

**GAUSSIAN PROCESSES**  
**EXERCISE SHEET 10: GAUSSIAN FREE FIELD II**

**Exercise 1.**

Consider the transition probability matrix  $(P_t)$  of the continuous-time random walk, defined by

$$(P_t)_{v,w} := \mathbb{P}_v(X_t = w), \quad v, w \in V \setminus V_\partial.$$

We have

$$P_t = e^{tL},$$

where  $L = -\Delta$  and  $\Delta$  is the discrete Laplacian on the graph.

Thus the Green's function can be written as

$$G = \int_0^\infty e^{tL} dt.$$

**Step 1: Symmetry.** Since  $L$  is symmetric, the matrix exponential  $e^{tL}$  is symmetric for all  $t$ . Hence

$$G = \int_0^\infty e^{tL} dt$$

is symmetric.

**Step 2: Covariance of the Gaussian Free Field.** Recall the identity

$$\frac{d}{dt}(e^{tL}L^{-1}) = e^{tL}.$$

Integrating from 0 to  $T$  yields

$$\int_0^T e^{tL} dt = (e^{TL} - I)L^{-1}.$$

Since  $e^{TL} \rightarrow 0$  as  $T \rightarrow \infty$  (equivalently,  $P_t \rightarrow 0$ ),

$$G = \int_0^\infty e^{tL} dt = -L^{-1} = \Delta^{-1}.$$

Therefore  $G$  is exactly the covariance matrix of the Gaussian Free Field.

**Step 3: Positive definiteness.** For any function  $\Gamma : V \rightarrow \mathbb{R}$  vanishing on  $V_\partial$ , we have

$$\sum_{v,w \in V \setminus V_\partial} \Gamma(v) \Delta_{v,w} \Gamma(w) = \frac{1}{2} \sum_{\substack{v,w \in V \\ v \sim w}} (\Gamma(v) - \Gamma(w))^2.$$

The right-hand side is zero if and only if  $\Gamma \equiv 0$ , so  $\Delta$  is positive definite, and therefore  $G = \Delta^{-1}$  is also positive definite.  $\square$

**Exercise 2.**

The discrete Poisson equation on  $V \setminus V_\partial$  is

$$\Delta F = f, \quad F|_{V_\partial} = 0.$$

Since  $G = \Delta^{-1}$ ,

$$F(v) = \sum_{w \in V \setminus V_\partial} G(v, w) f(w), \quad v \in V \setminus V_\partial.$$

Using the probabilistic representation of the Green's function, this becomes

$$F(v) = \mathbb{E}_v \left[ \int_0^\tau f(X_t) dt \right],$$

which is the unique solution. □

### Exercise 3.

#### (a) Covariance of the massive Gaussian free field.

The massive GFF  $\phi = (\phi_x)_{x \in V}$  is defined by the density

$$\mathbb{P}(\phi) \propto \exp \left[ -\frac{1}{2} \sum_{x \sim y} (\phi_x - \phi_y)^2 - \frac{m^2}{2} \sum_{x \in V} \phi_x^2 \right] \prod_{x \in V} d\phi_x.$$

Express the quadratic form in matrix notation. Since

$$\sum_{x \sim y} (\phi_x - \phi_y)^2 = \phi^\top \Delta \phi,$$

the density becomes

$$\mathbb{P}(\phi) \propto \exp \left[ -\frac{1}{2} \phi^\top (\Delta + m^2 I) \phi \right] \prod_{x \in V} d\phi_x.$$

Thus  $\phi$  is a centered Gaussian vector with covariance matrix

$$\text{Cov}(\phi) = (\Delta + m^2 I)^{-1},$$

i.e.

$$\text{Cov}(\phi_x, \phi_y) = (\Delta + m^2 I)^{-1}(x, y).$$

#### (b) Connection to the killed random walk.

Let  $(X_t)$  be the continuous-time simple random walk on  $G$ , and kill it independently at rate  $m^2$ . Let

$$G^m(x, y) = \int_0^\infty \mathbb{P}_x(X_t = y, \text{ walk not killed before } t) dt.$$

**Step 1: Green's function via a matrix exponential.** Note that if we add a weighted edge with weight  $m^2$  between a new death point  $v_\partial$  and all the points in  $V$ , we can directly use the idea in Exercise 1 for this problem. (The random walk is killed if and only if it gets to  $v_\partial$ .) Define

$$L_m := L - m^2 I = -(\Delta + m^2 I).$$

Then the transition matrix of the killed walk is

$$P_t^m = e^{tL_m} = e^{-t(\Delta + m^2 I)},$$

and therefore

$$G^m = \int_0^\infty P_t^m dt = \int_0^\infty e^{-t(\Delta + m^2 I)} dt.$$

**Step 2: Evaluate the integral.** Since  $\Delta + m^2 I$  is symmetric positive definite,

$$\int_0^\infty e^{-t(\Delta + m^2 I)} dt = (\Delta + m^2 I)^{-1}.$$

Thus

$$\text{Cov}(\phi_x, \phi_y) = G^m(x, y).$$

□