

Notes about Complex Systems' presentation

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Contents

1	RW with absorbing barriers: discrete time and space	1
1.1	Solutions for $r = 1/2$ and $r \neq 1/2$	3
1.2	Thermo limit	4
2	FP: Stationary diffusion with absorbing barriers	4
2.1	Stationary solution in the diffusive case	5
2.2	Probability to leave from X_1	7
3	Isotropic RW with a trap	8
3.1	First passage probability	8
3.2	Mean position	10
4	Anisotropic RW with a trap	10
4.1	First passage time	11
4.2	Average time before being trapped	12
5	Compact Direct Percolation (DCP)	13
5.1	Introduction	13
5.2	β'	14
5.3	β	14
5.4	θ	15
5.5	ν_{\perp} and ν_{\parallel}	15

1 RW with absorbing barriers: discrete time and space

Definition (Absorbing barrier). An absorbing barrier is a state i such that in the corresponding Markov Chain (MC) has $W_{ii} = 1$. Hence, from normalization, $W_{ji} = 0 \quad \forall j \neq i$.

Setup: we have a 1D random walker in a system of N discrete points. Both in 1 and N we have absorbing barriers. The corresponding transition probability matrix is the follow:

$$W = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & \cdots & 0 & 0 & 0 \\ 0 & 0 & 1-r & 0 & 0 & \cdots & 0 & 0 & 0 \\ 0 & r & 0 & 1-r & 0 & \cdots & 0 & 0 & 0 \\ 0 & 0 & r & 0 & 1-r & \cdots & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & 0 & \cdots & 0 & 0 & 1 \end{pmatrix}$$

where $W_{i+1,i} = r$ and $W_{i-1,i} = 1-r$. Note the absorbing conditions on 1 and N .

Remark. Also note the convention on indices: the row index, the first one from the left, is relative to the final state while the column index, the second one, is relative to the initial state.

For any $j \in [1, N] \subset \mathbb{N}$ the walker has probability 1 to be absorbed by one of the barriers¹. We can conclude that from the structure of W the MC is irreducible but, apart of the persistent states 1 and N , all the others are transient

Definition. A MC is said to be *irreducible* if all states are accessible from any other states. Formally, there exist a t such that $W_{j,i}^t > 0 \quad \forall j, i$. A state i is said to be *persistent* if $W_{i,i} = 1$. On the other hand, is said to be *transient* if $W_{i,i} < 1$.

Which is th probability that a RW starting in j will reach 1 without being absorbed by N ?

Let p_j = time-dependent prob to be trapped at 1 given that the RW started in j at $t = 0$. For $j \in [2, N-1]$ it obeys the equation:

$$\begin{aligned} p_j &= W_{j+1,j} p_{j+1} + W_{j-1,j} p_{j-1} \\ &= r p_{j+1} + (1-r) p_{j-1} \end{aligned} \tag{1}$$

which has the same formal structure of the update rule for the probability distribution of a RW on a ring but with the boundary conditions $p_1 = 1$ and $p_N = 0$, because of its definition. Note also that W is actually a transition rate when considering discrete and unitary time steps, as in our case.

A simple derivation of (1) is given in the book "Probability and Random Processes" by G. R. Grimmett and D. R. Stirzaken²: Let A be the event for which the walker is absorbed in 1 and B the event related to take a step on the right, a.k.a. $j+1$. Let $\mathbb{P}_k(A)$ the probability

¹See P. Brémaud, "Markov chains".

²Chap. 1.7, example 4, "Symmetric random walk" or "Gambler's ruin".

of A given that we are in the k state (position), hence we can write, denoting the universe space as Σ :

$$\begin{aligned}\mathbb{P}_k(A) &= \mathbb{P}_k(A \cap \Sigma) = \mathbb{P}_k(A \cap (B \cup B^c)) = \mathbb{P}_k((A \cap B) \cup (A \cap B^c)) \\ &= \mathbb{P}_k(A|B)\mathbb{P}_k(B) + \mathbb{P}_k(A|B^c)\mathbb{P}_k(B^c)\end{aligned}$$

Now we look into the detail of $\mathbb{P}_k(A|B)$, probability to be absorbed in 1 given that the walker jumped into $k+1$ state; basically, after the walker go to the state $k+1$ we have to compute the probability that, from $k+1$ position, the walker is absorbed in 1. But this is exactly $\mathbb{P}_{k+1}(A)$, thus $\mathbb{P}_k(A|B) = \mathbb{P}_{k+1}(A)$. Similarly $\mathbb{P}_k(A|B^c) = \mathbb{P}_{k-1}(A)$, given that B^c represent the event in which the walker goes to its left, a.k.a. $k-1$. Lastly, we can identify $\mathbb{P}_k(B) = W_{k+1,k}$ and this proves (1).

1.1 Solutions for $r = 1/2$ and $r \neq 1/2$

We assume $p_k = \psi^k$, i.e the parameter ψ to the k , hence

$$\psi^j = r\psi^{j+1} + (1-r)\psi^{j-1} \Rightarrow 0 = r\psi^2 - \psi + (1-r)$$

which has the solutions (just substitute):

- $\psi_1 = 1 \Rightarrow p_j^{(1)} = 1$
- $\psi_2 = (1-r)/r := s \Rightarrow p_j^{(2)} = s^j$

Remark. These solutions are such thanks to the ansatz $p_k = \psi^k$

These are degenerate when $r = 1/2$; in this case it appears another solution which is $p_j^{(2)} = j$ (here we are calling 2 the second non degenerate solution). This is a solution because the equation (1) becomes $p_j = (p_{j+1} + p_{j-1})/2$.

For $r \neq 1/2$, a.k.a. *asymmetric case*, the general solution of the equation is given by the linear combination of independent solutions, hence:

$$p_j^{as} = As^j + B$$

where A and B are constants which are fixed by the boundary conditions:

$$\begin{aligned}p_1^{as} &= As + B = 1 \\ p_N^{as} &= As^N + B = 0 \\ \Rightarrow A &= \frac{1}{s(1-s^{N-1})} \quad B = -\frac{s^{N-1}}{(1-s^{N-1})}\end{aligned}$$

Thus the solution for the asymmetric case is

$$\boxed{p_j^{as} = \frac{s^{j-1} - s^{N-1}}{1 - s^{N-1}}} \tag{2}$$

Remark. This is the one, together with the symmetric solution, that will be tested numerically

For $r = 1/2$, a.k.a. *symmetric case*, the general solution is, as before, a linear combination,

$$p_j^s = A j + B$$

and the boundary conditions gives:

$$A = -\frac{1}{N-1} \quad B = \frac{N}{N-1}$$

resulting the the solution for the symmetric case:

$$\boxed{p_j^s = \frac{N-j}{N-1}} \quad (3)$$

1.2 Thermo limit

For $r > 1/2$ (or $s < 1$) and in the limit $N \rightarrow \infty$, we have:

$$A = 1/s \quad B = 0 \quad \Rightarrow \quad \boxed{p_j = s^{j-1}} \quad (4)$$

hence the probability of being absorbed in 1 starting at j decreases exponentially with j because the system has a bias ($r > 1/2$) toward the other absorbing barrier with is at infinity. On the other hand, if $r < 1/2$ we have $p_j = 1 \quad \forall j$ because the system is driven toward the absorbing 1 state. Stated in other words, eventually the RW will remain trapped in 1.

The same conclusion can be drawn for $r = 1/2$ because the unbiased RW is such that any state is visited from any other in a finite amount of time, i.e. from every j eventually one arrives at 1. This can be seen by the symmetric solution or by showing that the unbiased RW is an irreducible MC. Hence we can conclude:

$$p_j = 1 \quad r \leq \frac{1}{2} \quad \text{and} \quad N \rightarrow \infty$$

2 FP: Stationary diffusion with absorbing barriers

In the context of continuous time ad space stochastic processes we can introduce as a continuous approximation, in space and time, for the master equation the Fokker-Planck equation. Without entering into details, the general form of the FP equation for the probability density $P(X, t)$ is the following:

$$\frac{\partial P(X, t)}{\partial t} = -\frac{\partial}{\partial X}(a(X, t)P(X, t)) + \frac{1}{2} \frac{\partial^2}{\partial X^2}(b^2(X, t)P(X, t))$$

where $a(X, t)$ is the generalized drift component while $b(X, t)^2/2$ is the generalized diffusion coefficient.

The FP can be express as a continuity eq:

$$\frac{\partial P(X, t)}{\partial t} + \frac{\partial J(X, t)}{\partial X} = 0 \quad (5)$$

where

$$J(X, t) = a(X, t)P(X, t) - \frac{1}{2} \frac{\partial}{\partial X} (b^2(X, t)P(X, t))$$

which can be interpreted as a probability current. Let $X \in \mathbb{R}$, we consider an interval $I = [X_1, X_2]$ and define the probability that the stochastic process described by (5) is in I , $\mathcal{P}(t)$,

$$\mathcal{P}(t) = \int_I P(X, t) dX$$

Integrating (5) in X , we obtain

$$\frac{\partial \mathcal{P}(t)}{\partial t} = J(X_1, t) - J(X_2, t)$$

This form is quite useful because, usually, in numerical simulation one woks with stochastic processes confined in finite intervals. To this situation, we can apply different boundary conditions:

- Reflective barriers
E.g. $J(X_1, t) = J(X_2, t) = 0 \quad \forall t$, i.e. zero probability flow outside as well as inside the interval, which correspond to a conservation of the total probability in that interval.
- Absorbing barriers
Hence $P(X_1, t) = P(X_2, t) = 0 \quad \forall t$.
- Periodic Boundaries
Hence $P(X_1, t) = P(X_2, t)$ and $J(X_1, t) = J(X_2, t)$.

2.1 Stationary solution in the diffusive case

The stationary distribution $P^*(X)$ is defined as:

$$\begin{aligned} \frac{\partial P(X, t)}{\partial t} &\Rightarrow \frac{\partial J(X, t)}{\partial X} = 0 \\ &\Rightarrow \frac{d}{dX} (a(X)P^*(X)) - \frac{1}{2} \frac{d^2}{dX^2} (b^2(X)P^*(X)) = 0 \end{aligned} \quad (6)$$

Note that here we have only total derivatives. Moreover, it was assumed that $a(X)$ and $b(X)$ don't depend on time anymore. Now we consider the case in which $a = 0$ and $b^2 = 2D$ a.k.a. the *diffusive case*; the general solution is given integrating two times w.r.t. X obtaining:

$$P^*(X) = C_1 X + C_2 \quad (7)$$

where C_1, C_2 are constant to be determine by boundary and normalization conditions.

Let's consider *reflective boundaries*.

The continuity equation reads:

$$J^*(X_i) = -D \frac{d}{dX} P^*(X_i) = 0$$

for $i = 1, 2$. From (7) we have $C_1 = 0$ and considering the normalization in I

$$\int_{X_1}^{X_2} P^*(X) dX = C_2(X_2 - X_1) = 1 \quad \Rightarrow \quad C_2 = (X_2 - X_1)^{-1}$$

Hence, in this case, the stationary probability is a constant.

Let's consider *absorbing barriers*

If both extrema are absorbing, then P^* is identically zero. This because we have the system:

$$\begin{aligned} C_1 X_1 + C_2 &= 0 \\ C_1 X_2 + C_2 &= 0 \end{aligned}$$

which has solution $C_1 = C_2 = 0$ and it's the only one because the coefficient matrix is non-singular. This reflect the fact that, at infinite time, the random walk eventually encounters an absorbing barrier.

A more interesting situation is when we consider a set of random walkers. Basically we inject at each time step a random walker in $X_0 \in I$, generating a flux of random walkers. In this case, the general solution for the stationary distribution is

$$P^*(X) = \begin{cases} C_1 (X - X_1), & \text{for } X < X_0 \\ C_2 (X_2 - X), & \text{for } X > X_0 \end{cases}$$

because it satisfy the boundary conditions $P^*(X_1) = P^*(X_2) = 0$. The probability distribution has to be continuous in X_0 , hence the following must be satisfied:

$$C_1 (X_0 - X_1) = C_2 (X_2 - X_0)$$

The extra flux has to be taken into account in the continuity equation:

$$\frac{\partial P(X, t)}{\partial t} + \frac{\partial J(X, t)}{\partial X} = F \delta(X - X_0)$$

In the stationary case, $\partial_t P(X, t) = 0$, we have that the flux must satisfy:

$$\begin{aligned} F &= \int_{X_0-\varepsilon}^{X_0+\varepsilon} \partial_X J dX \\ &= -D \left[\frac{dP^*}{dX} \Big|_{X_0+\varepsilon} - \frac{dP^*}{dX} \Big|_{X_0-\varepsilon} \right] \\ &= -D[C_1 + C_2] \end{aligned}$$

The latter along with the continuity constrain for the P^* gives a non-homogeneous system for C_1 and C_2 which has solutions:

$$C_1 = \frac{F}{D} \frac{X_2 - X_0}{X_2 - X_1} \quad C_2 = \frac{F}{D} \frac{X_0 - X_1}{X_2 - X_1}$$

Finally the stationary solutions for the density probability and for the current are, defining $L_+ = X_2 - X_0$ and $L_- = X_0 - X_1$ and $L = X_2 - X_1$:

$$\begin{aligned} P^*(X) &= \begin{cases} \frac{F}{D} \frac{L_+}{L} (X - X_1) \\ \frac{F}{D} \frac{L_-}{L} (X_2 - X) \end{cases} \\ J^*(X) &= \begin{cases} -F \frac{L_+}{L} := -J_-, & \text{for } X < X_0 \\ F \frac{L_-}{L} := J_+, & \text{for } X > X_0 \end{cases} \end{aligned}$$

Basically, J_- and J_+ represent the fluxes of random walkers being absorbed at X_1 and X_2 respectively which satisfy the conservation of "particles" $J_+ + J_- = F$.

2.2 Probability to leave from X_1

In this framework we can introduce the probability that a particle, starting from X_0 , is absorbed at X_1 :

$$\Pi(X_1|X_0) = \frac{J_-}{F} = \frac{L_+}{L} = \frac{X_2 - X_0}{X_2 - X_1} \quad (8)$$

which is basically the exit flux from X_1 divided by the injected flux F .

Remark. We can notice that what we obtained is a restatement of (3) obtained considering continuous approach given by the FP equation.

Moreover one can obtain the average exit time from X_1 by considering the total number of random walkers divided by the total current flow out, a.k.a F .

3 Isotropic RW with a trap³

The FP associated to a 1D RW with probability density distribution $p(x, t)$ is

$$\frac{\partial p}{\partial t} = D \frac{\partial^2 p}{\partial x^2}$$

With the initial condition $p(x, 0) = \delta(x - x_0)$ where $x_0 = x(0)$, the solution to the latter is a gaussian:

$$p_{x_0}(x, t) = \frac{1}{\sqrt{4\pi Dt}} \exp\left(-\frac{(x - x_0)^2}{4Dt}\right)$$

Note: this can be obtained by Fourier transforming the diffusion equation into the \vec{k} space, impose the initial condition and go back to the real space.

Let's introduce a trap in $x_t = 0$.

Definition. A trap in x_t , is defined as the point in which $p(x_t, t) = 0 \quad \forall t$.

This condition is satisfied by taking a linear combination of the previous solution such that:

$$p(x, t) = \frac{1}{\sqrt{4\pi Dt}} \left[\exp\left(-\frac{(x - x_0)^2}{4Dt}\right) - \exp\left(-\frac{(x + x_0)^2}{4Dt}\right) \right] \quad (9)$$

which satisfy the definition. A plot for a fixed time is the one in figure 1.

3.1 First passage probability

The first passage probability $f(t)$ is the probability that the particle is trapped in $x_t = 0$ in the time interval $[t, t + dt]$ and it's equal to the module of the current $|J|$ evaluated at x_t . Basically, this as a similar interpretation to the stationary probability for a set of random walkers to be absorbed in an absorbing barrier, i.e. equation (8). Hence using (9):

$$f(t) = \left| J(x, t) \Big|_{x=0} \right| = \left| -D \frac{\partial p(x, t)}{\partial x} \Big|_{x=0} \right| = \frac{x_0}{\sqrt{4\pi Dt^3}} \exp\left(-\frac{x_0^2}{4Dt}\right)$$

A quick check, by integrating over time from 0 to ∞ , confirm that the probability is normalized.

Note: the average trapping time, time in which the particle is *not* absorbed, $\langle t_t \rangle$ is divergent because for large $t \Rightarrow t f(t) \sim 1/\sqrt{t}$ and the definition is:

$$\langle t_t \rangle = \int_0^\infty t f(t) dt \sim \infty$$

³It's a *trap*!

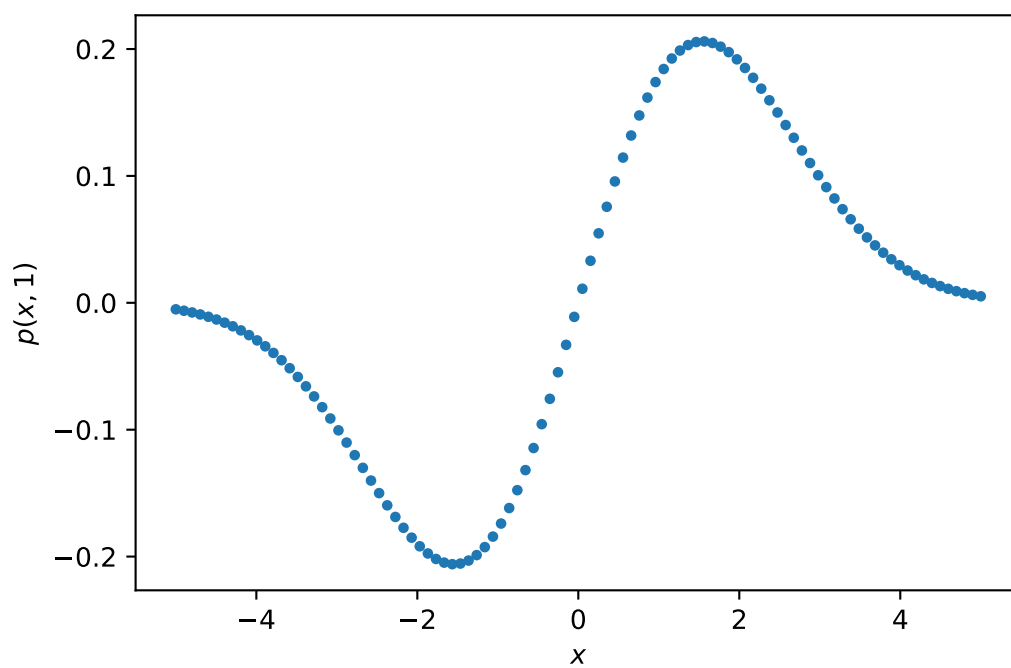


Figure 1: Probability distribution for the Isotropic RW with a trap in $x_t = 0$.

Nevertheless, the *median* trapping time t_{med} , defined as:

$$\int_{t_{med}}^{\infty} f(t) dt = \int_0^{t_{med}} f(t) dt$$

it is finite and scales as x_0^2/D (??).

3.2 Mean position

The mean position $\langle x \rangle$ for this process can be computed as follows:

$$\begin{aligned} \langle x(t) \rangle &= \int_0^{\infty} dx \frac{x}{\sqrt{4\pi Dt}} \left[\exp\left(-\frac{(x-x_0)^2}{4Dt}\right) - \exp\left(-\frac{(x+x_0)^2}{4Dt}\right) \right] \\ &= \int_0^{\infty} dx \frac{x}{\sqrt{4\pi Dt}} \exp\left(-\frac{(x-x_0)^2}{4Dt}\right) + \int_{-\infty}^0 dx \frac{x}{\sqrt{4\pi Dt}} \exp\left(-\frac{(x-x_0)^2}{4Dt}\right) \\ &= \int_{-\infty}^{\infty} dx \frac{x}{\sqrt{4\pi Dt}} \exp\left(-\frac{(x-x_0)^2}{4Dt}\right) \\ &= x_0 \end{aligned}$$

where we performed the change of variable $x \rightarrow -x$ in the second integral of the first equality.

Remark. Notice that the mean position over time doesn't change with the presence of a trap. Hence it remains constant $\boxed{\langle x(t) \rangle = x_0}$.

4 Anisotropic RW with a trap⁴

We consider a RW with a negative drift term, i.e. $-v$ where $v > 0$ hence the RW tends to move toward left. This is described by the FP:

$$\frac{\partial p}{\partial t} = v \frac{\partial p}{\partial x} + D \frac{\partial^2 p}{\partial x^2}$$

Assuming the initial condition $p(x, 0) = \delta(x - x_0)$ and no traps, the solution is given by:

$$p_{x_0}(x, t) = \frac{1}{\sqrt{4\pi Dt}} \exp\left(-\frac{(x - x_0 + vt)^2}{4Dt}\right) \quad (10)$$

This can be obtained as in the case of the diffusion equation.

Let's introduce a trap in $x_t = 0$. In this case the solution can be obtained as in the previous

⁴Oh no, again.

chapter, by linearly combining the solution (10) to satisfy the condition $p(0, t) = 0$ for any time. Hence:

$$p(x, t) = \frac{1}{\sqrt{4\pi Dt}} \left[\exp\left(-\frac{(x - x_0 + vt)^2}{4Dt}\right) - \exp\left(\frac{vx_0}{D}\right) \exp\left(-\frac{(x + x_0 + vt)^2}{4Dt}\right) \right]$$

Note the different signs in x_0 .

4.1 First passage time

As before, the probability to be absorbed in the time interval $[t, t + dt]$ $f(t)$ is defined as the module of the current evaluated in $x_t = 0$. Since we have a drift term, the current is, see equation (6):

$$J(x, t) = -v p(x, t) - D \frac{\partial p}{\partial x}$$

Thus:

$$\begin{aligned} f(t) &= \left| J(x, t) \right|_{x=0} = D \left| \frac{\partial p}{\partial x} \right|_{x=0} \\ &= \frac{D}{\sqrt{4\pi Dt}} \left[\exp\left(-\frac{(x_0 - vt)^2}{4Dt}\right) \frac{(x_0 - vt)}{2Dt} + \exp\left(\frac{vx_0}{D}\right) \exp\left(-\frac{(x_0 + vt)^2}{4Dt}\right) \frac{(x_0 + vt)}{2Dt} \right] \\ &= \frac{x_0}{\sqrt{4\pi Dt^3}} \exp\left(-\frac{(x_0 - vt)^2}{4Dt}\right) \end{aligned}$$

Now we want to compute the probability that the particle is absorbed within t , which is basically the integral from 0 to t of $f(t)$. Hence:

$$\begin{aligned} \int_0^t f(t) dt &= \frac{x_0}{\sqrt{4\pi D}} \exp\left(\frac{x_0 v}{2D}\right) \int_0^t \frac{dt'}{(t')^{3/2}} \exp\left(-\left(\frac{x_0^2}{4D} \frac{1}{t'} + \frac{v^2}{4D} t'\right)\right) \\ &= \frac{2}{\sqrt{\pi}} e^c \int_{\bar{u}}^{\infty} du \exp\left(-\left(u^2 + \frac{c^2}{4u^2}\right)\right) \end{aligned}$$

where $\bar{u} = x_0/\sqrt{4Dt}$ and $c = vx_0/2D$ and the last passage is done using the substitution $u^2 = x_0^2/4Dt'$. For the latter integral, we introduce the *error function* $\text{erf}(t)$ which is defined for $c > 0$:

$$\int du \exp\left(-\left(u^2 + \frac{c^2}{4u^2}\right)\right) = \frac{\sqrt{\pi}}{4} \left[e^c \text{erf}\left(\frac{c}{2u} + u\right) - e^{-c} \text{erf}\left(\frac{c}{2u} - u\right) \right]$$

Hence we get:

$$\begin{aligned} \int_0^t f(t) dt &= \frac{x_0}{\sqrt{4\pi Dt^3}} \exp\left(-\frac{(x_0 - vt)^2}{4Dt}\right) \\ &= \frac{e^c}{2} \left\{ e^c \left[1 - \operatorname{erf}\left(\frac{c}{2\bar{u}} + \bar{u}\right) \right] + e^{-c} \left[1 + \operatorname{erf}\left(\frac{c}{2\bar{u}} - \bar{u}\right) \right] \right\} \end{aligned} \quad (11)$$

where $\operatorname{erf}(\infty) = 1$ is used. For $t \rightarrow \infty$ we have $\bar{u} \rightarrow 0$, hence the previous becomes $\int_0^\infty f(t) dt = 1$, as it should be. Note, this is true for every v , i.e. for every $c > 0$.

We now want to consider the equation (11) in the $t \gg 1$ limit.

In this case $\bar{u} \rightarrow \infty$ and the argument of the $\operatorname{erf}(x)$ function diverges. We use the asymptotic behavior:

$$\operatorname{erf}(x) \simeq 1 - \frac{e^{-x^2}}{\sqrt{\pi}x}, \quad \text{for } x \rightarrow +\infty$$

Getting in the limit of $\bar{u} \rightarrow 0$,

$$\begin{aligned} \int_0^t dt' f(t') &\simeq 1 + \frac{e^c}{2\sqrt{\pi}} \exp\left(-\frac{c^2}{4\bar{u}^2}\right) \left[\frac{1}{\frac{c}{2\bar{u}} + \bar{u}} - \frac{1}{\frac{c}{2\bar{u}} - \bar{u}} \right] \\ &\simeq 1 - \frac{4}{\sqrt{\pi}} \frac{e^c}{c^2} \bar{u}^3 \exp\left(-\frac{c^2}{4\bar{u}^2}\right) \end{aligned}$$

The survival probability $S(t)$ can be obtained as follow, making explicit the expression for c and \bar{u} ,

$$S(t) := 1 - \int_0^t f(t') dt' \simeq \frac{2}{\sqrt{\pi}} \frac{x_0 \sqrt{D}}{v^2} \frac{e^{\frac{vx_0}{2D}}}{t^{3/2}} \exp\left(-\frac{v^2}{4D}t\right)$$

4.2 Average time before being trapped

Let's compute the average time before being trapped:

$$\begin{aligned} \langle t_{\text{rr}} \rangle &= \int_0^\infty dt t f(t) = \frac{x_0}{\sqrt{4\pi D}} \int_0^\infty \frac{dt}{\sqrt{t}} \exp\left(-\frac{(x_0 - vt)^2}{4Dt}\right) \\ &= \frac{x_0}{\sqrt{4\pi D}} \exp\left(\frac{x_0 v}{2D}\right) \int_0^\infty \frac{dt}{\sqrt{t}} \exp\left(-\left(\frac{x_0^2}{4D} \frac{1}{t} + \frac{v^2}{4D} t\right)\right) \\ &= \sqrt{\frac{x_0^3}{\pi D v}} \exp\left(\frac{x_0 v}{2D}\right) K_{1/2}\left(\frac{x_0 v}{2D}\right) \end{aligned}$$

where $K_{1/2}(x)$ is the modified Bessel function of the second kind, where the integral representation used is the following:

$$K_\nu(z) = \frac{z^\nu}{2^{\nu+1}} \int_0^\infty t^{-\nu-1} e^{-t-z^2/4t} dt$$

was used⁵

For small asymmetry in the RW, i.e. small v , we have that $K_{1/2}(x) \simeq \sqrt{\pi/(2x)}$. Hence in this case we have:

$$\boxed{\langle t_{\text{rr}} \rangle \simeq \frac{x_0}{v} \exp\left(\frac{x_0 v}{2D}\right)} \quad \text{for } v \rightarrow 0 \quad (12)$$

5 Compact Direct Percolation (DCP)

5.1 Introduction

This section is not self-consistent, please refer to the course's notes or the reference book for more details. This is meant to summing up the computation of critical exponents in the universality class of CDP, which is different from the one of DP, using the tools developed in the previous sections.

The DCP evolution can be mapped to the evolution of the domain walls thanks to the fact that $P(0|1, 1) = 0$, because $q = 0$ and where $P(i|i-1, i+1)$ is the conditional probability of having a certain $s_i(t)$ given the states of $s_{i-1}(t-1)$ and $s_{i+1}(t-1)$ ⁶. Let $L(t)$ be the number of consecutive active sites at time t ⁷ and let the transition probabilities related to the evolution of the CDP be the following:

$$\begin{aligned} P(1|1, 1) &= P(0|0, 0) = 1, & P(0|1, 1) &= P(1|0, 0) = 0, \\ P(1|1, 0) &= P(1|0, 1) = p \end{aligned}$$

Graphically, and try to draw some example, it's easy to see that the update rule for $L(t)$ is:

$$L(t+1) = \begin{cases} L(t) + 1, & \text{with probability } p^2 \\ L(t) - 1, & \text{with probability } (1-p)^2 \\ L(t), & \text{with probability } 2p(1-p) \end{cases} \quad (13)$$

$L(t)$, which represent the behavior of domain walls, performs a random walk. In particular we are interested only in the case in which $L(t)$ changes its value. Hence we can defined a random walk for the number of consecutive active sites as a RW within

$$\begin{aligned} \text{Probability to increase by 1 (move to the right)} &= r = \frac{p^2}{p^2 + (1-p)^2} \\ \text{Probability to decrease by 1 (move to the left)} &= 1 - r = \frac{(1-p)^2}{p^2 + (1-p)^2} \end{aligned}$$

where the denominator is used for normalization. Note that since $P(0|0, 0) = 1$, the state where all the sites are inactive is an absorbing state, because the system remain trapped in

⁵See this post <https://tinyurl.com/yx49fb9t>.

⁶For more details on the notation see the course's notes or the Livi and Politi's book.

⁷This is also called $N(t) := \sum_{i=0}^{\infty} \langle s_i(t) \rangle$.

that state; the same can be drawn for the state in which all sites are active.

In the RW performed by $L(t)$ this translates into an *absorbing state* for $L(t) = 0$. Hence we want to apply all the theory previously developed.

Remark. For each exponent, related to some observable of the process, we have particular initial conditions (IC), which gives a corresponding meaning to the quantities under investigation.

5.2 β'

In this context we use as IC only a single sites active at time t ; hence $L(t)$ represent the width of the cluster at time t .

We define the order parameter $P(t)$:

$$P(t) = \langle 1 - \prod_{i=0}^{\infty} (1 - s_i(t)) \rangle$$

which represent the probability average over different realizations, which have the same IC, to have at least one active site. Rephrasing it the probability to have a cluster at time t . In the active phase, i.e. $p \geq p_c = 1/2$ ⁸, and near the critical point in the $t \rightarrow \infty$ limit we have the following behavior:

$$P(\infty) \sim (p - p_c)^{\beta'}$$

Now that we have the behavior in the thermo limit we ask: *which is the probability to grow an infinite cluster in time?* This in the context of the RW performed by $L(t)$ translates into: what is the probability to not be absorbed by the state $L(t) = 0$, a.k.a. all inactive sites?

The answer is given by the discrete biased RW studied in the first section in the thermo limit, basically using the relation (4). Hence the result is:

$$P(\infty) = 1 - p_2 = 1 - s^1 = 1 - \left(\frac{1-r}{r} \right) = \frac{2}{p^2} \left(p - \frac{1}{2} \right) = \frac{2}{p^2} (p - p_c)$$

Note, the absorbing state in (4) was in 1, hence the use of p_2 . Comparing with the definition of the critical behavior, we have that, close to criticality, $\beta' = 1$.

5.3 β

The fraction of active sites is an order parameter defined as

$$\rho(t) = \frac{N(t)}{L} = \frac{\sum_{i=0}^{\infty} \langle s_i(t) \rangle}{L}$$

⁸ $P_c = 1/2$ because if we apply the transformation $p \rightarrow 1 - p$ and viceversa we obtain another process. But exchanging active and inactive site we have that the system is the same of the beginning, i.e. we have a symmetry. Hence the critical point must be the same, thus $p_c = 1 - p_c \Rightarrow p_c = 1/2$.

Using as IC an homogeneous state with a finite $\rho(0)$, a.k.a. choosing a set of random active sites at $t = 0$ ⁹, we have near criticality and for $t \rightarrow \infty$ the following behavior for the fraction of active sites:

$$\rho(\infty) \sim (p - p_c)^\beta$$

Always from $p \geq p_c$. This latter condition is important because in the large time limit this means that the activity dominates the system leading to all active sites, hence $\rho(\infty) = 1$. This let us to conclude that $\beta = 0$.

Remark. The fact that β and β' are different is an hallmark for the different universality class respect to the DP class.

5.4 θ

For simplicity we use as IC the one active state. The θ exponent is defined as the time dependence for large time of $L(t)$ at criticality, i.e.

$$L(t) \stackrel{t \gg 1}{\sim} t^\theta$$

This can be solve looking into the continuum case, as treated in section 3 because The random walk is isotropic, since at criticality $p = p_c = 1/2$.

The IC $L(0) = 1$ translate into $x_0 = 1$ while the absorbing state $L(t) = 0$ is represented by $x_t = 0$. The average of $L(t)$ correspond to the average position of the walker which, as demonstrated in 3.2, despite the absorbing trap is constant with time. Hence $\theta = 0$.

5.5 ν_\perp and ν_\parallel

Near criticality, the space (\perp) and time (\parallel) correlation lengths diverge as power laws, respectively:

$$\begin{aligned} \xi_\perp &\sim (p - p_c)^{\nu_\perp} \\ \xi_\parallel &\sim (p - p_c)^{\nu_\parallel} \end{aligned}$$

In the context of single active site IC and pp_c , ξ_\perp represent the maximum distance attained before the process dies out and ξ_\parallel represent the average lifetime of the walker.

We start by looking at ξ_\parallel : this can be interpreted as the average time before the walker is trapped. This was computed in 4.2 in the continuum and in the limit of small drift, a.k.a. small asymmetry toward left.¹⁰ In this approximation we obtained relation (12) which reads:

$$\langle t_{\text{tr}} \rangle = \xi_\parallel \sim v^{-1}$$

⁹This IC are the one used when dealing with this order parameter.

¹⁰Note that the walk is asymmetric in this case because we are looking near the criticality.

The drift term is proportional to the asymmetry δ defined as the difference between the prob to hop to the right and the prob to hop to the left:

$$\delta = \frac{p^2 - (1-p)^2}{p^2 + (1-p)^2} = -\frac{1-2p}{1+2p(1-p)} \stackrel{p \rightarrow p_c}{\simeq} -(p_c - p)$$

Thus

$$\xi_{\parallel} \sim v^{-1} \sim |\delta|^{-1} \simeq (p_c - p)^{-1}$$

hence $\nu_{\parallel} = 1$.

On the other hand, the walker perform a RW, hence the relation between the averaged traveled distance ξ_{\perp} and the time ξ_{\parallel} is

$$\xi_{\perp} \sim \xi_{\parallel}^{-1/2} \sim (p - p_c)^{-1/2}$$

implying $\nu_{\perp} = 1/2$.

Remark. Most of the exponent obtained are equal to the one of the mean field approach of the DP process, with the notable exception $\beta' \neq \beta$.