

# Dynamically Emergent Correlations

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**Happy b'day Peter!**



## Non-intersecting Brownian walkers and Yang–Mills theory on the sphere

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## Reunion Probability of $N$ Vicious Walkers: Typical and Large Fluctuations for Large $N$

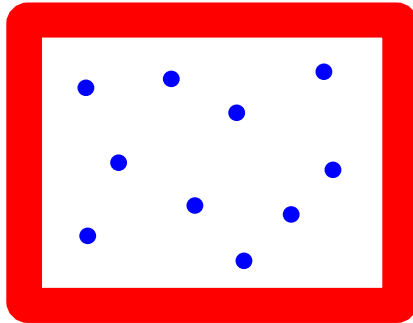
Grégory Schehr · Satya N. Majumdar · Alain Comtet ·  
Peter J. Forrester

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- **Dynamically emergent correlations** between independent particles via **stochastic** fluctuation of the environment  
⇒ **general setting**
- A simple model: **Switching** confining potential (box/trap)
- Exact **nonequilibrium stationary state (NESS)**  
⇒ **Conditionally independent and identically distributed (CIID)**
- Recent experiments using **optical tweezers**
- Generalisation to other models with **stochastically driven** environment
- Summary and Conclusion

# Dynamically **emergent** correlations: general setting

environment

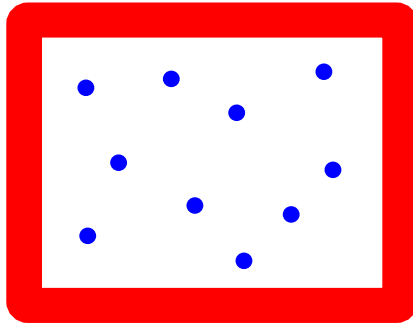


Particles: **non-interacting** (say Brownian)

Environment: **stochastic** (independent of particle motions)

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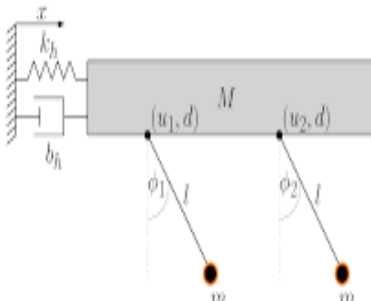
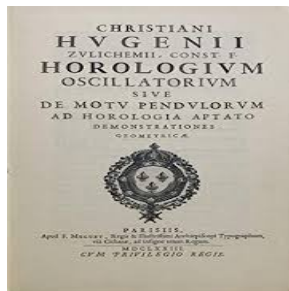
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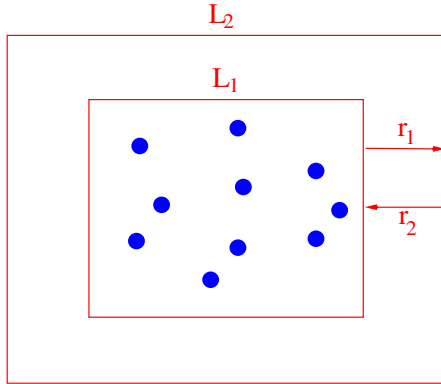
**Stochastic dynamics** of the **environment** induces **strong correlations** between particles

⇒ **Emergent correlations**

# Christiaan Huygens (1629-1695)



# A simple model: stochastically switching box size



**Particles**  $\Rightarrow$  **independent** Brownian with reflecting bc. at the walls

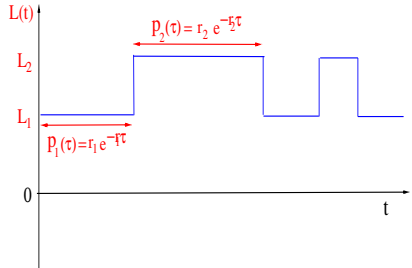
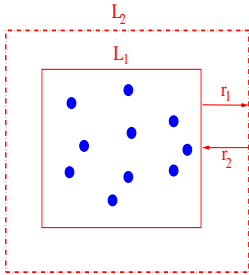
**Box size**  $\Rightarrow$  **stochastically** switching between two values  $L_1$  and  $L_2$

$$L_1 \xrightarrow{r_1} L_2$$

$$L_1 \xleftarrow{r_2} L_2$$

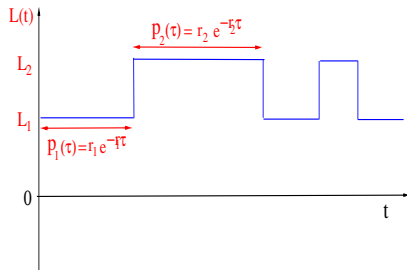
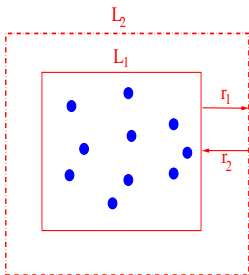


# A simple model: stochastically switching box size



Box size  $L(t) \Rightarrow$  a **dichotomous telegraphic** process

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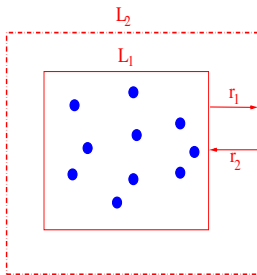
Box size  $L(t) \Rightarrow$  a **dichotomous telegraphic** process

**Goal:** to compute the joint PDF  $P(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N | t)$  at time  $t$

In particular, the stationary state (if it exists)

$$P^{\text{st}}(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N) = P(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N | \mathbf{t} \rightarrow \infty)$$

# An interesting limiting case

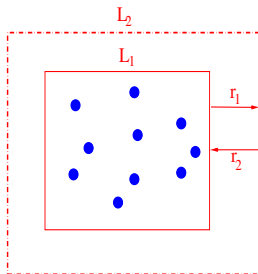


A limiting case:

$$L_1 = 0 \text{ and } L_2 \rightarrow \infty$$

$$r_1 \rightarrow \infty \text{ and } r_2 = \mathbf{r}$$

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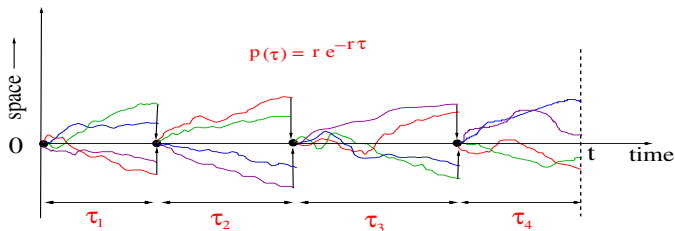


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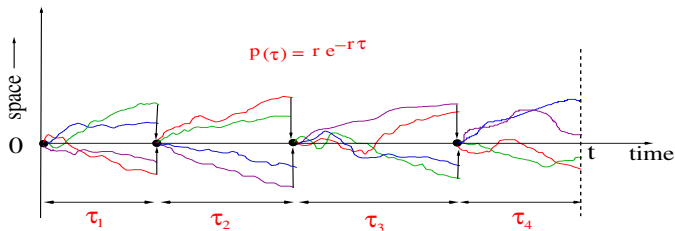
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$N$  independent Brownian motions  $\Rightarrow$  **simultaneously reset**  
(**instantaneously**) to the origin with rate  $\mathbf{r}$



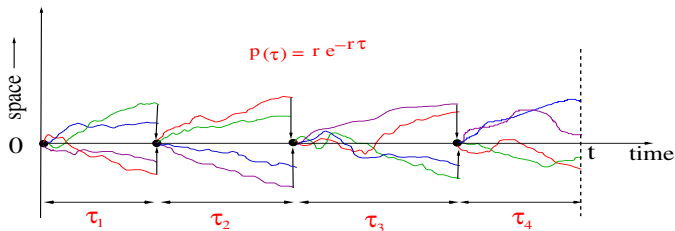
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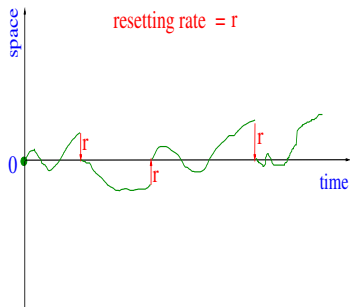


$\Rightarrow$  A **multiparticle** generalization of **stochastic resetting** of a  
single  **$N = 1$**  diffusing particle

[M.R. Evans & S.M., PRL, 106, 160601 (2011)]

## Single particle ( $N = 1$ )

# $N = 1 \Rightarrow$ a single Brownian particle in $d = 1$



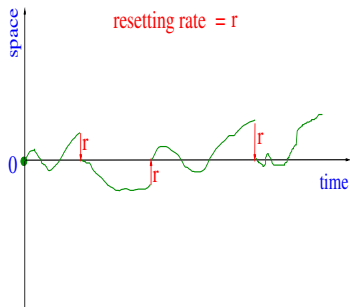
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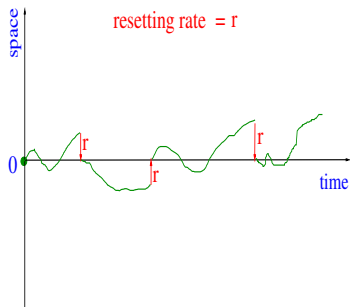
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**Dynamics:** In a small time interval  $\Delta t$

$$x(t + \Delta t) = 0 \quad \text{with prob. } r\Delta t \quad (\text{resetting})$$

$$= x(t) + \eta(t) \Delta t \quad \text{with prob. } 1 - r\Delta t \quad (\text{diffusion})$$

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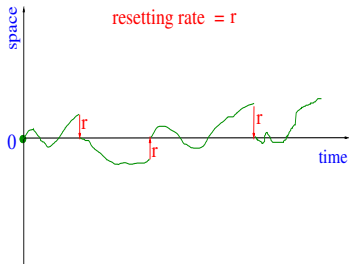
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$\eta(t) \rightarrow$  Gaussian white noise:  $\langle \eta(t) \rangle = 0$  and  $\langle \eta(t)\eta(t') \rangle = 2D\delta(t - t')$

[M.R. Evans & S.M., PRL, 106, 160601 (2011)]

# Prob. density $p_r(x, t)$ with resetting rate $r > 0$

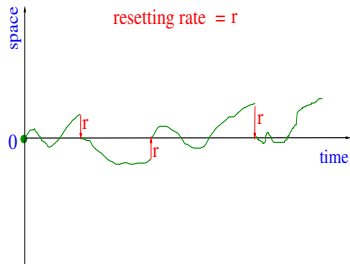


$p_r(x, t) \rightarrow$  prob. density at time  $t$ ,  
given  $p_r(x, 0) = \delta(x)$

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$$p_0(x, t) = \frac{1}{\sqrt{4\pi Dt}} \exp[-x^2/4Dt]$$

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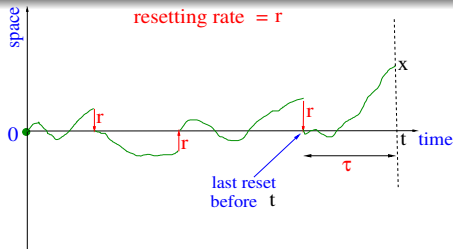
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- In the presence of resetting ( $r > 0$ ):

$$p_r(x, t) = ?$$

# Exact solution valid at all times $t$



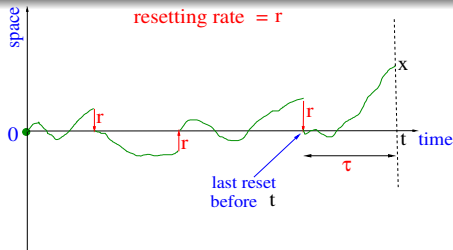
- Exact solution at all times  $t$ :

$$p_r(x|t) = e^{-rt} p_0(x|t) + \int_0^t d\tau (r e^{-r\tau}) p_0(x|\tau)$$

where  $p_0(x|\tau)$  = diffusion propagator =  $\frac{1}{\sqrt{4\pi D \tau}} \exp[-x^2/4D\tau]$

Renewal interpretation:  $\tau \rightarrow$  time since the last resetting during which  
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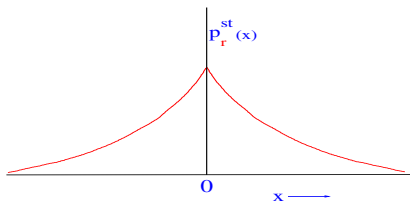
- As  $t \rightarrow \infty$ ,  $p_r^{\text{st}}(x) = r \int_0^\infty p_0(x|\tau) e^{-r\tau} d\tau = \frac{\alpha_0}{2} \exp[-\alpha_0 |x|]$   
where  $\alpha_0 = \sqrt{r/D}$

# Stationary State

Exact solution  $\rightarrow$   $p_r^{\text{st}}(x) = \frac{\alpha_0}{2} \exp[-\alpha_0 |x|]$  with  $\alpha_0 = \sqrt{r/D}$

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$\rightarrow$  nonequilibrium stationary state (NESS)

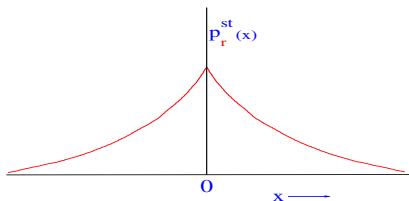
$\Rightarrow$  current carrying with detailed balance  $\rightarrow$  violated

$p_r^{\text{st}}(x) = \alpha_0 \exp[-V_{\text{eff}}(x)]$   
effective potential:  $\alpha_0|x|$



# Stationary State

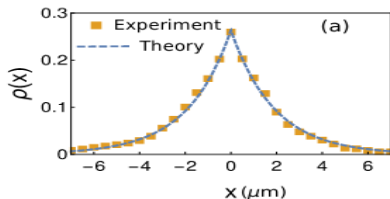
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Experimental verification using holographic optical tweezers

Tal-Friedman, Pal, Sekhon, Reuveni, & Roichman  
J. Phys. Chem. Lett. 11, 7350 (2020)

# Toy model $\Rightarrow$ explosion of activities

- Enzymatic reactions in biology ([Michaelis-Menten](#) reaction)
  - Diffusion in a confining potential/box
  - Lévy flights, Lévy walks, fractional BM with resetting
  - Space-time dependent resetting rate  $r(x, t)$
  - Search via nonequilibrium reset dynamics vs. equilibrium dynamics
  - Resetting dynamics of extended systems
  - Memory dependent reset
  - Quantum dynamics with reset
  - Active particles with reset
  - Cost of resetting
  - Optimization of random search algorithms
  - Optimal strategy for animal movements
- ...  $\Rightarrow$  a long list !

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Reviews: “Stochastic resetting and applications” ,

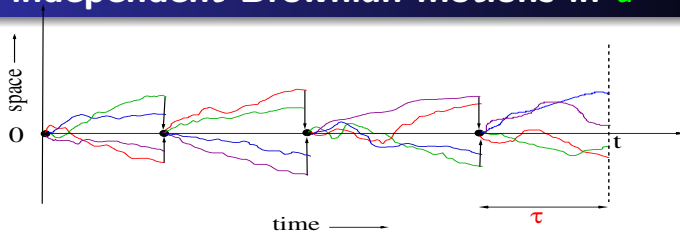
M.R. Evans, S.M., & G. Schehr, J. Phys. A : Math. Theor. 53, 193001 (2020)

“The inspection paradox in stochastic resetting” ,

A. Pal, S. Kostinski & S. Reuveni, J. Phys. A : Math. Theor. 55, 021001 (2022)

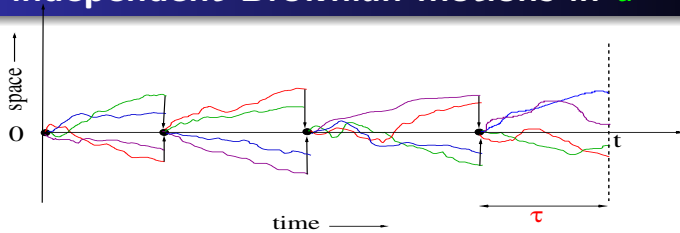
# Multiparticle Generalization ( $N > 1$ )

# $N > 1$ independent Brownian motions in $d = 1$



Consider  $N$  Brownian motions (**independent**) that are **simultaneously** reset with rate  $r$  to the origin

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Joint distribution at any time  $t$ :

$$P_r(\{x_i\}|t) = e^{-rt} \prod_{i=1}^N p_0(x_i|t) + r \int_0^t d\tau e^{-r\tau} \prod_{i=1}^N p_0(x_i|\tau)$$

$$\text{where } p_0(x|\tau) = \frac{1}{\sqrt{4\pi D\tau}} e^{-x^2/4D\tau}$$

M. Biroli, H. Larralde, S. M., G. Schehr, PRL, 130, 207101 (2023)

# Dynamically emergent correlations

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The joint distribution does not **factorize**  $\Rightarrow$  **correlated** resetting gas

In this model, **interactions** between particles are **not built-in**, but the correlations are generated by the dynamics (**simultaneous resetting**), that persist all the way to the stationary state

$\rightarrow$  **dynamically emergent correlations**

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Correlation function at time  $t$  for any pair  $i \neq j$ :

By symmetry,  $\langle x_i x_j \rangle - \langle x_i \rangle \langle x_j \rangle = 0$  at all times  $t$



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One needs to go to higher order to detect the correlations

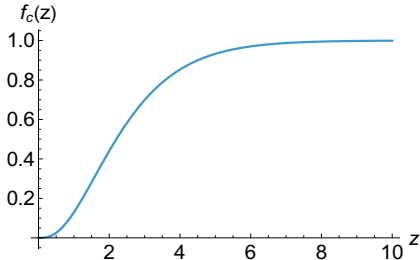
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Higher order correlation function for  $i \neq j$

$$C_{i,j}(t) = \langle x_i^2 x_j^2 \rangle - \langle x_i^2 \rangle \langle x_j^2 \rangle = \frac{4D^2}{r^2} f_c(r t)$$

where

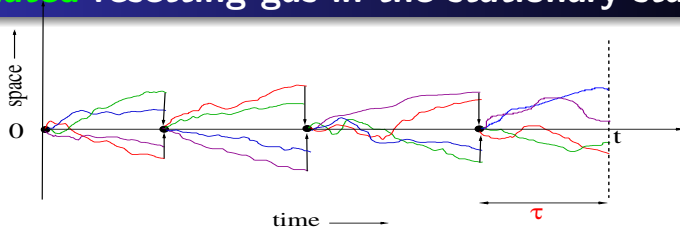
$$f_c(z) = 1 - 2z e^{-z} - e^{-2z}$$



$$f_c(z) \approx \begin{cases} \frac{1}{3} z^3 & z \rightarrow 0 \\ 1 - 2z e^{-z} & z \rightarrow \infty \end{cases}$$

correlations **grow with time**

# Correlated resetting gas in the stationary state



$N$  Brownian motions (**independent**) that are **simultaneously** reset with rate  $r$  to the origin

The joint position distribution approaches a **nonequilibrium stationary state** (**NESS**) at long times  $t \rightarrow \infty$

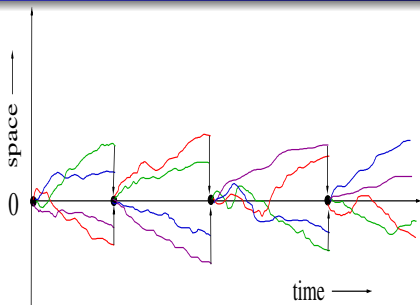
$$P_r^{\text{st}}(\{x_i\}) = r \int_0^\infty d\tau e^{-r\tau} \prod_{i=1}^N \frac{1}{\sqrt{4\pi D\tau}} e^{-x_i^2/4D\tau}$$

The joint distribution does not **factorize** even in the NESS

$$\langle x_i^2 x_j^2 \rangle - \langle x_i^2 \rangle \langle x_j^2 \rangle = 4 \frac{D^2}{r^2} \Rightarrow \text{attractive all-to-all interaction}$$

$\Rightarrow$  **strongly correlated** resetting gas

# Solvable Correlated Gas



Joint distribution:

$$P_r^{\text{st}}(\{x_i\}) = r \int_0^\infty d\tau e^{-r\tau} \prod_{i=1}^N p_0(x_i|\tau)$$

$$p_0(x|\tau) = \frac{1}{\sqrt{4\pi D\tau}} e^{-x^2/4D\tau}$$

The stationary joint distribution has a **CIID** structure  $\Rightarrow$  Solvable

$$P_r^{\text{st}}(x_1, x_2, \dots, x_N) = \int_{-\infty}^{\infty} du h(u) \prod_{i=1}^N p_0(x_i|u)$$

Here  $u \equiv \tau$  and  $h(u) \equiv r e^{-r u} \theta(u)$

**CIID**  $\Rightarrow$  Conditionally Independent and Identically Distributed

# Super-statistical ensembles in random matrix theory

Examples of **deformed** Gaussian random matrix ensembles:

$$P(\mathbf{X}) = \int_0^\infty du h(u) \exp \left[ -\frac{1}{2u} \text{Tr}(\mathbf{X}^2) \right]$$

The variance  $u \Rightarrow$  random variable drawn from  $h(u)$

Abdul-Magd (2005); Bohigas, Carvalho, Pato (2008);

Abdul-Magd, Akemann, Vivo (2009)

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However, no **microscopic dynamics** leading to these ensembles

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Despite the presence of **strong correlations**, several physical observables can be computed **exactly** in the **NESS** due to the **CIID** structure

- (1) Compute any observable for the **ideal** gas  $\Rightarrow$  **I.I.D** variables with distribution  $p_0(x|\tau)$  parametrized by  $\tau \Rightarrow$  **easy**
- (2) Average over the **random** parameter  $\tau$  using  $h(\tau) = r e^{-r\tau}$

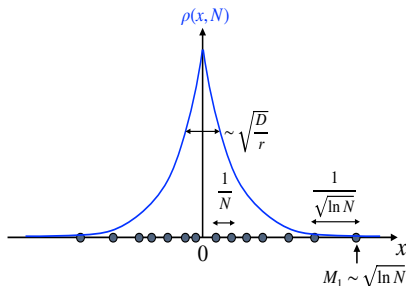


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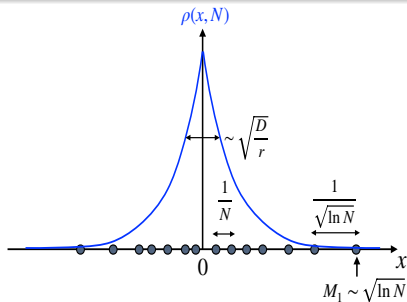


Examples:

- Average density
- Distribution of the  $k$ -th maximum:  
**Order statistics**
- Spacing distribution
- Full Counting Statistics

# Explicit Results

# Average Density



Joint distribution:

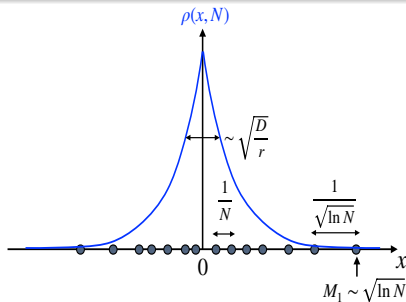
$$P_r^{\text{st}}(\{x_i\}) = r \int_0^\infty d\tau e^{-r\tau} \prod_{i=1}^N p_0(x_i|\tau)$$

$$p_0(x|\tau) = \frac{1}{\sqrt{4\pi D\tau}} e^{-x_i^2/4D\tau}$$

Average density:

$$\rho(x, N) = \frac{1}{N} \sum_{i=1}^N \langle \delta(x_i - x) \rangle = \int P_r^{\text{st}}(x, x_2, \dots, x_N) dx_2 dx_3 \dots dx_N$$

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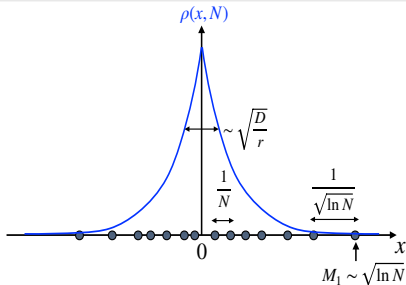
Average density:

$$\begin{aligned} \rho(x, N) &= \frac{1}{N} \sum_{i=1}^N \langle \delta(x_i - x) \rangle = \int P_r^{\text{st}}(x, x_2, \dots, x_N) dx_2 dx_3 \dots dx_N \\ &= r \int_0^\infty d\tau e^{-r\tau} p_0(x|\tau) = \frac{\alpha_0}{2} \exp[-\alpha_0 |x|] \end{aligned}$$

where  $\alpha_0 = \sqrt{r/D}$

$\Rightarrow$  same as the **single** particle position distribution

# Order Statistics



$M_k \Rightarrow k$ -th maximum

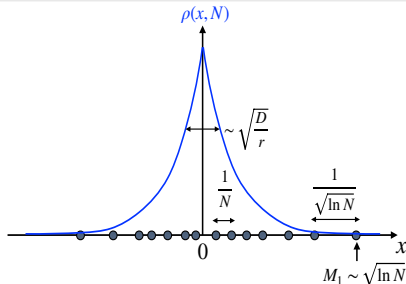
Set  $k = \alpha N$

$\alpha \sim O(1) \Rightarrow$  **bulk**

$\alpha \sim O(1/N) \Rightarrow$  **edge**

Symmetric around  $\alpha = 1/2$

# Order Statistics



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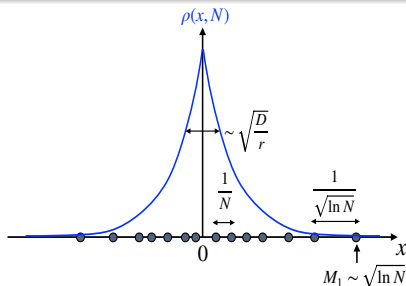
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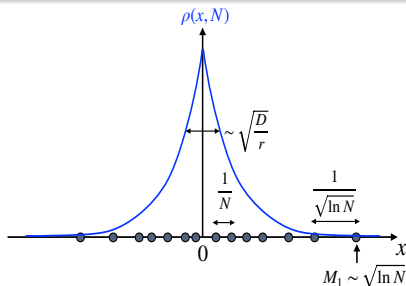
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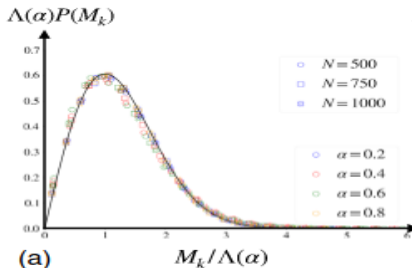
The scaling function

$$\mathbf{f(z) = 2z e^{-z^2} \theta(z)}$$

$\Rightarrow$  **universal** (indep. of  $\alpha \geq 1/2$ )



# Order Statistics



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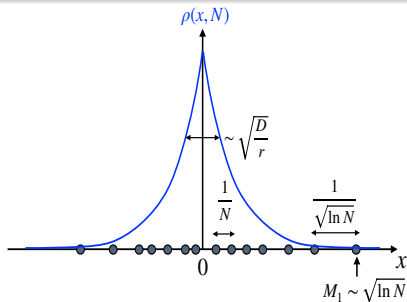
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M. Biroli, H. Larralde, S. M., G. Schehr, PRL, 130, 207101 (2023)

# Gap/Spacing Statistics



$M_k \Rightarrow k$ -th maximum

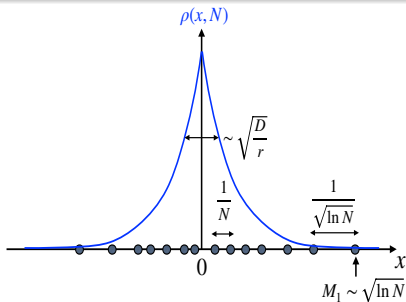
$k$ -th gap:  $d_k = M_k - M_{k+1}$

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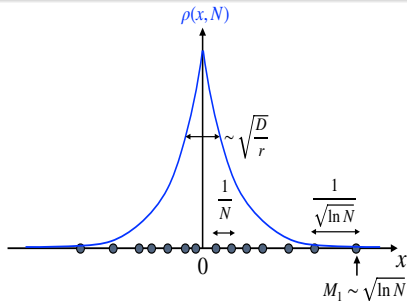
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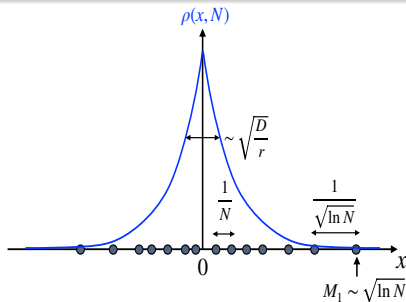
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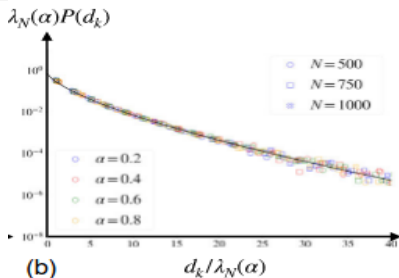
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The scaling function

$$h_{\text{gap}}(z) = 2 \int_0^\infty du e^{-u^2 - z/u} \quad (z \geq 0)$$

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# Gap/Spacing Statistics



The gap scaling function:

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$$\rightarrow \sqrt{\pi} \quad \text{as } z \rightarrow 0$$

$$\sim e^{-3(z/2)^{2/3}} \quad \text{as } z \rightarrow \infty$$

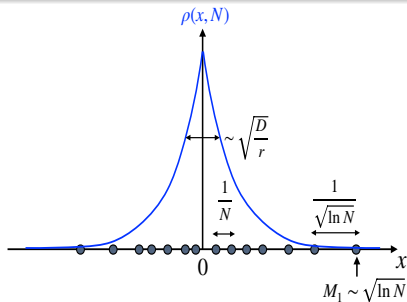
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# Full Counting Statistics

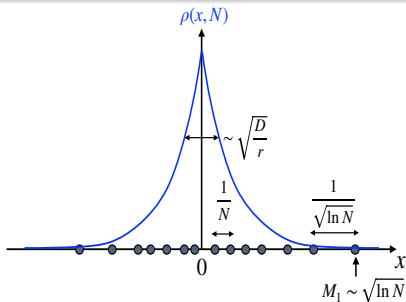


$N_L \Rightarrow$  number of particles in  $[-L, L]$

Clearly,  $0 \leq N_L \leq N$

$P(N_L, N) = ?$

# Full Counting Statistics



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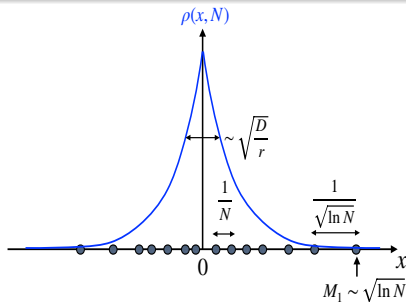
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Full Counting Statistics:  $P(N_L, N) \approx \frac{1}{N} H\left(\frac{N_L}{N} = \kappa\right) \quad (0 \leq \kappa \leq 1)$



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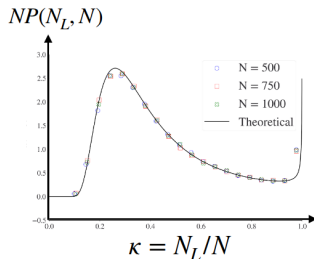
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where the scaling function:

$$H(\kappa) = \gamma \sqrt{\pi} [u(\kappa)]^{-3} \exp[-\gamma u^{-2}(\kappa) + u^2(\kappa)]$$

with  $\gamma = r L^2 / (4D)$  and  $u(\kappa) = \text{erf}^{-1}(\kappa)$

# Full Counting Statistics



The scaling function  $H(\kappa)$

$$H(\kappa) \rightarrow \frac{8\gamma}{\pi \kappa^3} \exp \left[ -\frac{4\gamma}{\pi \kappa^2} \right] \text{ as } \kappa \rightarrow 0$$

$$H(\kappa) \rightarrow \frac{\gamma \sqrt{\pi}}{(1-\kappa) [\ln(1-\kappa)]^{3/2}} \text{ as } \kappa \rightarrow 1$$

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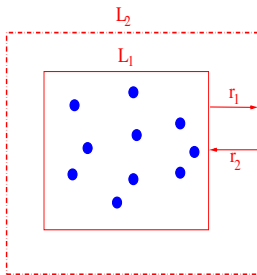
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M. Biroli, H. Larralde, S. M., G. Schehr, PRL, 130, 207101 (2023)

# Summary so far . . .

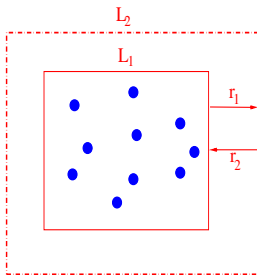


A limiting case:

$$L_1 = 0 \text{ and } L_2 \rightarrow \infty$$

$$r_1 \rightarrow \infty \text{ and } r_2 = \mathbf{r}$$

# Summary so far ...



A limiting case:

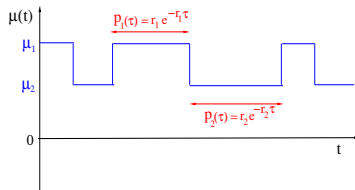
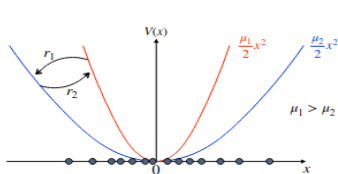
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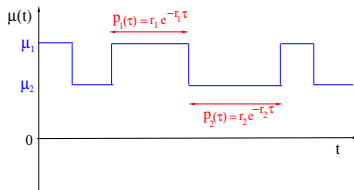
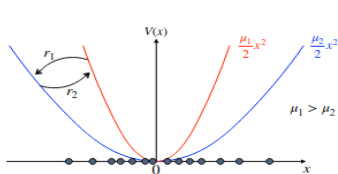
- This limiting case  $\Rightarrow$  **Exactly solvable** with a **strongly correlated NESS**
- The stationary joint distribution in the **NESS**  $\Rightarrow$  **CIID** structure  
 $\Rightarrow$  allows us to compute physical observables explicitly
- How generic is the **CIID** structure  $\Rightarrow$  going beyond this limiting case ?

# Switching Harmonic Trap

# $N$ Brownian particles in a **switching** harmonic trap



# N Brownian particles in a **switching** harmonic trap

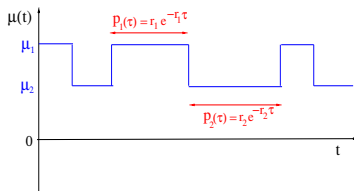
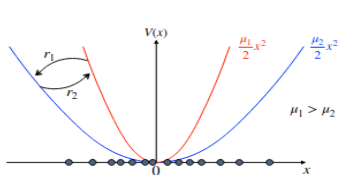


$$\frac{dx_i}{dt} = -\mu(t) x_i + \sqrt{2D} \eta_i(t)$$

$\eta_i(t) \longrightarrow$  Gaussian white noise with zero mean  
and correlator  $\langle \eta_i(t) \eta_j(t') \rangle = \delta_{i,j} \delta(t - t')$

The stiffness  $\mu(t)$  of the harmonic trap changes from  $\mu_1 \rightarrow \mu_2 < \mu_1$  with rate  $r_1$  and  $\mu_2 \rightarrow \mu_1$  with rate  $r_2 \implies$  **dichotomous** telegraphic noise

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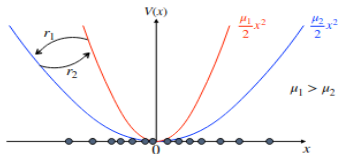
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$\implies$  drives the system into a **correlated NESS** with a stationary joint distribution  $P(x_1, x_2, \dots, x_N | t \rightarrow \infty) = P^{\text{st}}(\vec{x} | t \rightarrow \infty) = ?$

Biroli, Kulkarni, S.M., Schehr, PRE, 109, L032106 (2024)



# $N$ Brownian particles in a **switching** harmonic trap

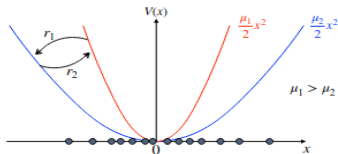


The limit

$$\mu_1 \rightarrow \infty, \mu_2 \rightarrow 0, \\ r_1 \rightarrow \infty \text{ and } r_2 = r$$

$\Rightarrow$  simultaneous resetting  
model

# $N$ Brownian particles in a **switching** harmonic trap



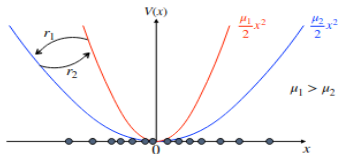
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$P_{1,2}(\vec{x}|t) \rightarrow$  Prob. that the position is  $\vec{x}$  and the stiffness is  $\mu_1$  (or  $\mu_2$ )  
at time  $t$

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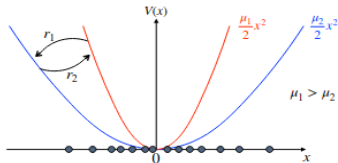
They satisfy a pair of **coupled Fokker-Planck** equations:

$$\partial_t P_1 = D \sum_{i=1}^N \partial_{x_i}^2 P_1 + \mu_1 \sum_{i=1}^N \partial_{x_i} (x_i P_1) - r_1 P_1 + r_2 P_2$$

$$\partial_t P_2 = D \sum_{i=1}^N \partial_{x_i}^2 P_2 + \mu_2 \sum_{i=1}^N \partial_{x_i} (x_i P_2) - r_2 P_2 + r_1 P_1$$

with initial conditions:  $P_1(\vec{x}|0) = \frac{1}{2} \delta(\vec{x})$  and  $P_2(\vec{x}|0) = \frac{1}{2} \delta(\vec{x})$

# Exact stationary solution



Fourier transforms:

$$\tilde{P}_{1,2}(\vec{k}|t) = \int P_{1,2}(\vec{x}|t) e^{i \vec{k} \cdot \vec{x}} d\vec{x}$$

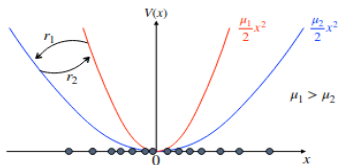
Rotational symmetry

$$\Rightarrow \tilde{P}_{1,2}(\vec{k}|t) = \tilde{P}_{1,2}(k|t)$$

where

$$k^2 = k_1^2 + k_2^2 + \dots + k_N^2$$

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Exact **stationary** solution in terms of  $R_1 = \frac{r_1}{2\mu_1}$  and  $R_2 = \frac{r_2}{2\mu_2}$

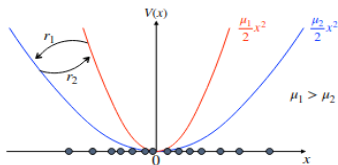
$$\tilde{P}_1^{\text{st}}(k) = \frac{r_2}{r_1+r_2} e^{-D k^2/(2\mu_1)} M\left(R_1, 1 + R_1 + R_2, -\frac{D k^2 (\mu_1 - \mu_2)}{2 \mu_1 \mu_2}\right)$$

$$\tilde{P}_2^{\text{st}}(k) = \frac{r_1}{r_1+r_2} e^{-D k^2/(2\mu_2)} M\left(R_2, 1 + R_1 + R_2, -\frac{D k^2 (\mu_2 - \mu_1)}{2 \mu_1 \mu_2}\right)$$

where  $M(a, b, z) = 1 + \frac{a}{b} z + \frac{a(a+1)}{b(b+1)} \frac{z^2}{2!} + \dots \rightarrow$  **Kummer's** function

$$R_1 = \frac{r_1}{2\mu_1} \text{ and } R_2 = \frac{r_2}{2\mu_2}$$

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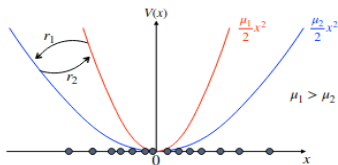
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The full solution:  $\tilde{P}^{\text{st}}(k) = \tilde{P}_1^{\text{st}}(k) + \tilde{P}_2^{\text{st}}(k)$

# Exact stationary solution



Fourier transforms:

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$$k^2 = k_1^2 + k_2^2 + \dots + k_N^2$$

Exact **stationary** solution in terms of  $R_1 = \frac{r_1}{2\mu_1}$  and  $R_2 = \frac{r_2}{2\mu_2}$

$$\tilde{P}_1^{\text{st}}(k) = \frac{r_2}{r_1+r_2} e^{-D k^2/(2\mu_1)} M\left(R_1, 1 + R_1 + R_2, -\frac{D k^2 (\mu_1 - \mu_2)}{2 \mu_1 \mu_2}\right)$$

$$\tilde{P}_2^{\text{st}}(k) = \frac{r_1}{r_1+r_2} e^{-D k^2/(2\mu_2)} M\left(R_2, 1 + R_1 + R_2, -\frac{D k^2 (\mu_2 - \mu_1)}{2 \mu_1 \mu_2}\right)$$

where  $M(a, b, z) = 1 + \frac{a}{b} z + \frac{a(a+1)}{b(b+1)} \frac{z^2}{2!} + \dots \rightarrow$  **Kummer's** function

$$R_1 = \frac{r_1}{2\mu_1} \text{ and } R_2 = \frac{r_2}{2\mu_2}$$

The full solution:  $\tilde{P}^{\text{st}}(k) = \tilde{P}_1^{\text{st}}(k) + \tilde{P}_2^{\text{st}}(k)$

Inverse Fourier transform  $P^{\text{st}}(\vec{x})$  has a **CIID** structure  $\rightarrow$  not **manifest**

# Fourier inversion

$$\tilde{P}_1^{\text{st}}(k) = \frac{r_2}{r_1+r_2} e^{-D k^2/(2\mu_1)} M\left(R_1, 1 + R_1 + R_2, -\frac{D k^2 (\mu_1 - \mu_2)}{2 \mu_1 \mu_2}\right)$$

$$\tilde{P}_2^{\text{st}}(k) = \frac{r_1}{r_1+r_2} e^{-D k^2/(2\mu_2)} M\left(R_2, 1 + R_1 + R_2, -\frac{D k^2 (\mu_2 - \mu_1)}{2 \mu_1 \mu_2}\right)$$



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Using the integral representation

$$M(a, b, -c k^2) = \frac{\Gamma(b)}{\Gamma(a)\Gamma(b-a)} \int_0^1 du u^{a-1} (1-u)^{b-a-1} e^{-c k^2 u}$$

One can invert  $\tilde{P}^{\text{st}}(k) = \tilde{P}_1^{\text{st}}(k) + \tilde{P}_2^{\text{st}}(k)$  explicitly  $\Rightarrow$

# 'Hidden' CIID structure of the stationary state

Inverting the Fourier transform one finds the CIID representation

$$P^{\text{st}}(\vec{x}) = \int_0^1 du h(u) \prod_{i=1}^N \frac{1}{\sqrt{2 V(u)}} e^{-x_i^2 / (2 V(u))}$$

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where

$$V(u) = D \left( \frac{u}{\mu_2} + \frac{1-u}{\mu_1} \right)$$

and

$$h(u) = A u^{R_1-1} (1-u)^{R_2-1} \left[ \frac{u}{\mu_2} + \frac{1-u}{\mu_1} \right]$$

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Since  $\int_0^1 h(u) du = 1$ , the function  $h(u)$  can be interpreted as the PDF of the random variable  $u \in [0, 1] \rightarrow$  the fraction of time each particle spends in  $\mu_2$  phase

Biroli, Kulkarni, S.M., Schehr, PRE, 109, L032106 (2024)

# All observables $\rightarrow$ exactly solvable

Using the explicit **CIID** structure of the stationary joint PDF

$$P^{\text{st}}(\vec{x}) = \int_0^1 du h(u) \prod_{i=1}^N p_0(x_i|u)$$

all observables in the correlated **NESS** can be computed explicitly and they exhibit rich and interesting behaviors

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$$P(M_1 = w, N) \rightarrow \frac{1}{\sqrt{\ln N}} f\left(\frac{w}{\sqrt{\ln N}}\right)$$

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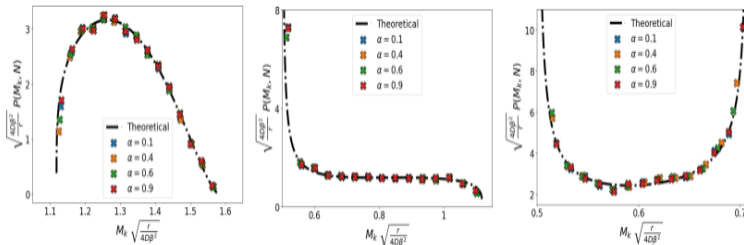
$$P(M_1 = w, N) \rightarrow \frac{1}{\sqrt{\ln N}} f\left(\frac{w}{\sqrt{\ln N}}\right)$$

where the exact scaling function (with  $R_1 = \frac{r_1}{2\mu_1}$  and  $R_2 = \frac{r_2}{2\mu_2}$ ):

$$f(z) = B z^3 \left(1 - \frac{z^2}{R_2}\right)^{R_2-1} \left(\frac{z^2}{R_1} - 1\right)^{R_1-1} \quad \text{with } \sqrt{R_1} \leq z \leq \sqrt{R_2}$$

$\rightarrow$  a new **extreme value distribution** of **strongly** correlated random variables with a **finite** support

# EVS with a finite support: **Universality**



The exact scaling function for the distribution of the scaled  $k$ -th maximum  $M_k$

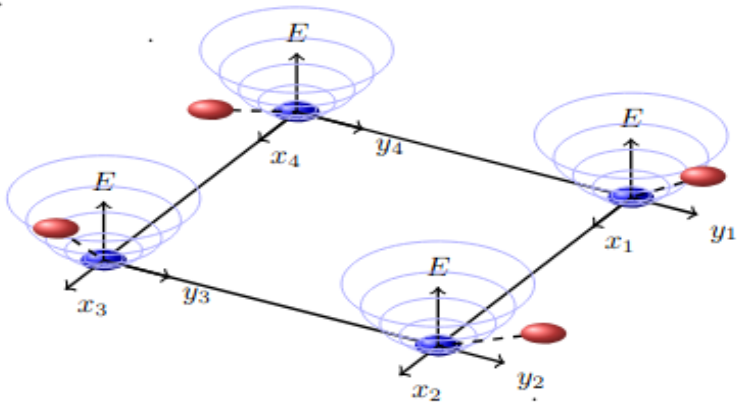
$$f(z) = B z^3 \left(1 - \frac{z^2}{R_2}\right)^{R_2-1} \left(\frac{z^2}{R_1} - 1\right)^{R_1-1} \text{ with } \sqrt{R_1} \leq z \leq \sqrt{R_2}$$

The scaling function  $f(z) \rightarrow$  **universal**, i.e., same for all  $M_k$ 's in  $d = 1$  and also for all  $\mathbf{d} \geq 1$



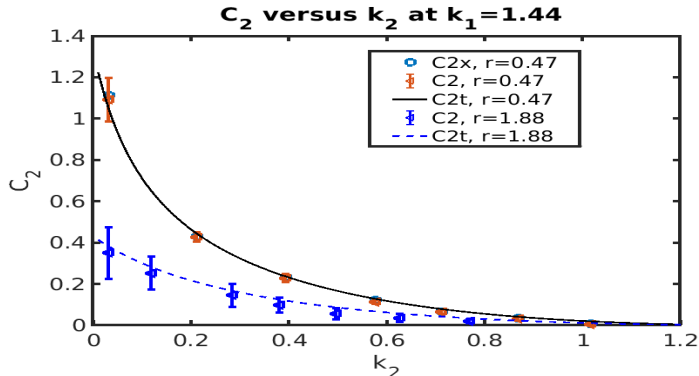
# Experimental results:

[set-up: S. Ciliberto]



- colloidal particles in **synchronized** harmonic traps
- particles are immersed in fluid  $\Rightarrow$  **long-range** hydrodynamic interaction (**neglected** in the theoretical model)

# Experimental results on $C_2$

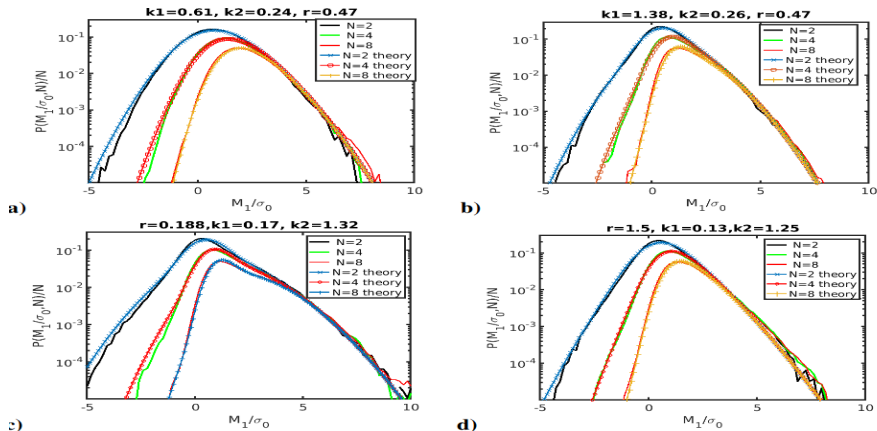


$$C_2 = \frac{\langle x_i^2 x_j^2 \rangle}{\langle x_i^2 \rangle \langle x_j^2 \rangle} - 1 \quad [\text{S. Ciliberto, unpublished data}]$$

Theoretical prediction [for  $r_1 = r_2 = r$  and with  $R_1 = \frac{r}{2\mu_1}$  and  $R_2 = \frac{r}{2\mu_2}$ ]:

$$C_2 = \frac{(\mu_2 - \mu_1)^2(2 + 3R_1 + 3R_2 + 4R_1R_2)}{(2 + R_1 + R_2)(2r + \mu_1 + \mu_2)^2}$$

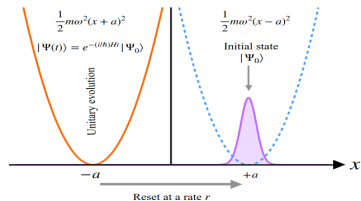
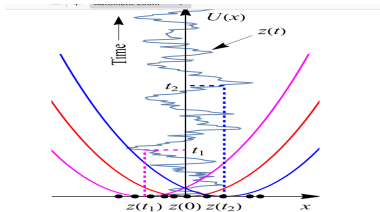
# Experimental results on $M_1$ :



Experiments with a finite number of colloidal particles in an optical trap

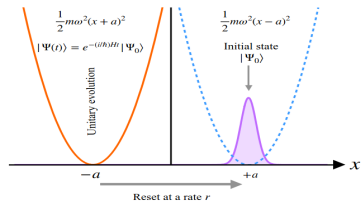
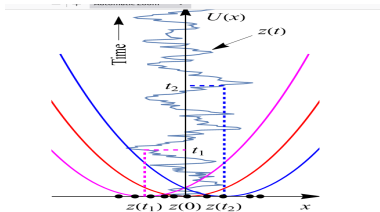
$\Rightarrow$  up to  $N = 8$  particles [S. Ciliberto, unpublished data]

# Two other models with CIID structure



$N$  noninteracting particles (bosons) in a harmonic trap

# Two other models with CIID structure

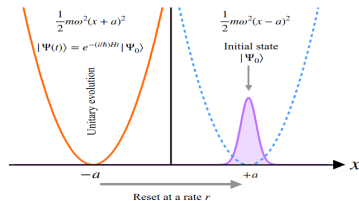
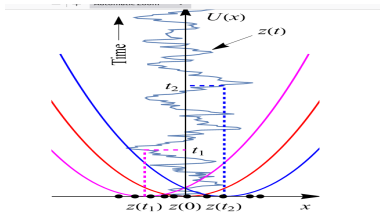


$N$  noninteracting particles (bosons) in a harmonic trap

(1) **Model 1 (Classical)**: The center of the harmonic trap performs a stochastic motion  $\Rightarrow$  drives the system into a **correlated NESS**

Sabhapandit & S.M. J. Phys. A.: Math. Theor. **57**, 335003 (2024)

# Two other models with CIID structure



$N$  noninteracting particles (bosons) in a harmonic trap

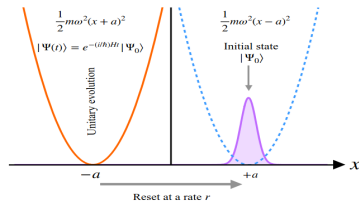
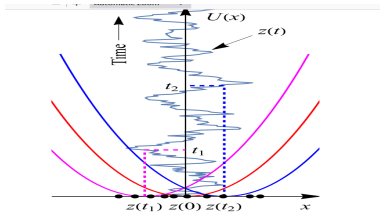
(1) **Model 1 (Classical)**: The center of the harmonic trap performs a stochastic motion  $\Rightarrow$  drives the system into a **correlated NESS**

Sabhapandit & S.M. J. Phys. A.: Math. Theor. **57**, 335003 (2024)

(2) **Model 2 (Quantum)**:  $N$  noninteracting bosons in the ground state of a harmonic trap whose center is **quenched** from  $+a$  to  $-a$ , evolves unitarily for a random time and then the state is **reset** to the ground state with center at  $+a$   $\Rightarrow$  drives the system into a **correlated NESS**

Kulkarni, S.M. & Sabhapandit, J. Phys. A: Math. Theor. **58**, 105003 (2025)

# Two other models with **CIID** structure



In both models, the NESS has the **CIID** (conditionally independent and identically distributed) structure

$$P^{\text{st}}(x_1, x_2, \dots, x_N) = \int_{-\infty}^{\infty} du h(u) \prod_{i=1}^N p_0(x_i | u)$$

This **CIID** structure makes the problem **solvable** for various observables such as average density, spacing distribution, extreme statistics, full counting statistics etc.

# Summary and Conclusion

- **Stochastic fluctuation** of the **common** environment  
⇒ **strong emerging** correlations between particles
- Exactly solvable example: **switching** harmonic trap
- The **NESS** has a **CIID** structure  
⇒ Several physical observables are exactly computable and have rich interesting behaviors, despite being a **strongly correlated** system
- Comparison with experiments on colloidal particles in an optical trap
- Easily generalisable to a whole new class of **solvable** correlated gases in their **nonequilibrium** stationary state → **ballistic** particles, **Lévy** flights, harmonic potential with a stochastic center, noninteracting bosons, ..  
⇒ all have this **CIID** structure ⇒ **Exactly solvable**



- Marco Biroli (LPTMS, Univ. Paris Saclay)
- Sergio Ciliberto (ENS, Lyon)
- Manas Kulkarni (ICTS, Bangalore)
- Hernan Larralde (UNAM, Mexico)
- Gabriele de Mauro (LPTMS, Orsay)
- Nikhil Mesquita (RRI, Bangalore)
- Sanjib Sabhapandit (RRI, Bangalore)
- Gregory Schehr (LPTHE, Univ. Sorbonne)

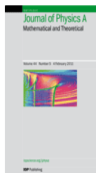
Earlier work on **Stochastic Resetting** with

- M. R. Evans (Edinburgh, UK)

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