

6^o LECTURE - ADVANCED STOCHASTIC PROCESSES

Def [Visits to a state - local time]: $\forall y \in S$, let

$$\forall m \geq 0: N_m(y) := \sum_{n=0}^{m-1} \mathbb{1}_{\{X_n=y\}}, \quad N(y) := \sum_{n=0}^{+\infty} \mathbb{1}_{\{X_n=y\}}$$

of visits to y , up to time m # of visits to y

and set $G(x,y) := \mathbb{E}_x(N(y)) = \sum_{n=0}^{+\infty} \mathbb{P}_x(X_n=y) = \sum_{n=0}^{+\infty} P_{x,y}^{(n)}$

\hookrightarrow called Green functions: $G: S \times S \rightarrow \bar{\mathbb{R}}_+$

Theorem 1 [Criterion for recurrence]: $\forall x, y \in S$, it holds

$$G(x,y) = \begin{cases} \frac{F(x,y)}{1 - F(y,y)} & \text{if } x \neq y \\ \frac{1}{1 - F(y,y)} & \text{if } x = y \end{cases}$$

Remark: Then y is recurrent ($F(y,y)=1$) $\iff G(y,y) = +\infty$.

Convenient, as $G(y,y) = \sum_{n=0}^{+\infty} P_{y,y}^{(n)}$, explicit sum.

The proof of the above theorem use the following Lemma

Lemma: $\forall y \in S$, define iteratively the k -th passage times in y

$$T_y^1 \equiv T_y, \quad T_y^k = \inf\{n > T_y^{k-1} : X_n = y\} \quad \forall k \geq 2$$

Then

$$\mathbb{P}_x(T_y^k < \infty) = F(x,y)F(y,y)^{k-1}$$

Proof [of Lemma]: By induction over k .

- for $k=1$, obvious from def. of $F(x,y)$;
- for $k \geq 2$, by the strong Markov property:

$$0 \leq \mathbb{P}_x(T_y^k < \infty) \leq \mathbb{P}_x(T_y^{k-1} < \infty) \mathbb{P}_y(T_y < \infty) = \mathbb{P}_x(T_y^{k-1} < \infty) F(y,y)$$

• for $n \geq 2$, by the strong Markov property:

$$P_x(\tau_y^k < \infty) = E_x \left(\underbrace{P_x(\tau_y^k < \infty | \mathcal{F}_{\tau_y^{k-1}})} \cdot \mathbb{1}_{\{\tau_y^{k-1} < \infty\}} \right)$$

$$\underbrace{X_{\tau_y^{k-1}} = y}_{\text{inductive hypothesis}} = E_x \left(P_y(\tau_y^k < \infty) \mathbb{1}_{\{\tau_y^{k-1} < \infty\}} \right) = F(y, y) P_x(\tau_y^{k-1} < \infty)$$

$$\stackrel{\text{inductive hypothesis}}{=} F(y, y) F(y, y)^{k-2} \cdot F(x, y) \quad \#$$

Proof [of Theorem - criterion for recurrence]:

Note that $N(y) = \mathbb{1}_{\{X_0=y\}} + \sum_{k=1}^{\infty} \mathbb{1}_{\{\tau_y^k < \infty\}}$. Averaging w.r.t. P_x

$$G(x, y) = E_x(N(y)) = \delta_x(y) + \sum_{k=1}^{\infty} P_x(\tau_y^k < \infty)$$

$$\text{(from Lemma)} = \delta_x(y) + \sum_{k=1}^{\infty} F(x, y) F(y, y)^{k-1} = \delta_x(y) + \frac{F(x, y)}{1 - F(y, y)} \quad \#$$

Consequences: If X is irreducible and $x \in S$ is recurrent, then:

1. all states $y \in S$ are recurrent (recurrence is a property of X)

Proof: We know that $G(x, x) = +\infty$.

For $y \in S$, let $l, k \in \mathbb{N}$ st. $P_{x,y}^{(l)} > 0$, $P_{y,x}^{(k)} > 0$ (irreduc.) so that

$$P_{y,y}^{(l+m+k)} \geq P_{y,x}^{(l)} \cdot P_{x,y}^{(m)} \cdot P_{y,y}^{(k)} \quad \rightarrow \quad \text{Then}$$

$$G(y, y) = \sum_{m \geq 0} P_{y,y}^{(m)} \geq \sum_{m \geq 0} P_{y,y}^{(l+m+k)} \geq P_{y,x}^{(l)} P_{x,y}^{(k)} \underbrace{\sum_{m \geq 0} P_{x,y}^{(m)}}_{G(x, y) = +\infty} = +\infty \quad \#$$

2. $F(x, y) = F(y, x) = 1 \quad \forall x, y \in S$

Proof: $0 = P_x(\tau_x = \infty) \geq \underbrace{P_x(\tau_x = \infty, X_k = y)}_{\forall k \in \mathbb{N}} = \underbrace{P_x(X_k = y)}_{\text{Markov Prop.}} P_y(\tau_x = \infty)$

$$= P_{x,y}^{(k)} (1 - F(y, x)) \iff F(y, x) = 1$$

for some $k \in \mathbb{N}$
due to irreduc.

(hence no x, y and y we get similarly $F(x, y) = 1$. ")

due to irreducibility
 Changing x and y , we get similarly $F(x,y)=1$. \neq

Corollary: If S is finite and X is irreducible then X is recurrent

Proof: $\forall x \in S: \sum_{y \in S} G(x,y) = \sum_{n=0}^{\infty} \sum_{y \in S} P_{x,y}^{(n)} = +\infty$

But since S is finite, this can be true only if $\exists y \in S$ st. $G(x,y) = \infty$. From the recurrence criterion, as $F(x,y) > 0$ from ir.,

$$G(x,y) = \frac{F(x,y)}{1 - F(y,y)} = \infty \iff F(y,y) = 1,$$

hence y (and all states in S) are recurrent. \neq

Existence of invariant measures and distribution

Theorem 2: If X is irreducible on S and transient \implies no invariant distribution

Proof: $G(x,y) = \sum_{n=0}^{\infty} P_{xy}^{(n)} < \infty \quad \forall x,y \in S$ (from transience and thm 1)

Then $P_{xy}^{(n)} \xrightarrow{n \rightarrow \infty} 0$ and hence $\sum_{x \in S} \mu(x) P_{xy}^{(n)} \xrightarrow{n \rightarrow \infty} 0$, $\forall \mu \in \mathcal{P}(S), \forall y \in S$

If μ would be an invariant dist., we would have

$$\mu P^n = \mu, \quad \forall n \in \mathbb{N}, \text{ and taking the limit } n \rightarrow \infty: \mu = 0,$$

which is a contradiction. \neq

Remark: Under the above hypotheses, there may exist invariant measures

\hookrightarrow Example 1: let $(S_n)_{n \geq 0}$ be a random walk on \mathbb{Z} as follows:

$$\begin{matrix} 1-p & p \\ \leftarrow & \rightarrow \end{matrix}$$

$$P_{j,j+1} = p, \quad P_{j,j-1} = 1-p, \quad \text{if } j \neq \pm 1$$

↳ Example 1: Let $(S_n)_{n \geq 0}$ be a random walk on \mathbb{Z} as follows:



If $p \neq \frac{1}{2}$, by Stirling approximation one gets

$$P_{0,0}^{(2n)} = \binom{2n}{n} p^n (1-p)^n \approx \frac{1}{\sqrt{\pi n}} \underbrace{[4p(1-p)]^n}_{< 1}$$

Hence, $G(0,0) = \sum_{n \geq 0} P_{0,0}^{(2n)} < \infty$, implying that (S_n) is transient

While, no invariant distribution exists, one can check that

The following measures are invariant:

- $\mu_1(x) = 1 \quad \forall x \in \mathbb{Z}$
 - $\mu_2(x) = \left(\frac{p}{1-p}\right)^x$
- (Verify that, $\forall x \in \mathbb{Z}$
 $p\mu(x-1) + (1-p)\mu(x+1) = \mu(x)$)

Theorem 3: If X is irreducible and recurrent, then, $\forall x \in S$,

$$\underline{\mu_x(y) := \mathbb{E}_x(N_{T_x}(y))}, \quad \forall y \in S$$

are invariant measures, and they are s.t. $\mu_x(y) > 0 \quad \forall x, y \in S$.

Moreover, invariant measures only differ from a multiplicative factor.

Remark:

1. Notice that the measure μ_x is not necessarily finite, hence the existence of an invariant distr. is not guaranteed.

Indeed:

$$\sum_{y \in S} \mu_x(y) = \mathbb{E}_x \left(\sum_{y \in S} N_{T_x}(y) \right) = \mathbb{E}_x(T_x)$$

- Under solely recurrence, $\mathbb{E}_x(T_x)$ may be $\infty \rightarrow$ no inv. distr.

- Under solely recurrence, $\mathbb{E}_x(\tau_x)$ may be $\infty \rightarrow$ no inv. distr
- Under positive recurrence of x : $\mathbb{E}_x(\tau_x) < \infty$, hence:

$$\pi_x(y) := \frac{\mu_x(y)}{\mathbb{E}_x(\tau_x)}, \quad y \in S \text{ is } \mathcal{P}(S)$$

and is s.t. $\pi_x(x) = \frac{1}{\mathbb{E}_x(\tau_x)}$

But since μ_x is unique up to multiplicative constant, i.e. $\mu_x = c \mu_y$ for some $c = c(x, y) > 0, \forall x, y \in S$, it turns out

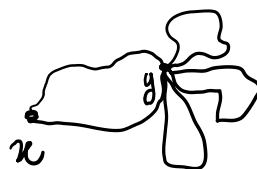
- $\mathbb{E}_y(\tau_y) < \infty \quad \forall y \in S$ (positive recurrence is a property of the MC)

- $\forall x, y \in S: \pi_x = \pi_y = \pi$ and then

$$\pi_y(y) = \frac{1}{\mathbb{E}_y(\tau_y)} = \frac{\mu_x(y)}{\mathbb{E}_x(\tau_x)} = \pi_x(y) \iff \mathbb{E}_x(\tau_x) = \mu_x(y) \mathbb{E}_y(\tau_y)$$

$$\iff \mathbb{E}_x(\tau_x) = \mathbb{E}_x(N_{\tau_x}(y)) \mathbb{E}_y(\tau_y)$$

idea \rightarrow



- 4 passage to y before τ_x
 - all of them take a time $\mathbb{E}_y(\tau_y)$

As a consequence of theorem 3 and of the Remark that followed it, we get the proof of the following result:

Theorem 4: If X is irreducible and positive recurrent, then exists a unique invariant distribution π , s.t.

$$\pi(x) = \frac{1}{\mathbb{E}_x(\tau_x)}, \quad \forall x \in S$$

Remark: This implies that if S is finite and X is irreducible,
Then the MC is positive recurrent (put things together).
