Bath-Mediated Interactions between Driven Tracers in Dense Single-Files

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Single-file transport, where particles cannot bypass each other, has been observed in various experimental setups. In such systems, the behaviour of a tracer particle (TP) is subdiffusive, which originates from strong correlations between particles. These correlations are especially marked when the TP is driven and leads to inhomogeneous density profiles. Determining the impact of this inhomogeneity when several TPs are driven in the system is a key question, related to the general issue of bath-mediated interactions, which are known to induce collective motion and lead to the formation of clusters or lanes in a variety of systems. Quantifying this collective behaviour, the emerging interactions and their dependence on the amplitude of forces driving the TPs, remains a challenging but largely unresolved issue. Here, considering dense single-file systems, we analytically determine the entire dynamics of the correlations and reveal out of equilibrium cooperativity and competition effects between driven TPs.

The motion of particles in narrow channels in which particles cannot bypass each other is known as single-file diffusion. Such systems have been studied in various experimental setups with zeolites [1, 2], micro- [3, 4] and nano-channels [5], and simulations of carbon nanotubes [6]. Key features, on the theoretical side, involve the existence of a subdiffusive scaling for the mean position of a given particle [7, 8], and strong correlations between particles [9].

The basic phenomenology of the single-file transport is well-captured by the symmetric exclusion process (SEP). In this paradigmatic model of crowded equilibrium systems, particles perform symmetric random walks on a one-dimensional lattice with the constraint of at most a single occupancy of each lattice site. Different facets of the SEP have been scrutinised (see Refs. [8, 10–15]), including several important extensions to out-of-equilibrium situations. In particular, the mean displacement [16], as well as all higher-order cumulants [17] of an unbiased tagged particle (TP) placed initially at the shock point of a step-like density profile have been determined. Moreover, for a SEP with a single biased TP (due to either an energy consumption or an external force), the mean displacement of the latter [18, 19] and the higher-order cumulants in the dense limit [20] have been calculated, and shown to grow sublinearly as \( \sqrt{t} \). Here, the particles accumulate in front of the TP and are depleted behind it, which results in an inhomogeneous, non-stationary spatial distribution of particles.

A general open question concerns situations when several biased TPs are introduced in an otherwise quiescent medium of bath particles. The TPs are then expected to entrain the bath particles in a directional motion, which brings the system out-of-equilibrium and gives rise to effective bath-mediated interactions (BMIs) between the TPs. Such BMIs potentially lead to self-organisation, as observed in systems as diverse as colloidal solutions [21, 30], nearly-critical fluid mixtures [31], dusty complex plasmas [32], pedestrian counter flows [33].

Quantifying the emerging interactions between biased TPs and the ensuing collective behaviour is thus a key issue which however remains largely unresolved. Here, modeling a host medium as a dense SEP, we analytically determine the temporal evolution of all correlation functions, reveal intrinsically out-of-equilibrium cooperativity and competition effects between multiple TPs and quantify BMIs.

The quiescent host medium is modelled as a SEP which involves a high density \( \rho (\rho \rightarrow 1) \) of hard-core bath particles performing symmetric random walks (with unit jump rate) on a one-dimensional lattice. We then tag \( N \) particles at initial positions \( X_j^0 \) (see Figure 1). These TPs are biased: the \( j \)-th TP jumps to the left (resp. right) with probability \( 1-s_j/2 \) (resp. \( 1+s_j/2 \)). The bias \( s_j \in (-1,1) \) may either be due to an “activity” of the particle or to an external force \( f_j \), in which case one has the detailed balance condition: \( e^{\beta f_j} = \frac{1+s_j}{1-s_j} \), \( \beta \) being the reciprocal temperature.

We aim at determining the correlations between the TPs, embodied in the so called cumulant-generating function \( \psi(k,t) \equiv \ln \langle e^{ikY(t)} \rangle \), where \( k = (k_1, \ldots, k_N) \), \( Y = (Y_1, \ldots, Y_N) \) and \( Y_j(t) = X_j(t) - X_j^0 \) is the displace-
ment of the $j$-th TP. The cumulants are denoted by $\langle \bullet \rangle_c$ and defined by the expansion

$$
\psi(k, t) = \sum_{p_1, \ldots, p_N = 0}^{\infty} \frac{(ik_1)^{p_1} \cdots (ik_N)^{p_N}}{(p_1 + \cdots + p_N)!} (Y_1^{p_1} \cdots Y_N^{p_N})_c.
$$

(1)

In densely populated single-files ($\rho \to 1$), the dynamics of the system can be reformulated in terms of independent vacancies. In essence, this amounts to neglecting events where two vacancies interact simultaneously with any TP [20]. Our approach is based first on considering an auxiliary problem involving a single vacancy, initially at position $U$. By counting all the interactions of this vacancy with the TPs, we determine the probability $p_U(Y, t)$ that the TPs have displacements $Y$ at time $t$. Then, for a density of vacancies $\rho_0 = 1 - \rho \to 0$, the cumulant-generating function reads [35]

$$
\lim_{\rho_0 \to 0} \frac{\psi(k, t)}{\rho_0} = \sum_{U \in \{X^q\}} [\tilde{p}_U(k, t) - 1]
$$

(2)

where $\tilde{p}_U(k, t)$ is the Fourier transform of $p_U(Y, t)$ and can be expressed in terms of first-passage quantities of simple random walks with or without absorbing sites [36]. Analysis of the explicit expression of the cumulant-generating function for densely populated single-files [35] allows us to draw a number of important conclusions, which we present below.

**Bath-mediated binding.** As expected, at short times, the TPs move independently subject to their own biases. Our first finding is that, at large time and at high particle density, they are moving as a single TP. More precisely, in the large-time limit the $N$-TPs cumulants are given by $\langle Y_1^{q_1} \cdots Y_N^{q_N} \rangle_c = \langle Z^{q_1 + \cdots + q_N} \rangle_c$ for positive integer $q_j$, where $Z = \sum_{j=1}^N Y_j/N$ is the displacement of the center of mass. At high density, even and odd cumulants of $Z$ satisfy

$$
\langle Z(t)^{2n} \rangle_c = \langle Z(t)^{2n+1} \rangle_c = \frac{2t}{\pi}.
$$

(3)

where $S = \tanh(\beta F/2)$ is the effective bias, $F = \sum_{j=1}^N f_j$ the effective force and the ratio $\langle \bullet \rangle / \rho_0$ is understood as the limit $\rho_0 \to 0$. Equations (3) implies in particular that for any number of TPs and arbitrary forces $\langle Y_j \rangle = \langle Z \rangle$, meaning that at large time all the TPs move like their center of mass with an effective force $F$ (in agreement with the hydrodynamic analysis of Ref. [37]).

In the following, we determine the full dynamics of the correlations between TPs, and focus for simplicity reasons on the case of two TPs.

**Bath-mediated entrainment.** We examine the case of a single biased TP ($s_2 \neq 0$) followed by an unbiased TP ($s_1 = 0$), initially separated by a distance $L$ (see Figure 2 top), which allows us to quantify the perturbation induced by a biased tracer in a quiescent medium. While the behaviour of $\langle Y_2 \rangle$ is known [20], we unveil an interesting scaling behaviour of $\langle Y_1 \rangle$ beyond the large-time regime. Indeed, in the limit $t \to \infty$ with $t/L^2$ constant, one finds

$$
\frac{\langle Y_2(t) \rangle}{\rho_0} = s_2 \sqrt{\frac{2t}{\pi}}, \quad \frac{\langle Y_1(t) \rangle}{\rho_0} = s_2 \sqrt{\frac{2t}{\pi}} g \left( \frac{L}{\sqrt{2t}} \right),
$$

(4)

$$
g(u) = e^{-u^2} - \sqrt{\pi} u \operatorname{erfc}(u).
$$

(5)

This provides the dynamics of the entrainment of the TP1 by TP2, which admits a typical time scale $L^2$ and leads to the final bound state discussed above (see Fig. 2).

The evolution towards the final regime can further be quantified by the dynamics of the two TPs even (resp. odd) cumulants $\kappa^e = \langle Y_1^p Y_2^q \rangle_c$ (resp. $\kappa^o = \langle Y_1^p Y_2^q \rangle_c$)
with \( p + q \) even (resp. odd), \( p, q \geq 1 \). They obey

\[
\frac{\kappa^c}{\rho_0} = \frac{\kappa^o}{\rho_0 s_2} = \sqrt{\frac{2t}{\pi}} g \left( \frac{L}{\sqrt{2t}} \right). \tag{6}
\]

Several comments are in order. (i) Equations (4) and (5) are similar to the expressions found for the random average process \([38, 39]\), which points towards their universality. (ii) The same scaling function \( g \) is involved in the expressions of \( \langle Y_1(t) \rangle \) and \( \langle Y_1 Y_2(t) \rangle \) (Eqs. (4) and (5)). This leads to the generalized fluctuation-dissipation relation

\[
\lim_{f_2 \to 0} \frac{2}{f_2} \langle Y_1 (f_1 = 0, f_2) \rangle = \langle Y_1 Y_2 \rangle (f_1 = f_2 = 0). \tag{7}
\]

Note that this relation holds in the opposite limit of a dilute (\( \rho \to 0 \)) SEP \([40]\). (iii) Our approach provides the time dependence of all cumulants of individual particles and the law of the distance between TPs (insets of Fig. 2) and \([35]\). The time, initial distance between TPs, and driving force dependences from numerical simulations are unambiguously captured by our theoretical expressions (Fig. 2).

**Both-mediated cooperatively and competition.** We now turn to the general case in which both TPs are biased (see Figure 3, top). The dynamics of effective interactions between TPs at the level of averages is conveniently analysed by introducing the rescaled instantaneous velocities

\[
A_j(t) = \frac{\sqrt{2\pi t \rho_0}}{\rho_0} \frac{d\langle Y_j \rangle}{dt}, \tag{8}
\]

which satisfy \( A_j(t) = s_j \) at small time and \( A_j(t) = S = \frac{s_1 + s_2}{1 + s_1 s_2} \), at large time. The full time dependency is found \([35]\) to be given by \( A_1 = H_{s_1, s_2(1+s_1), -s_1 s_2} (L/\sqrt{2t}) \) and \( A_2 = H_{s_2, s_1(1-s_2), s_1 s_2} (L/\sqrt{2t}) \) with

\[
H_{\beta_0, \beta_1, \beta_2} (u) = \sum_{n=0}^{\infty} (-s_1 s_2)^n \sum_{m=0}^{2 \beta_m e^{-[(2n+m)u]^2}} (9)
\]
while higher order cumulants follow

\[
\frac{\kappa^c}{\rho_0} = \frac{\kappa^o}{\rho_0 S} = \sqrt{\frac{2t}{\pi}} G_{s_1 s_2} \left( \frac{L}{\sqrt{2t}} \right),
\]

\[
G_{s}(u) = (1 + \sigma) \sum_{n=0}^{\infty} (-\sigma)^n g([2n + 1]u).
\]

These results fully quantify the dynamics of the BMIs between two biased TPs and reveal striking behaviors. (i) In the case of same sign biases, the TPs cooperate [37] (Fig. 3 left). At large times a pair of biased TPs moves faster than a single TP in agreement with the asymptotic result Eq. (3). Note that such an accelerated dynamics has been numerically observed in two-dimensional systems [22-24]. At intermediate times, we unveil an overshoot of the rescaled velocity of the trailing TP. (ii) When the biases act in opposite directions (say \(0 < -s_1 < s_2\)), each TP starts to move in the direction of its own force, and eventually both TPs move in the direction of the largest force (Fig. 3 left). This competing stage can be quantified from Eq. (9) by determining the U-turn time \(t^*\) at which the velocity of TP1 changes its sign. This time \(t^*\) vanishes when \(s_1\) is small, and diverges when \(s_1\) is close to \(-s_2\) according to the scaling laws

\[
\frac{t^*}{L^2} \sim 0, \quad \frac{1}{2\log(-s_1/s_1)}; \quad \frac{t^*}{L^2} \sim \frac{\gamma}{1 + s_1/s_2},
\]

with \(\gamma = 2(1 + s_2)^2/(1 - s_2)\). Figure 3 shows an excellent quantitative agreement between the analytical predictions and the numerical simulations.

Our approach can be extended to determine the dynamics of correlations in the case of an arbitrary number of driven TPs. Cooperativity and competition involve a complex cascade of time scales associated with the initial distances between TPs, fully captured by our approach, as exemplified in the case of 3 TPs in Figure 4 and 35.

**Bath-mediated interactions.** The BMIs between two bound biased TPs can be further analysed by associating the probability distribution of the variation of distance \(D = Y_2 - Y_1\) to an effective potential \(U(D)\) via \(P(D) = \exp(-\beta U(D))\) [11]. For identical forces \((f_1 = f_2 = f)\) and \(D\) sufficiently close to its average value, the two TPs are effectively bound by a harmonic potential

\[
U(D) \sim \frac{\kappa}{2} (D - \langle D \rangle)^2, \quad \langle D \rangle = \frac{1}{\cosh(\beta f) - 1},
\]

where the constant \(\kappa\) is explicitly given by

\[
\kappa = \frac{\cosh(\beta f)}{\beta \rho_0 L (1 + \cosh(\beta f))}.
\]

In the regime, \(D \gg \langle D \rangle\), the potential displays a weaker dependence on the distance \(U(D) \sim D(\ln D + \nu)/\beta\) (see 35 for the value of \(\nu\)). We remark that this qualitative change of regimes has been observed in 2d in numerical simulations of two biased TPs in a quiescent colloidal bath [24].

Altough, we determined the full dynamics of correlation functions in a paradigmatic model of non equilibrium statistical physics and entirely characterized the corresponding bath-mediated interactions.
See appendices for details of the numerical simulations and detailed calculations.

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Appendix A: Details of the calculations

1. Model and approximation

a. Model

Our model is based on the well known Symetric Exclusion Principle (SEP). Let us consider particles on a discrete one-dimensional line (the sites correspond to the integers). The mean density is denoted \( \rho \in [0,1] \). The particles follow a symmetric random walk with hard-core exclusions in continuous time. The rate of jumps of these particles to the left / right is \( 1/2 \). In addition to these particles, we consider two biased tagged particles (TPs), initially at positions \( X_1^0 = 0 \) and \( X_2^0 = L \). They can jump to the right with probability \( p_j = (1 + s_j)/2 \) and to the left with probability \( p_{-j} = (1 - s_j)/2 \), where \( s_1 \) and \( s_2 \) are the biases on the two TPs. See Fig. 1 on the article for a sketch of the system.

For simplicity reasons, we choose to present only the computations for two TPs. The treatment of three or more TPs is very similar but it leads to heavier computations. Results for three TPs will be presented in the next section.

b. Approximation and consequences

We focus on the limit of a dense system (\( \rho \to 1 \)) and we follow the evolution of the vacancies as was done in Ref. [20]. In this limit of small number of vacancies, one can approximate the motion of the TPs as being generated by the vacancies interacting independently (i.e. not simultaneously) with them: in the large density limit the events corresponding to two vacancies interacting simultaneously with the TPs happen only with negligible probability.

Let us consider a system of size \( N \) with \( M \) vacancies and denote by \( Y(t) = (X_1(t) - X_1^0, X_2(t) - X_2^0) \) the vector of the displacements of the two tracers. The probability \( P(t)(Y|\{Z_j\}) \) of having displacements \( Y \) at time \( t \) knowing that the \( M \) vacancies started at sites \( Z_1 \ldots Z_M \) is exactly given by:

\[
P(t)(Y|\{Z_j\}) = \sum_{Y_1, \ldots, Y_M} \delta_{Y,Y_1+\ldots+Y_M} P(t)(\{Y_j\}|\{Z_j\})
\]

(S1)

where \( P(t)(\{Y_j\}|\{Z_j\}) \) is the probability of displacements \( Y_j \) due to the vacancy \( j \), for all \( j \), knowing the initial positions of all the vacancies.

Making the approximation that we described above, we can link it to the probability \( p_{Z}(t)(Y) \) that the tracers have moved by \( Y \) at time \( t \) due to a single vacancy that was initially at site \( Z \):

\[
P(t)(\{Y_j\}|\{Z_j\}) \sim_{\rho \to 1} \prod_{j=1}^{M} p_{Z_j}(t)(Y_j)
\]

(S2)

We then have :

\[
P(t)(Y|\{Z_j\}) \sim_{\rho \to 1} \sum_{Y_1, \ldots, Y_M} \delta_{Y,Y_1+\ldots+Y_M} \prod_{j=1}^{M} p_{Z_j}(t)(Y_j)
\]

(S3)

We take the Fourier transform and we average over the initial positions of the vacancies:

\[
\tilde{p}^{(t)}(k) = \frac{1}{N-2} \sum_{Z \neq 0,L} \sum_{Y} p^{(t)}(Y) e^{ikY}
\]

(S4)

and mutatis mutandis for \( \tilde{P}^{(t)}(k) \). We obtain

\[
\tilde{P}^{(t)}(k) \sim_{\rho \to 1} \left[ \tilde{p}^{(t)}(k) \right]^M = \left[ \frac{1}{N-2} \sum_{Z \neq 0,L} \tilde{p}^{(t)}(k) \right]^M.
\]

(S5)

Furthermore we write \( \tilde{p}_{Z}^{(t)}(k) = 1 + \tilde{q}_{Z}^{(t)}(k) \) (\( q_Z \) corresponds to the deviation from a Dirac centered in 0) to get

\[
\tilde{P}^{(t)}(k) \sim_{\rho \to 1} \left[ 1 + \frac{1}{N-2} \sum_{Z \neq 0,L} \tilde{q}_{Z}^{(t)}(k) \right]^M.
\]

(S6)
We now take the limit $N,M \to \infty$ with $\rho_0 \equiv 1 - \rho = M/N$ (density of vacancies) remaining constant. The second characteristic function reads
\[
\psi^{(t)}(k) \equiv \ln \left[ \hat{P}^{(t)}(k) \right] \approx \rho_0 \sum_{Z \neq 0,L} \hat{q}^{(t)}_Z(k).
\] (S7)

$\psi$ gives all the cumulants: the cumulant associated to $Y_1^q Y_2^r$ is $(Y_1^q Y_2^r) = i^{-(q+r)} \frac{\partial^{q+r}}{\partial q \partial r} \psi |_{k=0}$.

Our goal is now to find an expression for $\psi^{(t)}(k)$.

2. Resolution

The approximation (S3) that leads to (S6) tells us that the case of a vanishing density of vacancies can be deduced from the case of a single vacancy. We first focus on the latter before using (S7) to get the cumulants.

a. Single vacancy

We consider the situation with a single vacancy. In the following we shall denote by $\nu = -1, +1, -2, +2$ the “special sites” respectively to the left of TP1, to the right of TP1, to the left of TP2, to the right of TP2. We use the vectors $e_{\pm 1} = (\pm 1, 0)$ and $e_{\pm 2} = (0, \pm 1)$.

The key is to introduce (conditional) first-passage time quantities $P^{(t)}_{\eta,A}$ denotes the probability that the vacancy that started from site $A$ ($A$ can be a special site) at time 0 arrives for the first time to the position of one of the tracers at time $t$ conditioned on the fact that it was on the special site $\eta$ at time $t-1$.

A subtlety is that this probability shall depend on a quantity called $\psi$.

Our goal is now to find an expression for $\hat{q}_Z$ which is the main quantity involved in (S7):

\[
\hat{p}^{(t)}_Z(Y) = \delta_{Y,0} \left( 1 - \sum_{j=0}^{t} \sum_{\nu=\pm 1, \pm 2} P^{(j)}_{\nu,Z} \right) + \sum_{j=0}^{t} \sum_{\nu=\pm 1, \pm 2} \hat{p}^{(t-j)}_{\nu,Z}(Y - e_{\nu}) P^{(j)}_{\nu,Z}.
\] (S8)

\[
\hat{p}^{(t)}_Z(k) = 1 - \sum_{j=0}^{t} \sum_{\nu=\pm 1, \pm 2} P^{(j)}_{\nu,Z} + \sum_{j=0}^{t} \sum_{\nu=\pm 1, \pm 2} \hat{p}^{(t-j),z}_{\nu,Z}(k) e^{ik} e_{\nu} P^{(j)}_{\nu,Z}
\] (S9)

\[
\hat{q}^{(t)}_Z(k) = - \sum_{j=0}^{t} \sum_{\nu=\pm 1, \pm 2} \left[ 1 - \left( 1 + \hat{q}^{(t-j),z}_{\nu,Z}(k) \right) e^{ik} e_{\nu} \right] P^{(j)}_{\nu,Z}.
\] (S10)

An exponent $z$ to a quantity means that this quantity is computed taking into account $z = \mathbb{I}(0 < Z < L)$.

We now need an expression for $\hat{q}^\eta_Z$ where $\eta$ is a special site. To do so we decompose the propagator of the displacements over the successive passages of the vacancy to the position of one of the tracers:

\[
\hat{p}^{(t)},z_{\eta}(Y) = \delta_{Y,0} \left( 1 - \sum_{j=0}^{t} \sum_{\mu} P^{(j)},z_{\mu,\eta} \right)
+ \sum_{p=1}^{\infty} \sum_{m_1, \ldots, m_p} \sum_{\nu_1, \ldots, \nu_p} \delta_{\nu_1, \ldots, \nu_p} \delta_{Y, \sum_{\nu=\pm 1, \pm 2} e_{\nu}} \left( 1 - \sum_{j=0}^{m_{p+1}} \sum_{\mu} P^{(j)},z_{\mu,-\nu}\right) P^{(m_p),z}_{\nu_p,-\nu_{p-1}} \ldots P^{(m_2),z}_{\nu_2,-\nu_1} P^{(m_1),z}_{\nu_1,\eta}
\] (S11)

the sums on $\mu$ and $\nu_i$ run over the special sites ($\pm 1, \pm 2$).
The discrete Laplace transform of a function of time $g(t)$ is $\hat{g}(\xi) \equiv \sum_{t=0}^{\infty} g(t) \xi^t$. We can now take both the Laplace and Fourier transforms of (S11) to get:

$$\hat{\rho}_\eta(Y, \xi) = \frac{1}{1-\xi} \left\{ \delta_{Y,0} \left( 1 - \sum_{\mu} \hat{F}_\mu^z \right) + \sum_{p=1}^{\infty} \sum_{\nu_1, \ldots, \nu_p} \delta_{Y, \sum_{i=1}^{\nu_i} \nu_i} \sum_{\mu} \left( 1 - \hat{F}_{\mu, -\nu_p}^z \hat{F}_{\nu_p, -\nu_{p-1}}^z \cdots \hat{F}_{\nu_2, -\nu_1}^z \hat{F}_{\nu_1}^z \right) \right\} \quad (S12)$$

$$\hat{q}_\eta^z(k, \xi) \equiv \hat{\rho}_\eta^z(k, \xi) - \frac{1}{1-\xi} = \frac{1}{1-\xi} \sum_{\mu, \nu} \left\{ [1 - T^z(k, \xi)]^{-1} \nu_{\mu} \times (1 - e^{-i k e_{0\nu}}) e^{i k e_{0\nu}} \hat{F}_{\mu, \eta}^z(\xi) \right\} \quad (S13)$$

The matrix $T$ is defined by $T^z(k, \xi)_{\nu \mu} = \hat{F}_{\nu, -\mu}(\xi) e^{i k e_{0\nu}}$.

b. Multiple vacancies

Introducing (S10) into (7), one gets an expression for the Laplace transform of the second characteristic function:

$$\hat{\psi}(k, \xi) = -\rho_0 \left\{ \sum_{\nu=-1,2} \left[ \frac{1}{1-\xi} - \left( \frac{1}{1-\xi} + \hat{q}_{\nu}^z(0)(k, \xi) \right) e^{i k e_{0\nu}} \right] h_{\nu}(\xi) + \sum_{\nu=1,2} \left[ \frac{1}{1-\xi} - \left( \frac{1}{1-\xi} + \hat{q}_{-\nu}^z(0)(k, \xi) \right) e^{i k e_{0\nu}} \right] h_{\nu}(\xi) \right\} \quad (S14)$$

$$h_{\nu}(\xi) = \sum_{\nu \neq 0, L} \hat{F}_{\nu, 0}(\xi) \quad (S15)$$

where $\hat{q}_{\nu}^z$ is given by (S13).

At the end of the day, the only quantities that need to be computed are: $\hat{F}_{-1,1}, \hat{F}_{2,2}$ (returning to the left of TP1 / to the right of TP2), $\hat{F}_{1,1}, \hat{F}_{-2,2}$ (returning to the left of TP1 without touching TP2, and vice-versa), $F_{1, -2}, F_{-2, 1}$ (arriving to the right of TP2 starting from the left of TP1 without returning, and vice-versa), and the sums $h_{-1}, h_{2}, h_{1}, h_{-2}$.

a. Computation of the “outside” quantities. We call “outside” quantities the ones related to the sites $-1$ and $2$: $\hat{F}_{-1, -1}(\xi), \hat{F}_{2, 2}(\xi), h_1(\xi)$ and $h_2(\xi)$. In this case, we do not care about the fact that there are two tracers: we simply consider a random walk on a half line in which site 1 is biased. Note that in this case, we do not care either about the distance between the tracers (ie the value of $z$).

We first compute $F_{2, 2}$: this amounts to the probability to return for the first time to the origin at time $t$ and being on site 1 at time $t - 1$ knowing that one was at site 1 at time 0. We can partition on the first step of the walk.

For $\eta$ a “special site”, we call $\tilde{p}_\eta = 1/(2p_\eta + 1)$ the probability for the vacancy to jump to the left ($\nu = 1$) or right ($\nu = -1$) of a tracer when it is next to it.

$$F_{2, 2}(t) = (1 - \tilde{p}_2) \delta_{t, 1} + \tilde{p}_2 \sum_{j=1}^{t} f_1(j - 1) F_{2, 2}(t - j) \quad (S16)$$

where $f_1(t)$ the first-passage time density at the origin at time $t$ of a symmetric Polya walk starting from site $l$. One knowns $f_1(\xi) = \alpha^{l_1}$ with $\alpha = \xi^{-1}(1 - \sqrt{1 - \xi^2})$ [30].

We take the Laplace transform and obtain:

$$\hat{F}_{2, 2}(\xi) = \frac{(1 - \tilde{p}_2)\xi}{1 - \tilde{p}_2\alpha\xi} \quad (S17)$$

Similarly

$$\hat{F}_{-1, -1}(\xi) = \frac{(1 - \tilde{p}_{-1})\xi}{1 - \tilde{p}_{-1}\alpha\xi} \quad (S18)$$
\( h_{-1} \) and \( h_2 \) are easily computed from the former quantities:

\[
F_{+2,Z}(t) = \Theta(Z - (L + 1)) \sum_{j=0}^{t} f_{Z-1}(j) F_{2,2}(t - j) \tag{S19}
\]

\[
\hat{F}_{+2,Z}(\xi) = \Theta(Z - (L + 1)) \hat{f}_{Z-1}(\xi) \hat{F}_{2,2}(\xi) = \Theta(Z - (L + 1)) \hat{F}_{2,2}(\xi) \alpha^{Z-L-1} \tag{S20}
\]

\[
h_2(\xi) = \sum_{Z\neq 0,L} \hat{F}_{+2,Z}(\xi) = \hat{F}_{2,2} \sum_{Z'=0}^{\infty} \alpha^{Z'} = \frac{\hat{F}_{2,2}(\xi)}{1 - \alpha} \tag{S21}
\]

\[
h_{-1}(\xi) = \frac{\hat{F}_{-1,-1}(\xi)}{1 - \alpha} \tag{S22}
\]

b. Computation of the “inside” quantities  We first solve the case with no bias \( s_1 = s_2 = 0 \). There are only two quantity to compute: \( \hat{F}^{(=)} \equiv \hat{F}_{+1,+1} = \hat{F}_{-2,-2} \) and \( \hat{F}^{(\neq)} \equiv \hat{F}_{+1,-2} = \hat{F}_{-2,+1} \). We also compute (for \( 0 < Z < L \)) \( \hat{F}^2_Z \equiv \hat{F}_{+1,Z} = \hat{F}_{-2,L-Z} \). (\( L \) is the distance between the tracers)

We recall the formula for the first passage time (of a Polya random walk) at site \( s_1 \) starting from \( s_0 \) and considering \( s_2 \) as an absorbing site \[36].

\[
\hat{f}^1(s_1|s_0, \xi) = \frac{\hat{f}_{s_1-s_0}(\xi) - \hat{f}_{s_1-s_2}(\xi) \hat{f}_{s_2-s_0}(\xi)}{1 - \hat{f}_{s_1-s_2}(\xi)^2} \tag{S23}
\]

with \( \hat{f}(\xi) = \alpha^{j\xi} \) in our 1-dimensional case. This gives us:

\[
\hat{F}^{(=)}(\xi, L) = \frac{\alpha - \alpha^{2L-1}}{1 - \alpha^{2L}} \tag{S24}
\]

\[
\hat{F}^{(\neq)}(\xi, L) = \frac{\alpha^{L-1} - \alpha^{L+1}}{1 - \alpha^{2L}} \tag{S25}
\]

\[
\hat{F}^2_Z(\xi, L) = \frac{\alpha^{Z} - \alpha^{2L-Z}}{1 - \alpha^{2L}} \tag{S26}
\]

The case \( s_1 \neq 0, s_2 \neq 0 \) can be deduced from the unbiased case for a distance \( L - 2 \) by partitioning on the first passage of the vacancy to either +1 or -2:

\[
F_{1,1}(t, L) = (1 - \tilde{p}_1)\delta_{t,1} + \tilde{p}_1 \sum_{j=1}^{t} F^{(=)}(j - 1, L - 2) F_{1,1}(t - j, L) + \sum_{j=1}^{t} F^{(\neq)}(j - 1, L - 2) F_{1,1}(t - j, L) \tag{S27}
\]

\[
\hat{F}_{1,1}(\xi, L) = (1 - \tilde{p}_1)\xi + \tilde{p}_1\xi \left( \hat{F}^{(=)}(L - 2) \hat{F}_{1,-2}(L) + \hat{F}^{(\neq)}(L - 2) \hat{F}_{1,-2}(L) \right) \tag{S28}
\]

The first term corresponds to a return to 1, the second to a passage to -2.

This leads us to a system of \( 2 \times 2 \) equations with \( 2 \times 2 \) unknowns.

\[
\left\{
\begin{align*}
(\xi\tilde{p}_1 \hat{F}^{(=)}(L - 2) - 1) \hat{F}_{1,1} + (\xi\tilde{p}_1 \hat{F}^{(\neq)}(L - 2) - 1) \hat{F}_{1,-2} + (1 - \tilde{p}_1)\xi & = 0 \\
(\xi\tilde{p}_2 \hat{F}^{(=)}(L - 2) - 1) \hat{F}_{1,-2} + (\xi\tilde{p}_2 \hat{F}^{(\neq)}(L - 2) - 1) \hat{F}_{1,1} & = 0 \\
(\xi\tilde{p}_2 \hat{F}^{(=)}(L - 2) - 1) \hat{F}_{-2,1} + (\xi\tilde{p}_2 \hat{F}^{(\neq)}(L - 2) - 1) \hat{F}_{-2,-2} + (1 - \tilde{p}_2)\xi & = 0 \\
(\xi\tilde{p}_1 \hat{F}^{(=)}(L - 2) - 1) \hat{F}_{-2,1} + (\xi\tilde{p}_1 \hat{F}^{(\neq)}(L - 2) - 1) \hat{F}_{-2,-2} & = 0
\end{align*}
\right. \tag{S29}
\]

The solution is:

\[
\left\{
\begin{align*}
\hat{F}_{1,1} &= \frac{(1 - \xi\tilde{p}_2 \hat{F}^{(=)}(L - 2)) (1 - \tilde{p}_1)\xi}{(1 - \xi\tilde{p}_2 \hat{F}^{(=)})(1 - \xi\tilde{p}_2 \hat{F}^{(\neq)})(1 - \xi\tilde{p}_1 \hat{F}^{(\neq)})(1 - \xi\tilde{p}_1 \hat{F}^{(\neq)})} \\
\hat{F}_{1,-2} &= \frac{\xi^2 \tilde{p}_2 (1 - \tilde{p}_1) \xi}{(1 - \xi\tilde{p}_2 \hat{F}^{(=)})(1 - \xi\tilde{p}_2 \hat{F}^{(\neq)})(1 - \xi\tilde{p}_1 \hat{F}^{(\neq)})(1 - \xi\tilde{p}_1 \hat{F}^{(\neq)})} \\
\hat{F}_{-2,1} &= \frac{\xi^2 \tilde{p}_1 (1 - \tilde{p}_2) \xi}{(1 - \xi\tilde{p}_2 \hat{F}^{(=)})(1 - \xi\tilde{p}_2 \hat{F}^{(\neq)})(1 - \xi\tilde{p}_1 \hat{F}^{(\neq)})(1 - \xi\tilde{p}_1 \hat{F}^{(\neq)})} \\
\hat{F}_{-2,-2} &= \frac{\xi^2 \tilde{p}_1 (1 - \tilde{p}_2) \xi}{(1 - \xi\tilde{p}_2 \hat{F}^{(=)})(1 - \xi\tilde{p}_2 \hat{F}^{(\neq)})(1 - \xi\tilde{p}_1 \hat{F}^{(\neq)})(1 - \xi\tilde{p}_1 \hat{F}^{(\neq)})}
\end{align*}
\right. \tag{S30}
\]
with \( \hat{F}(\ast) \) and \( \hat{F}(\neq) \) evaluated for \( L - 2 \).

For \( 0 < Z < L \), one can again partition on the first passage of the vacancy to +1 or -2.

\[
F_{+1,Z}(t, L) = \sum_{j=1}^{t} F_{Z-1}^{\ast}(j, L - 2) F_{1,1}(t - j, L) + \sum_{j=1}^{t} F_{L-Z-1}^{\ast}(j, L - 2) F_{-2,-2}(t - j, L) \quad (S31)
\]

\[
\hat{F}_{+1,Z}(\xi, L) = \hat{F}_{Z-1}(L-2) \hat{F}_{1,1}(L) + \hat{F}_{L-Z-1}(L-2) \hat{F}_{-2,-2}(L) \quad (S32)
\]

Using (S26), we obtain:

\[
h_{1}(\xi) = \sum_{Z=1}^{L-1} \hat{F}_{+1,Z} = \frac{(1 - \alpha^{L-2})(1 - \alpha^{L-1})}{1 - \alpha^{2(L-2)}} (\hat{F}_{+1,+1} + \hat{F}_{+1,-2}) \quad (S33)
\]

\[
h_{-2}(\xi) = \sum_{Z=1}^{L-1} \hat{F}_{-2,Z} = \frac{(1 - \alpha^{L-2})(1 - \alpha^{L-1})}{1 - \alpha^{2(L-2)}} (\hat{F}_{-2,-2} + \hat{F}_{-2,+1}) \quad (S34)
\]

3. Results

a. Symbolic calculation

We use a symbolic calculation software (Mathematica) to put all the bricks together according to the following graph. An asterix (*) indicates that one should be careful about which value of \( L \) (distance between the TPs) should be used.

\[
\hat{\psi}(k, \xi) \quad \hat{\hat{\psi}}(k, \xi) \quad \hat{\hat{\hat{\psi}}}(k, \xi) \quad \hat{\hat{\hat{\hat{\psi}}}}(k, \xi) \quad \hat{\hat{\hat{\hat{\hat{\psi}}}}}(k, \xi) \quad \hat{\hat{\hat{\hat{\hat{\hat{\psi}}}}}}(k, \xi)
\]

b. Scaling limit

We define a continuous Laplace variable \( p = 1 - \xi \). When \( p \) goes to zero, \( \alpha = (1 - \sqrt{1 - \xi^2})/\xi = 1 - \sqrt{2p} + \mathcal{O}(p) \).

We will be interested in the limit of large time \( t \) and large distance \( L \), with \( \tau = t/L^2 \) constant. This amounts to writing \( p = \tilde{p}/L^2 \) with \( L \to \infty \) and \( \tilde{p} \) constant. The inverse Laplace transform of a function \( \hat{f}(p, L) = L^2 \hat{g}(\tilde{p}) \) is \( f(t, L) = g(\tau) \) with \( g \) the inverse Laplace transform of \( \hat{g} \).

Finally, our scaling limit in Laplace space amounts to the following substitutions that can be performed with a numerical software (in this order).

\[
1 - \xi \mapsto \tilde{p}/L^2 \quad 1 - \alpha \mapsto \sqrt{2p}/L \quad \alpha^L \approx \left(1 - \sqrt{2p}/L\right)^L \mapsto e^{-\sqrt{2p}} \quad \alpha \mapsto 1 \quad (S35)
\]

Appendix B: Detailed results

1. One TP

We start by recalling the results of Ref. [20] for a single TP of bias \( s \) in the limit of high density (fraction of vacancies \( \rho_{0} \to 0 \)), at large time (small \( p \)).

\[
\lim_{\rho_{0} \to 0} \frac{\hat{\psi}(k, p)}{\rho_{0}} = \frac{1}{\sqrt{2p}^{1/2}} \left\{ \cos k - 1 + s \sin k \right\} \quad (S1)
\]

\[
\lim_{\rho_{0} \to 0} \frac{\hat{\psi}(k, t)}{\rho_{0}} = \sqrt{\frac{2t}{\pi}} \left\{ \cos k - 1 + s \sin k \right\} \quad (S2)
\]
The even (resp odd) cumulants of the displacement $Y$ are equal.

$$\frac{\langle Y^{2n}\rangle_c}{\rho_0} = \sqrt{\frac{2\tau}{\pi}} \quad \quad \frac{\langle Y^{2n+1}\rangle_c}{\rho_0} = \sqrt{\frac{2\tau}{\pi}}$$ (S3)

The fraction (●) is understood as a limit $\rho_0 \to 0$.

2. Two TPs

We now focus on the case of two TPs with biases $s_1$ and $s_2$, initially separated by a distance $L$. The Laplace transform of the cumulant generating function reads

$$\lim_{\rho_0 \to 0} \frac{\hat{\psi}(k, p = \hat{p}/L^2)}{\rho_0} = \frac{L^3}{\sqrt{2p^3/2}} \left\{ \hat{K}^{c,2}(\hat{p})(\cos(k_1 + k_2) - 1) + \hat{K}^{o,2}(\hat{p}) \sin(k_1 + k_2) \right. $$

$$\left. + \sum_{i=1}^{2} \left[ \hat{K}_i^{c,1}(\hat{p})(\cos k_i - 1) + \hat{K}_i^{o,1}(\hat{p}) \sin k_i \right] \right\}. \quad \text{(S4)}$$

with $\hat{K}^{\alpha,\beta}$ given in the following. In real time, this gives

$$\frac{1}{L} \lim_{\rho_0 \to 0} \frac{\psi(k, t = L^2\tau)}{\rho_0} = \sqrt{\frac{2\tau}{\pi}} \left\{ K^{c,2}(\tau)(\cos(k_1 + k_2) - 1) + K^{o,2}(\tau) \sin(k_1 + k_2) \right. $$

$$\left. + \sum_{i=1}^{2} \left[ K_i^{c,1}(\tau)(\cos k_i - 1) + K_i^{o,1}(\tau) \sin k_i \right] \right\}. \quad \text{(S5)}$$

Note that $K(\tau)$ is related but not equal to the inverse Laplace transform of $\hat{K}(\hat{p})$.

From the previous expression one can deduce all the cumulants $(i = 1, 2)$, including those of the variation of distance $D = Y_2 - Y_1$.

$$\frac{\langle Y_1^2 Y_2^{2n-j}\rangle_c}{\rho_0 L} = K^{c,2}(\tau) \sqrt{\frac{2\tau}{\pi}} \quad \quad \frac{\langle Y_1^2 Y_2^{2n+1-j}\rangle_c}{\rho_0 L} = K^{o,2}(\tau) \sqrt{\frac{2\tau}{\pi}}$$ (S6)

$$\frac{\langle Y_2^{2n}\rangle_c}{\rho_0 L} = [K^{c,2}(\tau) + K_1^{c,1}(\tau)] \sqrt{\frac{2\tau}{\pi}} \quad \quad \frac{\langle Y_2^{2n+1}\rangle_c}{\rho_0 L} = [K^{o,2}(\tau) + K_1^{o,1}(\tau)] \sqrt{\frac{2\tau}{\pi}}$$ (S7)

$$\frac{\langle D^{2n}\rangle_c}{\rho_0 L} = [K_1^{c,1}(\tau) + K_2^{c,1}(\tau)] \sqrt{\frac{2\tau}{\pi}} \quad \quad \frac{\langle D^{2n+1}\rangle_c}{\rho_0 L} = [K_2^{o,1}(\tau) - K_1^{o,1}(\tau)] \sqrt{\frac{2\tau}{\pi}}$$ (S8)

Again, these results correspond to a limit $\rho_0 \to 0$.

We now give the expressions of the quantities $\hat{K}(p)$ and $K(t)$ both in the cases $s_1 = 0, s_2 \neq 0$ and $s_1, s_2 \neq 0$.

| Case $s_1 = 0, s_2 \neq 0$ | Case $s_1, s_2 \neq 0$ |
|-----------------------------|-----------------------------|
| $\hat{K}^{c,2}(\hat{p})$   | $v$                         |
| $\hat{K}^{o,2}(\hat{p})$   | $(1 + s_1 s_2 v)/d_2$     |
| $\hat{K}_1^{c,1}(\hat{p})$ | $s_2 v$                    |
| $\hat{K}_1^{o,1}(\hat{p})$ | $(1 - v)(1 + s_2 v)$       |
| $\hat{K}_2^{c,1}(\hat{p})$ | $0$                        |
| $\hat{K}_2^{o,1}(\hat{p})$ | $s_2 (1 - v)$              |

with $v = e^{-\sqrt{2\tau}}$ and $d_2 = 1 + s_1 s_2 v^2$. 

From this we deduce that

\[ K^{e,2}(s) = g_1(s) \quad \text{and} \quad K^{o,2}(s) = s_2 g_1(s) \]

\[ K^{e,1}(s) = 1 + (s_2 - 1)g_1(s) + s_2 g_2(s) \quad \text{and} \quad K^{o,1}(s) = 0 \]

\[ K^{e,1}(s) = 1 - g_1(s) \quad \text{and} \quad K^{o,1}(s) = s_2(1 - g_1(s)) \]

with

\[ g(u) = e^{-u^2} - \sqrt{\pi} u \text{erfc} u \] \hspace{1cm} (S9)

\[ g_n(s) = g(n/\sqrt{2s}) \] \hspace{1cm} (S10)

\[ G_{\alpha,\beta,\gamma}(s) = \sum_{n=0}^{\infty} (-s_1 s_2)^n \left[ \alpha g_{2n}(s) + \beta g_{2n+1}(s) + \gamma g_{2n+2}(s) \right] \] \hspace{1cm} (S11)

Note that at small \( s \), \( G_{\alpha,\beta,\gamma} = \alpha \) while at large \( s \), \( G_{\alpha,\beta,\gamma} = (\alpha + \beta + \gamma)/(1 + s_1 s_2) \).

The time evolution of the cumulants is given by Eqs. (S6), (S7). The rescaled velocities studied in the main text are obtained by taking the time derivative of the average displacements.

### 3. U-turn time for two TPs with opposite biases

We consider two TPs with biases \( s_2 > 0 \) and \(-s_2 < s_1 < 0\). We focus on the rescaled velocity of particle 1 defined as.

\[ A_1(t) = \frac{\sqrt{2\pi t} \, d(\rho_0)}{\rho_0} \] \hspace{1cm} (S12)

From (S7),

\[ \langle Y_1 \rangle = \left[ G_{0,s_1+s_2,0}(s) + G_{s_1,s_1(s_2-1),-s_1 s_2}(s) \right] \sqrt{\frac{2s}{\pi}} = G_{s_1,s_2(1+s_1),-s_1 s_2}(s) \sqrt{\frac{2s}{\pi}} \] \hspace{1cm} (S13)

From this we deduce that

\[ A_1(s) = H_{s_1,s_2(1+s_1),-s_1 s_2}(1/\sqrt{2s}) \] \hspace{1cm} (S14)

\[ H_{\beta_0,\beta_1,\beta_2}(s) = \sum_{n=0}^{\infty} \left( -s_1 s_2 \right)^n \left\{ \beta_0 e^{-[2n+1]s} + \beta_1 e^{-[(2n+1)s]^2} + \beta_2 e^{-[(2n+2)s]^2} \right\} \] \hspace{1cm} (S15)

At short time \( A_1 = s_1 < 0 \) while at large time \( A_1 = (s_1 + s_2)/(1 + s_1 s_2) > 0 \). There exists a rescaled time \( \tau^* \) such that \( A_1(\tau^*) = 0 \). One can solve numerically for \( \tau^* \). Here we derive the asymptotic behavior of \( \tau^* \) when \( s_1/s_2 \to 0 \) and when \( s_1/s_2 \to -1 \).

When \( s_1/s_2 \) is small, \( \tau^* \) is small so \( (\tau^*)^{-1/2} \) is large. We can keep only the first two terms in the sum (S15),

\[ A_1(\tau^*) \approx s_1 + s_2(1 + s_1)e^{-\frac{1}{\tau^*}} \approx s_1 + s_2 e^{-\frac{1}{\tau^*}} = 0, \] \hspace{1cm} (S16)

\[ \tau^* \approx \left[ 2 \ln \left( \frac{s_2}{-s_1} \right) \right]^{-1}. \] \hspace{1cm} (S17)

In the opposite limit, we write \( s_1 = -s_2(1 - \epsilon) \) with \( \epsilon \ll 1 \). We guess that \( (2\tau^*)^{-1} = \eta \epsilon \) with \( \eta \) depending only on
s_2. The expansion gives

\[ A_1 = \sum_{n=0}^{\infty} (-s_1 s_2)^n \left\{ s_1 (e^{-1/(2\tau)} (2n)^2 + s_2 (1 + s_1) (e^{-1/(2\tau)} (2n+1)^2 - s_1 s_2 (e^{-1/(2\tau)} (2n+2)^2) \right\} \]  \hspace{1cm} (S18)

\[ = \sum_{n=0}^{\infty} s_2^{2n} (1 - n\epsilon) \times \]

\[ \{ -s_2 (1 - \epsilon) (1 - (2n)^2 \eta \epsilon) + s_2 (1 - s_2 + s_2 \epsilon) (1 - (2n + 1)^2 \eta \epsilon) + s_2^2 (1 - \epsilon) (1 - (2n + 2)^2 \eta \epsilon) \} \]  \hspace{1cm} (S19)

\[ = 0 + \epsilon \sum_{n=0}^{\infty} s_2^{2n} \{ s_2 (1 + (2n)^2 \eta) + s_2^2 - s_2 (1 - s_2) (2n + 1)^2 \eta - s_2^2 - s_2^2 (2n + 2)^2 \} \]  \hspace{1cm} (S20)

\[ = \epsilon \sum_{n=0}^{\infty} s_2^{2n+1} + \eta \epsilon \sum_{n=0}^{\infty} s_2^{2n} \{ s_2 (2n)^2 - s_2 (1 - s_2) (2n + 1)^2 - s_2^2 (2n + 2)^2 \} \]  \hspace{1cm} (S21)

\[ = \epsilon \left\{ \frac{s_2}{1 - s_2^2} \eta (1 - s_2^2) \right\} \]  \hspace{1cm} (S22)

\[ A_1 = 0 \]  \hspace{1cm} gives

\[ \eta = \frac{s_2 (1 - s_2)^2}{1 - s_2^2 s_2 (1 + s_2)} = \frac{1 - s_2}{(1 + s_2)^2} \]  \hspace{1cm} (S23)

At the end of the day,

\[ \tau^* \approx \frac{1}{2\eta \epsilon} = \frac{(1 + s_2)^2}{2(1 - s_2) \frac{1}{1 + \frac{s_2}{s_2}}} \]  \hspace{1cm} (S24)

Eqs (S17) and (S24) correspond to the asymptote of the inset of Fig. 3, (b.ii) of the main text.

4. Single trajectories

We note that for large time and large distances, our results for cooperativity and competition (Eqs. (4) and (9) of the main text) actually hold at the very level of a single trajectory. Numerical evidence is shown on Fig. S1.

5. Three TPs

We now turn to the case of three TPs with biases s_1, s_2, s_3. We denote L_1 = X_2^0 - X_1^0 and L_2 = X_3^0 - X_2^0 and L = L_1 + L_2. Our result for the Laplace transform of the cumulant-generating function is

\[ \lim_{\rho_0 \to 0} \frac{\hat{\psi}(k, \rho = \hat{\rho}/L^2)}{\rho_0} = \frac{L^3}{\sqrt{2} \rho^{3/2}} \left\{ \hat{K}^{e,3}(\hat{\rho}) (\cos(k_1 + k_2 + k_3) - 1) + \hat{K}^{o,3}(\hat{\rho}) \sin(k_1 + k_2 + k_3) \right\} \]

\[ + \sum_{i=1}^{2} \left[ \hat{K}^{e,2}_{i,i+1}(\hat{\rho}) (\cos(k_i + k_{i+1}) - 1) + \hat{K}^{o,2}_{i,i+1}(\hat{\rho}) \sin(k_i + k_{i+1}) \right] \]

\[ + \sum_{i=1}^{3} \left[ \hat{K}^{e,1}_{i}(\hat{\rho}) (\cos k_i - 1) + \hat{K}^{o,1}_{i}(\hat{\rho}) \sin k_i \right] \]  \hspace{1cm} (S25)

with

\[ K^{\alpha,n}(\hat{\rho}) = \frac{\sum_{a,b=0}^{2} Q^{\alpha,n}(a, b) v_1^a v_2^b}{1 + s_1 s_2 v_1^2 + s_2 s_3 v_2^2 + s_1 s_3 v_1^2 v_2^2} \]  \hspace{1cm} (S26)
with \( v_1 = e^{-(L_1/L)\sqrt{\rho}} \), \( v_2 = e^{-(L_2/L)\sqrt{\rho}} \). For completeness, we give the 12 matrices \( Q \) of coefficients.

\[
Q^{e,3} = (1 + s_1 s_2 + s_2 s_3 + s_1 s_3) \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \quad Q^{v,3} = (s_1 + s_2 + s_3 + s_1 s_2 s_3) \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}
\]

\[
Q_{1,2}^{e,2} = (1 + s_1 s_2) \begin{pmatrix} 0 & 0 & 0 \\ 1 & s_3 - 1 & -s_3 \\ 0 & 0 & 0 \end{pmatrix} \quad Q_{1,2}^{v,2} = \frac{s_1 + s_2}{1 + s_1 s_2} Q_{1,2}^{e,2}
\]

\[
Q_{2,3}^{e,2} = (1 + s_2 s_3) \begin{pmatrix} 0 & 1 & 0 \\ 0 & -1 - s_1 & 0 \\ 0 & s_1 & 0 \end{pmatrix} \quad Q_{2,3}^{v,2} = \frac{s_2 + s_3}{1 + s_2 s_3} Q_{2,3}^{e,2}
\]

\[
Q_1^{e,1} = \begin{pmatrix} 1 & 0 & s_2 s_3 \\ s_2 - 1 & 0 & s_3 (1 - s_2) \\ -s_2 & 0 & -s_3 \end{pmatrix} \quad Q_1^{v,1} = s_1 Q_1^{e,1}
\]

\[
Q_2^{e,1} = \begin{pmatrix} 1 & s_3 - 1 \\ -1 - s_1 & 1 + s_1 - s_3 - s_1 s_3 & s_3 (1 + s_1) \\ s_1 & s_1 (s_3 - 1) & -s_1 s_3 \end{pmatrix} \quad Q_2^{v,1} = s_2 Q_2^{e,1}
\]

\[
Q_3^{e,1} = \begin{pmatrix} 1 & -1 - s_2 \\ 0 & 0 & 0 \\ s_1 s_3 - s_2 s_3 & 0 & 0 \\ 0 & s_1 s_3 - s_2 s_3 & s_1 + (1 + s_2) \end{pmatrix} \quad Q_3^{v,1} = s_3 Q_3^{e,1}
\]

The Laplace transform (\( S25 \)) can be inverted numerically to obtain the time evolution of the cumulants. One notices that the sum of all the coefficients of the matrices \( Q_1 \) and \( Q_2 \) is zero. This means that for \( \rho \to 0 \), (\( S25 \)) simplifies into

\[
\lim_{\rho_0 \to 0} \frac{\tilde{v}(k, \rho)}{\rho_0} = \frac{1}{\sqrt{2\rho}} \left\{ \cos(k_1 k_2 + k_3) - 1 + \frac{s_1 + s_2 + s_3 + s_1 s_2 s_3}{1 + s_1 s_2 + s_1 s_3 + s_2 s_3} \sin(k_1 k_2 + k_3) \right\}
\]
This corresponds to a single TP with effective bias \( S = \frac{s_1 + s_2 + s_3 + s_4 + s_5}{1 + s_1 + s_2 + s_3 + s_4 + s_5} \). This behavior generalises to an arbitrary number of TPs, the effective bias being

\[
S = \frac{\prod_{i=1}^{N}(1 + s_i) - \prod_{i=1}^{N}(1 - s_i)}{\prod_{i=1}^{N}(1 + s_i) - \prod_{i=1}^{N}(1 - s_i)} = \tanh \left( \frac{1}{2} \sum_{i=1}^{N} f_i \right)
\]

with \( s_i = \tanh(f_i/2) \).

6. Law of the distance between two TPs

a. Two TPs

We denote \( D = Y_2 - Y_1 \) the variation of distance between two TPs. The cumulant-generating function of \( D \) is given by

\[
\psi_D(q, t) \equiv \ln \left\langle e^{iqD(t)} \right\rangle = \psi(k_1 = -q, k_2 = q, t)
\]

with \( \psi \) given by Eq. (S5). Thus the Laplace transform of \( \psi_D \) reads:

\[
\lim_{\rho_0 \to 0} \frac{\hat{\psi}_D(q, p = \tilde{p} / L^2)}{\rho_0} = \frac{L^3}{\sqrt{2p^{3/2}}} \left\{ \left[ \hat{K}_{e,1}^{c,1}(\tilde{p}) + \hat{K}_{e,2}^{c,1}(\tilde{p}) \right] (\cos q - 1) + \left[ \hat{K}_{e,2}^{c,1}(\tilde{p}) - \hat{K}_{e,1}^{c,1}(\tilde{p}) \right] \sin q \right\}
\]

\[
= \frac{L^3}{\sqrt{2p^{3/2}}} \left\{ \left[ 2 + (s_2 - s_1)v \right] (\cos q - 1) + \left[ s_2 - s_1 - 2s_1s_2u \right] \sin q \right\}.
\]

At very-large time, ie when \( \tilde{p} \to 0, v = e^{-\sqrt{2p}} \approx 1 - \sqrt{2p} \). This lead us to the following stationary law \( \hat{\psi}_D^{\text{stat}} \) at large time:

\[
\lim_{\rho_0 \to 0} \frac{\hat{\psi}_D^{\text{stat}}(q, p)}{\rho_0} = \frac{L}{\rho} \left\{ \left[ 2 + (s_2 - s_1) \right] (\cos q - 1) + \left[ s_2 - s_1 - 2s_1s_2 \right] \sin q \right\}
\]

\[
= \frac{L}{\rho} \left\{ (1 - s_1)(1 + s_2) \left( e^{iq} - 1 \right) + (e^{-iq} - 1) \right\}
\]

\[
= \frac{2}{\rho} \left( e^{\beta f_1} + e^{-\beta f_2} \left( e^{iq} - 1 \right) + (e^{-iq} - 1) \right).
\]

We recall that the forces are defined by \( e^{\beta f_1} = (1 + s_1)/(1 - s_1) \). Eq. (S41) corresponds to a Skellam distribution \( P_D \) with parameters \( \mu_1 = 2\rho_0 L/(e^{\beta f_1} + e^{-\beta f_2}) \) and \( \mu_2 = \rho_0 L \).

\[
P_D(D) = e^{-(\mu_1 + \mu_2)} \left( \frac{\mu_1}{\mu_2} \right)^{k/2} I_k(2\sqrt{\mu_1\mu_2})
\]

The average of \( D \) is \( \mu_1 - \mu_2 \) and the variance is \( \mu_1 + \mu_2 \).

b. Effective potential and Gaussian limit

We focus on the case \( f_1 = f_2 = f \): \( \mu_1 = \rho_L [\cosh(\beta f)]^{-1} \), \( \mu_2 = \rho_0 L \). In this case, we can associate an effective potential \( U(D) \) to the probability law as

\[
P_D(D) \sim e^{-\beta U(D)}.
\]

For \( D \) close to its average value \( \mu_1 - \mu_2 \), the Skellam law is well approximated by a Gaussian

\[
P_D(D) \approx \frac{1}{(\mu_1 + \mu_2)^{3/2} \sqrt{2\pi}} e^{-\left( \frac{(D - \mu_1 + \mu_2)^2}{(\mu_1 + \mu_2)^2} \right)} \sim e^{-\beta U(D)}
\]
with the effective potential $U(D)$ given by
\begin{equation}
U(D) \approx \frac{\kappa}{2} (D - \langle D \rangle)^2 \tag{S45}
\end{equation}
\begin{equation}
\kappa = \frac{1}{\beta(\mu_1 + \mu_2)} = \frac{\cosh(\beta f)}{\beta \rho_0 L [1 + \cosh(\beta f)]} \tag{S46}
\end{equation}

\textit{c. Large deviation function and large distance scaling}

Now, we want to derive the behavior of $U(D)$ at large $D$. We first determine the large deviation function of the problem. We have
\begin{equation}
P_{\text{D}}^{\text{stat}}(D) = e^{-\rho_0 L (\tilde{\mu}_1 + 1)} \tilde{\mu}_1^{D/2} I_D(2\rho_0 L \sqrt{\tilde{\mu}_1}), \tag{S47}
\end{equation}
\begin{equation}
\ln P_{\text{D}}^{\text{stat}}(D) = -\rho_0 L (\tilde{\mu}_1 + 1) + \frac{D}{2} \ln \tilde{\mu}_1 + \ln I_D(2\rho_0 L \sqrt{\tilde{\mu}_1}) \tag{S48}
\end{equation}

with $\tilde{\mu}_1 = \mu_1/(\rho_0 L) = \cosh(\beta f)^{-1}$. We use the expansion of the Bessel function at large order and large argument (here large $D \sim \rho_0 L$).
\begin{equation}
I_D(Dz) \approx \frac{1}{\sqrt{2\pi D}} e^{\eta D} (1 + z^2)^{1/4} \tag{S49}
\end{equation}
\begin{equation}
\ln I_D(Dz) \approx \eta D \quad \ln I_D(2\rho_0 L z) \approx 2 \rho_0 L \eta \tag{S50}
\end{equation}

Here $z = \sqrt{\tilde{\mu}_1}/x$. Thus
\begin{equation}
\eta = z \sqrt{1 + \frac{1}{z^2}} - \ln \left( \frac{1}{z} + \frac{1}{x^2} \right) = \frac{1}{x} \sqrt{x^2 + \tilde{\mu}_1} - \ln(x + \sqrt{x^2 + \tilde{\mu}_1}) + \frac{1}{2} \ln(x + \sqrt{\tilde{\mu}_1}) \tag{S51}
\end{equation}

At the end of the day,
\begin{equation}
\frac{\ln P_{\text{D}}^{\text{stat}}(D = 2\rho_0 L x)}{2\rho_0 L} = -\frac{\tilde{\mu}_1 + 1}{2} + \frac{x}{2} \ln \tilde{\mu}_1 + x \eta \tag{S52}
\end{equation}
\begin{equation}
= -\frac{\tilde{\mu}_1 + 1}{2} + \sqrt{x^2 + \tilde{\mu}_1} + x \ln \tilde{\mu}_1 - x \ln(x + \sqrt{x^2 + \tilde{\mu}_1}) \tag{S53}
\end{equation}
\begin{equation}
= \phi(x) \tag{S54}
\end{equation}

$\phi(x)$ is a large deviation function. The large $D$ behavior corresponds to the limit $x \to \infty$.
\begin{equation}
\phi(x) \approx -x \ln x + (1 - \ln 2 + \ln \tilde{\mu}_1) x, \tag{S55}
\end{equation}
\begin{equation}
\ln P_{\text{D}}^{\text{stat}}(D) \approx D \left( 1 - \ln \frac{D}{\rho_0 L \tilde{\mu}_1} \right) \approx D \left( 1 - \ln \frac{D \cosh(\beta f)}{\rho_0 L} \right) \tag{S56}
\end{equation}
\begin{equation}
U(D) = -\frac{1}{\beta} \ln P_{\text{D}}^{\text{stat}}(D) \approx \frac{D}{\beta} \left[ \ln D + \ln \frac{\cosh(\beta f)}{\rho_0 L} - 1 \right] = \frac{D}{\beta} [\ln D + \nu] \tag{S57}
\end{equation}

with $\nu = \ln \frac{\cosh(\beta f)}{\rho_0 L} - 1$.

\textit{d. Three TPs and non-additivity of the interactions}

We consider three TPs undergoing forces $f_1$, $f_2$ and $f_3$. We denote $L_1 = X_2(t=0) - X_1(t=0)$ the initial distance between TPs 1 and 2. Let us call $D_{12} = Y_2 - Y_1$ the variation of distance between TPs 1 and 2. From [S25], we show that its cumulant generating function at large time reads
\begin{equation}
\psi_{D_{12}}(q) \equiv \ln(e^{iqD_{12}}) \tag{S58}
\end{equation}
\begin{equation}
\frac{\psi_{D_{12}}^{\text{stat}}(q)}{\rho_0} = L_1 \left\{ \frac{2}{e^{\beta f_1} + e^{-\beta(f_2 + f_3)}} (e^{iq} - 1) + (e^{-iq} - 1) \right\}. \tag{S59}
\end{equation}
This still corresponds to a Skellam law. The presence of a force $f_3$ modifies the stationary law. We see that the interactions between TPs are not additive.

Similarly, if we denote $L_2 = X_3(t = 0) - X_2(t = 0)$ and $D_{23} = Y_3 - Y_2$, we have

$$
\psi_{D_{23}}^{\text{stat}}(q) = L_2 \left\{ \frac{2}{e^{\beta(f_1 + f_3)} + e^{-\beta f_3}(e^{iq} - 1) + (e^{-iq} - 1)} \right\}.
$$

(S60)

Appendix C: Numerical simulations

1. Simulations

We do simulations of the SEP as follows. We consider a periodic line of $N$ sites. We first place $n = 1, 2$ or 3 tagged particles (TPs) at initial deterministic positions. Then, we place $M - n$ bath particles uniformly at random on the remaining sites. For the figures of the main text we used $N = 2500$, $M = 2475$ ($\rho_0 = M/N = 0.01$).

At each time step we draw a particle at random and try to move it either to the left or to the right according to its probability $p = (1 + \varepsilon)/2$ for the TPs, $p = 1/2$ for the bath particles. The continuous time is then incremented with a exponentially distributed random number with characteristic time $1/M$. This accounts for the exponentially distributed jump rates of the particles. At equally spaced continuous times we compute the observables: the moments with 1 to $n$ TPs.

We perform the average over multiple simulations to get good statistics. Typically we perform 50000 simulations (except for the Figure for a single realization).

2. Numerical effective biases

We now discuss how to obtain the numerical curves for the rescaled velocities defined as

$$
A(t) = \sqrt{2\pi t} \frac{dX}{dt}
$$

(S1)

from the numerical data $X$ versus $t$. Let us define

$$
B(t) = \sqrt{\frac{\pi}{2t}} X(t)
$$

(S2)

which is easy to compute from the data. One sees that

$$
A(t) = B(t) + 2t \frac{dB}{dt}.
$$

(S3)

Now, we write $w = \ln t$ which gives us

$$
A(w) = B(w) + 2 \frac{dB}{dw}.
$$

(S4)

Our procedure is as follow: we first gather the data into bins of logarithmic size over which we average. This gives us $B(w)$. We then compute the derivative $dB/dw$ using a Savisky-Golay filter. We finally obtain $A(w)$ that we can plot on our graphs.