

Maximal height of non-intersecting Brownian motions

G. Schehr

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Collaborators:

- A. Comtet (LPTMS, Orsay)
- P. J. Forrester (Dept. of Math., Melbourne)
- S. N. Majumdar (LPTMS, Orsay)
- J. Rambeau (Inst. Theor. Phys., Cologne)
- J. Randon-Furling (Univ. Paris 1, Paris)

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References:

- G. S., S. N. Majumdar, A. Comtet, J. Randon-Furling, Phys. Rev. Lett. **101**, 150601 (2008)
- J. Rambeau, G. S., Europhys. Lett. **91**, 60006 (2010); Phys. Rev. E **83**, 061146 (2011)
- P. J. Forrester, S. N. Majumdar, G. S., Nucl. Phys. B **844**, 500 (2011)

Height of rooted planar tree

THE AVERAGE HEIGHT OF PLANTED PLANE TREES

N. G. de Bruijn

Technological University
Eindhoven, The Netherlands

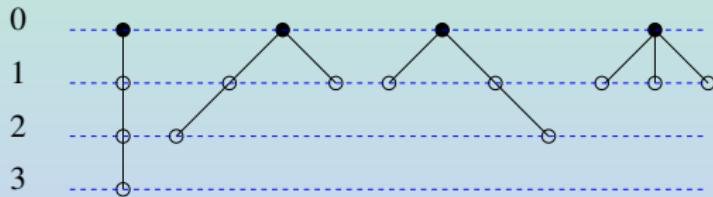
D. E. Knuth^t

Stanford University
Stanford, California

S. O. Rice

Bell Telephone Laboratories, Inc.
Murray Hill, New Jersey

- Height of rooted plane trees $H_{1,n}$ with $n+1$ nodes



$$H_{1,n=3} = 3$$

$$H_{1,n=3} = 2$$

$$H_{1,n=3} = 2$$

$$H_{1,n=3} = 1$$

Q : $\langle H_{1,n} \rangle$ for large n ?

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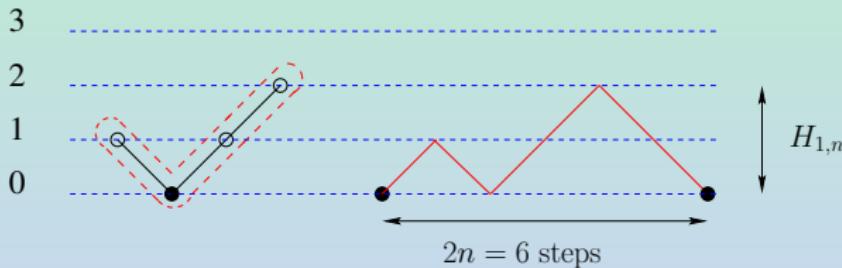
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- Mapping between rooted plane trees and Dyck paths



$$\lim_{n \rightarrow \infty} \frac{\langle H_{1,n} \rangle}{\sqrt{2n}} = \langle H_1 \rangle = \sqrt{\frac{\pi}{2}}$$

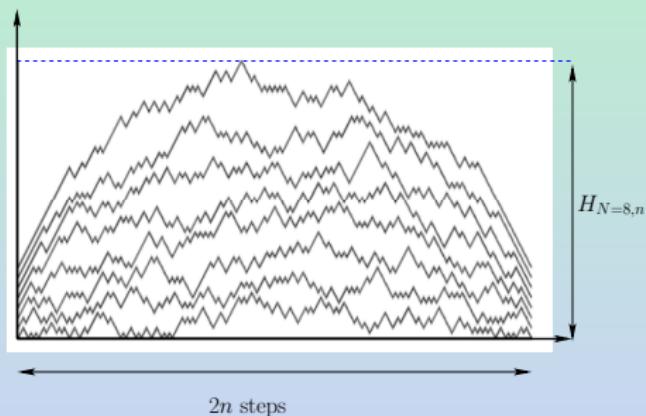
Generalization to N non-intersecting Dyck paths

Watermelon uniform random generation with applications

Nicolas Bonichon*, Mohamed Mosbah

LaBRI-Université Bordeaux 1, 351 Cours de la Libération, 33405 Talence, France

Theoretical Computer Science 307 (2003) 241–256



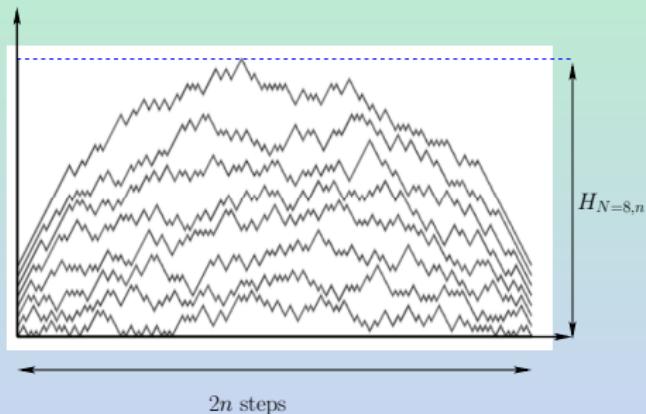
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⇒ numerical estimate for $\langle H_{N=8,n} \rangle$

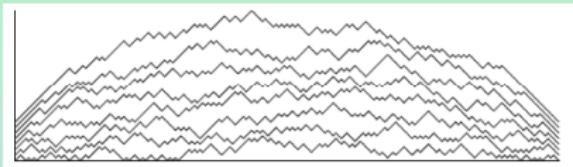
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$$\frac{\langle H_{N,n} \rangle_{\text{num}}}{\sqrt{2n}} = \langle H_N \rangle_{\text{num}} \sim \sqrt{1.67 N}$$

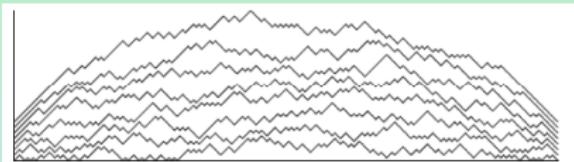
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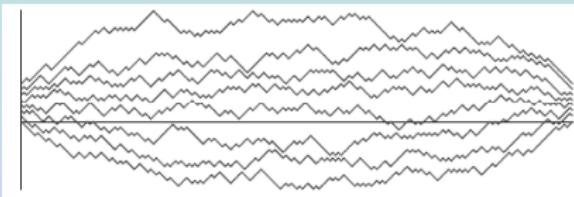
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$$\frac{\langle H_{N,n} \rangle_{\text{num}}}{\sqrt{2n}} = \langle H_N \rangle_{\text{num}} \sim \sqrt{1.67 N}$$



$$\frac{\langle H_{N,n} \rangle_{\text{num}}}{\sqrt{2n}} = \langle H_N \rangle_{\text{num}} \sim \sqrt{0.82 N}$$

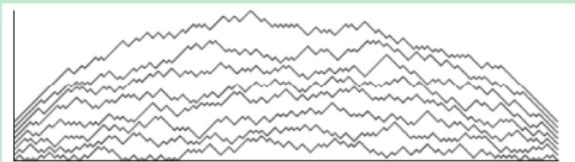
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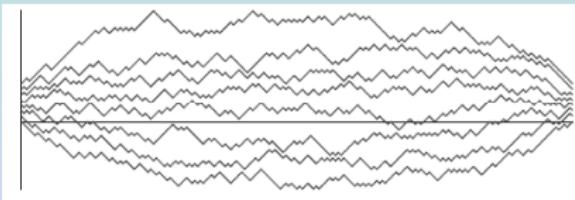
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$$\frac{\langle H_{N,n} \rangle_{\text{num}}}{\sqrt{2n}} = \langle H_N \rangle_{\text{num}} \sim \sqrt{1.67 N}$$



$$\frac{\langle H_{N,n} \rangle_{\text{num}}}{\sqrt{2n}} = \langle H_N \rangle_{\text{num}} \sim \sqrt{0.82 N}$$

Q : can one compute $\langle H_N \rangle$?

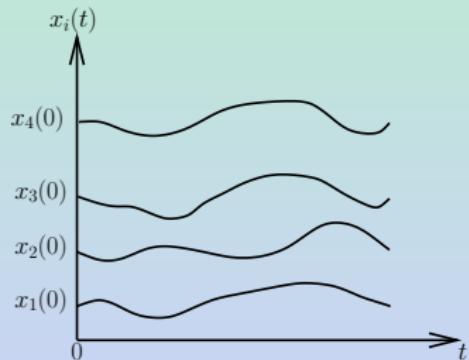
Non-intersecting Brownian motions in 1d

- N Brownian motions in one-dimension

$$\dot{x}_i(t) = \zeta_i(t), \langle \zeta_i(t)\zeta_j(t') \rangle = \delta_{i,j}\delta(t-t')$$
$$x_1(0) < x_2(0) < \dots < x_N(0)$$

- Non-intersecting condition

$$x_1(t) < x_2(t) < \dots < x_N(t), \quad \forall t \geq 0$$



Non-intersecting Brownian motions in 1d

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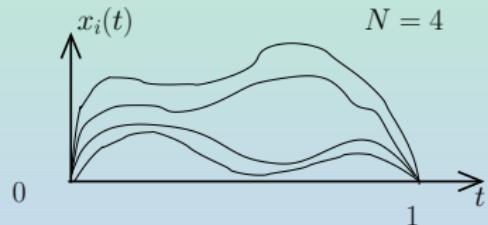
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- Non-intersecting condition

$$x_1(t) < x_2(t) < \dots < x_N(t), \\ \forall t \geq 0$$



watermelons "with a wall"

Vicious walkers in physics

THE JOURNAL OF CHEMICAL PHYSICS

VOLUME 48, NUMBER 5

1 MARCH 1968

Soluble Model for Fibrous Structures with Steric Constraints

P.-G. DE GENNES

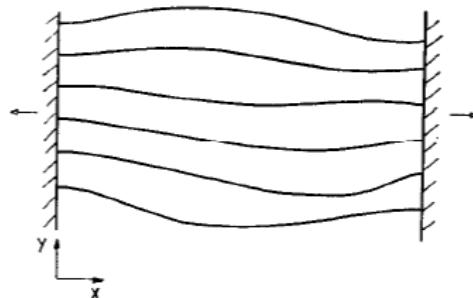
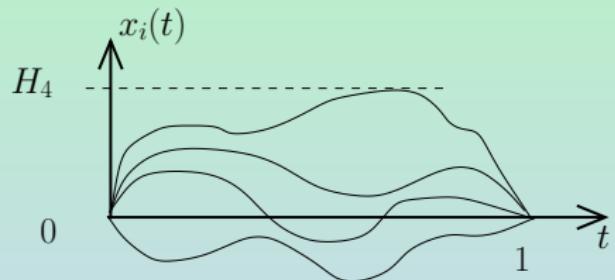


FIG. 1. Model for a two-dimensional fiber structure. The component chains are assumed to be attached to two plates I and F and placed under tension. The chains are bent by thermal fluctuations. Different chains cannot intersect each other.

Vicious walkers in physics and maths

- P. G. de Gennes, *Soluble Models for fibrous structures with steric constraints* (1968)
- M. E. Fisher, *Walks, Walls, Wetting and Melting* (1984)
- D. J. Grabiner, *Brownian motion in a Weyl chamber, non-colliding particles, and random matrices* (1999)
- C. Krattenthaler, A. J. Guttmann, X. G. Viennot, *Vicious walkers, friendly walkers and Young tableaux* (2000)
- P. L. Ferrari, K. Johansson, N. O'Connell, M. Praehofer, H. Spohn, C. Tracy, H. Widom... *Stochastic growth models, directed polymers* (from 2000)
- ...

Extreme statistics of vicious walkers



Maximal height of watermelons

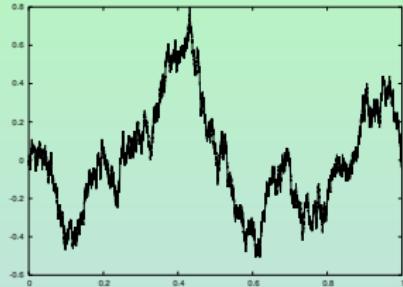
$$x_1(t) < x_2(t) < \dots < x_N(t)$$

$$H_N = \max_{\tau} [x_N(\tau), 0 \leq \tau \leq 1]$$

$$\langle H_N \rangle = ?$$

Extreme statistics of Brownian motion

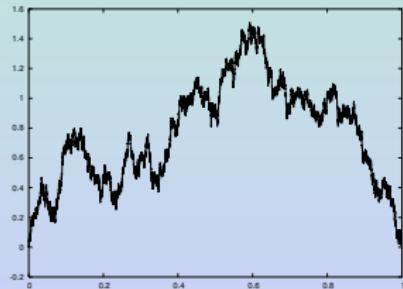
- Brownian bridge



$$H_1 = \max_{\tau} [x(\tau), 0 \leq \tau \leq 1]$$

$$\langle H_1 \rangle = \sqrt{\frac{\pi}{8}}$$

- Brownian excursion



$$H_1 = \max_{\tau} [x(\tau), 0 \leq \tau \leq 1]$$

$$\langle H_1 \rangle = \sqrt{\frac{\pi}{2}}$$

de Bruijn, Knuth, Rice '72

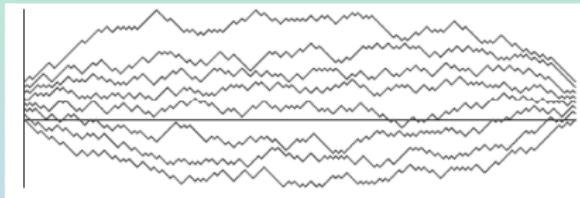
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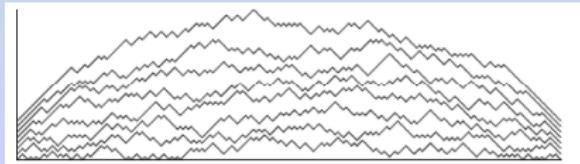
LaBRI-Université Bordeaux 1, 351 Cours de la Libération, 33405 Talence, France

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$$H_N = \max_{\tau} [x_N(\tau), 0 \leq \tau \leq 1]$$



$$\langle H_N \rangle_{\text{num}} \sim \sqrt{0.82 N}$$

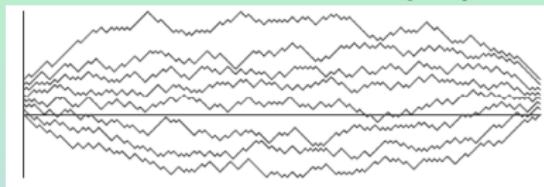


$$\langle H_N \rangle_{\text{num}} \sim \sqrt{1.67 N}$$

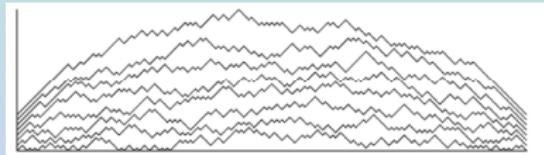
Main results

1 Connection between watermelons and random matrices

⇒ exact asymptotic results for $\langle H_N \rangle$ for $N \gg 1$



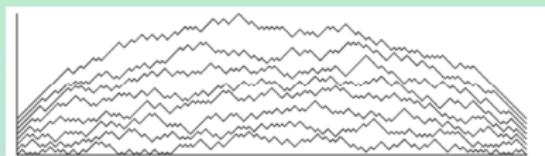
$$\langle H_N \rangle \sim \sqrt{N}$$



$$\langle H_N \rangle \sim \sqrt{2N}$$

Main results

2 Exact results for the full distribution of H_N



Cumulative distribution

$$F_N(M) = \text{Proba}[H_N \leq M]$$

$$F_N(M) \rightarrow \mathcal{F}_1 \left(2^{11/6} N^{1/6} \left| M - \sqrt{2N} \right| \right), \quad N \rightarrow \infty$$

\mathcal{F}_1 is the Tracy-Widom distribution for GOE random matrices

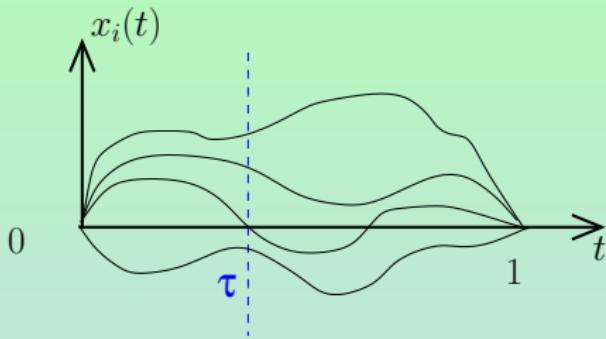
Outline

- 1 Vicious walkers and random matrices
- 2 Connection with stochastic growth models
- 3 Exact computation using Feynman-Kac formula
- 4 Large N limit using discrete orthogonal polynomials
- 5 Conclusion

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Non intersecting Brownian motions and RMT



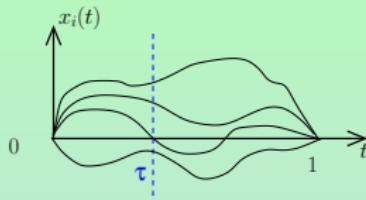
- Joint probability of $x_1(\tau), x_2(\tau), \dots, x_N(\tau)$ at fixed time τ

$$P_{\text{joint}}(x_1, x_2, \dots, x_N, \tau) \propto \prod_{i < j=1}^N (x_i - x_j)^2 e^{-\frac{1}{\sigma^2(\tau)} \sum_{i=1}^N x_i^2}$$

$$\sigma(\tau) = \sqrt{2\tau(1-\tau)}$$

- The rescaled positions $\frac{x_i}{\sigma(\tau)}$ are distributed like the eigenvalues of random matrices of the Gaussian Unitary Ensemble (GUE, $\beta = 2$)

Non intersecting Brownian motions and RMT



- $H \equiv H(t)$, $N \times N$ Hermitian matrices from GUE:

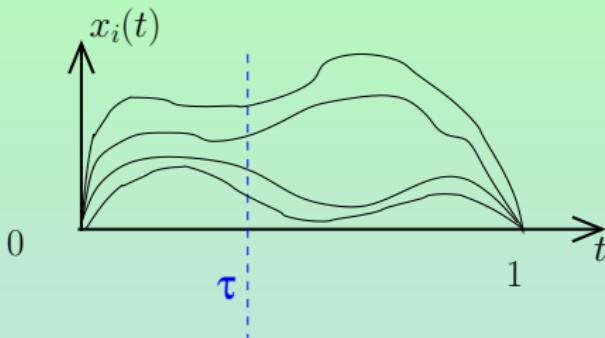
$$H_{mn}(t) = \begin{cases} \frac{1}{\sqrt{2}} \left(B_{mn}(t) + i \tilde{B}_{mn}(t) \right) , & m < n , \\ B_{mm}(t) , & m = n \\ \frac{1}{\sqrt{2}} \left(B_{nm}(t) - i \tilde{B}_{nm}(t) \right) , & m > n , \end{cases}$$

where $B_{m,n}, \tilde{B}_{m,n}$ independent Brownian bridges

- Eigenvalues of $H(t)$

$$\{\lambda_1(t) < \lambda_2(t) < \dots < \lambda_N(t)\} \stackrel{d}{=} \{x_1(t) < x_2(t) < \dots < x_N(t)\}$$

Non intersecting Brownian motions and RMT



- Joint probability of $x_1(\tau), x_2(\tau), \dots, x_N(\tau)$ at fixed time τ

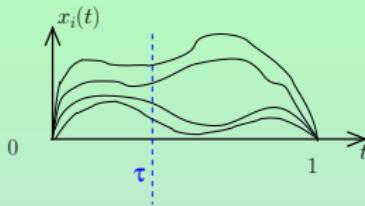
$$P_{\text{joint}}(\mathbf{x}, \tau) \propto \prod_{i=1}^N x_i^2 \prod_{1 \leq i < j \leq N} (x_i^2 - x_j^2)^2 e^{-\frac{\mathbf{x}^2}{\sigma^2(\tau)}}$$

- The rescaled positions $\frac{x_i}{\sigma(\tau)}$ are distributed like the eigenvalues of random matrices of the Bogoliubov-de Gennes type (class C)

A. Atland, M. R. Zirnbauer '96



Non intersecting Brownian motions and RMT



- $C(t)$, $2N \times 2N$ Herm. rand. mat. of **Bogoliubov-de-Gennes** type

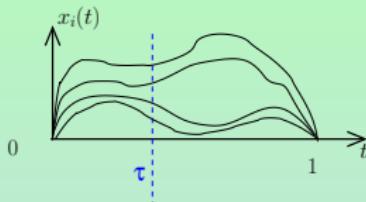
$$C(t) = \begin{pmatrix} H(t) & S(t) \\ S^\dagger(t) & -{}^tH(t) \end{pmatrix} \quad \text{A. Atland, M. R. Zirnbauer '96}$$

$H(t)$ is $N \times N$ Hermitian, $S(t)$ is $N \times N$ complex symmetric, where the entries are Brownian **bridges**

- $C(t)$ is symplectic, $C(t) \in \mathrm{sp}(2N)$, i.e.:

$${}^t C(t) J + J C(t) = 0, \quad J = \begin{pmatrix} \mathbb{O}_p & \mathbb{I}_p \\ -\mathbb{I}_p & \mathbb{O}_p \end{pmatrix}$$

Non intersecting Brownian motions and RMT



- $C(t)$, $2N \times 2N$ Herm. rand. mat. of Bologoliubov-de-Gennes type

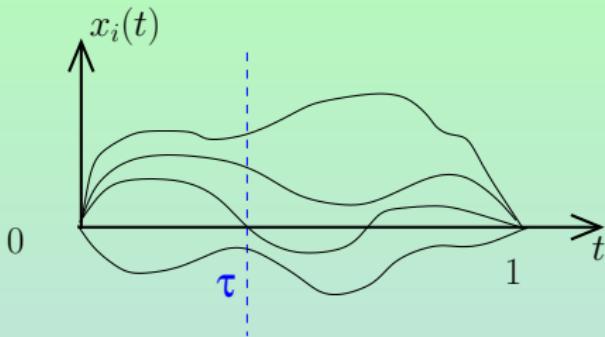
$$C(t) = \begin{pmatrix} H(t) & S(t) \\ S^\dagger(t) & -{}^tH(t) \end{pmatrix} \quad \text{A. Atland, M. R. Zirnbauer '96}$$

$H(t)$ is $N \times N$ Hermitian, $S(t)$ is $N \times N$ complex symmetric, where the entries are Brownian bridges

- Positive eigenvalues of $C(t) \stackrel{d}{=} \text{positions of walkers}$ M. Katori et al. '03

$$\lambda_N > \dots > \lambda_2 > \lambda_1 > -\lambda_1 > -\lambda_2 \dots > -\lambda_N , \quad \lambda_i > 0$$

Non intersecting Brownian motions and RMT



- Joint probability of $x_1(\tau), x_2(\tau), \dots, x_N(\tau)$ at fixed time τ

$$P_{\text{joint}}(x_1, x_2, \dots, x_N, \tau) \propto \prod_{i < j=1}^N (x_i - x_j)^2 e^{-\frac{1}{\sigma^2(\tau)} \sum_{i=1}^N x_i^2}$$

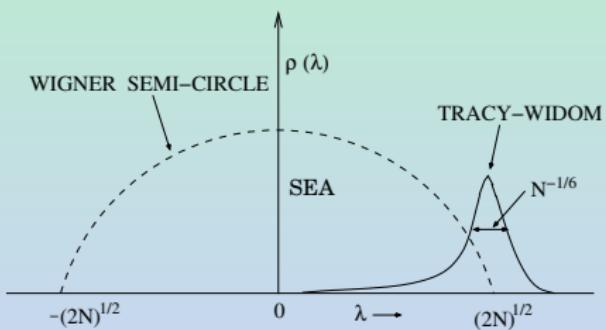
$$\sigma(\tau) = \sqrt{2\tau(1-\tau)}$$

- The rescaled positions $\frac{x_i}{\sigma(\tau)}$ are distributed like the eigenvalues of random matrices of the Gaussian Unitary Ensemble (GUE, $\beta = 2$)

Non intersecting Brownian motions and RMT

- The rescaled positions $\frac{x_i}{\sigma(\tau)}$ are distributed like the eigenvalues of random matrices of Gaussian Unitary Ensemble (GUE, $\beta = 2$)
- Mean density $\rho(\lambda)$ of eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_N$ for GUE

$$\rho(\lambda) = \frac{1}{N} \sum_{\alpha=1}^N \langle \delta(\lambda - \lambda_\alpha) \rangle$$



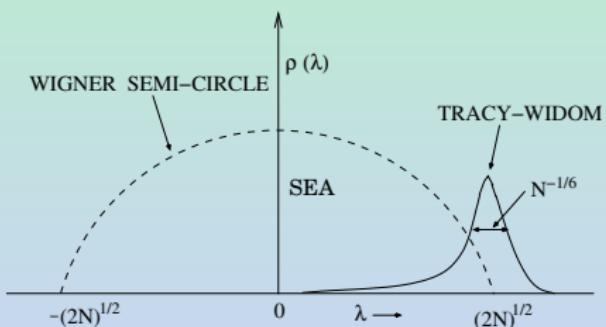
Non intersecting Brownian motions and RMT

- The rescaled positions $\frac{x_i}{\sigma(\tau)}$ are distributed like the eigenvalues of random matrices of Gaussian Unitary Ensemble (GUE, $\beta = 2$)
- Largest eigenvalue of random matrices from GUE

$$\begin{aligned}\lambda_{\max} &= \max_{1 \leq i \leq N} \lambda_i \\ &= \sqrt{2N} + \frac{N^{-\frac{1}{6}}}{\sqrt{2}} \chi_2\end{aligned}$$

$$\Pr[\chi_2 \leq \xi] = \mathcal{F}_2(\xi)$$

Tracy – Widom distribution



Non intersecting Brownian motions and RMT

- The rescaled positions $\frac{x_i}{\sigma(\tau)}$ are distributed like the eigenvalues of random matrices of Gaussian Unitary Ensemble (GUE, $\beta = 2$)
- Largest eigenvalue of random matrices from GUE

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$$\Pr[\chi_2 \leq \xi] = \mathcal{F}_2(\xi)$$

Tracy – Widom distribution

$$\mathcal{F}_2(\xi) = \exp \left[- \int_{\xi}^{\infty} (s - \xi) q^2(s) ds \right]$$

where $q(s)$ satisfies Painlevé II

$$\begin{aligned}q''(s) &= s q(s) + 2q^3(s) \\ q(s) &\sim \text{Ai}(s), s \rightarrow \infty\end{aligned}$$

C. Tracy, H. Widom '94

Watermelons in the limit of large N

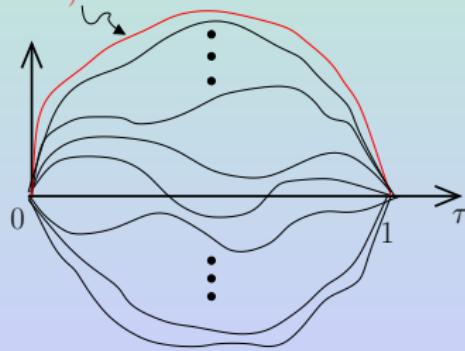
- Consequences for watermelons without wall for large N

$$\frac{x_N(\tau)}{\sqrt{2\tau(1-\tau)}} \sim \sqrt{2N} + \frac{N^{-1/6}}{\sqrt{2}} \chi_2$$

Proba[$\chi_2 \leq \xi$] = $\mathcal{F}_2(\xi)$, **Tracy-Widom distribution** for $\beta = 2$

- When $N \rightarrow \infty$, $x_N(\tau)$ reaches a deterministic elliptic shape

$$x_N(\tau) \sim 2\sqrt{N} \sqrt{\tau(1-\tau)}$$

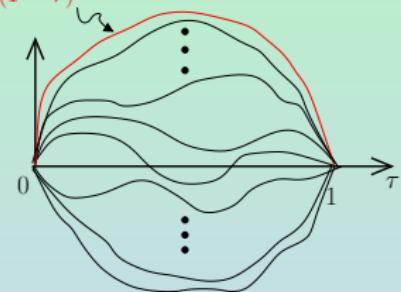
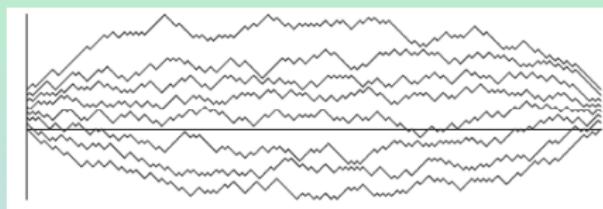


Asymptotic behavior of $\langle H_N \rangle$: without wall

- Consequences for watermelons without wall for large N

$$\frac{x_N(\tau)}{\sqrt{2\tau(1-\tau)}} \sim \sqrt{2N} + \frac{N^{-1/6}}{\sqrt{2}} \chi_2$$

$$x_N(\tau) \sim 2\sqrt{N}\sqrt{\tau(1-\tau)}$$



- The maximal height is reached for $\tau = 1/2$

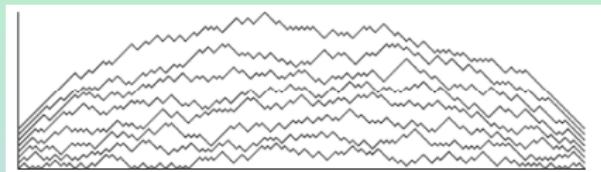
$$H_N = \max_{\tau} [x_N(\tau), 0 \leq \tau \leq 1]$$

$$\langle H_N \rangle = \langle x_N(\tau = \frac{1}{2}) \rangle \sim \sqrt{N} \quad \text{vs.} \quad \langle H_N \rangle_{\text{num}} \sim \sqrt{0.82 N}$$

Asymptotic behavior of $\langle H_N \rangle$: with wall

- Consequences for watermelons without wall for large N

$$\frac{x_N(\tau)}{\sqrt{2\tau(1-\tau)}} \sim 2\sqrt{N} + \frac{N^{-1/6}}{2^{2/3}} \chi_2$$



Proba[$\chi_2 \leq \xi$] = $\mathcal{F}_2(\xi)$,
Tracy-Widom distribution for $\beta = 2$

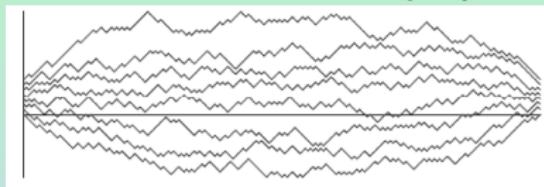
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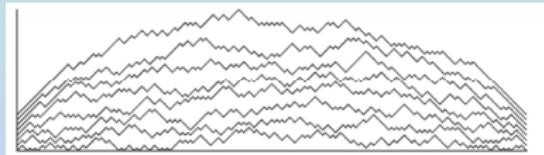
$$\langle H_N \rangle = \langle x_N(\tau = \frac{1}{2}) \rangle \sim \sqrt{2N} \quad vs. \quad \langle H_N \rangle_{\text{num}} \sim \sqrt{1.67 N}$$

Summary I

- Connection between **watermelons** and **random matrices**
 \implies exact asymptotic results for $\langle H_N \rangle$ for $N \gg 1$



$$\langle H_N \rangle \sim \sqrt{N}$$

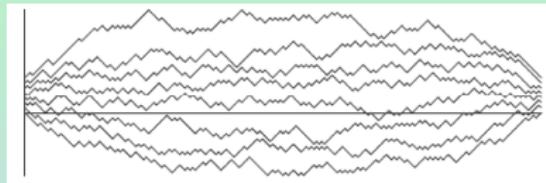


$$\langle H_N \rangle \sim \sqrt{2N}$$

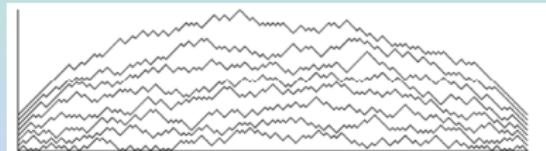
Summary I

- Connection between watermelons and random matrices

⇒ exact asymptotic results for $\langle H_N \rangle$ for $N \gg 1$



$$\langle H_N \rangle \sim \sqrt{N}$$



$$\langle H_N \rangle \sim \sqrt{2N}$$

What about the fluctuations of H_N ?

Fluctuations of H_N in the limit of large N

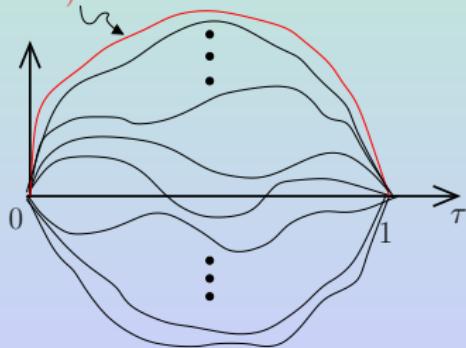
- Consequences for watermelons without wall for large N

$$\frac{x_N(\tau)}{\sqrt{2\tau(1-\tau)}} \sim \sqrt{2N} + \frac{N^{-1/6}}{\sqrt{2}} \chi_2$$

Proba[$\chi_2 \leq \xi$] = $\mathcal{F}_2(\xi)$, **Tracy-Widom distribution** for $\beta = 2$

- When $N \rightarrow \infty$, $x_N(\tau)$ reaches a deterministic elliptic shape

$$x_N(\tau) \sim 2\sqrt{N} \sqrt{\tau(1-\tau)}$$



Fluctuations

$$x_N(\tau = 1/2) - \sqrt{N} \sim N^{-\frac{1}{6}}$$

Outline

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- 2 Connection with stochastic growth models
- 3 Exact computation using Feynman-Kac formula
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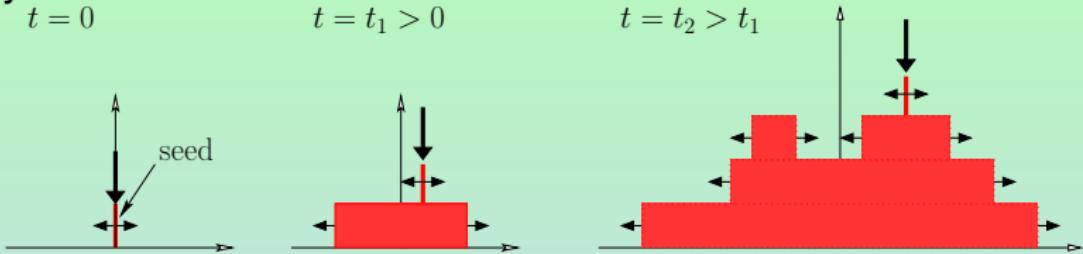
Curved growing interface : the **PNG** droplet

- **Polynuclear Growth Model**

$t = 0$

$t = t_1 > 0$

$t = t_2 > t_1$



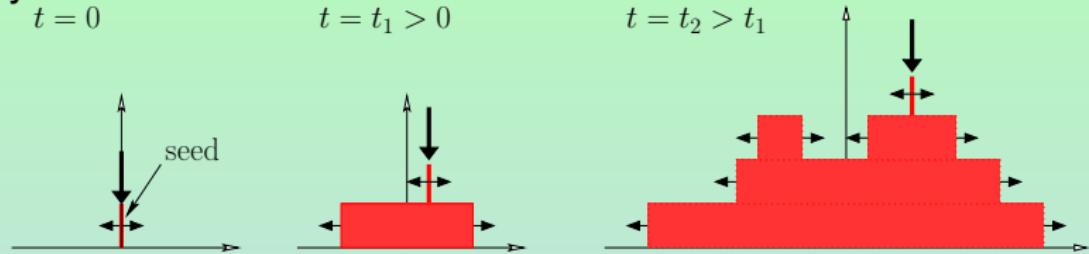
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- **Polynuclear Growth Model**

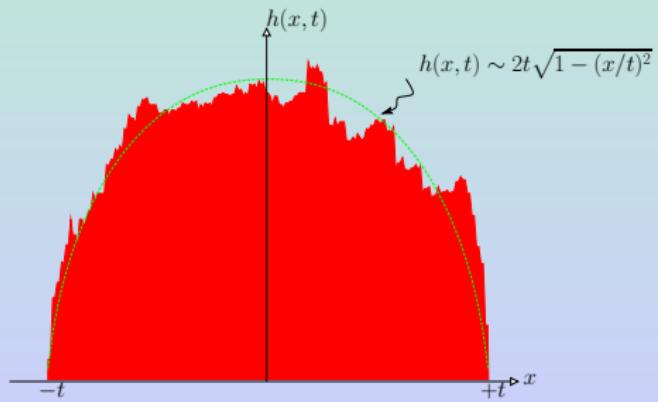
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- At large time t the profile becomes droplet-like



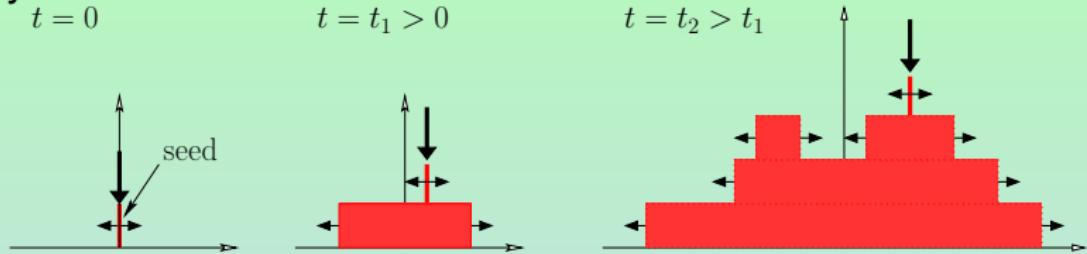
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- Polynuclear Growth Model

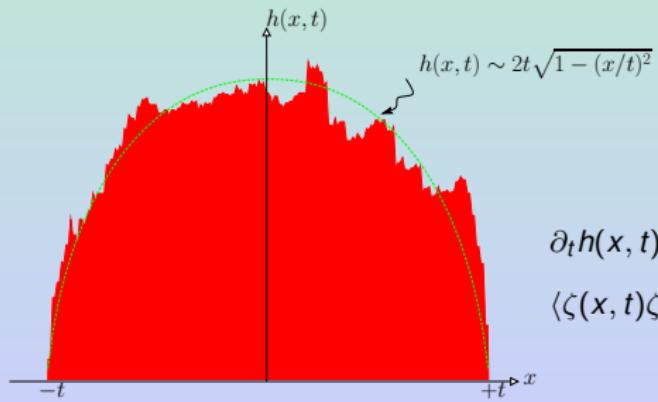
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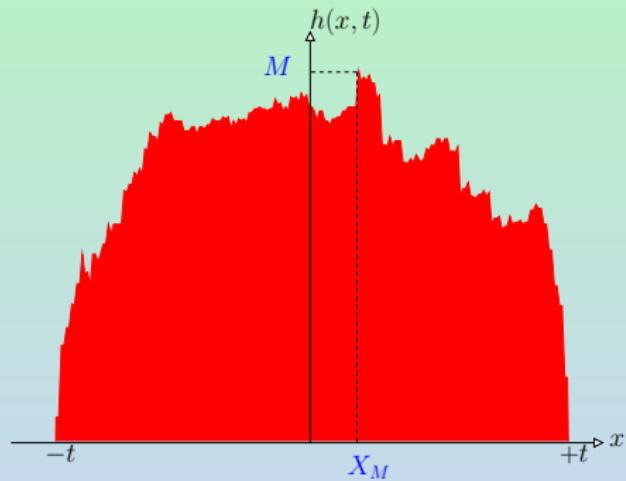


Fluctuations : KPZ
equation

$$\partial_t h(x, t) = \nu \nabla^2 h(x, t) + \frac{\lambda}{2} (\nabla h(x, t))^2 + \zeta(x, t)$$
$$\langle \zeta(x, t) \zeta(x', t') \rangle = D \delta(x - x') \delta(t - t')$$

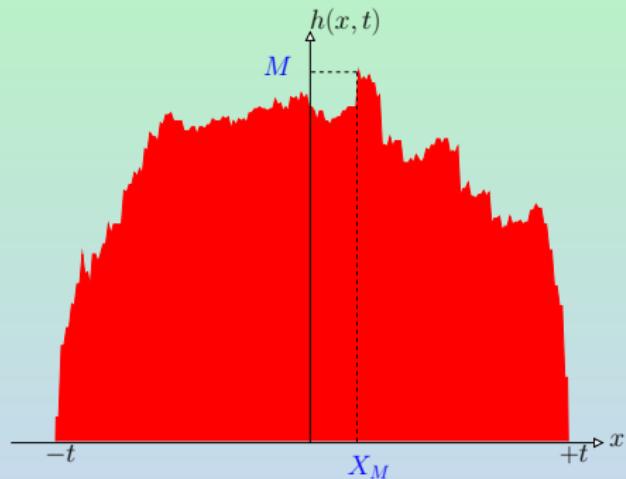
Curved growing interface : the PNG droplet

- Fluctuations : focus on **extreme statistics**



Curved growing interface : the PNG droplet

- Fluctuations : focus on **extreme statistics**



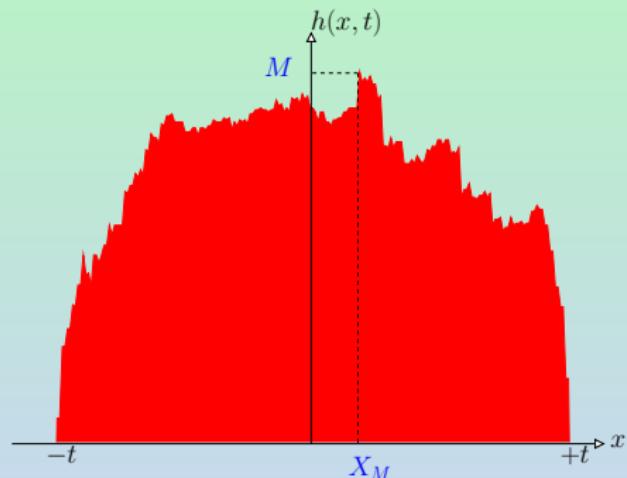
KPZ scaling

$$M - 2t \sim t^{1/3}$$

$$X_M \sim t^{2/3}$$

Curved growing interface : the **PNG** droplet

- Fluctuations : focus on **extreme statistics**



KPZ scaling

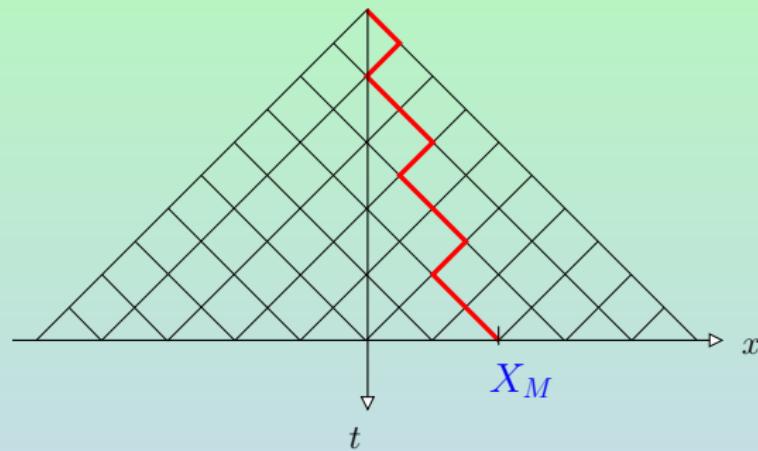
$$M - 2t \sim t^{1/3}$$

$$X_M \sim t^{2/3}$$

What is the joint distribution of M, X_M ?

Connection with the Directed Polymer (DPRM)

- DP in random media with one free end ("point to line")



$$E(\mathcal{C}) = \sum_{\langle i,j \rangle \in \mathcal{C}} \epsilon_{ij}$$

- $M \equiv -\text{Energy}$ of the optimal polymer
- $X_M \equiv \text{Transverse coordinate}$ of the optimal polymer

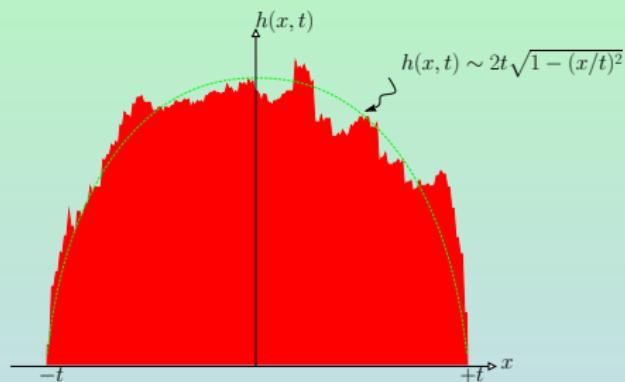
Vicious walkers and PNG droplet

watermelons

$$x_N(\tau) \sim 2\sqrt{N}\sqrt{\tau(1-\tau)}$$



PNG droplet



$$x_N \iff h$$

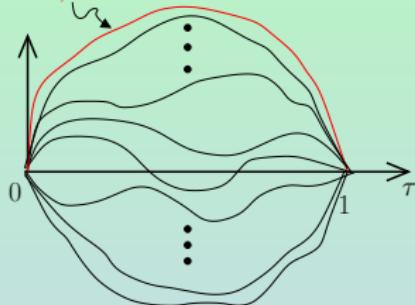
$$\tau \iff x$$

$$N \iff t$$

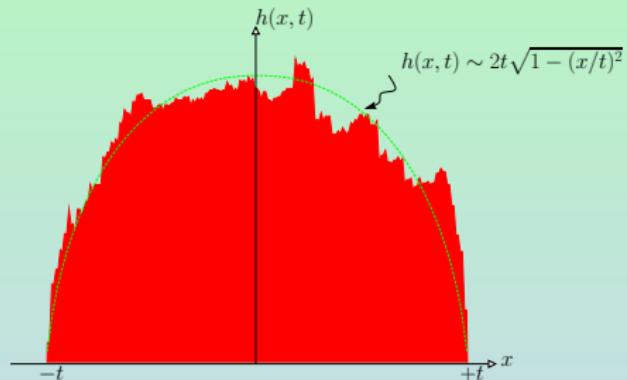
Vicious walkers and PNG droplet

watermelons

$$x_N(\tau) \sim 2\sqrt{N}\sqrt{\tau(1-\tau)}$$



PNG droplet



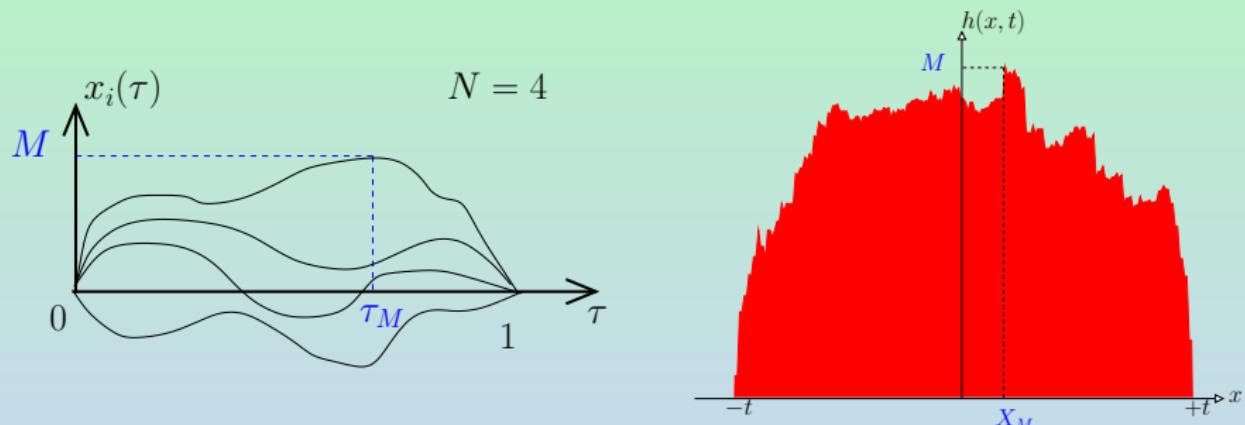
$$\frac{h(\textcolor{red}{u}t^{\frac{2}{3}}, t) - 2t}{t^{\frac{1}{3}}} \stackrel{d}{=} \frac{2 \left[x_N(\frac{1}{2} + \frac{\textcolor{red}{u}}{2} N^{-\frac{1}{3}}) - \sqrt{N} \right]}{N^{-\frac{1}{6}}} \stackrel{d}{=} \mathcal{A}_2(\textcolor{blue}{u}) - \textcolor{red}{u}^2$$

Prähofer & Spohn '00

$\mathcal{A}_2(u) \equiv \text{Airy}_2 \text{ process}$

Vicious walkers and PNG droplet

- Use this correspondence to study extreme statistics of PNG



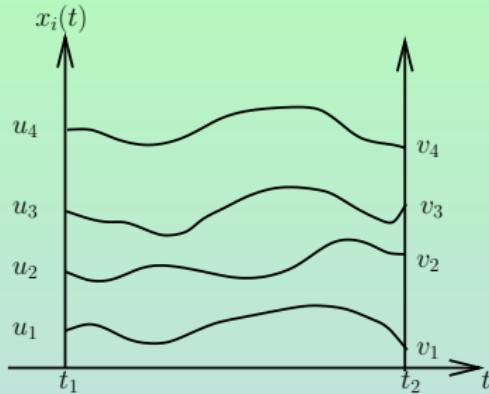
HERE: • exact computation of the distribution $P_N(M)$

- $P_N(M)$ in the $N \rightarrow \infty$ limit

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Transition probability



$$\mathcal{P}_N(v_1, v_2, \dots, v_N, t_2 | u_1, u_2, \dots, u_N, t_1) = ?$$

Karlin-Mc Gregor (1959), Lindström (1973),
Gessel-Viennot (1985)

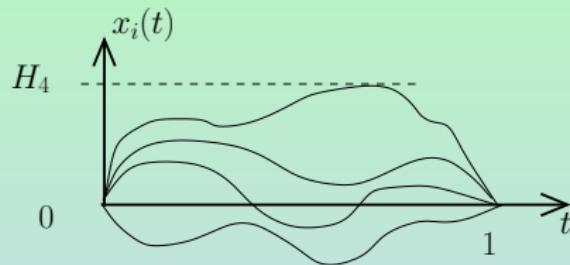
- The case $N = 2$: a reflection principle

$$\begin{aligned}\mathcal{P}_2(v_1, v_2, t_2 | u_1, u_2, t_1) &= \mathbf{P}_1(v_1, t_2 | u_1, t_1) \mathbf{P}_1(v_2, t_2 | u_2, t_1) \\ &\quad - \mathbf{P}_1(v_2, t_2 | u_1, t_1) \mathbf{P}_1(v_1, t_2 | u_2, t_1) \\ &= \mathbf{P}_2(\textcolor{green}{v_1}, \textcolor{red}{v_2}, t_2 | u_1, u_2, t_1) - \mathbf{P}_2(\textcolor{red}{v_2}, \textcolor{green}{v_1}, t_2 | u_1, u_2, t_1)\end{aligned}$$

where $\mathbf{P}_2(\cdot | \cdot)$ \equiv transition probability for two free particles (allowed to cross each other)

Watermelons configurations : regularization procedure

- Brownian motion has an infinite density of zero-crossings



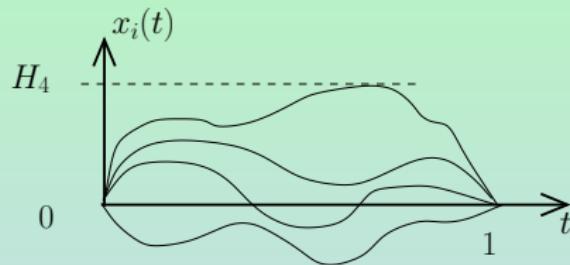
Such configurations are **ill-defined** for Brownian motion :

$$x_i(0) = x_{i+1}(0)$$

AND $x_i(t = 0^+) < x_{i+1}(t = 0^+)$

Watermelons configurations : regularization procedure

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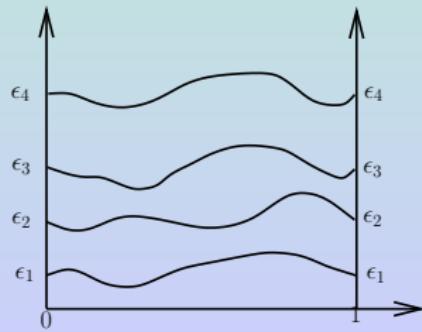


Such configurations are **ill-defined** for Brownian motion :

$$x_i(0) = x_{i+1}(0)$$

AND $x_i(t = 0^+) < x_{i+1}(t = 0^+)$

- A need for regularization : introduce cut-offs ϵ_i 's

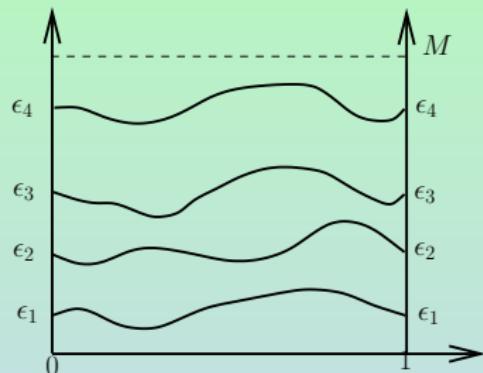


Only at the end take the limit
 $\epsilon_i \rightarrow 0$

Distribution of the maximal height : without wall

- Cumulative distribution of the maximal height

$$F_N(M) = \Pr [x_N(\tau) \leq M, \forall 0 \leq \tau \leq 1]$$



- Path integral (Feynman-Kac formula) for free fermions

$$F_N(M) = \frac{2^{-\binom{N}{2}}}{\prod_{j=0}^{N-1} j!} \det_{1 \leq i, j \leq N} \left[(-1)^{i-1} H_{i+j-2}(0) - H_{i+j-2}(\sqrt{2}M) e^{-2M^2} \right]$$

where $H_n(M) \equiv$ Hermite Polynomials

see also T. Feierl '08

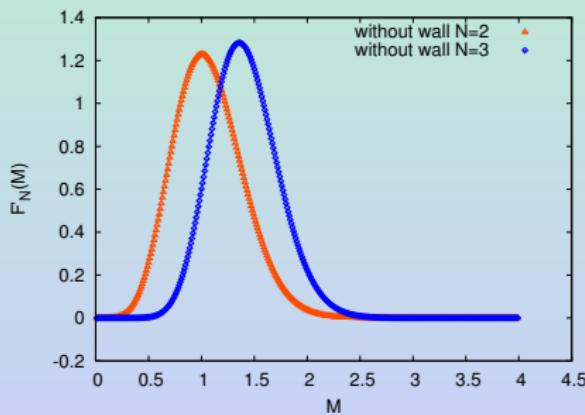
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- Shape and asymptotic behavior

$$F_N(M) \propto M^{N^2+N}, \quad M \rightarrow 0$$

$$1 - F_N(M) \sim e^{-2M^2}, \quad M \rightarrow \infty$$



Exact value of the $\langle H_N \rangle$

$$\langle H_1 \rangle = \frac{\sqrt{\pi}}{4} \sqrt{2}$$

$$\langle H_2 \rangle = \frac{\sqrt{\pi}}{4} (1 + \sqrt{2})$$

$$\langle H_3 \rangle = \frac{\sqrt{\pi}}{96} (45 + 36\sqrt{2} - 8\sqrt{6})$$

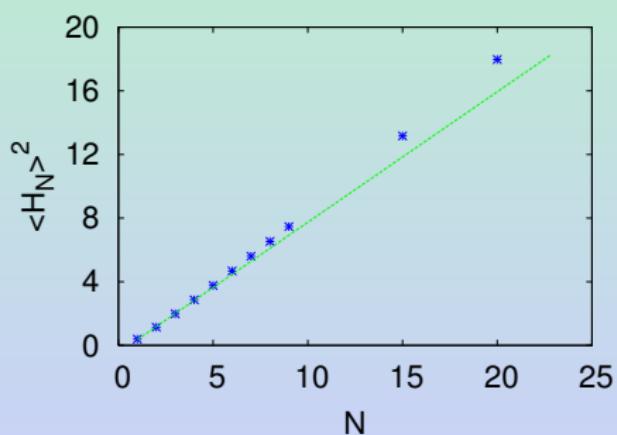
$$\langle H_4 \rangle = \frac{\sqrt{\pi}}{20736} (17091 + 5184\sqrt{2} - 1888\sqrt{6})$$

...

Comparison with numerics by Bonichon & Mosbah

Numerical estimate by Bonichon & Mosbah

$$\langle H_N \rangle_{\text{num}}^2 \sim 0.82N$$



Exact behavior at large N

$$\langle H_N \rangle^2 \sim N$$

Distribution of the maximal height : with a wall

- Cumulative distribution of the maximal height

G. S., S. N. Majumdar, A. Comtet, J. Randon-Furling '08

$$F_N(M) = \frac{A_N}{M^{2N^2+N}} \sum_{n_1, \dots, n_N=0}^{+\infty} \prod_{i=1}^N n_i^2 \prod_{1 \leq j < k \leq N} (n_j^2 - n_k^2)^2 e^{-\frac{\pi^2}{2M^2} \sum_{i=1}^N n_i^2}$$

$$A_N = \frac{\pi^{2N^2+N}}{2^{N^2-N/2} \prod_{j=0}^{N-1} \Gamma(2+j) \Gamma(\frac{3}{2}+j)} \text{ cf Selberg integral}$$

see also T. Feierl '08 and '12, N. Kobayashi *et al.* '08

Distribution of the maximal height : with a wall

- Cumulative distribution of the maximal height

G. S., S. N. Majumdar, A. Comtet, J. Randon-Furling '08

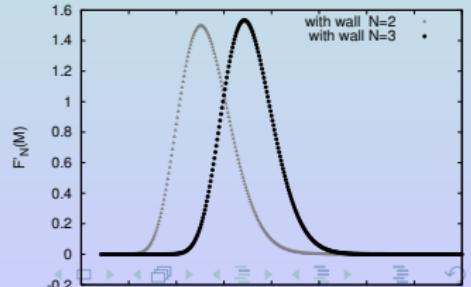
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see also T. Feierl '08 and '12, N. Kobayashi et al. '08

- Shape and asymptotic behavior

$$F_N(M) \sim \frac{\alpha_N}{M^{2N^2+N}} e^{-\frac{\pi^2}{12M^2} N(N+1)(2N+1)}, M \rightarrow 0$$

$$1 - F_N(M) \sim e^{-2M^2}, M \rightarrow \infty$$



Distribution of the maximal height : with a wall

- Cumulative distribution of the maximal height

G. S., S. N. Majumdar, A. Comtet, J. Randon-Furling '08

$$F_N(M) = \frac{A_N}{M^{2N^2+N}} \sum_{n_1, \dots, n_N=0}^{+\infty} \prod_{i=1}^N n_i^2 \prod_{1 \leq j < k \leq N} (n_j^2 - n_k^2)^2 e^{-\frac{\pi^2}{2M^2} \sum_{i=1}^N n_i^2}$$
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What about the asymptotic behavior of $F_N(M)$ for $N \rightarrow \infty$?

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Discrete orthogonal polynomials

- Large N analysis of

$$F_N(M) = \frac{A_N}{M^{2N^2+N}} \sum_{n_1, \dots, n_N=0}^{+\infty} \prod_{i=1}^N n_i^2 \prod_{1 \leq j < k \leq N} (n_j^2 - n_k^2)^2 e^{-\frac{\pi^2}{2M^2} \sum_{i=1}^N n_i^2}$$

- Discrete orthogonal polynomials

$$\sum_{n=-\infty}^{\infty} p_k(n) p_{k'}(n) e^{-\frac{\pi^2}{2M^2} n^2} = \delta_{k,k'} h_k ,$$
$$p_k(n) = n^k + \dots$$

- Useful expression for asymptotic analysis

$$F_N(M) = \frac{B_N}{M^{2N^2+N}} \prod_{k=1}^N h_{2k-1}$$

Large N limit for $F_N(M)$

- For large N , in the "double-scaling limit" P. J. Forrester, S. N. Majumdar, G. S. '11

$$\frac{d^2}{dt^2} \log F_N\left(\sqrt{2N}(1 + t/(2^{7/3}N^{2/3}))\right) = -\frac{1}{2}\left(q^2(t) - q'(t)\right)$$
$$q''(t) = 2q^3(t) + tq(t), \quad q(t) \sim \text{Ai}(t), \quad t \rightarrow \infty$$

i.e.

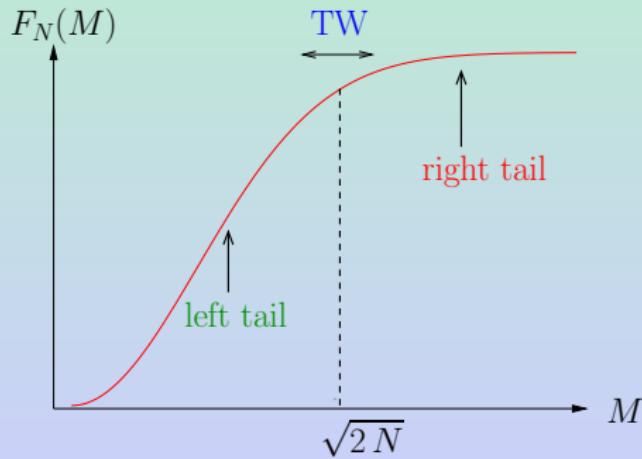
$$F_N(M) \rightarrow \mathcal{F}_1\left(2^{11/6}p^{1/6} \left|M - \sqrt{2N}\right|\right)$$
$$\mathcal{F}_1(t) = \exp\left(-\frac{1}{2} \int_t^\infty ((s-t)q^2(s) - q(s)) ds\right)$$
$$\equiv \text{Tracy-Widom distribution for GOE}$$

Large N limit for $F_N(M)$

$$F_N(M) \rightarrow \mathcal{F}_1 \left(2^{11/6} p^{1/6} \left| M - \sqrt{2N} \right| \right)$$

$$\mathcal{F}_1(t) = \exp \left(-\frac{1}{2} \int_t^\infty ((s-t) q^2(s) - q(s)) \, ds \right)$$

≡ Tracy-Widom distribution for GOE



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Conclusion

- Exact results for $\langle H_N \rangle$ for large N
- Exact result for distribution of the maximal height H_N using path integral techniques
- Connections with stochastic growth models for large N
- Large N limit

$$F_N(M) \rightarrow \mathcal{F}_1 \left(2^{11/6} N^{1/6} \left| M - \sqrt{2N} \right| \right), \quad N \rightarrow \infty$$

see also Liechty'11 for a recent rigorous proof