix} 1 & 6 & 11 & 6 \\ 1 & 2 & -1 & -2 \\ 1 & -2 & -1 & 2 \\ 1 & -6 & 11 & -6 \end{pmatrix}$$

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$$R(0, j) = S(n, n-j)$$

$S(n, j) := \#\{\sigma \in S_n \text{ with } j\text{-cycles}\}$ Stirling num. of 1st kind

Riffle Shuffle

Let $\{\sigma_1, \sigma_2, \dots\}$ ($\sigma_0 = id$), be the Markov chain on S_n induced by the repeated b -riffle shuffles on n -cards.

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Relation to Riffle Shuffles (Diaconis-Fulman, 2009)

$$\{C_k\}_{k=1}^{\infty} \stackrel{d}{=} \{d(\sigma_k)\}_{k=1}^{\infty}, \quad d(\sigma) : \text{the descent of } \sigma \in S_n.$$

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$$\{C_k\}_{k=1}^{\infty} \stackrel{d}{=} \{d(\sigma_k)\}_{k=1}^{\infty}, \quad d(\sigma) : \text{the descent of } \sigma \in S_n.$$

Since the stationary dist. of $\{\sigma_k\}$ is uniform on S_n ,

$$\begin{aligned} L_{0j} &= \lim_{k \rightarrow \infty} \mathbf{P}(C_k = j) = \lim_{k \rightarrow \infty} \mathbf{P}(d(\sigma_k) = j) \\ &= \mathbf{P}_{unif}(d(\sigma) = j) = E(n, j)/n! \end{aligned}$$

explaining why Eulerian num. appears.

Summary on Known Results

Amazing Matrix (the transition probability matrix P of the carries process) has the following properties.

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- (1) Eulerian num. appears in the stationary distribution.

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- (0) E-values depend only on b , and E-vectors depend only on n
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- (2) Left eigenvector matrix L equals to the Foulkes character table of S_n .

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(3) Stirling num. of 1st kind appears in the right eigenvector matrix R .

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- (0) E-values depend only on b , and E-vectors depend only on n
- (1) Eulerian num. appears in the stationary distribution.
- (2) Left eigenvector matrix L equals to the Foulkes character table of S_n .
- (3) Stirling num. of 1st kind appears in the right eigenvector matrix R .
- (4) carries process has the same distribution to the descent process of the riffle shuffle.

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Add n base- b numbers. Let $\frac{b-1}{p} \in \mathbf{N}$, $\frac{1}{p} + \frac{1}{p^*} = 1$.

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		\vdots		\vdots	\vdots
		$X_{n,k}$	\cdots	$X_{n,1}$	$X_{n,0}$
		$\frac{b-1}{p^*}$	\cdots	$\frac{b-1}{p^*}$	$\frac{b-1}{p^*}$
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Sum		S_k	\cdots	S_1	S_0

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		$\frac{b-1}{p^*}$	\cdots	$\frac{b-1}{p^*}$	$\frac{b-1}{p^*}$
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$\{C_k\}_{k=0}^{\infty}$ is called the (b, n, p) -carries process.

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		$\frac{b-1}{p^*}$	\cdots	$\frac{b-1}{p^*}$	$\frac{b-1}{p^*}$
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Choose $X_{j,k}$ uniformly at random from $D_b := \{0, 1, \dots, b-1\}$.
Given C_k , C_{k+1} is determined by

$$C_k + X_{1,k} + \cdots + X_{n,k} + \frac{b-1}{p^*} = C_{k+1}b + S_k, \quad S_k \in D_b.$$

$\{C_k\}_{k=0}^{\infty}$ is called the (b, n, p) -carries process.
($p = 1$: usual carries process)

Remarks

(1) If we generalize the usual carries process by changing the digit set such as

$$D_b = \{0, 1, \dots, b-1\} \implies D_b = \{d, d+1, \dots, d+b-1\},$$

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(2) $C_k \in \mathcal{C}_p(n)$ where $\mathcal{C}_p(n)$ is the carries set given by

$$\mathcal{C}_p(n) := \begin{cases} \{0, 1, \dots, n-1\} & (p = 1) \\ \{0, 1, \dots, n\} & (p \neq 1) \end{cases}$$

Left Eigenvectors

$\tilde{P} = \{\tilde{P}_{ij}\}$: Transition probability of (b, n, p) - process :

$$\tilde{P}_{ij} = \mathbf{P}(C_{k+1} = j \mid C_k = i).$$

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E-values/ E-vectors depend only on b / n .

$$\tilde{P} = L_p^{-1} D L_p, \quad D = \text{diag} \left(1, \frac{1}{b}, \dots, \frac{1}{b^{\#C_p(n)-1}} \right)$$

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Combinatorial meaning of L

[1] The stationary distribution $L_{0j}^{(p)}(n)$ gives

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[3] For $p \notin \mathbf{N}$, we do not know...

Examples of $L(n = 3)$

$$p = 1 : \begin{pmatrix} 1 & 4 & 1 \\ 1 & 0 & -1 \\ 1 & -2 & 1 \end{pmatrix} \quad p = 2 : \begin{pmatrix} 1 & 23 & 23 & 1 \\ 1 & 5 & -5 & -1 \\ 1 & -1 & -1 & 1 \\ 1 & -3 & 3 & -1 \end{pmatrix}$$

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$$p = 5/2 : \begin{pmatrix} 1 & \frac{311}{8} & \frac{101}{2} & \frac{27}{8} \\ 1 & \frac{33}{4} & -7 & -\frac{9}{4} \\ 1 & -\frac{1}{2} & -2 & \frac{3}{2} \\ 1 & -3 & 3 & -1 \end{pmatrix} \quad ?$$

No hits on OEIS...

Theorem 2

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- (2) $R_{ij}^{(p)}(n) = [x^{n-j}] \# \left\{ \sigma \in G_{p,n} \mid \sigma : (x, n, p)\text{-shuffle with } d(\sigma^{-1}) = i \right\}$

Colored Permutation Group

$$\Sigma := [n] \times \mathbf{Z}_p \quad ([n] := \{1, 2, \dots, n\}), \quad \underline{p} \in \mathbf{N}$$

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Descent on $G_{p,n}$

(1) Define a ordering on Σ

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(2) “ $\sigma \in G_{p,n}$ has a descent at i ”

$$\stackrel{\text{def}}{\iff} \begin{array}{l} \text{(i) } (\sigma(i), \sigma^c(i)) > (\sigma(i+1), \sigma^c(i+1)) \text{ (for } 1 \leq i \leq n-1) \\ \text{(ii) } \sigma^c(n) \neq 0 \text{ (for } i = n). \end{array}$$

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(3) $d(\sigma)$: the number of descents of σ .

$$\text{Ex. } (p=3) : d((5,0) \searrow (3,0) \nearrow (2,1) \searrow (4,2) \nearrow (1,1) \searrow) = 3$$

Generalized
cary process
and riffle
shuffle

Fumihiko
NAKANO,
Taizo
SADAHIRO

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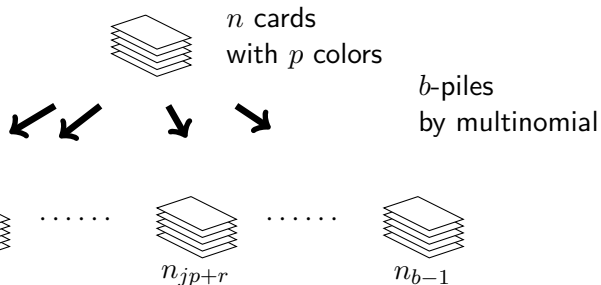
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Generalized Riffle Shuffle

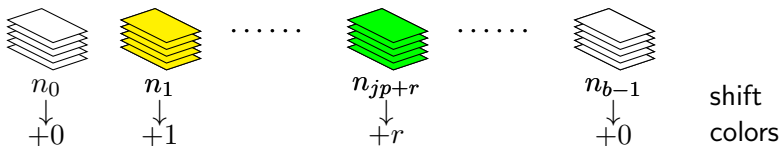
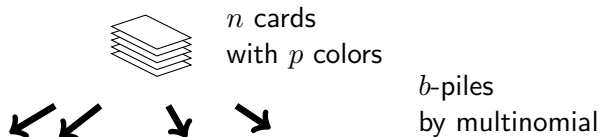


n cards
with p colors

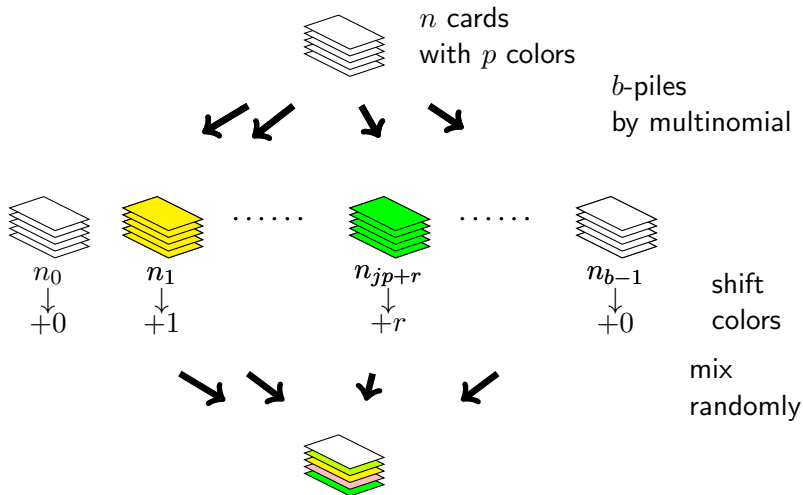
Generalized Riffle Shuffle



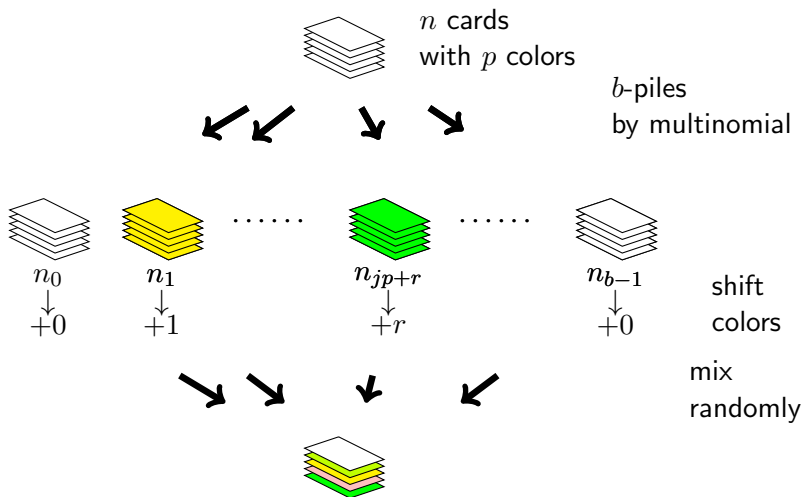
Generalized Riffle Shuffle



Generalized Riffle Shuffle



Generalized Riffle Shuffle



This process defines a Markov chain $\{\sigma_r\}_{r=0}^{\infty}$ on $G_{p,n}$.
(called the (b, n, p) -shuffle)

Carries Process and Riffle Shuffle

$$p \in \mathbf{N}, b \equiv 1 \pmod{p}$$
$$\{C_r\}_{r=1}^{\infty} : (b, n, p) - \text{process}$$

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(2) By Theorem 3, $\{d(\sigma_r)\}_r$ turns out to be a Markov chain.

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Any $x \in \mathbf{Z}$ can be expanded uniquely as

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\implies E-vectors of $\tilde{P}_{-b} : L_- = L_+, R_- = R_+.$

Dash - Descent on $G_{p,n}$

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$$\begin{aligned} \text{Ex. } (p = 3) \quad & d((5, 0) \searrow (3, 0) \nearrow (2, 1) \searrow (4, 2) \nearrow (1, 1) \searrow) = 3, \\ & d'((5, 0) \searrow (3, 0) \nearrow (2, 1) \nearrow (4, 2) \searrow (1, 1)) = 2. \end{aligned}$$

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$$p \in \mathbf{N}, (-b) \equiv 1 \pmod{p}$$
$$\{C_r^-\}_{r=1}^\infty : (-b, n, p) \text{ - process}$$

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Remark.

$\{d_r^-\}_r$ turns out to be a Markov chain.

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$$\Theta_i := \sum_{d(\sigma^{-1})=i} \sigma \in \mathbf{C}[G_{p,n}]$$

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(=Stirling Frobenius cycle number)

L : the left regular representation of $G_{p,n}$ on $\mathbf{C}[G_{p,n}]$.

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$$\Theta_i = \sum_k L_{ki} e_{n-k}, \quad e_{n-k} = \sum_i R_{ik} \Theta_i$$

Let Q : distribution of (b, n, p) -shuffle on $G_{p,n}$, and

$$m := \frac{3}{2} \log_b n + \log_b c,$$

$$\Phi(a) := \frac{1}{\sqrt{2\pi}} \int_{-\infty}^a e^{-\frac{x^2}{2}} dx.$$

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Theorem 6

$$\|Q^m - \text{Unif.}\|_{TV} = 1 - 2\Phi\left(-\frac{p}{4\sqrt{3}c}\right) + O(n^{-\frac{1}{2}})$$

Remark

$$1 - 2\Phi\left(-\frac{p}{4\sqrt{3}c}\right) \sim \begin{cases} \frac{p}{2c\sqrt{6\pi}} & (c \rightarrow \infty) \\ 1 - \frac{4c\sqrt{3}}{p\sqrt{2\pi}} \exp\left\{-\frac{1}{2}\left(\frac{-p}{4c\sqrt{3}}\right)^2\right\} & (c \rightarrow 0) \end{cases}$$

Limit Theorem

For **any** $p \geq 1$, and for $n \geq 2$, $k = 0, 1, \dots, n$, let

$$\left\langle \begin{matrix} n \\ k \end{matrix} \right\rangle_p := [x^k](A_{p,n}(x))$$

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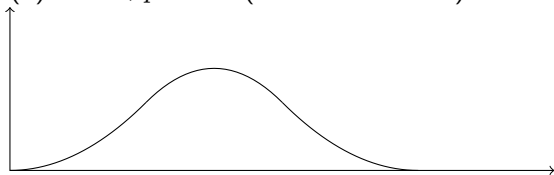
Theorem 5

$$\mathbf{P} \left(S_n \in \frac{1}{p} + [k - 1, k] \right) = \left\langle \begin{matrix} n \\ k \end{matrix} \right\rangle_p (p^n n!)^{-1}$$

for $k = 0, 1, \dots, n$.

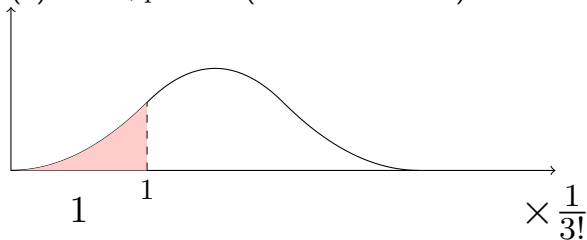
Example

(1) $n = 3, p = 1$: (Eulerian number)



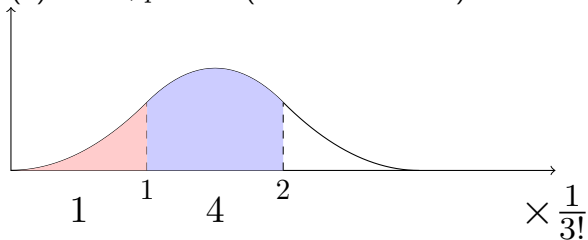
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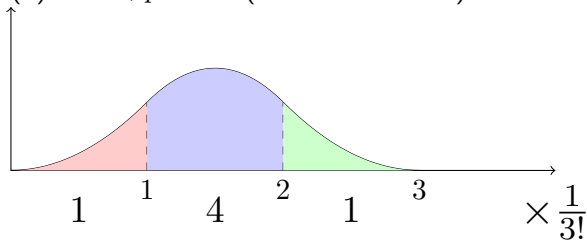
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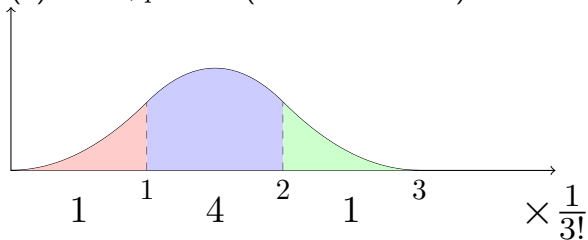
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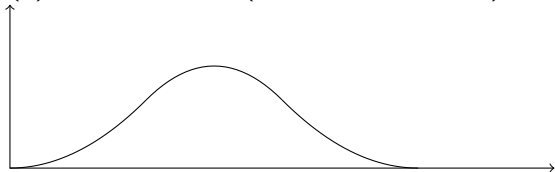


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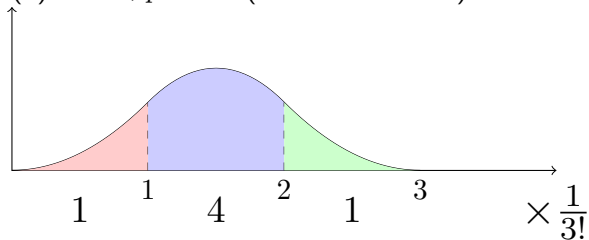


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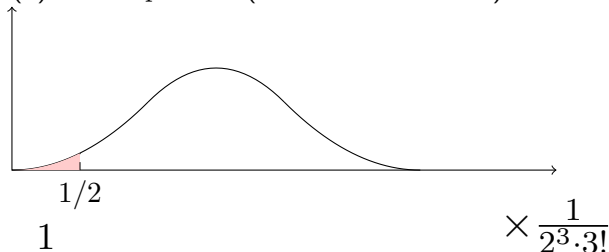


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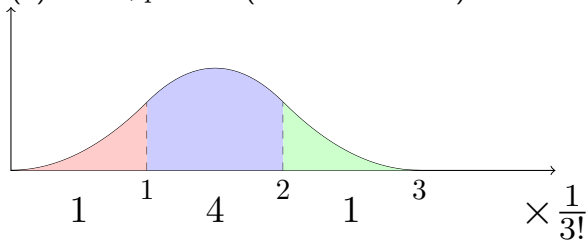


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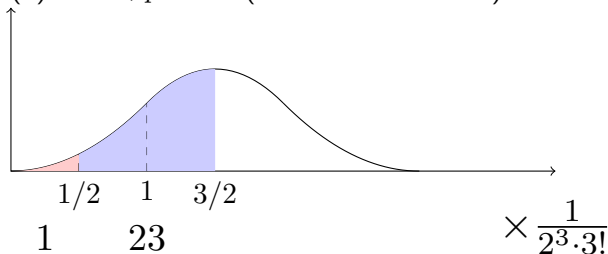


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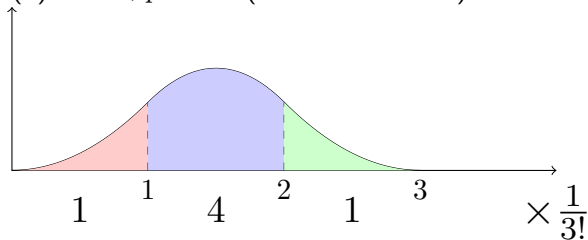


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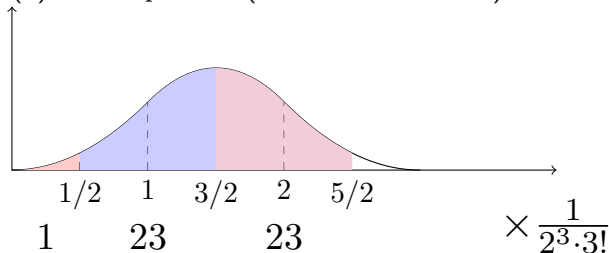


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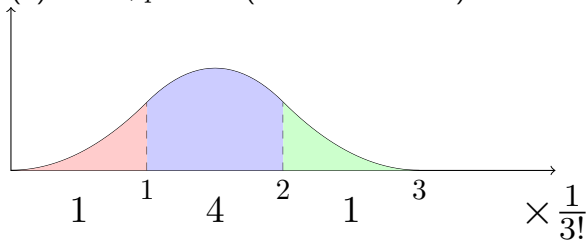


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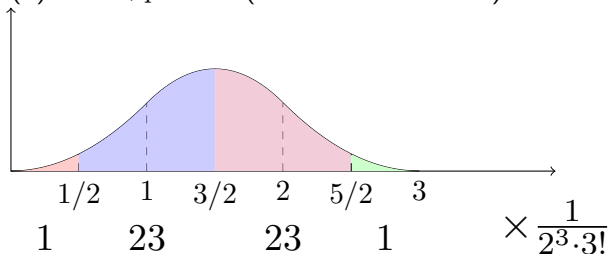


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Idea of Proof

Let X'_1, \dots, X'_m be independent, uniformly distributed r.v.'s on $[l, l + 1]$, and let $S'_m := X'_1 + \dots + X'_m$.

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$$\begin{array}{ccc} \downarrow & & \downarrow \\ \pi(j) & \mathbf{P}(S'_m \in [l, l + 1] + j) & \end{array}$$

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[3] for $p \notin \mathbf{N}$, **no** combinatorial meaning is known so far...

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