

Part II.

EVALUATION OF LARGE DEVIATION FUNCTIONS POPULATION DYNAMICS ALGORITHM

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Outline:

1. "Operator Formalism" and $\Psi(s)$ as a maximal eigenvalue
2. Evaluating $\Psi(s)$ using a "cloning algorithm"

1. "OPERATOR FORMALISM" & $\Psi(s)$ AS A MAXIMAL EIGENVALUE

1. a. Generic Observables. System with discrete configurations $\{\epsilon\}$
(ϵ even, a finite number)

* Dynamics of the system: given by transition rates $W(\epsilon \rightarrow \epsilon')$

i.e. b/w t & $t+dt$: (small dt)

$$\begin{cases} \epsilon \rightarrow \epsilon' & \text{with probability } dt W(\epsilon \rightarrow \epsilon') \\ \epsilon \rightarrow \epsilon & [\text{no event}] \end{cases} \quad 1 - \underbrace{\sum_{\epsilon'} dt W(\epsilon \rightarrow \epsilon')}$$

$$P(\epsilon, t+dt) = \sum_{\epsilon'} dt W(\epsilon' \rightarrow \epsilon) P(\epsilon', t) + \left(1 - dt \sum_{\epsilon'} W(\epsilon \rightarrow \epsilon')\right) P(\epsilon, t)$$

$$dt \rightarrow 0 \quad \frac{d}{dt} P(\epsilon, t) = \sum_{\epsilon'} W(\epsilon' \rightarrow \epsilon) P(\epsilon', t) - \underbrace{n(\epsilon)}_{= \sum_{\epsilon'} W(\epsilon \rightarrow \epsilon')} P(\epsilon, t) \quad [n(\epsilon) \text{ is called "escaperate"}]$$

* Description of the dynamics [see Tibor's lecture]

. the system stays in config ϵ for a duration Δt with prob. $n(\epsilon) e^{-\Delta t n(\epsilon)}$

. the system then jumps from $\epsilon \rightarrow \epsilon'$ with prob. $\frac{W(\epsilon \rightarrow \epsilon')}{n(\epsilon)}$

* Generic observable: $A = \sum_{k=0} \alpha(\epsilon_k \rightarrow \epsilon_{k+1})$

for a history $\epsilon_0 \dots \epsilon_K$ of the system;

i.e. at each "jump" (or event) $A \mapsto A + \alpha(\epsilon \rightarrow \epsilon')$

1.6 - The simplest case : $A = K$ (activity)

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i.e., upon each event $K \mapsto K+1$

- * evolution of the probability density of being in ϵ at time t , having observed a value K of the activity $P(\epsilon, K, t)$ before the "jump"

$$\partial_t P(\epsilon, K, t) = \sum_{\epsilon'} P(\epsilon', K-1, t) W(\epsilon' \rightarrow \epsilon) - n(\epsilon) P(\epsilon, K, t) \quad \textcircled{R}$$

the problem is difficult to solve
because this term is non-diagonal
along direction K

however note that if one obtains $P(K, t) = \sum_{\epsilon} P(\epsilon, K, t)$

As Tobias said: this def. of the generating function is akin to Fourier transforms.

- * Going to the s -ensemble : $\hat{P}(\epsilon, s, t) = \sum_{K \geq 0} e^{-sk} P(\epsilon, K, t)$

Remarks - this is a "Laplace transform" (\leftrightarrow generating functions)
- by definition $\langle e^{-sk} \rangle = \sum_{K \geq 0} \sum_{\epsilon} e^{-sk} P(\epsilon, K, t) = \sum_{\epsilon} \hat{P}(\epsilon, s, t)$ not normalized!

- More generally, $\langle \Theta(\epsilon) e^{-sk} \rangle = \sum_{\epsilon} \Theta(\epsilon) \hat{P}(\epsilon, s, t)$

From \textcircled{R} above :
$$\partial_t \hat{P}(\epsilon, s, t) = \sum_{\epsilon'} e^{-s} W(\epsilon' \rightarrow \epsilon) \hat{P}(\epsilon', s, t) - n(\epsilon) \hat{P}(\epsilon, s, t)$$

vector of components $\hat{P}(\epsilon, s, t)$ this is the only difference with the $s=0$ dynamics
 $\hat{W}(s)$ matrix

Linear evolution:
$$\partial_t |\hat{P}(s, t)\rangle = W(s) |\hat{P}(s, t)\rangle$$

$$(W(s))_{\epsilon \epsilon'} = e^{-s} W(\epsilon' \rightarrow \epsilon) - n(\epsilon) \delta_{\epsilon \epsilon'}$$

- * Eigenvectors & eigenvalue problem: of $\mathbf{W}(s)$
- let's call $\psi(s)$ the largest eigenvalue¹, of eigenvector $|\hat{P}_0(s)\rangle$ Complexity
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- In the long-time limit: $[\partial_t |\hat{P}\rangle = \mathbf{W} |\hat{P}\rangle \text{ is solved by } |\hat{P}\rangle = e^{t\mathbf{W}} |\hat{P}\rangle_{t=0}]$
- $|\hat{P}(s,t)\rangle = e^{t\mathbf{W}(s)} |\hat{P}(s,0)\rangle \xrightarrow[t \rightarrow \infty]{} e^{t\psi(s)} |\hat{P}_0(s)\rangle$ [See remark 2 below]
for a demonstration] ⊗
- Hence: all components $\hat{P}(r,s,t)$ evolve exponentially in time $\sim e^{t\psi(s)}$
- $\hookrightarrow \langle e^{-sk} \rangle \sim e^{t\psi(s)}, \psi(s) = \max_{\lambda} \text{Sp } \mathbf{W}(s)$
- $= \sum_k \hat{P}(k,s,t)$, see p II.2 Maximal eigenvalue

The problem of determining the fluctuations of k
 thus amounts to finding the extremal eigenvalue of a matrix.

One can thus borrow methods from quantum mechanics & mathematics where this sort of problems occurs rather regularly.

- * Remark 1: everything we wrote also works for the l.d.f. $\psi(s)$ of quantities $A = \sum_{k=0}^{\infty} \alpha(k \rightarrow k+1)$:
- One simply has to replace $e^{-s} W(r,s,r')$ by $e^{-s \alpha(k \rightarrow k')} W(r \rightarrow r')$
- * Remark 2: To show ⊗ above, decompose $|\hat{P}(s,0)\rangle$ on a basis where \mathbf{W} is diagonal: $|\hat{P}(s,0)\rangle = \alpha_0 |\hat{P}_0(s)\rangle + \alpha_1 |\hat{P}_1(s)\rangle + \dots + \alpha_k |\hat{P}_k(s)\rangle$ where $(\psi(s) = \lambda_0 > \lambda_1 > \dots > \lambda_k > \dots)$ are eigv. of \mathbf{W} . Then $e^{t\mathbf{W}(s)} |\hat{P}(s,0)\rangle = \underbrace{\alpha_0 e^{t\psi(s)} |\hat{P}_0(s)\rangle}_{\text{this finite sum is dominated by } e^{t\psi(s)} \text{ as } t \rightarrow \infty} + \alpha_1 e^{t\lambda_1(s)} |\hat{P}_1(s)\rangle + \dots + \alpha_k e^{t\lambda_k(s)} |\hat{P}_k(s)\rangle \dots$

2. EVALUATING $\Psi(s)$ THROUGH A "LEAVING" ALGORITHM

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Consider the equation of evolution of $\hat{P}(e, s, t) \quad \left\{ \begin{array}{l} n(e) = \sum_{e'} W(e \rightarrow e') \\ = W_s(e \rightarrow e) \end{array} \right.$

$$\partial_t \hat{P}(e, s, t) = \sum_{e'} e^{-s} W(e' \rightarrow e) \hat{P}(e', s, t) - n(e) \hat{P}(e, s, t)$$

It does not correspond to a Markov evolution of a probability $\hat{P}(e, s, t)$
because: $n(e) = \sum_{e'} W(e \rightarrow e')$ and not $\sum_{e'} W_s(e \rightarrow e')$

(And indeed: $\sum_e \hat{P}(e, s, t) \underset{t \rightarrow \infty}{\sim} e^{t \Psi(s)}$: "probability is not conserved")

2-a. Rewriting in terms of a population dynamics:

$$\partial_t \hat{P}(e, s, t) = \sum_{e'} W_s(e' \rightarrow e) \hat{P}(e', s, t) - n_s(e) \hat{P}(e, s, t) + [n_s(e) - n(e)] \hat{P}(e, s, t)$$

(Probability-preserving) evolution with
modified rates $W_s(e' \rightarrow e)$ reproduction at
a rate $n_s(e) - n(e)$

"mutations"

"selection"

Physical interpretation: [don't forget we study a large deviation]
characterized by s

- s -dependent transition rates slightly biasing the trajectories ("mutations" allowing to probe the available space of configurations)
- selection rules actually favor the configurations, which render the large deviation (determined by s) typical.

↳ In case of phase coexistence, this allows to select a phase and to study its properties

2.6. Implementation in continuous time :

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- . Start from a large number N_0 of copies of the system
 - evolve each copy of the system with rates $W_s(\epsilon \rightarrow \epsilon')$
 - ↳ . time before one exits from a configuration ϵ distributed exponentially with rate $n_s(\epsilon)$
 - . after this, change $\epsilon \rightarrow \epsilon'$ with probability $\frac{W_s(\epsilon \rightarrow \epsilon')}{n_s(\epsilon)}$
 - "clone" with rate $n_s(\epsilon) - n(\epsilon)$:
 - on each interval Δt during which ϵ does not change, replicate the copy by a factor $e^{\Delta t \cdot (n_s(\epsilon) - n(\epsilon))}$.
 - ↳ with these rules, the number $N(\epsilon, t)$ of copies of the system in configuration ϵ evolves with the same eq. as $\hat{P}(\epsilon, s, t)$
- Hence $N(\epsilon, t) \sim e^{t\psi(s)}$
- \nearrow measure the exponential evolution
 its rate yields $\psi(s)$
- Besides, average among the set of copies yields $\langle O \rangle_s$.

Remark: in practice, one uses tricks to keep total population constant!

[linked with Wright-Fisher process -
 see Tobias Galla's lecture.]

2-c Implementation in discrete time:

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$$\hat{P}(\epsilon, s, t+dt) = \underbrace{\sum_{\epsilon'} dt W_s(\epsilon' \rightarrow \epsilon) \hat{P}(\epsilon', s, t)}_{\text{change } \epsilon' \rightarrow \epsilon \text{ with probability } dt W_s(\epsilon' \rightarrow \epsilon)} + \underbrace{\left[1 - dt n_s(\epsilon)\right] \hat{P}(\epsilon, s, t) + dt \left[n_s(\epsilon) \cdot n(\epsilon)\right] \hat{P}}_{\substack{\text{make no change with probability } \\ 1 - dt n_s(\epsilon)}} + \underbrace{dt \left[n_s(\epsilon) \cdot n(\epsilon)\right]}_{\text{in any case, duplicate the copy with rate } dt \left[n_s(\epsilon) \cdot n(\epsilon)\right]}$$

Δ CAVEAT: depending on s , the choice of dt might require a very small dt !
 ↳ better to use continuous-time -