

Lecture 18

Wed Mar/27/2024

Last time

- ▷ Recap
- ▷ Recurrence and transience

Today

- ▷ Stationary measures
- ▷ Existence
- ▷ Uniqueness

Stationary Measures

Let S be the state space of a Markov chain. Today we will cover stationary or invariant measures.

Def: Let X_n be a MC with countable state space S . A measure μ on S is stationary if

$$\sum_{x \in S} \mu(x) p(x, y) = \mu(y)$$

"Prob." of getting to y after initializing with μ and running one step of the MC.

If μ is a prob. measure $\Rightarrow \mu$ is a stationary dist.

Lemma: If μ is stationary distribution,
 $\Rightarrow P_\mu(X_n = y) = \mu(y) \quad \forall n.$

Proof: Exercise.

□

Example (Asymmetric random walk): Let

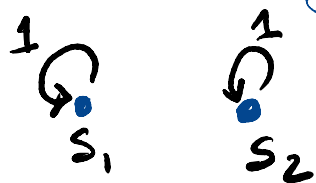
$$S_n = \sum_{k=1}^n \xi_k \quad \text{with} \quad P(\xi_k = 1) = p = 1 - P(\xi_k = -1).$$

Let $p > q := 1 - p$. Then $\mu(x) = \left(\frac{p}{q}\right)^x$ is a stationary measure. Note that

$$\begin{aligned} \sum_{x \in S} \left(\frac{p}{q}\right)^x p(x, y) &= \mu(y+1) p(y+1, y) + \mu(y-1) p(y-1, y) \\ &= \left(\frac{p}{q}\right)^{y+1} q + \left(\frac{p}{q}\right)^{y-1} p \\ &= \left(\frac{p}{q}\right)^y (p + q) \\ &= \left(\frac{p}{q}\right)^y. \end{aligned}$$

However, μ is not a dist. since $\sum_x \mu(x) = \infty$. \rightarrow

Example (Lack of uniqueness): Consider



Both $\mu_1(x) = \mathbb{1}_{x=s_1}$ and $\mu_2(x) = \mathbb{1}_{x=s_2}$

are stationary distributions. \rightarrow

Example (Lack of existence): Consider



Assume μ is a non zero stationary measure, then

$$\exists y \text{ s.t. } \mu(y) > 0,$$

$$\mu(y) = \sum \mu(x) p(x, y) = \mu(y-1) p(y-1, y) \stackrel{\text{induction}}{=} \prod_{x=0}^y p(x, x+1) \mu(0)$$

But also, $\mu(0) = \sum \mu(x) p(x, 0) = 0 \quad \Downarrow \quad \perp$

Existence

Theorem Let X_n be a MC with countable state space. Assume that $\exists x$ recurrent. Let $T_x = \inf \{n \geq 1 : X_n = x\}$, then $\mu_x(y) = \mathbb{E}_x \left(\sum_{n=0}^{T_x} \mathbb{1}_{\{X_n=y\}} \right) = \sum_{n=0}^{\infty} \mathbb{P}(X_n=y, T_x > n)$ defines a stationary \uparrow measure. \rightarrow

Number of times we visit y in one cycle.

Remark: When all states are transient, a stationary measure may or may not exist (see examples above).

Proof: Define $\bar{p}_n(x, y) = \mathbb{P}_x(X_n=y, T_x > n)$. By Fubini's

$$\begin{aligned} \sum_{y \in S} \mu_x(y) p(y, z) &= \sum_{y \in S} \sum_{n=0}^{\infty} \bar{p}_n(x, y) p(y, z) \\ &= \sum_{n=0}^{\infty} \sum_{y \in S} \bar{p}_n(x, y) p(y, z) = \textcircled{*}. \end{aligned}$$

We want to show that $\textcircled{*} = \mu_x(z)$.

Consider two cases

Case 1: $z \neq x$

$$\sum_y p_n(x, y) p(y, z) = \sum P_x(X_n = y, X_{n+1} = z, T_x > n)$$

Since $T > n$ and we are summing every possibility. $\bar{p}_{n+1}(x, z) = P_x(X_{n+1} = z, T_x > n+1)$

$$\text{Then } (*) = \sum_{n=0}^{\infty} \bar{p}_{n+1}(x, z) = \sum_{n=0}^{\infty} \bar{p}_n(x, z)$$
$$= \mu_x(z)$$

$\bar{p}_0(x, z) = 0$ since $x \neq z$.

Case 2: $z = x$

$$\sum_y p_n(x, y) p(y, x) = \sum P_x(X_n = y, T_x > n, X_{n+1} = x)$$
$$= P(T_x = n+1).$$

$$\text{Then, } (*) = \sum_{n=0}^{\infty} P(T_x = n+1) = \sum_{n=0}^{\infty} P(T_x = n)$$
$$= 1 = \mu_x(x).$$

$P(T_x = 0) = 0$.

□

Uniqueness

We saw an example where we had two different stationary measures, it is easy to see that their

conic hull would also be stationary.

Exercise (♥): Assume that μ_1 and μ_2 are stationary \Rightarrow for any $\alpha, \beta \in \mathbb{R}$ we have that $\nu = \alpha\mu_1 + \beta\mu_2$ is stationary provided that $\nu(x) \geq 0 \quad \forall x$ and $\exists x \quad \nu(x) > 0$ \dashv

Broadly speaking, uniqueness fails because of two reasons:

- Multiple irreducible classes (↻ ↻).
- Lack of recurrence (In the asymmetric walk example $\mu(x) = 1$ is stationary).

The following Theorem formalizes this claim.

Theorem: If a MC is irreducible and recurrent (all states are recurrent), then, there is a unique stationary measure up to constants.

Proof: Fix $x \in S$, let μ be a stationary measure, we shall prove that $\mu = \mu_x$ (up to constants). WLOG set $\mu(x) = 1$

Let $y \neq x$, by assumption

$$\mu(y) = \sum_{z_0 \in S} \mu(z_0) p(z_0, y)$$

$$\rightarrow = p(x, y) + \sum_{z_0 \neq x} \mu(z_0) p(z_0, y)$$

$$\mu(x)=1 = \mathbb{P}_x(X_1=y, T_x > 1) + \underbrace{\sum_{z_0 \neq x} \mu(z_0) p(z_0, y)}_{T_1}$$

Repeating the argument

$$T_1 = \sum_{z_0 \neq x} \left(\sum_{z_1 \in S} \mu(z_1) p(z_1, z_0) \right) p(z_0, y)$$

$$= \sum_{z_0 \neq x} p(x, z_0) p(z_0, y)$$

$$+ \sum_{z_0, z_1 \neq x} \mu(z_1) p(z_1, z_0) p(z_0, y)$$

$$= \mathbb{P}_x(X_2=y, T > 2) +$$

$$\sum_{z_0, z_1 \neq x} \mu(z_1) p(z_1, z_0) p(z_0, y)$$

Inductively we obtain

$$\mu(y) \geq \sum_{m=1}^n \mathbb{P}[X_m=y, T_x > m] \uparrow \mu_x(y).$$

Searching contradiction assume that

$$\mu(y) - \mu_x(y) > 0 \text{ for some } y.$$

Then by Exercise (b) we get that $\nu = \mu - \mu_x$ is stationary. Further since $\mu(x) = \mu_x(x) = 1$, $\nu(x) = 0$.

Let k be such that $p^{(k)}(y, x) > 0$.

Then,

$$0 = \nu(x) = \sum_{z \in S} \nu(z) p^{(k)}(z, x) = \nu(y) p^{(k)}(y, x) > 0.$$

So $\mu = \mu_x$.

□