

Poisson processes.

The current workshop is devoted to rare events in dynamics and the aim of my minicourse is to discuss several problems there the ~~statements~~ statements do not involve rare events but the proofs relies crucially on the rare event theory.

However, before discussing these examples I would like to provide some background on Poisson processes. Historically, Poisson limit theorem was among first results concerning rare events.

Recall that a random variable X has Poisson distribution with parameter λ denoted $X \sim \text{Pois}(\lambda)$ if $P(X=k) = e^{-\lambda} \frac{\lambda^k}{k!}$.

Below I recall the basic properties of Poisson random variables which should be familiar from ~~the~~ an introductory probability course and can be verified by direct computations.

(1) The ~~generating~~ generating function $M_X(z) := E(z^X) = \exp(\lambda(z-1))$

(2) Consequently the characteristic function is $\varphi_X(t) = E(e^{itX}) = \exp(\lambda(e^{it} - 1))$

(3) If X_1, X_2 are independent $X_i \sim \text{Pois}(\lambda_i)$ then $X = X_1 + X_2 \sim \text{Pois}(\lambda_1 + \lambda_2)$

(4) If X_1, X_2, X_3 are as in (3) then conditionally on X_1, X_2 have binomial distribution

$$P(X_3 = k \mid X_1 + X_2 = n) = \binom{n}{k} p^k (1-p)^{n-k}$$

where $p = \frac{\lambda_1}{\lambda_1 + \lambda_2}$

(5) (Pruning) If $X \sim \text{Pois}(\lambda)$ and each point from X is removed with probability p independently of the others then the resulting random variable Y has Poisson distribution with parameter (λp)

(6) Factorial moments

$$P(X(X-1)\dots(X-k+1)) = \lambda^k.$$

~~Exercise 1~~

Exercise 1 Prove properties (1) - (6).

Next we discuss Poisson processes.

~~Let (\mathcal{X}, μ) be a probability space.~~ Let (\mathcal{X}, μ) be a probability space. Let $\xi_1, \xi_2, \dots, \xi_n$ be a random sequence of points in \mathcal{X} . We say that $\{\xi_n\}$ forms a Poisson process with parameters (λ, μ) if denoting by $N(A)$ the number of points in A we have the following properties

$$(A) N(A) \sim \text{Pois}(\mu(A))$$

(B) $N(A_1)$ and $N(A_2)$ are independent if A_1 and A_2 are disjoint

Note that this definition is consistent due to property (B) of the Poisson process. We now discuss several properties of the Poisson process

(PP1) If $\xi \sim \text{Pois}(X, \mu)$ and $\pi: X \rightarrow Y$ is a measurable map then

$$\pi(\xi) \sim \text{Pois}(Y, \pi^* \mu)$$

Indeed if $B \subset Y$ and $N(B)$ is the number of points from $\pi\xi$ in B then

$$P(N(B) = k) = P(N(\pi^{-1}(B)) = k) =$$

$$= \frac{\lambda^k}{k!} e^{-\lambda} \quad \text{where } \lambda = \mu(\pi^{-1}(B)) = \pi^* \mu(B)$$

(PP2) If $\xi' \sim \text{Pois}(X, \mu')$, $\xi'' \sim \text{Pois}(X, \mu'')$ and ξ', ξ'' are independent then $\xi' \vee \xi'' \sim \text{Pois}(X, \mu' + \mu'')$

Exercise 2 Deduce (PP2) from property 3 of the Poisson random variable.

(PP3) Suppose that $\xi \sim \text{Pois}(X, \mu)$ and suppose that we color each point from ξ in one of k colors so that for each point color j is chosen with probability p_j independently of the others. Then

if $\xi_j^{(i)}$ are points of color j , then

$\xi_j^{(1)}, \xi_j^{(2)}, \dots, \xi_j^{(k)}$ are independent poisson processes with parameters (\mathcal{X}, μ_j) .

Exercise 3 Prove (PP3) (cf. property (5) of the Poisson random variable).

(PP4) If $\{\xi_n\} \sim \text{Poisson}(\mathcal{X}, \mu)$ and Y_n are independent identically distributed \mathcal{Y} -valued random variables with ~~Poisson~~ law P then $\{(\xi_n, Y_n)\} \sim \text{Pois}(\mathcal{X} \times \mathcal{Y}, \mu \times P)$

Exercise 4 Derive (PP4) from (PP3) in the case Y_n take values $\{1, \dots, k\}$ so that $P(Y_n = j) = P_j$

(PP5) ~~Conditioned~~ Conditioned on $N(A) = n$ the points in A are independent and distributed according to $P_A = \frac{P}{\mu(A)}$

That is $P(N(B) = k | N(A) = n) = \binom{n}{k} p^k (1-p)^{n-k}$ where $p = \frac{\mu(B)}{\mu(A)}$ for $B \subset A$.

Exercise 5 Derive (PP5) from property (4) of the Poisson random variable.

PP6 (Mayer formula) Let $\psi: X \rightarrow \mathbb{R}$ be a measurable function and

$$S = \sum_{\xi} \psi(\xi). \text{ Then}$$

$$E(e^{itS}) = \exp \left[\int_X [e^{it\psi(x)} - 1] d\mu(x) \right]$$

Exercise 6 Prove (PP6)

Hint Approximate ψ by piecewise constant functions.

One application of Poisson processes appears in relation to stable laws. We recall the corresponding facts. We say that a random variable U has

stable law if for each $n \in \mathbb{N}$ $\exists a_n, c_n$ s.t.

if U_1, U_2, \dots, U_n are independent identically distributed

(iid) random variable having the same law as U

then $U_1 + U_2 + \dots + U_n \stackrel{\text{law}}{=} a_n U + c_n$

One can show that in this case $a_n = n^{1/\alpha}$ where

$\alpha \in (0, 2]$ is called the index of U .

Classification of the stable laws has been accomplished by Levi and Khintchine. For $\alpha=2$, U is normal while for $\alpha \neq 2$,

$$U = L + U_+ - L_- U_- + L_0 \text{ where}$$

$\alpha_+, \alpha_-, \alpha_0$ are are ~~constant~~ constant
 $\alpha_{\pm} \geq 0$ and U_+ and U_- are independent and

$$E(e^{itU_{\pm}}) = \exp\left(\int_0^{\alpha_{\pm}} \left[e^{itx} - 1 - itx \right] \frac{s dx}{x^{s+1}} + \int_1^{\infty} (e^{itx} - 1) \frac{s dx}{x^{s+1}}\right)$$

Note that that for $s \in (0, 1)$

U_{\pm} have the law of $\sum_n \xi_n$ up to normalization

where $\{\xi_n\} \sim \text{Pois}(\mathbb{R}_+, \kappa_s)$

$\frac{d\kappa(u)}{du} = \frac{s}{u^{s+1}}$ while for $s \in (2, \infty)$ it is

the limit $\lim_{\varepsilon \rightarrow 0} \left(\sum_n \xi_n^{\varepsilon} - \frac{c_{\varepsilon}}{\varepsilon} \right)$ where $\xi_n^{\varepsilon} \sim \text{Pois}(c_{\varepsilon}, \infty) / \kappa_s$

and $c_{\varepsilon} = \varepsilon^{-s}$.

Exercise 7 For $s \in (0, 1)$ show that $\sum_n \xi_n$ is stable using (PP1) and (PP2).

Exercise 8 Let $\{\xi_n\} \sim \text{Pois}(\mathbb{R}, \text{Leb})$. Let

θ_n be iid symmetric with compact support.

For $\alpha > 1/2$ show that $\sum_n \frac{\theta_n}{\xi_n^{\alpha}}$ has stable law of index $1/2 + \alpha$.

$U = \sum_n \frac{\theta_n}{\xi_n^{\alpha}}$ (and U is symmetric).

Next we discuss proving the Poisson limit theorem for dynamical systems.

We say that a dynamical system $f: M \rightarrow M$ preserving a smooth measure μ is exponentially mixing of order r on $C^m(M)$ if

$$\left| \int \prod_{j=1}^r A_j(f^{n_j} x) d\mu(x) - \prod_{j=1}^r \int A_j d\mu \right| \leq$$

$$K_{m,r} e^{-\beta L} \prod_{j=1}^r \|A_j\|_{C^m(M)}$$

where β is a positive constant and

$$L = \min_{i \neq j} |n_i - n_j|.$$

Exercise 9 Show that $\forall m', m'' > 0$ \Leftrightarrow f is exponentially mixing of order r on $C^{m'}(M) \Leftrightarrow f$ is exponentially mixing of order r on $C^{m''}(M)$.

This justifies dropping $C^m(M)$ at the end and saying in the above scenario that f is exponentially mixing of order r .

Thm 1 (B. Fayad - S. Liu - D.) If f is exponentially mixing of all orders then for almost every x_0 the following holds

$$\text{Let } S_p^{(k)} = \sum_{k=0}^{N_p} \mathbb{1}_{B(x_0, p)}(f^k x) \text{ with } N_p = \frac{2}{\mu(B(x_0, r))}$$

Then if $p \rightarrow 0$ then $S_p \Rightarrow \text{Pois}(2)$.

Remark In all known examples of exponentially mixing systems one has a stronger result

$S_r \Rightarrow \text{Pois}(2)$ provided that x_0 is aperiodic.

However ~~our~~ our proof does not give this ~~result~~ if we only assume multiple exponential mixing.

Thus the description of exceptional points in Theorem 1 remains an interesting open problem.

The proof of Thm 1 consists of two parts.

We say that x_0 is slowly recurrent if

$\forall A, K \exists r_0$ such that for $r \leq r_0$ we have

$$\mu(B(x_0, r) \cap T^{-n} B(x_0, r)) \leq \frac{\mu(B(x_0, r))}{|\ln r|^A}$$

for all $n \leq K |\ln r|$.

Prop 2 If x_0 slowly recurrent then $S_r \Rightarrow \text{Pois}(2)$

Prop 3 Almost every $x_0 \in M$ is slowly recurrent

We start with Prop. 2. We use the following expression for the factorial moments of the number of visits

Let $S = \sum_{k=1}^N \mathbb{1}_{A_k}$. Then

$$\mathbb{E} \binom{S}{r} = \mathbb{E} \left[\frac{S(S-1)\dots(S-r+1)}{r!} \right] = \sum_{r_1 < r_2 < \dots < r_r} P(A_{r_1} A_{r_2} \dots A_{r_r})$$

Exercise 10 Multiple exponential mixing implies multiple exponential mixing on balls, that is

$\exists \kappa, \beta$ s.t.

$$\left| \mu \left(\prod_{k=1}^r \mathbb{1}_{B(x_0, r)} \right) (f^{\kappa x}) \right| - \mu(B(x_0, r)) \leq C \left(\frac{1}{r} \right)^\kappa e^{-\beta L} \quad L = \min_{i \neq j} |n_i - n_j|$$

We now show that ~~the main contribution to~~

$$E \left(\frac{s(s-1) \dots (s-r+1)}{r^r} \right) \xrightarrow{p \rightarrow 0} 2^r$$

We use (A). Take a constant $R \gg 1$. We claim that the main contribution to (1) comes from well separated tuples, that is, the tuples s.t.

$$\min_{i \neq j} |n_i - n_j| > R \ln r.$$

For well separated tuples

$$\mu \left(\prod_{j=1}^r \mathbb{1}_{B(x_0, r)} \right) (f^{n_j} x) = \mu(B(x_0, r)) (1 + o(1))$$

Since the number of tuples is $\frac{N^r}{r!} (1 + o(1))$ we conclude that the contribution of well separated tuples to $E \left(\frac{s(s-1) \dots (s-r+1)}{r!} \right)$ is $2^r (1 + o(1))$.

It remains to control the tuples which are not well separated. 9

By cluster decomposition we mean partition
 $\{n_1, n_2, \dots, n_r\} = \bigcup_{j=1}^s C_j$ where the clusters C_1, \dots, C_s
 satisfy

(a) $D := \max_{n', n'' \in C_j} |n' - n''| \leq \mathcal{K}(R, r) \ln p.$

(b) $\min_{n', n'' \text{ in different clusters}} |n' - n''| > R \max(D, \ln r)$

Proceeding recursively starting from the partition
 $\{n_1\} \cup \{n_2, \dots, n_r\}$ we see that each tuple
 admits ~~a~~ cluster partition and
 $s = r$ iff the tuple is well separated.

Let ψ_p be a function so that

$$\psi_p(x) = 1 \text{ if } x \in B(x_0, p)$$

$$\psi_p(x) = 0 \text{ if } x \notin B(x_0, 2p)$$

$$\text{and } \|\psi\|_{C^1} \leq K/p.$$

By definition of cluster partition

$$K \prod_{i=1}^r \mathbb{1}_{B(x_0, p)}(p^{n_i} x) \leq \mu \left(\prod_{j=1}^r \psi_p(p^{n_j} x) \right)$$

$$= \prod_{j=1}^s \mu(\varphi_{C_j}) + O(N^{-10r})$$

where $\varphi_{C_j} = \prod_{k \in C_j} \psi_p(p^{n_k} x)$

Note that if C_j has size 1 then

$$\mu(\mathcal{C}_{C_j}) \leq \mu(B(x_0, \rho)) \leq K \rho^d \text{ where}$$

d is the dimension of the phase space while if the size of C_j is greater than 1 then

$$\begin{aligned} \mu(\mathcal{C}_{C_j}) &\leq \mu(B(x_0, \rho) \cap \rho^{-L} B(x_0, \rho)) \\ &\leq \frac{K \rho^d}{|\ln \rho|^{100r}} \end{aligned}$$

by definition of slow recurrence where $L = n'' - n'$

for $n', n'' \in C_j$.

It follows that the contribution of each non well separated

term is $O(\rho^{ds} / |\ln \rho|^{100r})$. Since the number of

terms is $O(N^s |\ln \rho|^{r-s}) \leq O(\bar{\rho}^{sd} |\ln \rho|^r)$

the contribution of not well separated tuples is negligible completing the proof of the proposition \square

Next we prove Prop. 3. ~~Given~~ Given A take large B

and let $N = B \ln |\ln \rho|$, $\hat{\rho} = |\ln \rho|^{-A}$

If $k \geq N$ then

$$\begin{aligned} \mu(d(x, \rho^k x) \leq \rho) &\leq \mu(d(x, \rho^k x) \leq \hat{\rho}) \stackrel{(\Delta)}{\leq} C \mu(\{(x, y) \in \mathcal{M} : d(x, y) \leq \hat{\rho}\}) \\ &\leq \bar{C} \hat{\rho}^2 \end{aligned}$$

by exponential mixing

Exercise 11 prove (Δ) by approximating $\mathbb{1}_{d(x, y) \leq \hat{\rho}}$ by a smooth function.

Now we consider $k \in \mathbb{N}$. Suppose

$$d(x, T^k x) \leq \rho. \text{ Then}$$

$$d(x, T^{kL} x) \leq (\rho L)^{Lk} \text{ where } L = \max_x \|dT(x)\|$$

~~Then~~ Taking an appropriate L we get that

$\exists \hat{N} \in (\mathbb{N}, 2\mathbb{N})$ such that

$$d(x, T^{\hat{N}} x) \leq C \rho L^{\hat{N}} \leq \sqrt{\rho}$$

Then the same argument as before shows that

$$\mu(x, T^k x) \leq \rho \leq \mu(x, T^{\hat{N}} x \leq \sqrt{\rho}) \leq C \rho^{\hat{N}/2}$$

Next if x is not slowly recurrent then

$\forall j_0 \exists j \geq j_0 \exists k \leq K_j$ such that

~~$$\mu(B(x, 2^{-j}) \cap T^{-k} B(x, 2^{-j})) \geq \frac{1}{jA} \quad (*)$$~~

Call the set of x satisfying (*) $H_{j,k}$.

By the foregoing discussion

$$\mu(H_{j,k}) \leq \frac{1}{j^2 A}$$

and so Borel-Cantelli Lemma shows that a.e. point is (K, A) slowly recurrent.

Since (K, A) are arbitrary the proposition follows. \square

The proof of Thm 1 uses a standard argument common to most proofs of Poisson limit theorems in dynamics.

Namely the problem is reduced to studying the probability that the orbit of x

visits $B(x_0, \rho)$ at times n_1, n_2, \dots, n_r . If ~~the~~ n_i 's are far from each other (long returns) then the required estimate $\mu(B(x, \rho))^{-1}$ follows by mixing.

The case where there are short returns ($\exists i, j$ with $n_i - n_j$ small) requires a separate geometric argument.

We also note that the clustering argument used in the proof of ~~the~~ Thm 1 can be used to obtain the following result.

Let $S_N(x) = \sum_{n=0}^{N-1} A(T^n x)$, where $A \in C^k(\mathbb{M})$

Thm 4 (Björklund-Gorodnik) If T is multiply exponentially mixing then

$\frac{S_N - N\mu(A)}{\sqrt{N}}$ converges to a normal distribution.

We refer to the paper of Björklund-Gorodnik for detailed proof ~~and also~~ which is valid not only for \mathbb{Z} actions but for other groups of subexponential growth.