

Partial exam

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Reminder

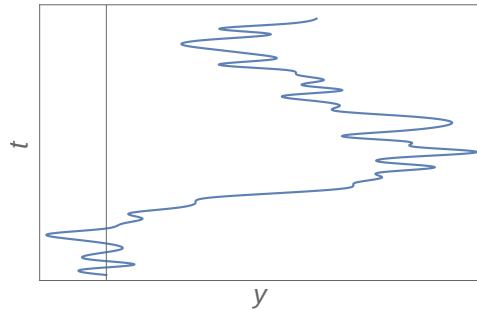
We say that a Brownian motion W_x is a Brownian motion of coordinate x and diffusion constant D when $W_0 = 0$ and

$$\langle (W_x - W_{x'})^2 \rangle = D|x - x'| \quad (1)$$

This 2-point correlation function fully characterizes the distribution of W_x . Explain why.

1 Fluctuations for a Kardar-Parisi-Zhang interface (in the directed polymer approach)

Consider a one-dimensional ‘interface’ or ‘polymer’: a trajectory $y(t)$ starting at the origin ($y(0) = 0$).



One is interested in the behaviour of the average variance at large scale t

$$B(t) = \langle y(t)^2 \rangle \quad (2)$$

It depends on the distribution of the trajectory. (Here and below we assume a symmetric distribution: $\langle y(t) \rangle = 0$).

1.1 Without disorder

A good model for an continuous interface living in an environment without disorder is to consider that $y(t)$ is a Brownian motion of coordinate t and diffusion coefficient equal to 1.

1. Determine $B(t)$. Find the *roughness exponent* ζ (defined by $B(t) \sim Ct^{2\zeta}$ as $t \rightarrow \infty$, where C is a constant).
2. What is the distribution $P(y, t)$ of the “arrival point” y of the interface at scale t ?

1.2 With disorder

To implement the influence of disorder in the environment seen by the interface, one might justify that a good model is given by the following distribution of the ‘arrival point’ y of the interface at fixed scale t :

$$P_V(y, t) \propto e^{-\frac{1}{2}\frac{y^2}{t} + V(y)} \quad (3)$$

Here $V(y)$ is a Brownian motion of coordinate y and diffusion constant D . It can be understood as a “disorder” seen by the extremity y of the interface at scale t

1. When $D = 0$, does V play a role? Do we recover the model “without disorder” of §.1.1?
2. Justify in details that the roughness, averaged over disorder, writes

$$B(t) = \left\langle \frac{\int dy y^2 e^{-\frac{1}{2}\frac{y^2}{t} + V(y)}}{\int dy e^{-\frac{1}{2}\frac{y^2}{t} + V(y)}} \right\rangle_V \quad (4)$$

where $\langle \dots \rangle_V$ represents the average over V

3. Consider a generic Brownian motion W_x of coordinate x and diffusion constant D . Using the Reminder around (1), show that

$$\langle F[W_{a\bar{x}}] \rangle_W = \langle F[a^{1/2}\bar{W}_{\bar{x}}] \rangle_{\bar{W}} \quad (5)$$

Where $\bar{W}_{\bar{x}}$ is a Brownian motion of coordinate \bar{x} and diffusion constant D . In this expression F is a function (or “functional”) which depends on many values of W_x (for instance $F[W_x] = \int dx W_x$)

4. Using the results of the two previous questions, perform in (4) the change of variable $y = t^z \bar{y}$ and find the exponent z such that (4) takes the form

$$B(t) = \left\langle \frac{\int d\bar{y} (t^z \bar{y})^2 e^{t^\alpha [-\frac{1}{2}\bar{y}^2 + \bar{V}(\bar{y})]}}{\int d\bar{y} e^{t^\alpha [-\frac{1}{2}\bar{y}^2 + \bar{V}(\bar{y})]}} \right\rangle_{\bar{V}} \quad (6)$$

where you have to determine the exponent α .

5. We admit that the following *saddle-point evaluation* is valid:

$$\int d\bar{y} f(\bar{y}) e^{t^\alpha [-\frac{1}{2}\bar{y}^2 + \bar{V}(\bar{y})]} \sim f(\bar{y}^*) e^{t^\alpha [-\frac{1}{2}(\bar{y}^*)^2 + \bar{V}(\bar{y}^*)]} \quad \text{as } t \rightarrow \infty \quad (7)$$

where \bar{y}^* is the point where $t^\alpha [-\frac{1}{2}\bar{y}^2 + \bar{V}(\bar{y})]$ reaches its maximum. Does \bar{y}^* depends on t ? Does \bar{y}^* depends on V ? We admit that \bar{y}^* does not depends on f in our case.

6. Use this result in (6) so as to obtain the roughness exponent ζ in

$$B(t) \sim Ct^{2\zeta} \quad \text{as } t \rightarrow \infty, \quad (8)$$

where C is a constant. How does ζ compare to the case without disorder? Provide a physical interpretation.

7. **[Long]** Following a similar route (*i.e.* using the Reminder around (1) so as to express $B(t)$ in a form similar to (6), but now with \bar{V} of diffusion constant 1, and D appearing elsewhere, thanks to a rescaling of the form $y = At^z \bar{y}$, with A a well-chosen constant), find how the constant C in (8) depends on D .

2 Large deviations in Markov chains

We consider a finite set of configurations \mathcal{C} and a continuous time Markov chain described by transition rates $W(\mathcal{C} \rightarrow \mathcal{C}')$.

2.1 Statistics of the activity K

We focus on histories of duration t and call K the number of events (or “jumps between configurations”) during the time window $[0, t]$. We are interested in the statistics of K .

1. Justify that

$$\partial_t P(\mathcal{C}, K, t) = \sum_{\mathcal{C}'} W(\mathcal{C}' \rightarrow \mathcal{C}) P(\mathcal{C}', K-1, t) - r(\mathcal{C}) P(\mathcal{C}, K, t) \quad (9)$$

2. From this, show in details that

$$\partial_t \langle K \rangle = \langle r(\mathcal{C}) \rangle \quad (10)$$

where you have to define the averages on both sides of this equality.

[Hint: multiply (9) by K and sum over K ; note also that $\sum_K P(\mathcal{C}, K, t) = P(\mathcal{C}, t)$ and explain why this equality holds.]

3. We introduce the Laplace transform

$$\hat{P}(\mathcal{C}, s, t) = \sum_K e^{-sK} P(\mathcal{C}, K, t) \quad (11)$$

Show that

$$\langle e^{-sK} \rangle = \sum_{\mathcal{C}} \hat{P}(\mathcal{C}, s, t) \quad (12)$$

where you have to specify what is the average on the left hand side.

4. We thus see that the study of the time behaviour of $\hat{P}(\mathcal{C}, s, t)$ gives access to $Z(s, t)$ defined as $Z(s, t) = \langle e^{-sK} \rangle$. We introduce the “cumulant generating function” $\psi(s)$ as

$$\psi(s) = \frac{1}{t} \log Z(s, t) \quad (13)$$

Show that $\psi(0) = 0$, $\psi'(0) = -\frac{1}{t} \langle K \rangle$, and $\psi''(0) = \frac{1}{t} [\langle K^2 \rangle - \langle K \rangle^2]$.

The function $\psi(s)$ is called the “cumulant generating function” (its successive derivatives in 0 provide the cumulants of K). It describes the statistics of K .

5. From (9) and (11), write the equation of evolution of $\hat{P}(\mathcal{C}, s, t)$ as

$$\partial_t \hat{P}(\mathcal{C}, s, t) = \dots \quad (14)$$

6. Introducing the vector $|\hat{P}(s, t)\rangle = \sum_{\mathcal{C}} \hat{P}(\mathcal{C}, s, t) |\mathcal{C}\rangle$ whose components are the $\hat{P}(\mathcal{C}, s, t)$, rewrite the previous evolution equation in a linear form

$$\partial_t |\hat{P}(s, t)\rangle = \mathbb{W}(s) |\hat{P}(s, t)\rangle \quad (15)$$

and show in details that the components of the *deformed* evolution operator $\mathbb{W}(s)$ are

$$\mathbb{W}(s)_{\mathcal{C}\mathcal{C}'} = e^{-s} W(\mathcal{C}' \rightarrow \mathcal{C}) - r(\mathcal{C}) \delta_{\mathcal{C}\mathcal{C}'} \quad (16)$$

7. Does $\mathbb{W}(s)$ preserve probability?
8. Show that $Z(s, t) = \langle \mathbf{1} | e^{t\mathbb{W}(s)} | P_0 \rangle$ where P_0 is the initial distribution and $\langle \mathbf{1} | = \sum_{\mathcal{C}} \langle \mathcal{C} |$.
9. Show in details that in the large time limit $\psi(s)$ is the maximal eigenvalue of $\mathbb{W}(s)$.

2.2 Statistics of the time-integrated escape rate R

We are now interested in the statistics of the time-integrated escape rate R , on a time window $[0, t]$

$$R(t) = \int_0^t d\tau r(\mathcal{C}(\tau)) \quad (17)$$

where $\mathcal{C}(\tau)$ represents the configuration at time τ .

1. [**Long** – you can admit the result.] Using the explicit expression of the probability of a trajectory (obtained by time-ordered exponential method), show that

$$\langle e^{-\sigma R} \rangle = \langle \mathbf{1} | e^{t\bar{\mathbb{W}}(\sigma)} | P_0 \rangle \quad (18)$$

where $\bar{\mathbb{W}}(\sigma)$ is a matrix of components

$$\bar{\mathbb{W}}(\sigma)_{\mathcal{C}\mathcal{C}'} = W(\mathcal{C}' \rightarrow \mathcal{C}) - (1 + \sigma)r(\mathcal{C})\delta_{\mathcal{C}\mathcal{C}'} \quad (19)$$

and P_0 the initial distribution.

2. Using this result, explain why $\bar{\psi}(\sigma)$, the cumulant generating function for R ,

$$\bar{\psi}(\sigma) = \frac{1}{t} \log \langle e^{-\sigma R} \rangle \quad (20)$$

is given by the largest eigenvalue of $\bar{\mathbb{W}}(\sigma)$.

3. Find a relation of the form $\mathbb{W}(s) = a(s)\bar{\mathbb{W}}(b(s))$ where $a(s)$ and $b(s)$ are two functions of s that you have to determine. Deduce from it a relation of the form $\psi(s) = (\dots)\bar{\psi}(\dots)$
4. What does this mean for the relation between the statistics of K and the statistics of R ? Relate the 2 first cumulants of K and the 2 first cumulants of R .