

LECTURE 18:

Thm: (Uniqueness of stationary measure) If S irreducible, all states recurrent $\Rightarrow \exists!$ non-zero stationary μ up to rescaling. (i.e. all = $c\mu$)

Pf: let $\mu \neq 0$ stationary, $x \in S$ w/ $\mu(x) \neq 0$. WLOG $\mu(x) = 1$, will show then $\mu = \mu_x$ (recall $\mu_x(x) = 1$).

Will see below:
~~($\mu_x(y) > 0 \forall y$)~~ $\mu(y) > 0$ also.

Consider some $y \neq x$:

$$\mu(y) = \sum_{z_0} \mu(z_0) p(z_0, y) = \underbrace{p(x, y)}_{\mathbb{P}_x[Y_1=y, T_x > 1]} + \underbrace{\sum_{z_0 \neq x} \mu(z_0) p(z_0, y)}_{(*)}$$

$$(*) = \sum_{z_0 \neq x, z_1} \mu(z_1) p(z_1, z_0) p(z_0, y) = \underbrace{\sum_{z_0 \neq x} p(x, z_0) p(z_0, y)}_{\mathbb{P}_x[Y_2=y, T_x > 2]} + \underbrace{\sum_{z_0, z_1 \neq x} \mu(z_1) p(z_1, z_0) p(z_0, y)}_{(*)}$$

... repeat $\Rightarrow \forall N, \mu(y) \geq \sum_{n=1}^N \mathbb{P}_x[Y_n=y, T_x > n]$

$$\Rightarrow \mu(y) \geq \sum_{n=1}^{\infty} \mathbb{P}_x[Y_n=y, T_x > n] = \mu_x(y) \Rightarrow \boxed{\mu(y) \geq \mu_x(y)}$$

$\mu' := \mu - \mu_x$ stationary, $\mu'(x) = 0$.

But, S irreducible $\Rightarrow \forall y, p_{yx} > 0 \Rightarrow p^{(k)}(y, x) > 0$ for some k .

$$0 = \mu'(x) = \sum_z \mu'(z) p^{(k)}(z, x) \geq \mu'(y) p^{(k)}(y, x) \Rightarrow \mu'(y) = 0 \quad \forall y$$

$$\Rightarrow \mu' = 0 \Rightarrow \mu = \mu_x. \blacksquare$$

Cor: S irreducible + recurrent, μ^* stationary $\Rightarrow \mu(x) > 0 \forall x$.

② Lemma: (Partial converse) If μ stationary ^{prob measure}, $\mu(x) > 0 \implies x$ recurrent.

PF: $\forall n, \mu(x) = \sum_y \mu(y) p^{(n)}(y, x)$.

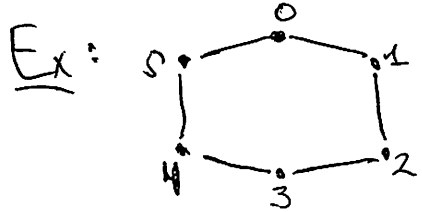
OTOH, $\sum_{n=1}^{\infty} p^{(n)}(x, y) = \sum_{n=1}^{\infty} P_x[Y_n = y] = \mathbb{E}_x N(y) = \frac{p_{xy}}{1 - p_{yy}}$ (# visits to y)

$\infty = \sum_{n=1}^{\infty} \mu(x) = \sum_y \mu(y) \frac{p_{xy}}{1 - p_{yy}} \leq \frac{1}{1 - p_{yy}} \implies p_{yy} = 1$.

Rk: -- $\curvearrowright \curvearrowright \curvearrowright \curvearrowright \curvearrowright$ example shows μ must be finite.

Convergence: recall, $\mu_{x,n}(y) := P_x[Y_n = y]$. If μ stationary, $\mu_{x,n}(y) \rightarrow \mu(y)$.

One obstruction: periodicity:



SRW on this graph $\implies \begin{cases} P_0[Y_n = 1] = 0, \\ P_0[Y_{2n+1} = 1] \rightarrow \frac{1}{3}. \end{cases}$
 $\mu(x) = \frac{1}{6}$ stationary, recurrent

Def: MC aperiodic if $\forall x \in S^v \exists n_0$ s.t. $p^{(n)}(x, x) > 0 \forall n \geq n_0$.

Exc: Aperiodic $\iff \forall x \in S^v$, there is no $k \geq 2$ divisor of all $\{n : p^{(n)}(x, x) > 0\}$.

Thm: Suppose MC is irreducible + aperiodic, and μ is stationary prob. measure.

Then, $\forall x \in S, \mu_{x,n} \xrightarrow[n \rightarrow \infty]{(weak)}$ μ .

Cor: ~~_____~~

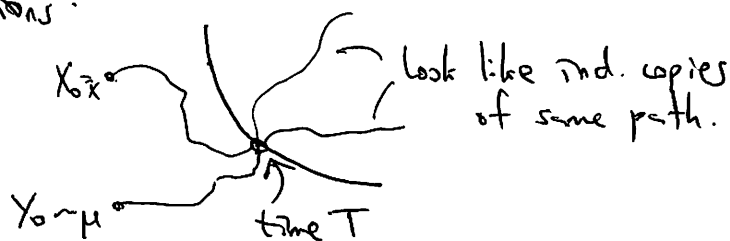
same weak convergence for $\forall y, \mu_{\mu_0}(y) = P_{\mu_0}[Y_1 = y] \forall \mu_0$.

③ Pf idea: Coupling: compare (X_n) started from $X_0 = x$ versus

(Y_n) from $Y_0 \sim \mu$ ~~Observations~~ Observations:

(1) Law $(X_n) = \mu_{X,n}$

(2) Law $(Y_n) = \mu$



(3) $T := \min \{ n \geq 1 : X_n = Y_n \} \rightarrow$ "after T look identical":

Law $(X_{T+n}) = \text{Law}(Y_{T+n}) = \mu$.

Pf: Define kernel on S^2 : $q((x_1, y_1), (x_2, y_2)) = p(x_1, x_2) p(y_1, y_2)$.

• q irreducible: p irred. $\Rightarrow p^{(k)}(x_1, x_2), p^{(l)}(y_1, y_2) > 0$

~~aperiodic~~ aperiodic $\Rightarrow \exists m$ s.t. $p^{(k+m)}(x_1, x_2), p^{(l+m)}(y_1, y_2) > 0$.

$$\Rightarrow q^{(k+l+m)}((x_1, y_1), (x_2, y_2)) \geq p^{(k)}(x_1, x_2) p^{(l+m)}(x_1, x_2) p^{(l)}(y_1, y_2) p^{(k+m)}(y_1, y_2) > 0.$$

• $v((x, y)) := v(x) v(y)$ stationary for q , and prob. measure.

• Lem \Rightarrow all states recurrent for q . (S recurrent $\Rightarrow \mu > 0; v = \mu^{\otimes 2}$)

Let (X_n, Y_n) be MC w/ transition kernel q , init. from $\delta_x \otimes \mu$.

Claim: $O_n \{T \leq n\}$, X_n and Y_n have same law. I.e.:

$$\begin{aligned} P[X_n = z, T \leq n] &= \sum_{m=0}^n \sum_y P[T = m, X_m = y, X_n = z] \\ &= \sum_{m=0}^n \sum_y P[T = m, X_m = y] \cdot P[X_n = z | X_m = y, T = m] \xrightarrow{\text{Markov prop.}} \\ &= \sum_{m=0}^n \sum_y P[T = m, Y_m = y] \cdot P[Y_n = z | Y_m = y] = \dots = P[Y_n = z, T \leq n]. \end{aligned}$$

④ Finally, compare distributions of interest:

$$\begin{aligned} \mu_{x,n}(y) &= P[X_n = y] \\ &= P[X_n = y, T \leq n] + P[X_n = y, T > n] \\ &= P[Y_n = y, T \leq n] + P[X_n = y, T > n] \\ &= \mu(y) - P[Y_n = y, T > n] + P[X_n = y, T > n]. \end{aligned}$$

$$\Rightarrow |\mu_{x,n}(y) - \mu(y)| \leq P[X_n = y, T > n] + P[Y_n = y, T > n]$$

$$\Rightarrow \sum_y |\mu_{x,n}(y) - \mu(y)| \leq 2P[T > n] \quad \text{Exc: Condition on starting } y \sim \mu.$$

\int^2 recurrent + irreducible $\Rightarrow P[\text{ever reach } (z, z)] = 1 \Rightarrow T < \infty$ a.s.

$$\Rightarrow \sum_y |\mu_{x,n}(y) - \mu(y)| \xrightarrow[n \uparrow \infty]{\delta_x \otimes \mu} 0 \quad \text{Rk: } \mu(y) > 0 \text{ by earlier Cor.}$$

$$\sum_y \left| \frac{\mu_{x,n}(y)}{\mu(y)} - 1 \right| \mu(y) \geq d_1(\mu, \mu_{x,n}) \quad (\text{HW 2, Prob 4}) \\ = \text{"total variation" distance.}$$

$$\Rightarrow \mu_{x,n} \xrightarrow{(\text{weak})} \mu. \quad \blacksquare$$

Rk: Important applications in Markov Chain Monte Carlo \rightarrow given μ ,
design p making μ stationary.