

Monotonicity and phase transition for the VRJP and the ERRW

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Definition (Diaconis, Coppersmith)

Let $\mathcal{G} = (V, E)$ be a locally-finite, connected, non-directed graph and $x_0 \in V$ a vertex of this graph. The edge-reinforced random walk with initial weights $(a_e)_{e \in E}$ is the process $(X_n)_{n \in \mathbf{N}}$ defined by $X_0 = x_0$ and:

$$\mathbb{P}(X_{n+1} = y | X_0, \dots, X_n) = \mathbf{1}_{\{X_n, y\} \in E} \frac{a_{\{X_n, y\}} + N_n(\{X_n, y\})}{\sum_{z, \{X_n, z\} \in E} a_{\{X_n, z\}} + N_n(\{X_n, z\})},$$

where

$$N_n(\{x, y\}) = \sum_{i=0}^{n-1} \mathbf{1}_{\{X_i, X_{i+1}\} = \{x, y\}}.$$

Theorem (Pemantle Merkl,Rolles Sabot,Zeng)

For $d \in \{1, 2\}$, the ERRW is recurrent for any initial weight a .

Theorem (Sabot,Tarrès Angel,Crawford,Kozma Disertori,Sabot,Tarrès)

For any $d \geq 3$, there exists $a_r, a_t \in (0, \infty)$ such that for an initial weight a the ERRW in Z^d is recurrent if $a < a_r$ and transient if $a > a_t$.

Theorem (P)

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Definition (Davis, Volkov)

Let $\mathcal{G} = (V, E)$ be a locally-finite, connected, non-directed graph and $x_0 \in V$ a vertex of this graph. The vertex reinforced jump process with initial weights $(W_e)_{e \in E}$ is the process $(Y_t)_{t \in \mathbb{R}}$ that starts at x_0 and jumps to a neighbour vertex y at a rate

$$1_{\{X_{t,y}\} \in E} W_{\{X_{t,y}\}} (1 + l_y(t)),$$

where

$$l_y(t) = \int_{s=0}^t 1_{X_s=y} ds.$$

The model

The link between ERRW and VRJP

The ERRW and the VRJP have similar behaviours. This is explained by the following result.

Proposition (Sabot, Tarrès 2013)

The ERRW on a locally finite $\mathcal{G} = (V, E)$ with initial weights $(a_e)_{e \in E}$ is a mixture of discrete time VRJP where the initial weights $(W_e)_{e \in E}$ are independent gamma random variables of parameter a_e : $W_e \sim \Gamma(a_e)$.

Proposition (Disertori, Spencer, Zirnbauer Sabot, Tarrès)

(i) The probability measure μ_n^W on $\mathcal{H}_{i_0}^n = \{u \in \mathbf{R}^n, u_{i_0} = 0\}$ is defined by the density:

$$\mu_n^{W, i_0}(du) := \left(\frac{1}{2\pi}\right)^{\frac{n-1}{2}} e^{-\sum u_i} e^{-\frac{1}{2} \sum_{i \sim j} W_{\{i,j\}} (e^{u_i - u_j} + e^{u_j - u_i} - 2)} \sqrt{|H_{W,u}|_{n-1}} du_1 \dots du_{n-1},$$

where $H_{W,u}(i, i) = \sum_{j \sim i} W_{\{i,j\}} e^{u_i + u_j}$, $H_{W,u}(i, j) = -W_{\{i,j\}} e^{u_i + u_j}$ and $|H_{W,u}|_{n-1}$ is the determinant of any minor of $H_{W,u}$.

(ii) The VRJP on a finite graph (V, E) with weights $(W_e)_{e \in E}$ is a time-changed random walk in random reversible environments. The environment is given by conductances $W_{\{x,y\}} e^{U_x + U_y}$ where the random variable U has a probability distribution given by $\mu_{|V|}^W$.

The model

A first link with the GFF

We can see that for small values of u , the density

$$\mu_n^W(du) := \left(\frac{1}{2\pi}\right)^{\frac{n-1}{2}} e^{-\sum u_i} e^{-\frac{1}{2} \sum_{i \sim j} W_{\{i,j\}} (e^{u_i - u_j} + e^{u_j - u_i} - 2)} \sqrt{|H_{W,u}|_{n-1}} du_1 \dots du_{n-1},$$

is similar to that of the GFF $(Y_x)_{x \in V}$ where we impose $Y_{i_0} = 0$:

$$\mathbf{g}_n^W(dy) := \left(\frac{1}{2\pi}\right)^{\frac{n-1}{2}} e^{-\frac{1}{2} \sum_{i \sim j} W_{\{i,j\}} (y_i - y_j)^2} \sqrt{|H_W|_{n-1}} dy_1 \dots dy_{n-1},$$

where $H_W(i, i) := \sum_{j \sim i} W_{\{i,j\}} = H_{W,0}(i, i)$ and $H_W(i, j) := -W_{\{i,j\}} = H_{W,0}(i, j)$.

Proposition

For any weights W and any vertex j we have:

$$\mathbb{E}_{\mu_n^W}(du)(e^{u_j}) = 1.$$

Proposition

For any choice of W , the law of $\frac{1}{2} \sum_{i \sim j} W_{\{i,j\}} (e^{U_i - U_j} + e^{U_j - U_i} - 2)$ is a Gamma of parameter $(n - 1)/2$.

Similarly for the GFF, the random variable $\frac{1}{2} \sum_{i \sim j} W_{\{i,j\}} (Y_i - Y_j)^2$ is also a Gamma of parameter $(n - 1)/2$.

This representation is not always practical so a new one was introduced.

Definition

Let γ be a Gamma random variable of parameters $(1/2, 1/2)$ independent of the rest.
The β -field $(\beta_x)_{x \in V}$ is defined by:

$$\forall x \in V, \beta_x := \sum_{y \sim x} W_{\{x,y\}} e^{U_y - U_x} + \gamma \mathbf{1}_{x=i_0}$$

The equivalent for the GFF is the vector $(B_x)_{x \in V}$ defined by:

$$\forall x \in V, B_x := \sum_{y \sim x} W_{\{x,y\}} (Y_y - Y_x).$$

This can also be written as $B = H_W Y$.

The law of the β -field is characterized by the following density.

Definition (Sabot, Tarrès, Zeng – Letac)

Set $n \in \mathbf{N}$ and non-negative weights $(W_{i,j})_{1 \leq i,j \leq n}$. Let $\eta \in [0, \infty)^n$ be a vector and let $\mathbf{1}_n \in \mathbf{R}^n$ be the vector with only ones. We can define the probability measure $\nu_n^{W,\eta}$ on \mathbf{R}^n by the density:

$$\nu_n^{W,\eta}(d\beta) := \left(\frac{1}{2\pi}\right)^{\frac{n}{2}} e^{-\frac{1}{2}(\mathbf{1}_n H_\beta \mathbf{1}_n + \sum \eta_j H_\beta^{-1} \eta_j - 2 \sum \eta_j)} \frac{1}{\sqrt{\text{Det}(H_\beta)}} \mathbf{1}_{H_\beta > 0} d\beta_1 \dots d\beta_n,$$

where $H_\beta(i, i) = \beta_i$, $H_\beta(i, j) = -W_{\{i,j\}}$ and $H_\beta > 0$ means that H_β is positive definite.

For the GFF, the vector $(B_x)_{x \in V}$ is a degenerate centered gaussian vector of covariance matrix H_W such that $\sum B_x = 0$.

The model

Properties of the β -field

Proposition

The law of the β -field does not depend on the starting point.

Proposition

For any non-negative coefficients $(\lambda_x)_{x \in V}$:

$$\mathbb{E} \left(e^{-\sum_{x \in V} \lambda_x \beta_x} \right) = e^{-\sum_{\{x,y\} \in E} W_{\{x,y\}} (\sqrt{1+2\lambda_x} \sqrt{1+2\lambda_y} - 1)} \prod_{x \in V} \frac{1}{\sqrt{1+2\lambda_x}}.$$

Proposition

Let V_1, V_2 be two subset of V such that for any $(x, y) \in V_1 \times V_2$, $d(x, y) \geq 2$. The beta-fields in V_1 and V_2 are independent.

Proposition

Let V_1 be a subset of V . The marginal law of the β -field on V_1 is $\nu_{|V_1|}^{W,\eta}$ for some η .

Proposition

Let V_1, V_2 be a partition of V . Let W_e be non-negative weights. Let β be distributed according to $\nu_{|V|}^W$, the law of $(\beta_x)_{x \in V_1}$ knowing $(\beta_y)_{y \in V_2}$ is $\nu_{|V_1|}^{W+W'}$ where W' are non-negative weights that do not depend on $(\beta_x)_{x \in V_1}$.

Let V_1, V_2 be a partition of V . Let W_e be non-negative weights. Let B be a centered gaussian vector of covariance matrix H_W with:

$$\begin{pmatrix} H_W^1 & -W^{12} \\ -W^{21} & H_W^2 \end{pmatrix}.$$

Let B_1, B_2 be the restriction of B to V_1 and V_2 respectively. Let \bar{B}_1 be defined by :

$$\bar{B}_1 := B_1 + W^{12} (H_W^2)^{-1} B_2.$$

The vectors \bar{B}_1 and B_2 are independent and centered. Furthermore, the value of \bar{B}_1 only depends on the GFF on V_1 .

Theorem (Sabot, Tarrès, Zeng)

Set a finite graph $\mathcal{G} = (V, E)$ and non-negative weights $(W_e)_{e \in E}$. Let H_β be distributed according to ν_n^W and let G_β be its inverse. For any $i_0 \in V$, the random walk in random reversible environment given by the random conductances $(\omega_e)_{e \in E}$ defined by:

$$\omega_{x,y} := W_{x,y} G_\beta(i_0, x) G_\beta(i_0, y)$$

has the same law as a time-change of the VRJP with initial weights $(W_e)_{e \in E}$ and starting point i_0 .

This allows us to get back U from β .

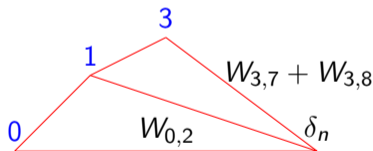
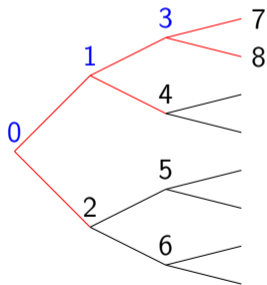
Proposition

For any $i_0, x \in V$, $\mathbb{E} \left(\frac{G_\beta(i_0, x)}{G_\beta(i_0, i_0)} \right) = 1$.

The martingale ψ

Definition

For an infinite, connected, locally finite graph $\mathcal{G} = (V, E)$ and initial weights $(W_e)_{e \in E}$ we look at the sequence of graphs $\mathcal{G}_n = (V_n, E_n)$ obtained by keeping a finite subset of \mathcal{G} and collapsing all other vertices into one vertex δ_n . We define $\psi_n(x) := \frac{G_\beta(\delta_n, x)}{G_\beta(\delta_n, \delta_n)}$.



Proposition

For some choice of coupling of the β -fields on a sequence of graphs \mathcal{G}_n that is increasing, for any $x \in V$, for n large enough:

$$\mathbb{E}(\psi_{n+1}(x) | \psi_n(x)) = \psi_n(x).$$

Since $\psi_n(x) \geq 0$, there exists a random variable $\psi_\infty(x)$ such that a.s

$$\psi_n(x) \rightarrow \psi_\infty(x).$$

Theorem (Sabot, Zeng)

If $\psi_\infty(0) = 0$ the VRJP starting at 0 is recurrent, otherwise it is transient.

Theorem ((P))

Let $G = (V, E)$ be a finite graph and $(W_e^-)_{e \in E}$ and $(W_e^+)_{e \in E}$ two families of weights such that for all $e \in E$, $0 < W_e^- \leq W_e^+$. For any $x \in V$, there exists $(\beta_i^-)_{i \in V}$ and $(\beta_i^+)_{i \in V}$ distributed according to $\nu_{|V|}^{W^-}$ and $\nu_{|V|}^{W^+}$ respectively such that:

$$\forall y \in V, \mathbb{E} \left(\frac{G_{\beta^-}(x, y)}{G_{\beta^-}(x, x)} \middle| \beta^+ \right) = \frac{G_{\beta^+}(x, y)}{G_{\beta^+}(x, x)}.$$

This is the same as saying that for any convex function f :

$$\mathbb{E}_{W^-} \left(f \left(\frac{G_{\beta}(x, y)}{G_{\beta}(x, x)} \right) \right) \geq \mathbb{E}_{W^+} \left(f \left(\frac{G_{\beta}(x, y)}{G_{\beta}(x, x)} \right) \right).$$

The proof

Reduction to 2 points

The idea is to look at what happens when we decrease a weight $W_{\{x,y\}}$.
We look at the following partition of V : $V_1 := \{x, y\}$ and $V_2 := V \setminus \{x, y\}$.

Proposition

The law of the β -field on V_2 does not depend on the weight $W_{\{x,y\}}$.

This means that we can condition on the value of the β -field on V_2 and look at what the impact of $W_{\{x,y\}}$ is on V_1 .

For the GFF, the covariance matrix of B is given by $H_{W,0}$ with $H_{W,0}(i, i) = \sum_{j \sim i} W_{\{i,j\}}$

and $H_{W,u}(i, j) = -W_{\{i,j\}}$. So the law of B on V_2 does not depend on $W_{\{x,y\}}$.

The proof

What happens for 2 points

We will look at the simple case when the graph has only two points i_0, x joined by an edge of weight w . In the general case the starting point is a mixture of the two points.

$$\frac{1}{2\pi} e^{-\sum u_x} e^{-\frac{1}{2}w(e^{u_x} + e^{-u_x} - 2)} \sqrt{we^{u_x}} du_x.$$

The law of $K := \frac{w}{2}(e^{U_x} + e^{-U_x} - 2)$ is a gamma of parameters $(1/2, 1)$.

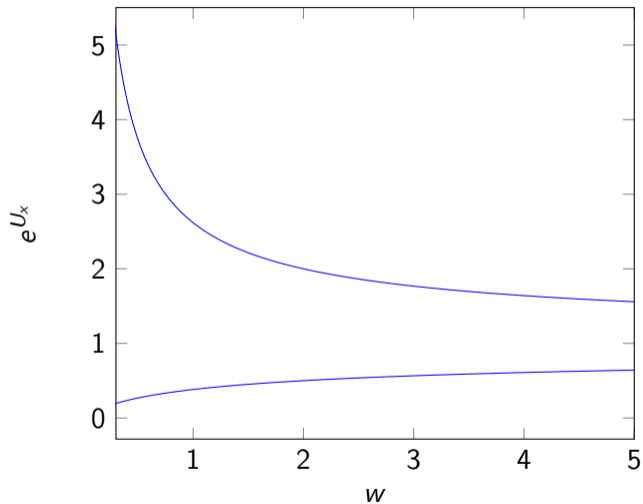
Knowing K , the expectation of e^{U_x} is 1.

Knowing K , the smaller w is, the further away from 1 the random variable U_x is.

For a gaussian Y of variance σ^2 , the equivalent of k is the random variable $\frac{1}{2\sigma^2} Y^2$ which is also a gamma of parameters $(1/2, 1)$. For the coupling of two centered gaussians Y_+ and Y_- of variances σ_+^2 and σ_-^2 , this would be the same as sending Y_- onto $\frac{\sigma_+}{\sigma_-} Y_-$ and $-\frac{\sigma_+}{\sigma_-} Y_-$.

The proof

What happens for 2 points



A few other consequences

Theorem (P)

If the simple walk on a graph (V, E) with conductances $(W_e)_{e \in E}$ is recurrent then so is the ERRW and the VRJP with initial weights $(W_e)_{e \in E}$.

Theorem (Sabot, Zeng)

On a graph (V, E) the ERRW and the VRJP with initial weights $(W_e)_{e \in E}$ are recurrent with probability 0 or 1 (almost every environment are recurrent or almost every environment are transient) if the graph and the weights are invariant by translation.

Theorem (P)

On a graph (V, E) the ERRW and the VRJP with initial weights $(W_e)_{e \in E}$ are recurrent with probability 0 or 1.

Recurrence on recurrent graphs

