

# Chapter 5

## Self diffusion.

### 5.1 Motion of a tagged particle.

Let us look at the simple exclusion process in equilibrium on  $\mathbb{Z}^d$  at density  $\rho$ . The distribution is the Bernoulli distribution  $\mu_\rho$  defined by  $\mu_\rho[\eta(x) = 1] = \rho$  with  $\{\eta(x) : x \in \mathbb{Z}^d\}$  being independent. Let us suppose that at time 0, there is a particle at 0 which is tagged and observed. It is convenient to move the origin with that particle. The simple exclusion process now acts only on  $\mathbb{Z}^d - \{0\}$  and is the environment as seen by the particle. The environment changes in two different ways. When one of the other particles currently at  $x$  moves to  $y$ . The generator for this part is

$$\mathcal{A}_1 = \frac{1}{2} \sum_{x,y \neq 0} \pi(y-x)[f(\eta^{x,y}) - f(\eta)] \quad (5.1)$$

Or the tagged particle moves from 0 to  $z$  and then the origin is shifted to  $z$ . This is a transformation  $T_z$  that acts when  $\eta(z) = 0$  and the new configuration on  $\mathbb{Z}^d - \{0\}$  is given by

$$(T_z \eta)(x) = \eta(x+z) \text{ if } x \neq -z, 0; (T_z \eta)(-z) = 0$$

contributing to the generator the term

$$\mathcal{A}_2 = \sum_z \pi(z)(1 - \eta(z))[f(T_z \eta) - f(\eta)] \quad (5.2)$$

The full generator is therefore

$$\begin{aligned} (\mathcal{A}f)(\eta) &= \frac{1}{2} \sum_{x,y \neq 0} \pi(y-x)[f(\eta^{x,y}) - f(\eta)] \\ &\quad + \sum_z \pi(z)(1 - \eta(z))[f(T_z \eta) - f(\eta)] \\ &= (\mathcal{A}_1 f)(\eta) + (\mathcal{A}_2 f)(\eta) \end{aligned} \quad (5.3)$$

It is not difficult to check that the probability distribution  $\mu_\rho$  on  $\mathbb{Z}^d - \{0\}$  is a reversible invariant distribution for  $\mathcal{A}$  given by 5.3. The jumps  $x \rightarrow y$  and  $y \rightarrow x$  as well as  $T_z$  and  $T_{-z}$  provide pairs with detail balance. The rates are same in either direction and  $\mu_\rho$  is invariant under the transitions.

Our main tool is a central limit theorem for additive functions of a reversible Markov process. Given a real valued function  $f$  on a space  $\mathcal{X}$ , a Markov process on that space with generator  $\mathcal{A}$  and a reversible ergodic invariant measure  $\mu$  for  $\mathcal{A}$  satisfying with  $E^\mu[f(x)] = 0$  under suitable conditions we will show that

$$\int_0^t f(x(s))ds = M(t) + a(t)$$

where  $M(t)$  is a square integrable Martingale with stationary increments and  $a(t)$  is negligible. If  $\mathcal{A}$  is the self adjoint generator of the process  $-\mathcal{A}$  has a spectral resolution  $-\mathcal{A} = \int_0^\infty \sigma E(d\sigma)$ . We have the Dirichlet form  $\mathcal{D}(f) = \langle -\mathcal{A}f, f \rangle_{L_2(\mu)}$  associated with  $\mathcal{A}$ . The space  $\mathcal{H}_1$  is the abstract Hilbert space obtained by completing the space of square integrals functions with respect to the Dirichlet inner product. One might start with functions  $u$  in the domain of  $\mathcal{A}$ , ensuring the finiteness of  $\mathcal{D}(f)$ . The completion will be an abstract space  $\mathcal{H}_1$ . There will be a dual  $H_{-1}$  to  $\mathcal{H}_1$  relative to the inner product of  $\mathcal{H}_0 = L_2(\mu)$ . Formally  $\|u\|_{-1} = \langle (-\mathcal{A})^{-1}u, u \rangle$  and

$$\|u\|_{-1}^2 = \sup_f 2\langle u, f \rangle - \mathcal{D}(f)$$

We have the following theorem.

**Theorem 5.1.1.** *If  $f$  is in  $L_2$  with spectral resolution  $\langle E(d\sigma)f, f \rangle$ , and*

$$\langle (-\mathcal{A})^{-1}f, f \rangle = \int_0^\infty \sigma^{-1} \langle E(d\sigma)f, f \rangle < \infty = \sigma^2 < \infty$$

*there is a square integrable Martingale  $M(t)$  with stationary increments such that  $E^P[M(t)^2] = 2\sigma^2 t$  and*

$$\int_0^t f(x(s))ds = M(t) + a(t)$$

*with  $E[|a(t)|^2] = o(t)$  as  $t \rightarrow \infty$ . The central limi theorem follows. Moreover*

$$P[\sup_{0 \leq t \leq T} |a(t)| \geq c\sqrt{T}] \rightarrow 0$$

*for every  $c > 0$  implying the functional CLT.*

For the proof of the theorem we need two lemmas.

**Lemma 5.1.2.** *Let  $P$  be a reversible stationary Markov process with invariant measure  $\mu$  and generator  $\mathcal{A}$ . Let  $u \in L_2[\mu]$  with  $\mathcal{D}(u) = \langle -\mathcal{A}u, u \rangle < \infty$ . Then*

$$P\left[\sup_{0 \leq t \leq T} |u(x(t))| \geq \ell\right] \leq \frac{e}{\ell} \sqrt{T\mathcal{D}(u) + \|u\|_2^2}$$

*Proof.* Since  $\mathcal{D}(|u|) \leq \mathcal{D}(u)$  we can assume that  $u \geq 0$ . If  $x(t)$  is a Markov process and  $\tau$  is the exit time from  $G$ , then  $E_x[e^{-\sigma\tau}] = v(\sigma, x)$  is the solution of

$$\sigma v(x) - (\mathcal{A}v)(x) = 0 \quad \text{for } x \in G; \quad v = 1 \quad \text{on } G^c$$

The function  $v$  is also the minimizer of

$$\sigma \|v\|_2^2 + \mathcal{D}(v)$$

over  $v$  such that  $v = 1$  on  $G^c$ . Therefore the solution  $v_\sigma$  satisfies

$$\sigma \|v_\sigma\|_2^2 \leq \inf_{v: v=1 \text{ on } G^c} [\sigma \|v\|_2^2 + \mathcal{D}(v)]$$

If we take for  $G$  the set  $u(x) < \ell$ , the function  $v = \frac{u\wedge\ell}{\ell}$  is an admissible choice for  $v$ . Therefore with  $\sigma = T^{-1}$ ,

$$\|v_\sigma\|_1 \leq \frac{1}{\ell} \sqrt{\|u\|_2^2 + T\mathcal{D}(u)}$$

We obtain the estimate

$$\int P_x[\tau < T] d\mu \leq e^{\sigma T} \int E_x[e^{-\sigma\tau}] d\mu = e \|v_\sigma\|_1 \leq \frac{e}{\ell} \sqrt{\|u\|_2^2 + T\mathcal{D}(u)}$$

□

This lemma quantifies the statement that set of singularities of a function  $u$  on  $\mathbb{R}^d$  that is in the Sobolev space  $W_2^1(\mathbb{R}^d)$  has capacity 0. In other words even if  $u$  has singularities, a Brownian path will not see it, i.e.  $u(\beta(t))$  is almost surely continuous.

**Lemma 5.1.3.** *Let  $\|u\|_2$  and  $\mathcal{D}(u)$  be finite. Then for any  $c > 0$*

$$\limsup_{T \rightarrow \infty} P\left[\sup_{0 \leq t \leq T} |u(x(t))| \geq c\sqrt{T}\right] = 0$$

*Proof.* For any given  $\delta > 0$  find  $u' \in L_\infty$  such that  $\|u - u'\|_2^2 \leq \delta$  and  $\mathcal{D}(u - u') \leq \delta$ . Clearly

$$\limsup_{T \rightarrow \infty} P\left[\sup_{0 \leq t \leq T} |u'(x(t))| \geq c\sqrt{T}\right] = 0$$

and

$$\limsup_{T \rightarrow \infty} P\left[\sup_{0 \leq t \leq T} |(u - u')(x(t))| \geq c\sqrt{T}\right] \leq \frac{e\sqrt{\delta}}{c}$$

and  $\delta$  can be made arbitrarily small the lemma is proved. □

*Proof.* Now we return to complete the proof of Theorem 5.1.1. First let us note that the condition is natural. An elementary calculation shows that

$$\begin{aligned}
\frac{1}{t}E^P[|\int_0^t f(x(s))ds|^2] &= \frac{1}{t}E^P[\int_0^t \int_0^t f(x(s))f(x(s'))dsds'] \\
&= \frac{2}{t} \int_{0 \leq s \leq s' \leq t} \langle T_{s'-s}f, f \rangle dsds' \\
&= 2 \int_0^t (1 - \frac{s}{t}) \langle T_s f, f \rangle ds \\
&\simeq 2 \int_0^\infty \langle T_s f, f \rangle ds \\
&= 2 \langle (-\mathcal{A})^{-1} f, f \rangle \\
&= 2\sigma^2
\end{aligned}$$

Since  $\langle T_t f, f \rangle \geq 0$ , the convergence has to be absolute. Let us solve the resolvent equation

$$\lambda u_\lambda - \mathcal{A}u_\lambda = f$$

Then  $\mathcal{A}u_\lambda = \lambda u_\lambda - f$  and

$$u_\lambda(x(t)) - u_\lambda(x(0)) - \int_0^t \lambda u_\lambda(x(s))ds + \int_0^t f(x(s))ds = M_\lambda(t)$$

where  $M_\lambda(t)$  is a martingale with

$$\frac{1}{t}E[M_\lambda(t)^2] = 2\mathcal{D}(u_\lambda) = 2\langle -\mathcal{A}u_\lambda, u_\lambda \rangle = 2 \int_0^\infty \frac{2\sigma}{(\lambda + \sigma)^2} \langle E(d\sigma)f, f \rangle$$

An easy computation shows that  $(\sigma + \lambda)^{-1} \rightarrow \sigma^{-1}$  and is dominated by  $\sigma^{-1}$  which is integrable with respect to  $\langle E(d\sigma)f, f \rangle$ . The martingales  $M_\lambda(t)$  have a limit in  $L_2(P)$ .  $\lambda u_\lambda \rightarrow 0$  in  $L_2(\mu)$ . Therefore  $a_\lambda(t) = u_\lambda(x(0)) - u_\lambda(x(t))$  has a limit  $a(t)$  and

$$\int_0^t f(x(s))ds = M(t) + a(t)$$

We will show that  $E[|a(t)|^2] = o(t)$ . Then martingale CLT will imply our result. This is again a spectral calculation.

$$E^P[|a(t)|^2] = 2 \lim_{\lambda \rightarrow 0} \int_0^\infty \frac{1 - e^{-t\sigma}}{(\lambda + \sigma)^2} \langle E(d\sigma)f, f \rangle = 2 \int_0^\infty \frac{1 - e^{-t\sigma}}{\sigma^2} \langle E(d\sigma)f, f \rangle$$

Since  $\frac{1 - e^{-t\sigma}}{t} \leq \sigma$  and  $\int_0^\infty \frac{1}{\sigma} \langle E(d\sigma)f, f \rangle < \infty$ , the dominated convergence theorem implies that

$$\lim_{t \rightarrow \infty} \frac{1}{t} E^P[|a(t)|^2] = 0$$

To prove the functional CLT, we need to consider

$$\frac{1}{\sqrt{T}} \int_0^{tT} f(x(s)) ds = \frac{1}{\sqrt{T}} M_\lambda(Tt) + \frac{1}{\sqrt{T}} \int_0^{tT} \lambda u_\lambda(s) ds - \frac{1}{\sqrt{T}} [u_\lambda(x(tT)) - u_\lambda(x(0))]$$

The functional CLT holds for  $\frac{1}{\sqrt{T}} M_\lambda(tT)$  and uniformly so as  $\lambda \rightarrow 0$  because  $M_\lambda(t) \rightarrow M(t)$  in mean square. We note that with the help of the dominated convergence theorem,

$$\lambda \|u_\lambda\|^2 = \int_0^\infty \frac{\lambda}{(\lambda + \sigma)^2} \langle E(d\sigma) f, f \rangle \rightarrow 0$$

as  $\lambda \rightarrow 0$ . Clearly with the choice of  $\lambda = T^{-1}$ ,

$$\xi_T = \sup_{0 \leq t \leq 1} \left| \frac{1}{\sqrt{T}} \int_0^{tT} \lambda u_\lambda(s) ds \right| \leq \frac{1}{T} \int_0^T \sqrt{\lambda} |u_\lambda(x(s))| ds$$

and  $E[|\xi_T|^2] \rightarrow 0$  as  $t \rightarrow \infty$ . To complete the proof we need to show that, with  $\lambda = T^{-1}$

$$P\left[ \sup_{0 \leq t \leq T} |u_{\frac{1}{T}}(x(s))| \geq c\sqrt{T} \right] \rightarrow 0$$

We can represent  $u_{\frac{1}{T}}$  as  $u_\delta + (u_{\frac{1}{T}} - u_\delta)$ . By lemma 5.1.3, we have for any  $\delta > 0$ ,

$$\limsup_{T \rightarrow \infty} P\left[ \sup_{0 \leq s \leq T} |u_\delta(x(s))| \geq c\sqrt{T} \right] = 0$$

Moreover

$$\lim_{T \rightarrow \infty} \frac{1}{T} \|u_{\frac{1}{T}} - u_\delta\|_2^2 = 0$$

and

$$\lim_{\substack{T \rightarrow \infty \\ \delta \rightarrow 0}} D(u_{\frac{1}{T}} - u_\delta) = 0$$

They imply that

$$\limsup_{\delta \rightarrow 0} \limsup_{T \rightarrow \infty} P\left[ \sup_{0 \leq t \leq T} |u_{\frac{1}{T}}(x(s)) - u_\delta(x(s))| \geq c\sqrt{T} \right] = 0$$

□

We now return to the motion of the tagged particle. We need to keep track of its motion as well as the changing environment seen by it. If  $w \in \mathbb{Z}^d$  is the location of the tagged particle in the original reference frame, then jointly the generator for  $w(t) \in \mathbb{Z}^d$  and  $\eta(\cdot) \in \{0, 1\}^{\mathbb{Z}^d - \{0\}}$  is

$$(\tilde{\mathcal{A}}f)(\eta) = \sum_z \pi(z) (1 - \eta(z)) [f(w + z, T_z \eta) - f(w, \eta)] + (\mathcal{A}f)(w, \eta) \quad (5.4)$$

with  $\mathcal{A}$  acting only on  $\eta$  for each  $w$ . The  $\eta(t, \cdot)$  part is a Markov process by itself and is in equilibrium at density  $\rho$ , the distribution being  $\mu_\rho$ . We are interested in establishing a central limit theorem for  $w(t)$ . We note that

$$w(t) - w(0) = \int_0^t \sum_z z \pi(z) (1 - \eta(s, z)) ds + M(t)$$

where  $M(t)$  is a Martingale with the decomposition

$$M(t) = \sum_z z M_z(t)$$

and

$$M_z(t) = N_z(t) - \pi(z) \int_0^t (1 - \eta(s, z)) ds$$

The quantity

$$V(\eta(\cdot)) = \sum_z z \pi(z) (1 - \eta(z))$$

has mean 0 in equilibrium and one may expect a CLT for

$$\int_0^t V(\eta(s, \cdot)) ds$$

We will prove a decomposition of the form

$$\int_0^t V(\eta(s, \cdot)) ds = N(t) + a(t)$$

where  $N(t)$  is a martingale and  $a(t)$  is negligible. Then

$$w(t) - w(0) = M(t) + N(t) + a(t)$$

and since central limit theorems for martingales are automatic the result will follow. The quantities here are vectors and the equations are for each component or they are interpreted as

$$\langle w(t) - w(0), \xi \rangle = \langle M(t), \xi \rangle + \langle N(t), \xi \rangle + \langle a(t), \xi \rangle$$

for  $\xi \in \mathbb{R}^d$ . We have now the main theorem.

**Theorem 5.1.4.** *The position  $w(t)$  of the tagged particle satisfies a functional CLT, with positive definite covariance matrix  $S(\rho)$  given by*

$$\langle S(\rho)\xi, \xi \rangle = \inf_f \left[ \int \left[ \sum_z \pi(z) (1 - \eta(z)) (\tau_z f - f - \langle \xi, z \rangle)^2 + \frac{1}{2} \sum_{x,y} \pi(y-x) (f(\eta^{x,y}) - f(\eta))^2 \right] d\mu_\rho \right]$$

First we need to prove, for each vector  $\xi \in \mathbb{R}^d$ , a bound of the form

$$\left| \int \sum_z \langle z, \xi \rangle (1 - \eta(z)) \pi(z) f(\eta) d\mu_\rho \right| \leq \sqrt{C(\xi)} \sqrt{D_\rho(f)}$$

We can rewrite, after combining the  $z$  and  $-z$  terms and symmetrizing

$$\begin{aligned} & E^{\mu_\rho} \left[ \sum_z \langle z, \xi \rangle (1 - \eta(z)) \pi(z) f(\eta) \right] \\ &= \frac{1}{2} E^{\mu_\rho} \left[ \sum_z \langle z, \xi \rangle [(1 - \eta(z)) - (1 - \eta(-z))] \pi(z) f(\eta) \right] \\ &= \frac{1}{2} E^{\mu_\rho} \left[ \sum_z \langle z, \xi \rangle [1 - \eta(z)] \pi(z) [f(\eta) - f(T_z \eta)] \right] \\ &\leq \frac{1}{2} \left[ E^{\mu_\rho} \left[ \sum_z |\langle z, \xi \rangle|^2 [1 - \eta(z)] \pi(z) \right] \right]^{\frac{1}{2}} \left[ E^{\mu_\rho} \left[ \sum_z [1 - \eta(z)] \pi(z) [f(\eta) - f(T_z \eta)]^2 \right] \right]^{\frac{1}{2}} \\ &\leq \sqrt{C(\xi)} \sqrt{D_\rho(f)} \end{aligned}$$

with

$$C(\xi) = \frac{1 - \rho}{4} \sum_z |\langle z, \xi \rangle|^2 \pi(z)$$

This proves the validity of functional CLT for  $w(t)$  with an upper bound on the variance.

The next step is to establish the formula and a lower bound for it. Let us compute  $\langle S(\rho)\xi, \xi \rangle$ . The minimizer  $f = f_\xi$  may not exist. The space  $H_1$  of functions  $u \in L_2$ , with the Dirichlet inner product, when completed, will admit objects that are not in  $L_2(\mu_\rho)$ . There is no Poincaré inequality available. Abstractly the space consists of collections of functions  $\{g^{x,y}(\eta)\}, \{g_z\}$ , that are the limits in  $H_1$  of  $\{f(\eta^{x,y}) - f(\eta)\}, (1 - \eta(z))[f(T_z \eta) - f(\eta)]$ . The functions  $g^{x,y}(\eta), g_z$  satisfy identities.  $g^{x,y}$  is 0 unless  $\eta(x) \neq \eta(y)$  and satisfies  $\eta^{x,y} + \eta^{y,x} = 0$ . Similarly  $g_z$  is nonzero only when  $\eta(z) = 0$  and  $(1 - \eta(z))g_z(\eta) + (1 - \eta(-z))g_{-z}(T_{-z}\eta) = 0$ . The Euler equation for the variational problem is

$$\begin{aligned} & E^{\mu_\rho} \left[ \sum_z \pi(z) (1 - \eta(z)) [g_z(\eta) - \langle \xi, z \rangle] [f(\tau_z \eta) - f(\eta)] \right] \\ &+ \frac{1}{2} E^{\mu_\rho} \left[ \sum_{x,y} \pi(y - x) g^{x,y}(\eta) [f(\eta^{x,y}) - f(\eta)] \right] = 0 \end{aligned}$$

for all  $f$ , which after a bit of calculation, takes the form

$$V(\eta) + \frac{1}{2} \sum_{x,y} \pi(y - x) g^{x,y} + \sum_z \pi(z) (1 - \eta(z)) g_z = 0$$

$w(t)$  now has the representation

$$\langle \xi, w(t) \rangle = \int_0^t \langle \xi, V(\eta(s)) \rangle ds + \sum_z \int_0^t \langle \xi, z \rangle (1 - \eta(z)) dM_z(t)$$

with

$$\int_0^t \langle \xi, V(\eta(s)) \rangle ds = a(t) + N(t)$$

and

$$N(t) = \sum_z \int_0^t g_z(\eta(s)) dM_z(s) + \sum_{x,y} \int_0^t g^{x,y}(\eta(s)) M_{x,y}(t)$$

with

$$M_{x,y}(t) = N_{x,y}(t) - \int_0^t \pi(y-x) \eta(s,x) (1 - \eta(y,s)) ds$$

Therefore

$$\begin{aligned} \langle \xi, w(t) \rangle &= \int_0^t \sum_z [\langle z, \xi \rangle - g_z(\eta(s))] dM_z(s) \\ &\quad - \int_0^t \sum_{x,y} g_{x,y}(\eta(s)) dN_{x,y}(s) + a(t) \\ &= M(t) + a(t) \end{aligned}$$

Computing the quadratic variation of the martingale  $M(t)$  proves the formula. Finally we will prove the non degeneracy of the quadratic form  $\langle S(\rho)\xi, \xi \rangle$ . We have to exclude the one dimensional nearest neighbor case, where  $S(\rho) \equiv 0$ . The proof depends on the following fact. We can obtain an estimate of the form

$$E^{\mu_\rho} [(\eta(z) - \eta(-z))f(\eta)] \leq C \sqrt{\mathcal{D}_2(f)}$$

in terms of the Dirichlet form

$$\mathcal{D}_1(u) = \langle -\mathcal{A}_1 u, u \rangle = \frac{1}{4} E^{\mu_\rho} \left[ \sum_{x,y} \pi(y-x) [f(\eta^{x,y}) - f(\eta)]^2 \right]$$

It is possible to shift a particle from  $z$  to  $-z$  without touching the tagged particle at 0. Jump over it or go around it. This provides an estimate of the form

$$E^{\mu_\rho} [(\eta(z) - \eta(-z))f(\eta)] \leq C(z) \left[ E^{\mu_\rho} \sum_{x,y} [\pi(y-x) [f(\eta^{x,y}) - f(\eta)]^2] \right]^{\frac{1}{2}} \quad (5.5)$$



We can estimate, for any  $a > 0$ ,

$$\begin{aligned} \langle (\lambda I - \mathcal{A})^{-1} \langle V, \xi \rangle, \langle V, \xi \rangle \rangle &\leq \sqrt{\langle (-\mathcal{A}_1(\lambda I - \mathcal{A})^{-1} \langle V, \xi \rangle, (\lambda I - \mathcal{A})^{-1} \langle V, \xi \rangle \rangle} \sqrt{\langle (-\mathcal{A}_1)^{-1} \langle V, \xi \rangle, \langle V, \xi \rangle \rangle} \\ &\leq \frac{a}{2} \langle (-\mathcal{A}_1(\lambda I - \mathcal{A})^{-1} \langle V, \xi \rangle, (\lambda I - \mathcal{A})^{-1} \langle V, \xi \rangle \rangle + \frac{1}{2a} \langle (-\mathcal{A}_1)^{-1} \langle V, \xi \rangle, \langle V, \xi \rangle \rangle \end{aligned}$$

Letting  $\lambda \rightarrow 0$ ,

$$\langle S(\rho)\xi, \xi \rangle \geq \langle (-\mathcal{A}_1(-\mathcal{A})^{-1} \langle V, \xi \rangle, (-\mathcal{A})^{-1} \langle V, \xi \rangle \rangle \geq \frac{2}{a} \langle (-\mathcal{A})^{-1} \langle V, \xi \rangle, \langle V, \xi \rangle \rangle - \frac{1}{a^2} \langle (-\mathcal{A}_1)^{-1} \langle V, \xi \rangle, \langle V, \xi \rangle \rangle$$

We can obtain a lower bound for the first quadratic form on the right from the variational formula

$$\langle (-\mathcal{A})^{-1} g, g \rangle = \sup_f [2\langle g, f \rangle - \langle -\mathcal{A}f, f \rangle]$$

and an upper bound for the second one from (5.5). Picking  $a$  large will do it.