

MIXING TIMES AND REPRESENTATION THEORY

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THE UNIVERSITY
OF BRITISH COLUMBIA

SIMONS
FOUNDATION

♠ : Motivation: different ways to mix a deck of cards

Example 1: riffle shuffle



Credit: Will Roya, Playingcarddeck.com

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Example 2: smooshing



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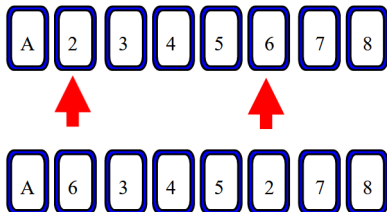
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Question: how long does it take to mix?

♠ : The random transposition shuffle

Method :

- ▶ Pick two cards uniformly and independently;
- ▶ If different, interchange them;
- ▶ If they are the same card, do nothing.



Credit: Elchanan Mossel

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Interpretation :

- ▶ Random walk on \mathfrak{S}_n with

$$P(\sigma, \sigma\tau) = \mu_n(\tau) = \begin{cases} 1/n & \text{if } \tau = id \\ 2/n^2 & \text{if } \tau \text{ is a transp.} \end{cases}$$

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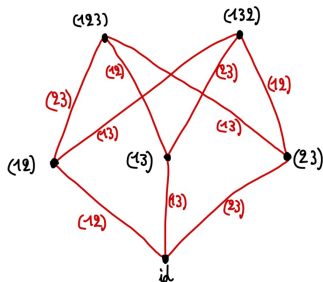
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Cayley graph for $n = 3$

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Définition

Distance to stationarity after t steps :

$$d_n(t) := d_{\text{TV}}(\nu_n(t), \text{Unif}_n).$$

where for probability measures μ and ν on \mathfrak{S}_n ,

$$d_{\text{TV}}(\mu, \nu) = \max_{A \subset \mathfrak{S}_n} |\mu(A) - \nu(A)| = \frac{1}{2} d_1(\mu, \nu).$$

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Question : How large should be take t so that $d_n(t) \approx 0$?

♠ : Cutoff for random transpositions

Theorem (DIACONIS AND SHAHSHAHANI, 1981)

It takes $\frac{1}{2}n \ln(n)$ steps to mix a deck of n cards by random transpositions.

For every $0 < \epsilon < 1$,

$$d_n \left((1 - \epsilon) \frac{1}{2} n \ln(n) \right) \xrightarrow{n \rightarrow +\infty} 1 \quad \& \quad d_n \left((1 + \epsilon) \frac{1}{2} n \ln(n) \right) \xrightarrow{n \rightarrow +\infty} 0$$

That is what is called the **cutoff phenomenon**.

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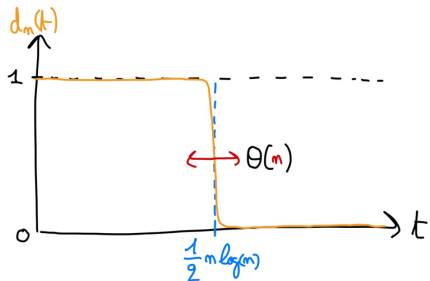
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More precisely, it takes $\frac{1}{2}n \ln(n) + \Theta(n)$ steps to mix.



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Answer: About 7.

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D. BAYER AND P. DIACONIS

TABLE 1
Total variation distance for m shuffles of 52 cards

m	1	2	3	4	5	6	7	8	9	10
$\ Q^m - U\ $	1.000	1.000	1.000	1.000	0.924	0.614	0.334	0.167	0.085	0.043

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New York Times, 9 Jan. 1990

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Answer: Bayer–Diaconis 1992: Precise estimates for $n = 52$, and cutoff profile:

Theorem (BAYER–DIACONIS, 1992)

For the riffle shuffle, we have for every $c \in \mathbb{R}$,

$$d_n \left(\frac{3}{2} \log_2(n) + c \right) \xrightarrow{n \rightarrow +\infty} p(c) := d_{\text{TV}} \left(\mathcal{N}(0, 1), \mathcal{N} \left(\frac{2^{-c}}{2\sqrt{3}}, 1 \right) \right).$$

(Written up to integer parts.)

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Question: and for transpositions, can we find the profile?

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Theorem (T., 2020)

For random transpositions, we have for every $c \in \mathbb{R}$,

$$d_n \left(\frac{1}{2} n \ln(n) + cn \right) \xrightarrow{n \rightarrow +\infty} p(c) := d_{\text{TV}}(\text{Pois}(1 + e^{-2c}), \text{Pois}(1)).$$

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- ▶ Several different types of profiles are known. For example with
 - ▶ **normal laws** for the riffle shuffle (Bayer–Diaconis, 1992), the random walk on $(\mathbb{Z}/2\mathbb{Z})^n$ (Diaconis–Graham–Morrison, 1990), or simple excursion process on the circle (Lacoin 2016),
 - ▶ **Poisson laws** for k -cycles ($k = o(n)$, Nestoridi–Olesker-Taylor, 2022) or more generally all conjugacy classes of the symmetric group (Olesker–Taylor T. 2024?),
 - ▶ **Tracy–Widom distributions** for the ASEP on a segment (Bufetov–Nejjar 2022),
 - ▶ **free Meixner laws** for the diffusion on O_N^+ (Freslon–T.–Wang, 2022).

♠ : Some results related to random transpositions

On random transpositions themselves :

Cutoff result : [Diaconis, Shahshahani](#), 1981, *PTRF*

Precise lower bound : [Matthews](#), 1988, *J. of Th. Prob.*

Phase transition result : [N. Berestycki, Durrett](#), 2006, *PTRF*

More precise estimates on the cutoff window : [Saloff-Coste–Zuniga](#), 2010, *AAP*

Probability of long cycles : [Alon, Kozma](#), 2013, *Duke*

Strong stationary time : [White](#), 2019

Cutoff profile : [T.](#), 2020, *Ann. Prob.*

Generalisations to other conjugacy classes :

Almost-precutoff for all conjugacy classes [Roichman](#), 1996, *Invent. Math.*

Some conjugacy classes with few fixed points [Lulov–Pak](#), 2002, *J. Alg. Comb.*

Precutoff for all conjugacy classes with few fixed points [Larsen–Shalev](#), 2008, *Invent. Math.*

Cutoff for k -cycles : [N. Berestycki, Schramm, Zeitouni](#), 2011, *Ann. Prob.*

Cutoff for conjugacy-invariant walks on \mathfrak{S}_n : [N. Berestycki, Şengül](#), 2014, *PTRF*

Profile for k -cycles : [Nestoridi, Olesker-Taylor](#), 2021, *PTRF*

Cutoff + profile for all conjugacy classes : [Olesker-Taylor–T.](#), 2024?

Some other generalisations :

Biaised random transpositions : [Matheau-Raven](#), 2020

Quantum random transpositions : [Freslon, T., Wang](#), 2021, *PTRF*

Star random transpositions : [Nestoridi](#), 2021

♡ : The non-commutative Fourier transform

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Inverse Fourier transform, isometry between Hilbert spaces, Parseval identity.

Pierre-Loïc Méliot, *Representation Theory of Symmetric Groups*, chap. 1.

♡ : A method to find cutoff profiles

For transpositions, we then apply the **inverse Fourier transform** on \mathfrak{S}_n to $f := v_n(t) - \text{Unif}_n$, and use that μ_n is **constant on conjugacy classes** (so by Schur's lemma each $\widehat{f}(\lambda)$ is a multiple of the identity (as a matrix)), to get

$$2d_n(t) = \frac{1}{n!} \sum_{\sigma \in \mathfrak{S}_n} \left| \sum_{\lambda \in \widehat{\mathfrak{S}_n}^*} d_\lambda s_\lambda^t \text{ch}^\lambda(\sigma) \right|.$$

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s_λ : eigenvalues of the (transition matrix of the) chain

d_λ : multiplicities

$\text{ch}^\lambda(\sigma)$: “eigenvectors”.

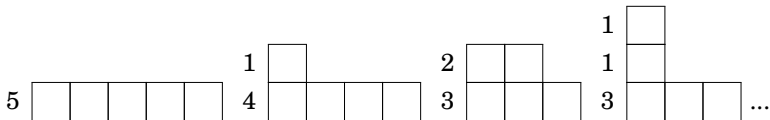
♥ : Representations of the symmetric group

- ▶ Irreducible representations λ of $\mathfrak{S}_n \longleftrightarrow$ Young diagrams of size n .
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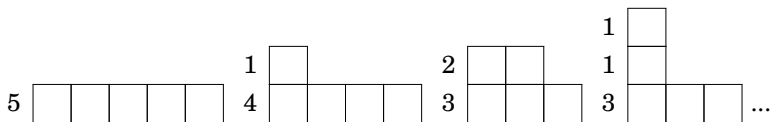
Example: $n = 5 = 4 + 1 = 3 + 2 = \dots$



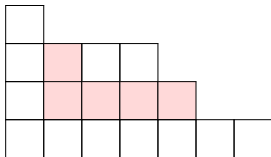
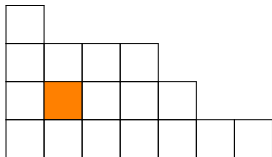
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- **Hook-length (of a box):** Example: $\lambda = [7, 5, 4, 1]$, u = orange box.
 The hook-length of u is $H(\lambda, u) = 5$.



◇ : The hook-length formula

The dimensions d_λ are easy to compute:

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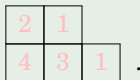
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Example

The hook-lengths of (the boxes of) $\lambda = [3, 2]$ are $4, 3, 1, 2, 1$, so

$$d_\lambda = \frac{|\lambda|!}{\prod_{u \in \lambda} H(\lambda, u)} = \frac{5!}{4 \cdot 3 \cdot 1 \cdot 2 \cdot 1} = 5.$$

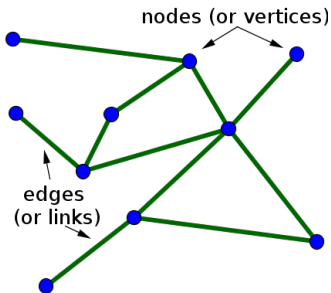


◇ : Another type of problems: the cover time problem

$\Gamma = (V, E)$: graph (finite, connected).

$(X_t)_{t \geq 0}$: simple random walk on Γ .

$T_x =$ “hitting time of the vertex x ” = $\min\{t : X_t = x\}$.



Credits: mathinsight.org

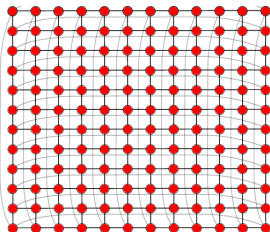
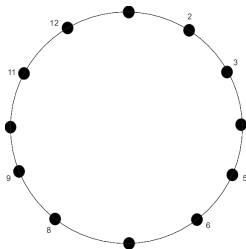
Cover time: $\tau_{\text{cov}} = \max_{x \in V} T_x =$ “first time at which all vertices have been visited”.

Research on cover times: active since the works of Aldous in the 80's.

Question: $\mathbb{E}[\tau_{\text{cov}}]$? Is τ_{cov} concentrated?

◇ : Cover time of tori

Example: the d -dimensional torus $(\mathbb{Z}/m\mathbb{Z})^d$. $n := m^d$.



Credits: Ljupco Kocarev ($d = 1$), Markus Quade ($d = 2$), researchgate

- ▶ $d = 1$: $\tau_{\text{cov}} \asymp n^2$, not concentrated.
- ▶ $d = 2$: cover time cutoff: $\tau_{\text{cov}} \sim c_2 n (\log n)^2$
(Dembo–Peres–Rosen–Zeitouni 2004, *Ann. Math.*)
- ▶ $d \geq 3$, cover time profile: $\tau_{\text{cov}} \approx c_d n (\log n + \chi)$, where $\chi \sim \text{Gumbel}$, i.e.
 $\mathbb{P}(\chi \leq s) = e^{-e^{-s}}$.
(Belius 2013, *Ann. Prob.* / De Prata 2012)

◇ : Cover time of vertex-transitive graphs

Our result: characterisation of Gumbel fluctuations.

$$t_{\text{hit}} = t_{\text{hit}}(\Gamma) = \max_{x,y \in V} \mathbb{E}_x T_y.$$

$$\text{Diam}(\Gamma) = \text{diameter of } \Gamma = \max_{x,y \in \Gamma} d(x,y).$$

$$n = n(\Gamma) := |V|.$$

Theorem (N. Berestycki–Hermon–T. 2023+)

For vertex-transitive graphs of (uniformly) bounded degree, we have

$$\frac{\tau_{\text{cov}}}{t_{\text{hit}}} - \log n \xrightarrow[n \rightarrow \infty]{} \chi \quad \text{if and only if} \quad \text{Diam}(\Gamma)^2 \log n = o(n).$$

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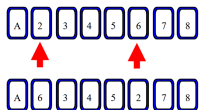
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- ▶ “Gumbel iff Γ is a bit more than 2-dimensional”
- ▶ Also iff the last points to be covered are “uniform”

◇ : Thank you for your attention!



$$d_n \left(\frac{3}{2} \log_2(n) + c \right) \xrightarrow{n \rightarrow +\infty} d_{\text{TV}} \left(\mathcal{N}(0, 1), \mathcal{N} \left(\frac{2^{-c}}{2\sqrt{3}}, 1 \right) \right)$$

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