

Generalized carry process and riffle shuffle

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11/03/2016

Carries in addition

Adding 2 numbers with randomly chosen digits,

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Carries in addition

Adding 2 numbers with randomly chosen digits,

01111	00001	00000	01101	11111	00000	1100
71578	52010	72216	15692	99689	80452	46312
20946	60874	82351	32516	23823	30046	06870
92525	12885	54567	48209	20513	10498	53182

0 and 1 seem to appear at equal rate.

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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
<hr/>		<hr/>		<hr/>		<hr/>	
	0		1		0		1
	0		0		1		1
<hr/>		<hr/>		<hr/>		<hr/>	
	0		1		1		0

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

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$$\begin{array}{c|c|c|c}
 \textcolor{red}{0} & 0 & \textcolor{red}{0} & 0 \\
 \hline
 & 0 & 1 & 0 \\
 & 0 & 0 & 1 \\
 \hline
 & 0 & 1 & 0
 \end{array}
 \quad \Rightarrow \quad (P_{0,0}, P_{0,1}) = \frac{1}{2^2} (3, 1)$$

For $b = 2, n = 2$

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

Transition Probability 2

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

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Transition Probability 2

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

$\begin{array}{c} 0 \quad 0 \\ \hline 0 \end{array}$	$\begin{array}{c} 0 \quad 0 \\ \hline 1 \end{array}$	$\begin{array}{c} 1 \quad 0 \\ \hline 1 \end{array}$	$\begin{array}{c} 1 \quad 0 \\ \hline 1 \end{array}$
$\begin{array}{c} 0 \\ 0 \end{array}$	$\begin{array}{c} 0 \\ 0 \end{array}$	$\begin{array}{c} 1 \\ 0 \end{array}$	$\begin{array}{c} 1 \\ 0 \end{array}$
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$$\Rightarrow (P_{0,0}, P_{0,1}, P_{0,2}) = \frac{1}{2^3} \cdot (4, 4, 0)$$

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

For $b = 2, n = 3$

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

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	C_{k-1}	\cdots	C_1	$C_0 = 0$
Addends		$X_{1,k}$	\cdots	$X_{1,2}$	$X_{1,1}$
		\vdots		\vdots	\vdots
		$X_{n,k}$	\cdots	$X_{n,2}$	$X_{n,1}$
Sum		S_k	\cdots	S_2	S_1

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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)

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

$$P_{ij} = b^{-n} \sum_{r=0}^{\lfloor z/b \rfloor} (-1)^r \binom{n+1}{r} \binom{n+z_{ij}-br}{n}$$

$$z_{ij} := (j+1)b - i - 1$$

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E-values and left E-vectors of Amazing Matrix

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

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

$$L_{ij} = v_{ij}(n) = \sum_{r=0}^j (-1)^r \binom{n+1}{r} (j-r+1)^{n-i}$$

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

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

$$E(3, 2) = \# \{ (\underline{321}) \} = 1.$$

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

Foulkes character

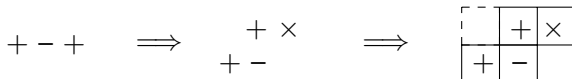
Example

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

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$$+ \ - \ + \quad \Rightarrow \quad \begin{array}{cc} + & \times \\ + & - \end{array} \quad \Rightarrow \quad \begin{array}{|c|c|c|} \hline & + & \times \\ \hline + & - & \\ \hline \end{array}$$

$$\begin{array}{|c|c|c|} \hline & & \\ \hline & & \\ \hline \end{array} \underset{\sim}{\overset{LR}} \begin{array}{|c|c|c|} \hline & & \\ \hline & & \\ \hline \end{array} \oplus \begin{array}{|c|c|} \hline & \\ \hline & \\ \hline \end{array}$$

dim = 3 dim = 2

Property of Right Eigenvectors

Right Eigenvector of P

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

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

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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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$$R(0, n-1-j) = S(n, j)$$

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

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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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.

Relation to Riffle Shuffles (Diaconis-Fulman, 2009)

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

In other words,

$$\begin{aligned} & \mathbf{P}(C_1 = j_1, C_2 = j_2, \dots, C_k = j_k \mid C_0 = 0) \\ &= \mathbf{P}(d(\sigma_1) = j_1, d(\sigma_2) = j_2, \dots, d(\sigma_k) = j_k \mid \sigma_0 = id). \end{aligned}$$

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

$$P_{0j} = \mathbf{P}_{unif}(d(\sigma) = j) \propto E(n, j),$$

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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- (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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We generalize the previous results by

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(1) taking the different digit set

$$\mathcal{D}_0 = \{0, 1, \dots, b-1\} \implies \mathcal{D}_d := \{\textcolor{red}{d}, d+1, \dots, d+b-1\}$$

Our aim

We generalize the previous results by

(1) taking the different digit set

$$\mathcal{D}_0 = \{0, 1, \dots, b-1\} \implies \mathcal{D}_d := \{\textcolor{red}{d}, d+1, \dots, d+b-1\}$$

and / or

(2) taking the negative base

$$b \implies \textcolor{red}{-b}$$

Generalized $(+b)$ -expansion

Take the digit set as follows.

$$\mathcal{D}_d := \{d, d+1, \dots, d+b-1\} \quad (\text{digit set})$$
$$d \leq 0, \quad d+b-1 \geq 0 \quad (\text{so that } 0 \in \mathcal{D}_d)$$

Holte's case corresponds to $d = 0$.

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Holte's case corresponds to $d = 0$.

Then $\forall x \in \mathbf{N}$ can be represented uniquely as

$$x = a_n(+b)^n + a_{n-1}(+b)^{n-1} + \dots + a_0, \quad a_k \in \mathcal{D}_d$$

A Generalized Carry Process

Add n base- b numbers in the representation above.

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$\{C_k\}$ ($C_k \in \mathcal{C}(b, n)$) is called a **(generalized) Carries Process**.

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(1) Let

$$l_+ := \frac{d}{b-1} \leq 0.$$

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Then the carry set $\mathcal{C}(b, n)$ is equal to

$$\mathcal{C}(b, n) = \{s, s+1, \dots, t\}$$

$$s := \lfloor (n-1)l_+ \rfloor, \quad t := \lceil (n-1)(l_+ + 1) \rceil$$

Carry set

(1) Let

$$l_+ := \frac{d}{b-1} \leq 0.$$

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$$(2) \# \mathcal{C}(b, n) = \begin{cases} n & (n-1)l_+ \in \mathbf{Z} \quad (\supset \text{ Holte's case }) \\ n+1 & (n-1)l_+ \notin \mathbf{Z} \end{cases}$$

Why ?

$$F := \left\{ \frac{a_1}{b} + \cdots + \frac{a_N}{b^N} \mid N \in \mathbf{N}, a_j \in \mathcal{D}_d \right\} \hookrightarrow (l_+, l_+ + 1)$$

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Why ?

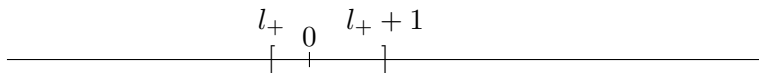
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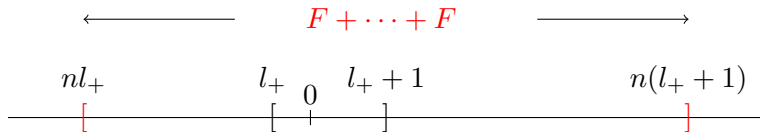
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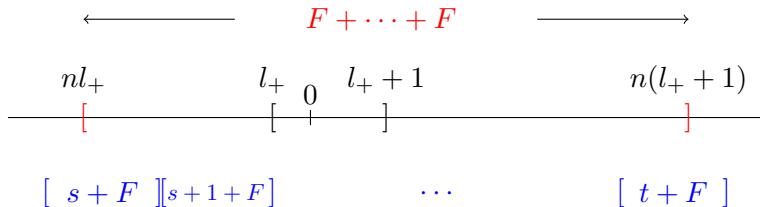
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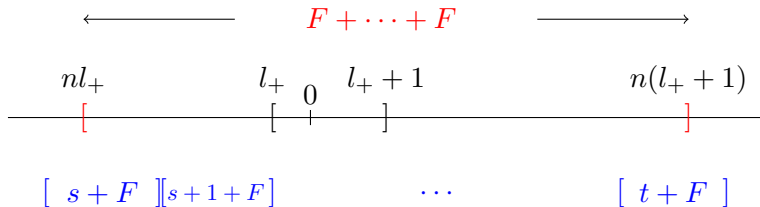
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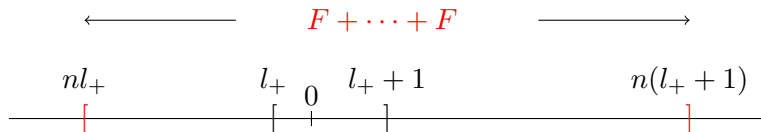


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$$[s + F] [s + 1 + F] \quad \cdots \quad [t + F]$$

$$s + l_+ \leq nl_+ \implies s = \lfloor (n-1)l_+ \rfloor$$

$$1 - \frac{1}{p} \quad \text{Define } p \text{ s.t. } 1 - \frac{1}{p} = nl_+ - (s + l_+)$$

(b, n, p) -process

Let $s := \min \mathcal{C}(b, n)$, $p := (1 - \{(n-1)l_+\})^{-1} \in \mathbf{Q}$, and let $\tilde{P} := \{\tilde{P}_{ij}\}_{i,j}$, $\tilde{P}_{ij} := \mathbf{P}(C_{k+1} - s = j \mid C_k - s = i)$.

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$$\tilde{P}_{ij} = b^{-n} \sum_{r=0}^{n+1} (-1)^r \binom{n+1}{r} \binom{n + A_p(i, j) - br}{n}$$

$$i, j = 0, 1, \dots, \#\mathcal{C}(b, n) - 1, \quad \#\mathcal{C}(b, n) = \begin{cases} n & (p = 1) \\ n + 1 & (p > 1) \end{cases}$$

Thus \tilde{P} is determined by (b, n, p) only.

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Remark : Holte's process $\implies p = 1$.

Left Eigenvectors

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

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

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D : independent of p

L : independent of b .

Combinatorial meaning of L

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

(1) $p = 1$: Eulerian number
(descent statistics of the permutation group)

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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...

Right Eigenvector

Theorem 2

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(2) $u_{ij}^{(p)}(n) =$

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

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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so we abuse to write $\sigma = ((4, 1), (1, 0), (2, 2), (3, 2))$.

In general, setting $(\sigma(i), \sigma^c(i)) := \sigma(i, 0) \in \Sigma, i = 1, 2, \dots, n,$

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$$\begin{array}{cccc} \Downarrow & \Downarrow & \Downarrow & \Downarrow \\ (4, 1) & (1, 0) & (2, 2) & (3, 2) \end{array} \implies \begin{array}{cc} \Downarrow & \Downarrow \\ (4, 2) & (2, 0) \end{array}$$

This σ is determined by $(4, 1) \ (1, 0) \ (2, 2) \ (3, 2)$.

so we abuse to write $\sigma = ((4, 1), (1, 0), (2, 2), (3, 2))$.

In general, setting $(\sigma(i), \sigma^c(i)) := \sigma(i, 0) \in \Sigma, i = 1, 2, \dots, n$,

we write $\sigma = ((\sigma(1), \sigma^c(1)), (\sigma(2), \sigma^c(2)), \dots, (\sigma(n), \sigma^c(n)))$.

Descent on $G_{p,n}$

Define a ordering on Σ

$$(1, 0) < (2, 0) < \cdots < (n, 0)$$

$$< (1, p-1) < (2, p-1) < \cdots < (n, p-1)$$

$$< (1, p-2) < (2, p-2) < \cdots < (n, p-2)$$

\dots

$$< (1, 1) < \cdots < (n, 1).$$

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

$\stackrel{def}{\iff}$

- (i) $(\sigma(i), \sigma^c(i)) > (\sigma(i+1), \sigma^c(i+1))$ (for $i = 1, 2, \dots, n-1$)
- (ii) $\sigma^c(n) \neq 0$ (for $i = n$).

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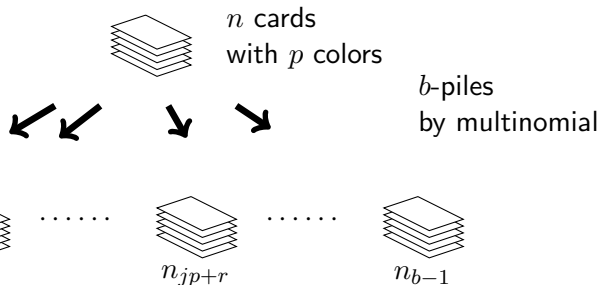
$d(\sigma)$: the number of descents of σ .

Generalized Riffle Shuffle

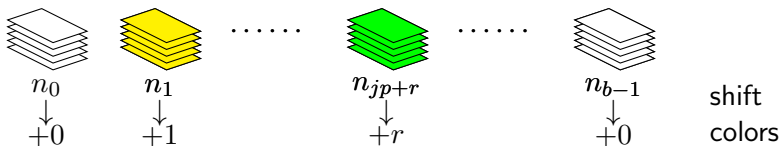
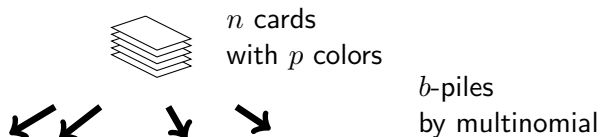


n cards
with p colors

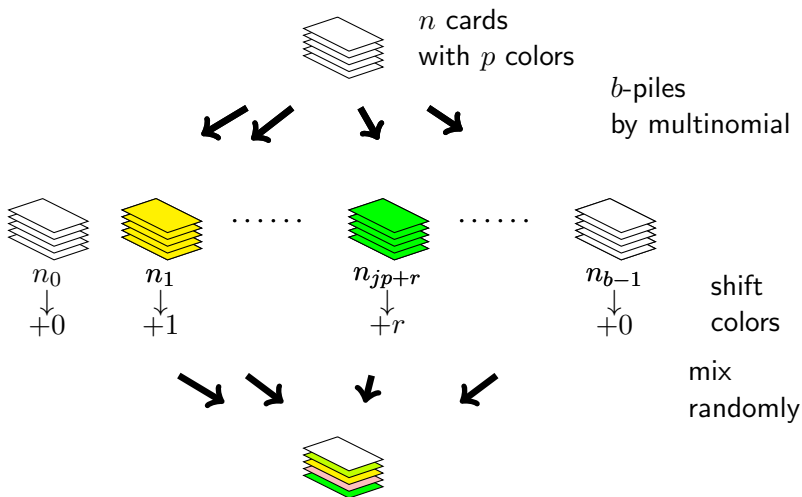
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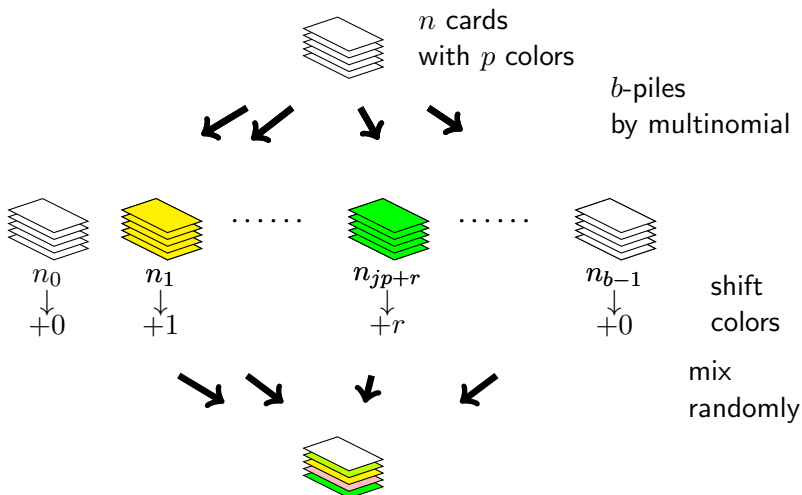
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Generalized Riffle Shuffle



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

$$\{\kappa_r := C_r - s\}_{r=1}^{\infty} : (b, n, p) - \text{process}$$

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Theorem 3

$$\{\kappa_r\} \stackrel{d}{=} \{d(\sigma_r)\}$$

In other words,

$$\begin{aligned} & \mathbf{P}(\kappa_1 = j_1, \kappa_2 = j_2, \dots, \kappa_k = j_k \mid \kappa_0 = 0) \\ &= \mathbf{P}(d(\sigma_1) = j_1, d(\sigma_2) = j_2, \dots, d(\sigma_k) = j_k \mid \sigma_0 = id) \end{aligned}$$

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Theorem 3 explains why the descent statistics of $G_{p,n}$ appears in the stationary distribution of $(b, n, p) - \text{process}$.

What about $(-b)$ -case ?

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

$$x = a_n(-b)^n + a_{n-1}(-b)^{n-1} + \cdots + a_0, \quad a_k \in \mathcal{D}_d$$

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$$\text{Ev} : 1, \frac{1}{b}, \frac{1}{b^2}, \cdots \implies 1, \left(-\frac{1}{b}\right), \left(-\frac{1}{b}\right)^2, \cdots,$$

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L_{\pm}, R_{\pm} have the same dependence on p .

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

(1) “ Dash - order ” $<'$ on Σ :

$$\begin{aligned} & (1, 0) <' (2, 0) <' \cdots <' (n, 0) \\ & <' (1, 1) <' (2, 1) <' \cdots <' (n, 1) \\ & <' \cdots \\ & <' (1, p-1) <' (2, p-1) <' \cdots <' (n, p-1) \end{aligned}$$

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

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$$\begin{aligned} & (\sigma(i), \sigma^c(i)) >' (\sigma(i+1), \sigma^c(i+1)) \text{ (for } i = 1, 2, \dots, n-1) \\ & \sigma^c(n) = p-1 \text{ (for } i = n). \end{aligned}$$

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(3) $d'(\sigma)$: the number of dash-descents of $\sigma \in G_{p,n}$.

$$\begin{aligned} & d(\sigma) = d'(\sigma) \text{ for } p = 1 \\ & E'_p(n, k) = E_p(n, n-k). \end{aligned}$$

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

$$\{\kappa_r^- = C_r^- - s^-\}_{r=1}^\infty : (-b, n, p) \text{ - process}$$

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$$d_r^- := \begin{cases} n - d'(\sigma_r) & (r : \text{ odd }) \\ d(\sigma_r) & (r : \text{ even }) \end{cases}$$

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Theorem 4

$$\{\kappa_r^-\}_r \stackrel{d}{=} \{d_r^-\}_r$$

Limit Theorem

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

$$\left\langle \begin{matrix} n \\ k \end{matrix} \right\rangle_p := \sum_{r=0}^k (-1)^r \binom{n+1}{r} \{p(k-r) + 1\}^n,$$

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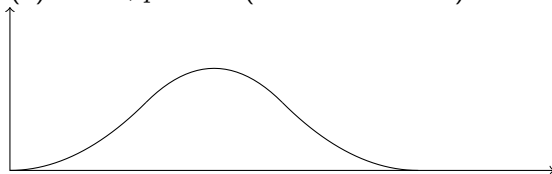
Theorem 5

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

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

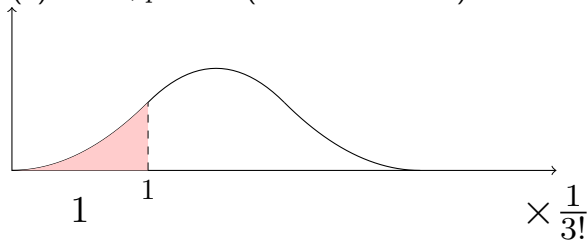
Example

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



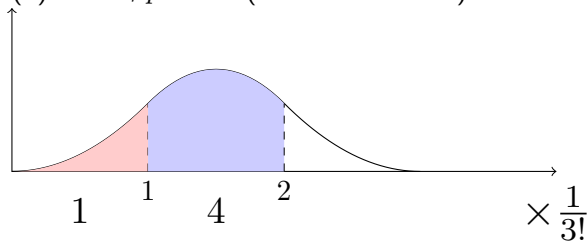
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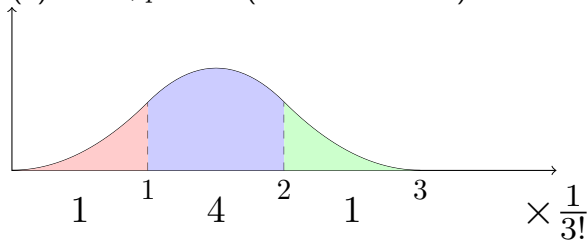
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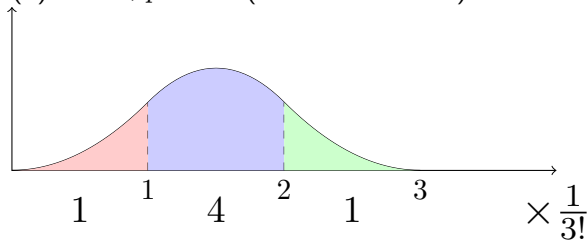
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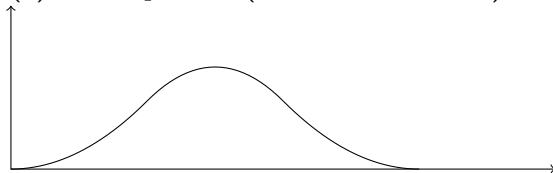


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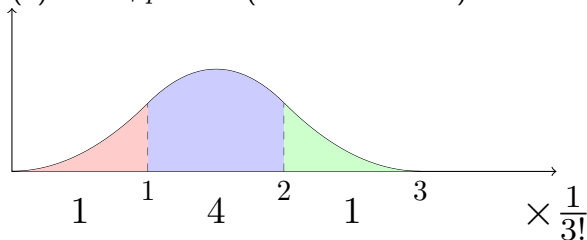


(2) $n = 3, p = 2$: (Macmahon number)

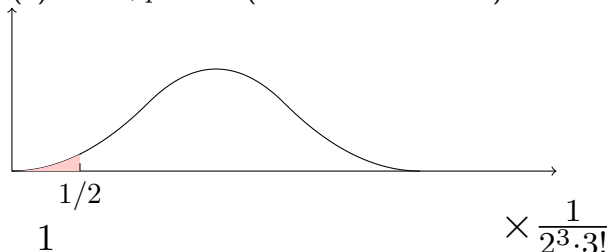


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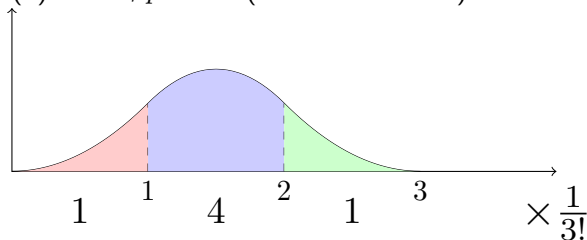


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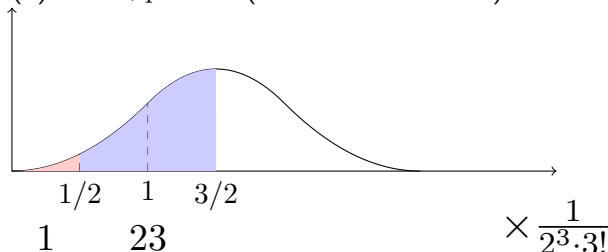


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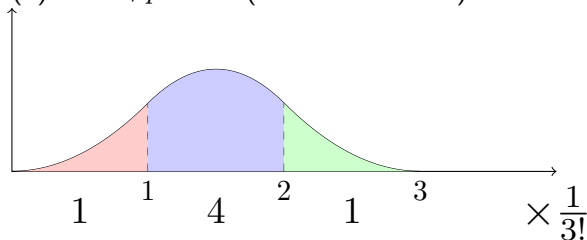


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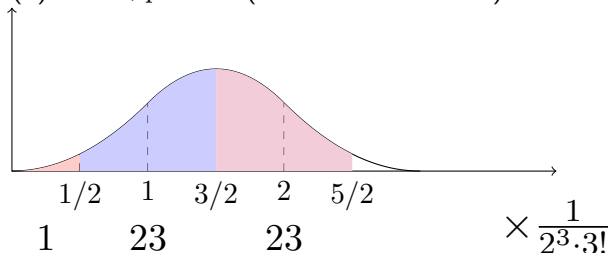


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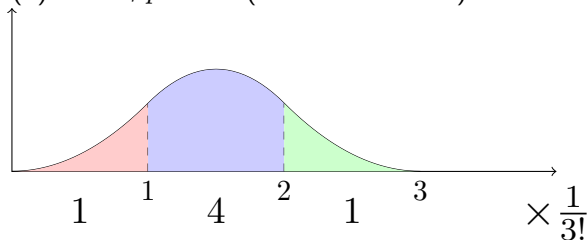


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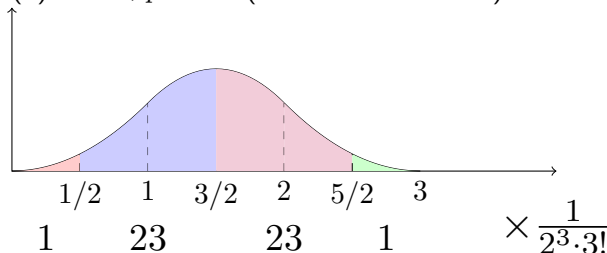


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Idea of Proof

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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Carry	C_k	C_{k-1}	\dots	C_1	C_0	
Addends		$X_{1,k}$	\dots	$X_{1,2}$	$X_{1,1}$	$= X_1^{(k)}$
		\vdots		\vdots	\vdots	\vdots
		$X_{m,k}$	\dots	$X_{m,2}$	$X_{m,1}$	$= X_m^{(k)}$
Sum		S_k	\dots	S_2	S_1	

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Since $X_i^{(k)} \xrightarrow{k \rightarrow \infty} X'_i$, $X_1^{(k)} + \dots + X_m^{(k)} \xrightarrow{k \rightarrow \infty} S'_m$.

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$$\mathbf{P}(C_k = j) = \mathbf{P}(X_1^{(k)} + \dots + X_m^{(k)} \in [l, l+1] + j)$$

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

Summary

[1] We study the generalization of the carries process $\{\kappa_r\}_r$, called $(\pm b, n, p)$ - process, and derived the left/right eigenvectors of its transition probability matrix.

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

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