# Finite-Time Current Probabilities in the Asymmetric Exclusion Process on a Ring

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#### Abstract

We calculate the time-dependent probability of non-zero current through a selected bond in the totally asymmetric exclusion process with periodic boundary conditions. For specific initial conditions corresponding to the minimal probability of non-zero current, we derive an explicit analytical expression which is valid for arbitrary time intervals.

#### 1 Introduction

The study of space-time correlations in stochastic models of interacting particles is a central subject of the non-equilibrium statistical mechanics [1]. Among a variety of correlations functions, the current characteristics are the most natural and important ones for physical applications. During the past decade, there has been considerable progress in the study of current fluctuations in the asymmetric exclusion process (ASEP) which is a paradigm for non-equilibrium many-particle systems [2, 3, 4, 5].

Two main quantities are used for the description of current, depending on the geometry of system. For the ring geometry and the fully asymmetric process, an adequate quantity is the total distance  $Y_t$  covered by all of the particles between time 0 and t [6, 7, 8]. For the infinite chain, the timeintegrated current can be measured by the number of particles  $Q_t$  which have crossed a particular bond up to time t [9]. For the finite chain which is in contact at its ends with two reservoirs,  $Q_t$  is the number of particles which have moved from the left reservoir into the system during time t [10].

Most of the known results obtained so far concern the limiting case of large time when the generating functions  $\langle e^{\alpha Y_t} \rangle$  and  $\langle e^{\alpha Q_t} \rangle$  increase exponentially with t,

$$\langle e^{\alpha Y_t} \rangle \sim e^{\lambda(\alpha)t}$$
  
 $\langle e^{\alpha Q_t} \rangle \sim e^{\mu(\alpha)t}$ 

and

where  $\lambda(\alpha)$  and  $\mu(\alpha)$  are the largest eigenvalues of the properly defined Markov matrices.

At the same time, much less is known about the finite-time behavior of  $Y_t$ and  $Q_t$ . The first exact result for the probability  $P(x_1, ..., x_P; t | a_1, ..., a_P; 0)$ of finding P particles on lattice sites  $x_1, ..., x_P$  at time t given that they were on sites  $x_1^0, ..., x_P^0$  at time 0, has been obtained in [11] (see also [12]) for the totally ASEP on the infinite chain. Based on this result, it became possible to find the probability distribution of the current  $Q_t(x)$ , i.e. the number of particles that have crossed the lattice bond (x - 1, x) up to time t for a specific boundary condition of the half filled infinite chain, when the sites from  $-\infty$  to 0 are occupied and the right half is empty at t = 0 [13].

The knowledge of  $P(x_1, ..., x_P; t | a_1, ..., a_P; 0)$  enables calculation of many other current properties for arbitrary time intervals. However, the infinite geometry is not sufficient for complete description of the relaxation phenomena because, in the case of an infinite lattice and a finite number of particles, the stationary state corresponds to zero density, so that the particles are non-interacting.

The probability  $P(x_1, ..., x_P; t | a_1, ..., a_P; 0)$  for the totally ASEP with P particles on a ring has been derived in [14]. This opens the prospect for studies of finite-time current probabilities during the whole process of relaxation from an initial configuration to a non-trivial steady state.

In this paper, we consider the current  $Q_t(0)$  on the ring of L sites which is defined as the number of particles that have crossed the bond (L-1,0)up to time t. Our goal will be to compute the probability  $Prob[Q_t(0) > 0]$  that at least one particle crosses the bond (L-1,0). In Section 2 we obtain a general expression for this probability assuming arbitrary initial positions of P particles on the ring. This result still contains summations of a determinant of a  $P \times P$  matrix over numbers of rotation of all the particles around the ring. In Section 3, we consider particular initial conditions  $a_1 = 0, a_2 = 1, \ldots, a_P = P - 1$  corresponding to the minimal current probability among all initial conditions, and derive an explicit analytical expression for  $Prob[Q_t(0) > 0]$ . Section 4 gives an analysis of the obtained formula.

#### 2 Current probabilities

Let C be a configuration of P particles on a ring of L sites, where the positions of particles are  $0 \le x_1 < x_2 < ... < x_P < L$ . The ASEP is defined by the master equation for the probability  $P_t(C)$  of finding the system in configuration C at time t,

$$\partial_t P_t(C) = \sum_{\{C'\}} [M_0(C, C') + M_1(C, C')] P_t(C'), \qquad (2.1)$$

with the initial condition that the system is in configuration  $C_0$  at time t. Here  $M_1(C, C')$  is the probability of going from configuration C' to C during a time interval dt, and  $M_0(C, C')$  is a diagonal matrix with diagonal elements

$$M_0(C,C) = -\sum_{\{C' \neq C\}} M_1(C',C).$$
(2.2)

The matrix elements of  $M_1(C, C')$  obey the exclusion rule that, during dt, each particle jumps with probability dt to its right provided that the target site is empty. Given the initial positions of particles  $0 \le a_1 < a_2 < ... < a_P < L$  at the moment t = 0,  $P_t(C)$  is the conditional probability  $P(x_1, ..., x_P; t | a_1, ..., a_P; 0)$  of finding P particles on the sites  $0 \le x_1 < ... < x_P < L$  at time t.

The solution of (2.1) is [14]:

$$P_t(C) = \sum_{n_1 = -\infty}^{\infty} \dots \sum_{n_P = -\infty}^{\infty} (-1)^{(P-1)\sum_{i=1}^P n_i} \det \mathbf{M}.$$
 (2.3)

Elements of the  $P \times P$  matrix **M** are

$$M_{ij} = F_{s_{ij}}(a_i, x_j + n_i L|t), (2.4)$$

where

$$s_{ij} = Pn_i - \sum_{k=1}^{P} n_k + j - i, \qquad (2.5)$$

and  $F_m(a, x|t)$  are functions introduced by Schütz [11]:

$$F_m(a,x|t) = \sum_{k=0}^{\infty} \left( \begin{array}{c} k+m-1\\ m-1 \end{array} \right) F_0(a-k,x|t),$$
(2.6)

if integer m > 0, and

$$F_m(a,x|t) = \sum_{k=0}^{-m} (-1)^k \begin{pmatrix} -m \\ k \end{pmatrix} F_0(a-k,x|t), \qquad (2.7)$$

if integer m < 0. For m = 0 and  $x \ge a$ ,

$$F_0(a, x|t) = \frac{e^{-t}t^K}{K!},$$
(2.8)

where K = x - a. For m = 0 and x < a

$$F_0(a, x|t) = 0. (2.9)$$

The derivation of (2.3) in ref.[14] contains, as an intermediate step, the evaluation of probabilities  $\psi_n(C;t|C_0;0)$  to reach configuration C from  $C_0$  for time t after making n visits of the origin  $0 \equiv L$  of the ring by each of the particles in turn, starting with the last. Thus, the probability  $P_t(C)$  is the sum

$$P_t(C) = \sum_{n=0}^{\infty} \psi_n(C; t | C_0; 0) = \sum_{n=0}^{\infty} \sum_{\{n_i\}_n} (-1)^{(P-1)\sum_{i=1}^{P} n_i} \det \mathbf{M}.$$
 (2.10)

where summation over  $n_i, i = 1, 2, ..., P$  is restricted by the condition  $n_1 + n_2 + \cdots + n_P = n$ .

To find  $Prob[Q_t(0) > 0]$ , we have to take the sum over all final configurations C which can be reached from  $C_0$  after at least one visit of the origin,

$$Prob[Q_t(0) > 0] = \sum_{n=1}^{\infty} \sum_{C} \psi_n(C; t | C_0; 0) =$$
(2.11)

$$\sum_{n=1}^{\infty} \sum_{0 \le x_1 < x_2 < \dots < x_P < L} \sum_{\{n_i\}_n} (-1)^{(P-1)\sum_{i=1}^P n_i} \det \mathbf{M},$$
(2.12)

or, in the explicit form,

$$Prob[Q_t(0) > 0] = \sum_{n=1}^{\infty} \sum_{x_P=P-1}^{L-1} \sum_{x_{P-1}=P-2}^{x_P-1} \cdots \sum_{x_2=1}^{x_3-1} \sum_{x_1=0}^{x_2-1} \sum_{\{n_i\}_n} (-1)^{(P-1)\sum_{i=1}^P n_i} \times (-1)^{(P-1)\sum_{i=1$$

$$\begin{array}{ccccc} F_{s_{11}}(a_1, x_1 + n_1L) & F_{s_{12}}(a_1, x_2 + n_1L) & \cdots & F_{s_{1P}}(a_1, x_P + n_1L) \\ F_{s_{21}}(a_2, x_1 + n_2L) & F_{s_{22}}(a_2, x_2 + n_2L) & \cdots & F_{s_{2P}}(a_2, x_P + n_2L) \\ \vdots & \vdots & & \vdots \\ F_{s_{P1}}(a_P, x_1 + n_PL) & F_{s_{P2}}(a_P, x_2 + n_PL) & \cdots & F_{s_{PP}}(a_P, x_P + n_PL) \end{array}$$

$$(2.13)$$

To evaluate these sums we proceed as in [15]. Using the identity

$$\sum_{x=x_1}^{x_2} F_s(a,x) = F_{s+1}(a,x_1) - F_{s+1}(a,x_2+1), \qquad (2.14)$$

the first column of the determinant becomes, after summation over  $x_1$ ,

$$F_{s_{11}+1}(a_1, n_1L) - F_{s_{11}+1}(a_1, x_2 + n_1L)$$

$$F_{s_{21}+1}(a_2, n_2L) - F_{s_{21}+1}(a_2, x_2 + n_2L)$$

$$\vdots$$

$$F_{s_{P1}+1}(a_P, n_PL) - F_{s_{P1}+1}(a_P, x_2 + n_PL).$$

It follows from (2.5) that

$$s_{i1} + 1 = s_{i2} \tag{2.15}$$

for all i = 1, 2, ..., P, and we can reduce the first column by adding the second to it. Continuing this process up to the sum over  $x_P$ , we get the first P - 1 columns in the form

$$F_{s_{1k}+1}(a_1, k - 1 + n_1L)$$

$$F_{s_{2k}+1}(a_2, k - 1 + n_2L)$$

$$\vdots$$

$$F_{s_{Pk}+1}(a_P, k - 1 + n_PL)$$
(2.16)

for k = 1, ..., P - 1, and only the last column remains nonreduced,

$$F_{s_{1P}+1}(a_1, P - 1 + n_1L) - F_{s_{1P}+1}(a_1, L + n_1L)$$

$$F_{s_{2P}+1}(a_2, P - 1 + n_2L) - F_{s_{2P}+1}(a_2, L + n_2L)$$

$$\vdots$$

$$F_{s_{PP}+1}(a_P, P - 1 + n_PL) - F_{s_{PP}+1}(a_P, L + n_PL).$$
(2.17)

Thus, the resulting determinant splits into two determinants  $D_1$  and  $D_2$  corresponding to two summands in the last column (2.17). The first determinant  $D_1$  has a convenient form

$$\begin{array}{ccccc} F_{s_{11}+1}(a_1, n_1L) & F_{s_{12}+1}(a_1, 1+n_1L) & \cdots & F_{s_{1P}+1}(a_1, P-1+n_1L) \\ F_{s_{21}+1}(a_2, n_2L) & F_{s_{22}+1}(a_2, 1+n_2L) & \cdots & F_{s_{2P}+1}(a_2, P-1+n_2L) \\ \vdots & \vdots & & \vdots \\ F_{s_{P1}+1}(a_P, n_PL) & F_{s_{P2}+1}(a_P, 1+n_PL) & \cdots & F_{s_{PP}+1}(a_P, P-1+n_PL) \end{array} \right| .$$

$$(2.18)$$

Consider the determinant  $D_2$ . Using the property

$$F_m(a,x) = \sum_{k=0}^{\infty} F_{m-1}(a,x+k)$$
(2.19)

we can write the i-th element of the first column as

$$F_{s_{i1}+1}(a_i, n_i L) = F_{s_{i1}}(a_i, n_i L) + F_{s_{i1}+1}(a_i, 1 + n_i L)$$
(2.20)

for all i = 1, 2, ..., P. We now prove that the contribution from the first term of Eq.(2.20) into the sum

$$\sum_{n=1}^{\infty} \sum_{\{n_i\}_n} (-1)^{(P-1)\sum_{i=1}^P n_i} \det D_2$$
(2.21)

vanishes.

Expanding the determinant in (2.21), we select among terms containing the first summand in (2.20) those which contain the *j*-th element of the last column:  $F_{s_{i1}}(a_i, n_iL) \times F_{s_{jP}+1}(a_j, L + n_jL)$ . Consider the unique "mirror" terms which coincide with the selected terms except two factors, one from the *j*-th element of the first column and the second from the *i*-th element of the last column:  $F_{s_{j1}}(a_j, n'_jL) \times F_{s_{iP}+1}(a_i, L + n'_iL)$ , where  $n'_i = n_i - 1$ and  $n'_j = n_i + 1$ . The indices  $s_{jk} = s_{jk}(\mathbf{n})$  are functions of the vector  $\mathbf{n} =$  $(n_1, n_2, \ldots, n_P)$ . We denote by  $\mathbf{n}'$  the vector obtained from  $\mathbf{n}$  by replacement  $n_i$  and  $n_j$  by  $n'_i$  and  $n'_j$  Taking into account that

$$s_{j1}(\mathbf{n}') = Pn'_{j} - \sum_{k=1}^{P} n'_{k} + 1 - j = Pn_{j} - \sum_{k=1}^{P} n_{k} + P - j + 1 = s_{jP}(\mathbf{n}) + 1 \quad (2.22)$$

and

$$s_{iP}(\mathbf{n}') = Pn'_{i} - \sum_{k=1}^{P} n'_{k} + P - i = Pn_{i} - \sum_{k=1}^{P} n_{k} - i = s_{i1}(\mathbf{n}) - 1, \quad (2.23)$$

we see that the two selected terms are equal and enter into (2.21) with opposite signs because the sum  $\sum n_k = \sum n'_k$  and the sign of the permutation of indices 1 and P is always negative. Thus, the set of terms containing the first summand in (2.20) splits into two subsets cancelling one another.

As the contribution from the first term of (2.20) vanishes, we obtain instead of  $D_2$  a determinant where the first two columns have the same arguments:

Again, using (2.19)

$$F_{s_{i2}+1}(a_i, n_i L) = F_{s_{i2}}(a_i, n_i L) + F_{s_{i2}+1}(a_i, 1 + n_i L)$$
(2.25)

for all i = 1, 2, ..., P, we obtain the sum of determinants

$$\begin{array}{cccc} F_{s_{11}+1}(a_1, 1+n_1L) & F_{s_{12}}(a_1, 1+n_1L) & F_{s_{13}+1}(a_1, 2+n_1L) & \cdots \\ F_{s_{21}+1}(a_2, 1+n_2L) & F_{s_{22}}(a_2, 1+n_2L) & F_{s_{23}+1}(a_2, 2+n_2L) & \cdots \\ \vdots & \vdots & \vdots \\ F_{s_{P1}+1}(a_P, 1+n_PL) & F_{s_{P2}}(a_P, 1+n_PL) & F_{s_{P3}+1}(a_P, 2+n_PL) & \cdots \\ \end{array}$$

$$(2.26)$$

and

$$\begin{array}{c|cccc} F_{s_{11}+1}(a_1, 1+n_1L) & F_{s_{12}+1}(a_1, 2+n_1L) & F_{s_{13}+1}(a_1, 2+n_1L) & \cdots \\ F_{s_{21}+1}(a_2, 1+n_2L) & F_{s_{22}+1}(a_2, 2+n_2L) & F_{s_{23}+1}(a_2, 2+n_2L) & \cdots \\ \vdots & \vdots & \vdots & & \vdots \\ F_{s_{P1}+1}(a_P, 1+n_PL) & F_{s_{P2}+1}(a_P, 2+n_PL) & F_{s_{P3}+1}(a_P, 2+n_PL) & \cdots \\ \end{array} \right| .$$

Two columns in the first determinant coincide because  $s_{i1} + 1 = s_{i2}$  for all i = 1, 2, ..., P and  $D_2$  gets reduced to the determinant (2.27) with equal arguments in the second and third columns. Continuing this procedure, we obtain finally

$$\begin{vmatrix} F_{s_{11}+1}(a_1, 1+n_1L) & \cdots \\ F_{s_{21}+1}(a_2, 1+n_2L) & \cdots \\ \vdots \\ F_{s_{P1}+1}(a_P, 1+n_PL) & \cdots \\ \cdots & F_{s_{1(P-1)}+1}(a_1, P-1+n_1L) & F_{s_{1P}+1}(a_1, L+n_1L) \\ \cdots & F_{s_{2(P-1)}+1}(a_2, P-1+n_2L) & F_{s_{2P}+1}(a_2, L+n_2L) \\ \vdots & \vdots \\ \cdots & F_{s_{P(P-1)}+1}(a_P, P-1+n_PL) & F_{s_{PP}+1}(a_P, L+n_PL) \end{vmatrix}$$
(2.28)

We expand the determinant (2.28) by the last column and consider the sum

$$\sum_{n=1}^{\infty} \sum_{\{n_i\}_n} (-1)^{(P-1)\sum n_k} \sum_{i=1}^{P} (-1)^{i+P} F_{s_{iP}+1}(a_i, (1+n_i)L) M_{iP}, \qquad (2.29)$$

where  $M_{iP}$  is a minor of the matrix in Eq.(2.28). Given the *i*-th element of the sum Eq.(2.29), we introduce a vector  $\mathbf{n}' = (n'_1, n'_2, \dots, n'_P)$  with  $n'_1 = n_1, \dots, n'_{i-1} = n_{i-1}, n'_i = n_i + 1, n'_{i+1} = n_{i+1}, \dots, n'_P = n_P$ , so that

$$\sum_{i=1}^{P} n_i' = \sum_{i=1}^{P} n_i + 1.$$
(2.30)

We have

$$s_{iP}(\mathbf{n}) = Pn_i - \sum n_k + P - i = Pn'_i - \sum n'_k + 1 - i = s_{i1}(\mathbf{n}') \quad (2.31)$$

and

$$s_{jm}(\mathbf{n}) = Pn_j - \sum n_k + m - j = Pn'_j - \sum n'_k + 1 + m - j = s_{j(m+1)}(\mathbf{n}') \quad (2.32)$$

for  $j \neq i$ . Then the sum (2.29) becomes

$$\sum_{n=2}^{\infty} (-1)^{P-1} \sum_{i=1}^{P} \sum_{\{n'_i\}_n} (-1)^{(P-1)\sum n'_k} (-1)^{i+P} F_{s_{i1}+1}(a_i, (n'_i)L) M_{iP} = \sum_{n=2}^{\infty} (-1)^{(P-1)} \sum_{\{n'_i\}_n} (-1)^{(P-1)\sum n'_k} \times$$

Performing a cyclic permutation in Eq.(2.33), we see that the sum (2.33) is similar to the sum  $\sim$ 

$$\sum_{n=1}^{\infty} \sum_{\{n_i\}_n} (-1)^{(P-1)\sum n_k} D_1, \qquad (2.34)$$

where  $D_1$  is given by Eq.(2.18). The only difference is in the ranges of summation over n. Remembering that  $D_1$  and  $D_2$  have opposite signs, we see that only terms obeying  $\sum n_k = 1$  remain and we obtain

$$Prob[Q_{t}(0) > 0] = \sum_{n_{1}+\dots+n_{P}=1} (-1)^{P-1} \begin{vmatrix} F_{s_{11}+1}(a_{1}, n_{1}L) \\ F_{s_{21}+1}(a_{2}, n_{2}L) \\ \vdots \\ F_{s_{12}+1}(a_{1}, 1+n_{1}L) & \cdots & F_{s_{1P}+1}(a_{1}, P-1+n_{1}L) \\ F_{s_{22}+1}(a_{2}, 1+n_{2}L) & \cdots & F_{s_{2P}+1}(a_{2}, P-1+n_{2}L) \\ \vdots & \vdots \\ F_{s_{P2}+1}(a_{P}, 1+n_{P}L) & \cdots & F_{s_{PP}+1}(a_{P}, P-1+n_{P}L) \end{vmatrix},$$

$$(2.35)$$

where  $s_{ij} = Pn_i - 1 + j - i$ .

### 3 Minimal current probability

The probability of the non-zero current through bond (L - 1, 0) depends on the initial configuration of particles. This probability is minimal if  $a_1 = 0, a_2 = 1, \ldots, a_P = P - 1$  because the particle at site 0 has a maximal obstacle to clear this site and the first particle which can cross the bond (L - 1, 0)has a maximal distance to the target site  $0 \equiv L$ .

Let  $\mathbb{P}(\mathcal{E}_t^{(P)})$  denote min  $Prob[Q_t(0) > 0]$  over all possible initial configurations. To simplify notations, we use the fact that functions  $F_m(a, x)$  depend only on the difference of their arguments and write  $F_m(a, x) \equiv F_m(x-a)$ . A further simplification comes from the observation that only the terms with  $n_1 = n_2 = \cdots = n_{i-1} = n_{i+1} = \cdots = n_P = 0, n_i = 1, i = 1, \ldots, P$  do not vanish in (2.35). Indeed, assume that  $n_i < 0$  for some  $i, 1 \le i \le P$ . Then, the *i*-th row in (2.35)

$$F_{s_{i1}+1}(a_i, n_i L), \dots, F_{s_{iP}+1}(a_i, P-1+n_i L)$$
(3.1)

vanishes owing to the condition  $F_{-m}(a, x) = 0$  if  $x - a < -m, m \ge 0$  and the inequalities  $s_{ik} + 1 = Pn_i + k - i > n_iL + k - 1 - a_i$  and  $a_i \ge i - 1$ .

Inserting the initial conditions  $a_1 = 0, a_2 = 1, \ldots, a_P = P - 1$  and the possible values of  $n_1, \ldots, n_P$  in (2.35) we obtain

$$\mathbb{P}(\mathcal{E}_{t}^{(P)}) = \sum_{i=1}^{P} (-1)^{P-1} \\ \times \begin{vmatrix} F_{0}(0) & F_{1}(1) & \cdots & F_{P-1}(P-1) \\ F_{-1}(-1) & F_{0}(0) & \cdots & F_{P-2}(P-2) \\ \vdots & \vdots & & \vdots \\ F_{P-i+1}(L-i+1) & F_{P-i+2}(L-i+2) & \cdots & F_{2P-i}(L+P-i) \\ \vdots & \vdots & & \vdots \\ F_{-P+1}(-P+1) & F_{-P+2}(-P+2) & \cdots & F_{0}(0) \end{vmatrix}$$

$$(3.2)$$

Using the fact that  $F_{-p}(-p) = (-1)^p F_0(0)$  and performing simple column operations, we can write this as

$$\mathbb{P}(\mathcal{E}_t^{(P)}) = \sum_{i=1}^{P} (-1)^{P-1} \Delta_P^{(i)}, \qquad (3.3)$$

where

$$\Delta_P^{(i)} = \begin{vmatrix} F_1(0) & F_2(1) & \cdots \\ 0 & F_1(0) & \cdots \\ \vdots & \vdots \\ F_{P-i+2}(L-i+1) & F_{P-i+3}(L-i+2) & \cdots \\ \vdots & \vdots \\ 0 & 0 & \cdots \end{vmatrix}$$

This can be further simplified to

$$\mathbb{P}(\mathcal{E}_t^{(P)}) = (-1)^{P-1} \left[ e^{-t} \sum_{i=2}^{P-1} \Delta_i^* + \Delta_P^{(1)} + \Delta_P^{(P)} \right], \qquad (3.5)$$

where

$$\Delta_P^{(1)} = e^{-t} F_{P+1}(L). \tag{3.6}$$

We now evaluate the determinant  $\Delta_i^*$  using the fact that

$$F_{n+1}(n) = \frac{t^n}{n!}.$$
(3.7)

Writing  $x_k = F_{P-i+k+1}(L-i+k)$  we have

$$\Delta_{i}^{*} = \begin{vmatrix} 1 & t & \cdots & \frac{1}{(i-2)!} t^{i-2} & \frac{1}{(i-1)!} t^{i-1} \\ 0 & 1 & \cdots & \frac{1}{(i-3)!} t^{i-3} & \frac{1}{(i-2)!} t^{i-2} \\ \vdots & \vdots & & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & t \\ x_{1} & x_{2} & \cdots & x_{i-1} & x_{i} \end{vmatrix} .$$
(3.8)

Applying row operations this can be reduced to

$$\Delta_{i}^{*} = \begin{vmatrix} 1 & 0 & \cdots & 0 & \frac{(-1)^{i}}{(i-1)!} t^{i-1} \\ 0 & 1 & \cdots & 0 & \frac{(-1)^{i-1}}{(i-2)!} t^{i-2} \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & t \\ x_{1} & x_{2} & \cdots & x_{i-1} & x_{i} \end{vmatrix} .$$
(3.9)

Indeed, after the bottom rows from i - k + 1 down to i - 1 have been cleared, we subtract these rows  $t^r/r!$  times from the (i - k)-th row (r = 1, ..., k - 1)to get

$$\frac{t^k}{k!} + \sum_{r=1}^{k-1} \frac{t^r}{r!} \frac{(-t)^{k-r}}{(k-r)!} = -\frac{(-t)^k}{k!}$$
(3.10)

in the last column. (The sum is the coefficient of  $t^k$  in the expansion of  $e^t e^{-t}$  except for the terms r = 0 and r = k.) The determinant now easily evaluates to

$$\Delta_i^* = \sum_{k=0}^{i-1} \frac{(-t)^k}{k!} F_{P+1-k}(L-k).$$
(3.11)

This sums to

$$\sum_{i=2}^{P-1} \Delta_i^* = (P-2)F_{P+1}(L) + \sum_{k=1}^{P-2} \frac{(-t)^k}{k!} (P-k-1)F_{P-k+1}(L-k). \quad (3.12)$$

The determinant  $\Delta_P^{(P)}$  can be treated similarly. It is given by

$$\Delta_{P}^{(P)} = \begin{vmatrix} 1 & t & \cdots & \frac{t^{P-2}}{(P-2)!} & F_{P-1}(P-1) \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & t & F_{2}(2) \\ 0 & 0 & \cdots & 1 & F_{1}(1) \\ F_{2}(L-P+1) & \cdots & \cdots & F_{P}(L-1) & F_{P}(L) \end{vmatrix} .$$
(3.13)

The entries in the last column are given by

$$F_{n+1}(n+1) = \sum_{k=0}^{n} (-1)^k \frac{t^{n-k}}{(n-k)!} + (-1)^{n+1} e^{-t}.$$
 (3.14)

A similar row reduction as for  $\Delta_i^*$  now yields

$$\Delta_P^{(P)} = \begin{vmatrix} 1 & 0 & \cdots & 0 & F_1(P-1) \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 & -F_1(2) \\ 0 & 0 & \cdots & 1 & F_1(1) \\ F_2(L-P+1) & \cdots & \cdots & F_P(L-1) & F_P(L) \end{vmatrix} .$$
(3.15)

Indeed, the reduction of the k-th row from the bottom leads to

$$F_k(k) + \sum_{r=1}^{k-1} (-1)^r \frac{t^{k-r}}{(k-r)!} F_1(r) = -(-1)^k F_1(k).$$
(3.16)

The result is

$$\Delta_P^{(P)} = F_P(L) + \sum_{k=1}^{P-1} (-1)^k F_1(k) F_{P-k+1}(L-k).$$
(3.17)

Using the relation

$$\sum_{k=r+1}^{P-1} (-1)^k F_{P-k+1}(L-k) = (-1)^{P-1} F_1(L-P+1) + (-1)^{r-1} F_{P-r}(L-r),$$
(3.18)

this can be written as

$$\Delta_{P}^{(P)} = (-1)^{P-1} F_{1}(L-P+1) + -e^{-t} \sum_{r=0}^{P-2} \frac{t^{r}}{r!} \left[ (-1)^{P-1} F_{1}(L-P+1) + (-1)^{r-1} F_{P-r}(L-r) \right].$$
(3.19)

Inserting into (3.5) we obtain the following expression for the probability of  $\mathcal{E}_t^{(P)}$ :

$$\mathbb{P}(\mathcal{E}_{t}^{(P)}) = F_{1}(P-1)F_{1}(L-P+1) + (-1)^{P-1}e^{-t}\sum_{k=0}^{P-2}\frac{(-t)^{k}}{k!}\left[(P-k-1)F_{P-k+1}(L-k) + F_{P-k}(L-k)\right].$$
(3.20)

Using the properties of functions  $F_p(n)$  and several combinatoric identities (see Appendix ), we obtain finally

$$\mathbb{P}(\mathcal{E}_{t}^{(P)}) = F_{1}(P-1)F_{1}(L-P+1) \\ -e^{-2t}\sum_{r=L}^{\infty} \frac{t^{r}}{r!} \left[ (P-1)\binom{L-1}{P} - \binom{L-1}{P-1} \\ -r\binom{L-2}{P-1} + \binom{r+1}{P} \right].$$
(3.21)

## 4 Analysis of the result

Figure 1 shows a plot of  $\mathbb{P}(\mathcal{E}_t^{(P)})$  for P = 2 and a number of values of L. It is clear that the probability increases from 0 to 1 as t increases, as it should.



Figure 1: The probability of at least one of two particles reaching the end of an interval of length L = 4, 6, 8, 10 and 12, as a function of time.

We can rewrite (3.21) in a more symmetric way as follows:

$$F_{1}(P-1)F_{1}(L-P+1) =$$

$$= \left(F_{1}(P) + \frac{t^{P-1}}{(P-1)!}e^{-t}\right) \left(F_{1}(L-P) - \frac{t^{L-P}}{(L-P)!}e^{-t}\right)$$

$$= F_{1}(P)F_{1}(L-P) + \frac{t^{P-1}}{(P-1)!}\sum_{k=L-P}^{\infty} \frac{t^{k}}{k!}e^{-2t}$$

$$- \frac{t^{L-P}}{(L-P)!}\sum_{k=P}^{\infty} \frac{t^{k}}{k!}e^{-2t}$$

$$= F_{1}(P)F_{1}(L-P) + \sum_{r=L}^{\infty} \frac{t^{r}}{r!} \left[\binom{r}{P-1} - \binom{r}{L-P}\right].$$

$$(4.1)$$

Inserting this, we get

$$\mathbb{P}(\mathcal{E}_{t}^{(P)}) =$$

$$= F_{1}(P)F_{1}(L-P) - e^{-2t}\sum_{r=L}^{\infty} \frac{t^{r}}{r!} \times \left[ \left( \frac{P(L-P)}{L} - 1 - \frac{P(L-P)}{L(L-1)}r \right) \binom{L}{P} + \binom{r}{P} + \binom{r}{L-P} \right].$$

$$(4.3)$$

$$(4.4)$$

This formula is manifestly symmetric under exchange of particles and holes, i.e.  $P \leftrightarrow L - P$ . As a particular case we have

$$\mathbb{P}(\mathcal{E}_t^{(L-1)}) = \mathbb{P}(\mathcal{E}_t^{(1)}) = F_1(L).$$
(4.5)

It is clear from (3.21) that  $\mathbb{P}(\mathcal{E}_t^{(P)})$  is bounded by 1. In fact,

$$(P-1)\binom{L-1}{P} - \binom{L-1}{P-1} - r\binom{L-2}{P-1} + \binom{r+1}{P} > 0$$
(4.6)

for  $r \geq L$ . This is easily seen by induction, as it is zero for r = L - 1and increases in r. The same relation is also useful to prove that  $\mathbb{P}(\mathcal{E}_t^{(P)})$  is increasing. Indeed, the derivative is given by

$$F_{0}(L-P)F_{1}(P-1) + F_{1}(L-P+1)F_{0}(P-2) + 2e^{-2t}\sum_{r=L}^{\infty} \frac{t^{r}}{r!} \left[ (P-1)\binom{L-1}{P} - \binom{L-1}{P-1} - r\binom{L-2}{P-1} + \binom{r+1}{P} \right] \\ -e^{-2t}\sum_{r=L}^{\infty} \frac{t^{r-1}}{(r-1)!} \left[ (P-1)\binom{L-1}{P} - \binom{L-1}{P-1} - \binom{L-1}{P-1} - r\binom{L-2}{P-1} + \binom{r+1}{P} \right] \\ = e^{-2t}\sum_{r=L-1}^{\infty} \frac{t^{r}}{r!} \left[ \binom{r}{L-P} + \binom{r}{P-2} \right] \\ +e^{-2t}\sum_{r=L}^{\infty} \frac{t^{r}}{r!} \left[ (P-1)\binom{L-1}{P} - \binom{L-1}{P-1} - \binom{r+1}{P-1} \right] \\ -(r-1)\binom{L-2}{P-1} + \binom{r+1}{P} - \binom{r+1}{P-1} \right] \\ -e^{-2t}\frac{t^{L-1}}{(L-1)!} \left[ (P-1)\binom{L-1}{P} - \binom{L-1}{P-1} - \binom{L-1}{P-1} - L\binom{L-2}{P-1} + \binom{L+1}{P} \right].$$
(4.7)

The term r = L - 1 in first line of the final expression compensates the last line, and the middle line is positive because

$$\binom{r+1}{P} - \binom{r+1}{P-1} = \binom{r}{P} - \binom{r}{P-2},\tag{4.8}$$

and the last combinatorial factor is also compensated by the expression in the first line. It is now clear that  $\mathbb{P}(\mathcal{E}_t^{(P)})$  must increase from 0 at t = 0 to 1 as  $t \to \infty$ .

It is natural to scale the time with L. It is not difficult to see that at constant P,  $\mathbb{P}(\mathcal{E}_{Lt}^{(P)})$  tends to a step function as  $L \to \infty$ . Indeed, the maximum term in

$$F_1(L - P + 1, Lt) = \sum_{k=L-P+1}^{\infty} \frac{(Lt)^k}{k!} e^{-Lt}$$
(4.9)

is attained for k = L - P + 1 if t < 1 and for  $k \approx Lt$  for t > 1 so that

$$\lim_{L \to \infty} F_1(L - P + 1, Lt) = \begin{cases} 0 & \text{if } t < 1, \\ 1 & \text{if } t > 1. \end{cases}$$
(4.10)

Moreover,  $F_1(P-1, Lt) \rightarrow 1$  and the second term tends to zero.

A more interesting limit is the thermodynamic limit, where both t and P scale with L. This can be analysed roughly as follows. We write  $t = L\tau$  and  $P = \rho L$ . Clearly,  $F_1(P-1) \sim 1_{\{\tau > \rho\}}$  and  $F_1(L-P+1) \sim 1_{\{\tau > 1-\rho\}}$  so

$$F_1(P-1)F_1(L-P+1) \sim 1_{\{\tau > \rho \lor 1-\rho\}}.$$
(4.11)

In analysing the second term of (3.21), we may assume  $L - P \ge P$ . We have seen that the second term is positive and therefore bounded by

$$e^{-2t} \sum_{r=L}^{\infty} \frac{t^r}{r!} \binom{r+1}{P} \sim e^{-2t} \left( \frac{t^P}{P!} + \frac{t^{P-1}}{(P-1)!} \right) \sum_{r=L-P}^{\infty} \frac{t^r}{r!} \\ \sim e^{-t} \left( \frac{t^P}{P!} + \frac{t^{P-1}}{(P-1)!} \right) \to 0$$
(4.12)

if  $\tau > 1 - \rho$ . Otherwise, the convergence is even faster.

The next interesting question is, what happens in the neighbourhood of  $\tau = 1 - \rho$  (assuming  $\rho < \frac{1}{2}$ ). The correct scaling is then presumably with  $\sqrt{L}$ . The following figure shows graphs of  $\mathbb{P}(\mathcal{E}_{(1-\rho)L+\sqrt{L\tau}}^{\rho L})$  as a function of  $\tau$  for  $\rho = 1/3$  and a number of values of L.



Figure 2: The probability  $\mathbb{P}(\mathcal{E}_t)$  for L = 6 (blue), 30 (red) and 90 (green), as a function of  $\tau$  where  $t = L\rho + \sqrt{L\tau}$  and P = L/3.

It suggests that there exists a constant  $\xi$  (depending on  $\rho$ ) such that

$$\mathbb{P}(\mathcal{E}^{\rho L}_{(1-\rho)L+\sqrt{L}\tau}) \to \int_{-\infty}^{\tau} e^{-t^2/2\xi} \frac{dt}{\sqrt{2\pi\xi}}.$$
(4.13)

Assuming  $\rho > \frac{1}{2}$ , we insert  $t = L\rho + \sqrt{L\tau}$  into  $F_1(P-1)F_1(L-P+1)$ . The second factor is very close to 1. The first factor can be approximated as follows:

$$e^{-t} \sum_{k=0}^{P} \frac{t^{k}}{k!} \approx e^{-L\rho - \sqrt{L\tau}} \sum_{n=0}^{L\rho} \frac{(L\rho + \sqrt{L\tau})^{L\rho - n}}{(L\rho - n)^{L\rho - n} e^{-L\rho + n} \sqrt{2\pi(L\rho - n)}}$$

$$= \sum_{n=0}^{L\rho} \left( \frac{L\rho + \sqrt{L\tau}}{L\rho - n} \right)^{L\rho - n} \frac{e^{-n - \sqrt{L\tau}}}{\sqrt{2\pi L\rho}}$$

$$\approx \sum_{n=0}^{L\rho} \exp\left[ (L\rho - n) \left( \frac{\tau}{\rho\sqrt{L}} + \frac{n}{\rho L} - \frac{\tau^{2}}{2\rho^{2}L} + \frac{n^{2}}{2\rho^{2}L^{2}} \right) \right] \frac{e^{-n - \sqrt{L\tau}}}{\sqrt{2\pi\rho L}}$$

$$\approx \frac{1}{\sqrt{2\pi\rho L}} \sum_{n=0}^{\infty} \exp\left[ -\frac{n\tau}{\sqrt{L\rho}} - \frac{n^{2}}{2L\rho} - \frac{\tau^{2}}{2\rho} \right]$$

$$\approx \int_{\tau}^{\infty} e^{-x^{2}/2\rho} \frac{dx}{\sqrt{2\pi\rho}}.$$
(4.14)

The second term in (3.21) still does not contribute in this limit, so (4.13) holds with  $\xi = \rho$ .

Notice that there is one exception: if  $\rho = \frac{1}{2}$  the both factors behave like (4.14), so the result for  $\mathbb{P}(\mathcal{E}^{\rho L}_{(1-\rho)L+\sqrt{L\tau}})$  is the square of the error function.

## 5 Appendix

Using the general formula

$$F_{p}(n) = \sum_{k=0}^{p-1} (-1)^{p-k+1} \frac{t^{k}}{k!} \binom{n-k-1}{p-k-1} + (-1)^{p} e^{-t} \sum_{k=0}^{n-p} \binom{n-k-1}{p-1} \frac{t^{k}}{k!}, \quad (5.1)$$

valid for  $n \ge p$ , we can rewrite the second term in (3.20) in a more convenient form. We have

$$(-1)^{P-1} \sum_{k=0}^{P-2} \frac{(-t)^{k}}{k!} (P-k-1) F_{P-k+1}(L-k)$$

$$= -\sum_{k=0}^{P-2} \frac{t^{k}}{k!} (P-k-1) \sum_{l=0}^{P-k} \frac{(-t)^{l}}{l!} \binom{L-k-l-1}{P-k-l}$$

$$+e^{-t} \sum_{k=0}^{P-2} \frac{t^{k}}{k!} (P-k-1) \sum_{l=0}^{L-P-1} \frac{t^{l}}{l!} \binom{L-k-l-1}{P-k}$$

$$= -\sum_{r=0}^{P} \frac{t^{r}}{r!} \binom{L-r-1}{P-r} \sum_{k=0}^{r\wedge(P-2)} (-1)^{r-k} (P-k-1) \binom{r}{k}$$

$$+e^{-t} \sum_{k=0}^{P} \frac{t^{k}}{k!} (P-k-1) \sum_{l=0}^{L-P-1} \frac{t^{l}}{l!} \binom{L-k-l-1}{P-k} + \frac{t^{P}}{P!} \sum_{l=0}^{L-P-1} \frac{t^{l}}{l!}.$$
(5.2)

In the first term we now use the simple identities

$$\sum_{k=0}^{r} (-1)^{r-k} \binom{r}{k} = 0 \tag{5.3}$$

if r > 0, and

$$\sum_{k=0}^{r} (-1)^{r-k} k \binom{r}{k} = 0$$
(5.4)

if r > 1 to write (for  $P \ge 2$ )

$$(-1)^{P-1} \sum_{k=0}^{P-2} \frac{(-t)^{k}}{k!} (P-k-1) F_{P-k+1}(L-k)$$

$$= -(P-1) \binom{L-1}{P} + \binom{L-2}{P-1} t - \frac{t^{P}}{P!}$$

$$+ e^{-t}(P-1) \sum_{r=0}^{L-1} \frac{t^{r}}{r!} \sum_{k=0 \lor (P+r+1-L)}^{P \land r} \binom{r}{k} \binom{L-r-1}{P-k}$$

$$- e^{-t} \sum_{k=1}^{P} \frac{t^{k}}{(k-1)!} \sum_{l=0}^{L-P-1} \frac{t^{l}}{l!} \binom{L-k-l-1}{P-k} + \frac{t^{P}}{P!} \sum_{l=0}^{L-P-1} \frac{t^{l}}{l!}. \quad (5.5)$$

Rewriting the last but one term as

$$\sum_{r=0}^{L-2} \frac{t^{r+1}}{r!} \sum_{k'=0 \lor (P+r+1-L)}^{r \land (P-1)} \binom{r}{k'} \binom{L-r-2}{P-k'-1},$$
(5.6)

and using the identity

$$\sum_{k=0\vee(p+r-n)}^{r\wedge p} \binom{n-r}{p-k} \binom{r}{k} = \binom{n}{p}$$
(5.7)

we obtain

$$(-1)^{P-1} \sum_{k=0}^{P-2} \frac{(-t)^k}{k!} (P-k-1) F_{P-k+1}(L-k)$$
  
=  $-(P-1) \binom{L-1}{P} + \binom{L-2}{P-1} t - \frac{t^P}{P!}$   
 $+e^{-t} \sum_{r=0}^{L-1} \frac{t^r}{r!} \left[ (P-1) \binom{L-1}{P} - r\binom{L-2}{P-1} \right] + \frac{t^P}{P!} \sum_{l=0}^{L-P-1} \frac{t^l}{l!}.$   
(5.8)

A similar analysis yields

$$(-1)^{P-1} \sum_{k=0}^{P-2} \frac{(-t)^k}{k!} F_{P-k}(L-k) = \binom{L-1}{P-1} - \frac{t^{P-1}}{(P-1)!} -e^{-t} \sum_{r=0}^{L-1} \frac{t^r}{r!} \binom{L-1}{P-1} + \frac{t^{P-1}}{(P-1)!} \sum_{l=0}^{L-P} \frac{t^l}{l!}.$$
(5.9)

The complete result for the second term of (3.20) is

$$(-1)^{P-1}e^{-t}\sum_{k=0}^{P-2}\frac{(-t)^{k}}{k!}[(P-k-1)F_{P-k+1}(L-k) + F_{P-k}(L-k)]$$

$$= -e^{-t}\left\{(P-1)\binom{L-1}{P} - \binom{L-1}{P-1} - \binom{L-2}{P-1}t + \frac{t^{P-1}}{(P-1)!} + \frac{t^{P}}{P!}\right\}$$

$$+e^{-2t}\sum_{r=0}^{L-1}\frac{t^{r}}{r!}\left[(P-1)\binom{L-1}{P} - r\binom{L-2}{P-1} - \binom{L-1}{P-1}\right]$$

$$+\frac{t^{P-1}}{(P-1)!}\sum_{l=0}^{L-P}\frac{t^{l}}{l!} + \frac{t^{P}}{P!}\sum_{l=0}^{L-P-1}\frac{t^{l}}{l!}.$$
(5.10)

Next we expand the  $e^{-t}$  term:

$$-e^{-t}\left\{ (P-1)\binom{L-1}{P} - \binom{L-1}{P-1} - \binom{L-2}{P-1}t + \frac{t^{P-1}}{(P-1)!} + \frac{t^P}{P!} \right\}$$

$$= -e^{-2t}\left(\sum_{r=0}^{L-1}\frac{t^r}{r!} + \sum_{r=L}^{\infty}\frac{t^r}{r!}\right) \left[ (P-1)\binom{L-1}{P} - \binom{L-1}{P-1} \right]$$

$$+e^{-2t}\binom{L-2}{P-1}\left(\sum_{r=0}^{L-2}\frac{t^{r+1}}{r!} + \sum_{r=L-P}^{\infty}\frac{t^r}{(r-1)!}\right)$$

$$-e^{-2t}\frac{t^P}{P!}\left(\sum_{r=0}^{L-P-1}\frac{t^r}{r!} + \sum_{r=L-P}^{\infty}\frac{t^r}{r!}\right)$$

$$-e^{-2t}\frac{t^{P-1}}{(P-1)!}\left(\sum_{r=0}^{L-P}\frac{t^r}{r!} + \sum_{r=L-P+1}^{\infty}\frac{t^r}{r!}\right).$$
(5.11)

It is clear that the terms up to order L-1 cancel the  $e^{-2t}$  contribution, and we find

$$\mathbb{P}(\mathcal{E}_{t}^{(P)}) = F_{1}(P-1)F_{1}(L-P+1) \\ -e^{-2t}\sum_{r=L}^{\infty}\frac{t^{r}}{r!}\left[(P-1)\binom{L-1}{P} - \binom{L-1}{P-1} \\ -r\binom{L-2}{P-1} + \binom{r+1}{P}\right].$$
(5.12)

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