

## Week 9

# Random walk in trap environment

LECTURER: SIVA ATHREYA

SCRIBE: SANCHAYAN BHOWAL, RAMKRISHNA SAMANTA

## 9.1 Continuous time random walk

In this model the random walker waits an exponential amount of time to perform a jump like a discrete time random walk. Consider  $\{V_i : i \geq 1\}$  to be a collection of independent  $\text{Exponential}(\lambda)$  random variables. Let  $\lambda = 1$ . Define  $T_k$  to be the sum of the first  $k$   $V_i$ 's. Also, define  $N_t$  to be the number of  $T_k$  less than  $t$ . Hence,

- $\mathbf{P}(V_i \leq t) = 1 - e^{-t}$
- $T_k = \sum_{i=1}^k V_i$
- $N_t = \sum_{k=1}^{\infty} \mathbb{1}(T_k \leq t)$
- $\{N_t = k\} = \{T_k \leq T_{k+1}\}$

**Theorem 9.1.1.** 1.  $N_t \sim \text{Poisson}(t)$

2.  $N_t - N_s$  is independent of  $N_r$  where  $r \leq s \leq t$ .

3. For  $0 \leq t_0 \leq t_1 \leq \dots \leq t_n$

$\{N_{t_{i+1}} - N_{t_i} : i = 0, \dots, n-1\}$  are independent

So,  $N_{t_{i+1}} - N_{t_i} \sim \text{Poisson}(t_{i+1} - t_i)$

**Definition 9.1.1.** Let  $U_n : n \geq 0$  be a random walk on  $(\Gamma, \mu)$ . Define, a continuous time random walk on  $(\Gamma, \mu)$  with rate 1 to be:

$$Y_t = U_{N_t} \forall t \geq 0$$

.

*Remark.* The random variable  $Y_t$  is a random step function which is right continuous with left limits.

## 9.2 Random walk in trap environment

### Continuous Time set-up

Consider the graph  $\mathbb{Z}^d$  with natural weights. Let  $\{X_t\}_{t \geq 0}$  be a continuous time random walk on  $\mathbb{Z}^d$  starting at 0, with rate  $\kappa$ . Now, let us set up the traps, i.e., for each  $y \in \mathbb{Z}^d$  let  $N_y \sim \text{Poisson}(\rho)$ . This  $N_y$  denote the number of *traps* at  $y$ . Each trap  $(Y^{j,y})$  perform a continuous time random walk  $(\{Y_t^{j,y}\}_{t \geq 0})$  with rate  $\nu$ ; where  $1 \leq j \leq N_y$ . The random walk gets killed if it meets a trap. There are two ways of killing viz,

Hard The walk gets killed upon intersection with any  $Y^{j,y}$ .

Soft At each site  $x$  at time  $t \geq 0$ , define

$$\xi(t, x) := \sum_{y \in \mathbb{Z}^d, 1 \leq j \leq N_y} \#\{Y^{j,y} \text{ at } x\}.$$

Now  $X_t$  gets killed at rate  $\gamma \xi(t, x)$  where  $\gamma \in \mathbb{R}$ .

*Remark.* Hard killing infact corresponds to  $\gamma = \infty$  case of soft killing.

The probability of survival is given by

$$Z_{\gamma, t} = \mathbf{E}^X[\exp(-\gamma \int_0^t \xi(s, X(s)) ds)]$$

### Discrete Time set-up

Let  $\{X_t\}_{t \geq 0}$  be a random walk on  $\mathbb{Z}^d$  with natural weights starting at 0. For each  $y \in \mathbb{Z}^d$  let  $N_y \sim \text{Poisson}(\rho)$  denotes the number of traps at  $y$ . Each trap  $(Y^{j,y})$  perform a lazy random walk  $(\{Y_t^{j,y}\}_{t \geq 0})$  on  $\mathbb{Z}^d$ ; where  $1 \leq j \leq N_y$ . The trap kills the random walk with probability  $q$  if it meets the random walk;  $q \in (0, 1)$ . Let  $\xi(n, x)$  denote the number of traps at location  $x$ , i.e.

$$\xi(n, x) = \sum_{y \in \mathbb{Z}^d, 1 \leq j \leq N_y} \delta_x(Y_n^{j,y}).$$

Assume  $X_k$  has survived till  $k \leq n$ . Given  $X_n$  the probability that  $X_n$  will survive at time  $n$  is  $(1 - q)^{\xi(n, X_n)}$ . Hence,

$$\begin{aligned} \sigma^X(n, \xi) &= \mathbf{P}(X \text{ has survived till time } n \text{ given } \{Y_m^{j,y}\}_{1 \leq j \leq m, y \in \mathbb{Z}^d} \text{ where } m \leq n) \\ &= (1 - q)^{\sum_{i=1}^n \xi(i, X_i)}. \end{aligned} \tag{9.1}$$

## 9.3 Pascal's Theorem

The average survival probability of a given trajectory  $X$  is given by  $\sigma^X(n) = \mathbf{E}^\xi[(1 - q)^{\sum_{i=1}^n \xi(i, X_i)}]$ .

**Theorem 9.3.1** (Pascal). *The survival probability is maximized by the trajectory  $\underline{0}$  where  $\underline{0}_k = 0$  for every  $k \in \mathbb{N} \cup 0$ , i.e.,*

$$\sigma^X(n) \leq \sigma^{\underline{0}}(n).$$

**Lemma 9.3.1.**  $\sigma^X(n) = \exp(-\lambda \sum_{y \in \mathbb{Z}^d} W_X(n, y))$  where  $W_X(n, y) = 1 - \mathbf{E}^y[1 - (1 - q)^{\sum_{i=1}^n \delta(Y_i^y)}]$ .  
The  $Y_i^y$  is a random variable with ditribution same as i.i.d.  $Y_i^{j,y}$ .

*Proof.* Let  $X : \mathbb{N} \cup 0 \rightarrow \mathbb{Z}^d$  with  $X_0 = 0$  be the trajectory. Now,

$$\begin{aligned}
\sigma^X(n) &= \mathbf{E}^\xi[(1 - q)^{\sum_{i=1}^n \xi(i, X_i)}] \\
&= \mathbf{E}^\xi[(1 - q)^{\sum_{i=1}^n \sum_{y \in \mathbb{Z}^d} \sum_{1 \leq j \leq N_y} \delta_{X_i}(Y_n^{j,y})}] \\
&= \prod_{y \in \mathbb{Z}^d} \mathbf{E}^\xi[\prod_{1 \leq j \leq N_y} (1 - q)^{\sum_{i=1}^n \delta_{X_i}(Y_n^{j,y})}] \\
&= \prod_{y \in \mathbb{Z}^d} \mathbf{E}^y \mathbf{E}^{N_y}[\prod_{1 \leq j \leq N_y} (1 - q)^{\sum_{i=1}^n \delta_{X_i}(Y_n^{j,y})}]
\end{aligned}$$

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