

# MATH 285: Stochastic Processes

[math-old.ucsd.edu/~ynemish/teaching/285](http://math-old.ucsd.edu/~ynemish/teaching/285)

## Today: Ergodic theorem

- Homework 3 is due on Friday, February 4, 11:59 PM

# Probability generating function

Def Let  $Y$  be a random variable with values in  $\{0, 1, 2, \dots\}$ .

We call the function

$$\varphi_Y(s) :=$$

The probability generating function of  $Y$ .

Properties:

(1)  $\varphi_Y(s)$  is analytic on  $(-1, 1)$ ;  $\varphi_Y^{(n)}(0) = n! \mathbb{P}[Y=k]$

(2)  $\varphi_Y(1) = 1$ ;  $\varphi_Y(0) = \mathbb{P}[Y=0]$

(3) For  $|s| < 1$ ,  $\varphi_Y'(s) = \sum_{k=1}^{\infty} k s^{k-1} \mathbb{P}[Y=k]$ ; if  $\mathbb{E}[Y] < \infty$ , then  $\varphi_Y'(1) = \mathbb{E}[Y]$

(4) For  $|s| < 1$ ,  $\varphi_Y''(s) = \sum_{k=2}^{\infty} k(k-1) s^{k-2} \mathbb{P}[Y=k]$ ; in particular, if  $\mathbb{P}[Y \geq 2] > 0$ , then  $\varphi_Y(s)$  is (strictly) convex on  $(0, 1)$

## Ergodic Theorem

Thm 11.3 Let  $(X_n)$  be an irreducible recurrent Markov chain with state space  $S$ . Let  $j \in S$ . Define

$$(\pi(j) = 0 \text{ if } \mathbb{E}_j[T_j] = \infty).$$

Let  $V_n(j) := \sum_{m=1}^n \mathbb{1}_{\{X_m=j\}}$  be the number of visits to state  $j$  up to time  $n$ . Then for any state  $i \in S$

and

Proof. (i)  $\mathbb{P}_i[V_n(j) \rightarrow \infty \text{ as } n \rightarrow \infty] = 1$

Otherwise  $\mathbb{P}_j[(X_n) \text{ visits } j \text{ finitely many times}] > 0$

# Ergodic Theorem

Denote by  $T_j^k$  the time of the  $k$ -th visit to state  $j$ .

(ii)  $T_j^{V_n(j)} \leq n \leq T_j^{V_n(j)+1}$

(iii)

Repeating the proof from Thm 10.2 we have that

$$\mathbb{P}_i \left[ \frac{T_j^k}{k} \rightarrow \mathbb{E}_j[T_j] \text{ as } k \rightarrow \infty \right] = 1$$

By (i)  $\lim_{n \rightarrow \infty} \frac{T_j^{V_n(j)}}{V_n(j)} = \lim_{k \rightarrow \infty} \frac{T_j^k}{k}$

(iv) By the Squeeze lemma

, therefore  
By definition  $\pi(j) = \frac{1}{\mathbb{E}_j[T_j]}$ .

# Ergodic Theorem

$$(v) \quad \frac{1}{n} \sum_{m=1}^n p_m(i,j) =$$

$$(vi) \quad \lim_{n \rightarrow \infty} \frac{1}{n} \mathbb{E}_i[V_n(j)] =$$

## Convergence theorem

Theorem 7.4 Let  $P$  be a transition matrix for a **finite-state**, irreducible, aperiodic Markov chain. Then there exists a unique stationary distribution  $\pi$ ,  $\pi = \pi P$ , and for any initial probability distribution  $\nu$

$$\lim_{n \rightarrow \infty} \nu P^n = \pi$$

Theorem 12.1 Let  $(X_n)$  be an irreducible aperiodic Markov chain possessing a stationary distribution  $\pi$ . Then for any states  $i, j$

Remark (1) Thm 12.1 implies that the stationary distribution of an irreducible aperiodic MC is unique.

(2) In fact any irreducible MC has at most one stationary distribution.

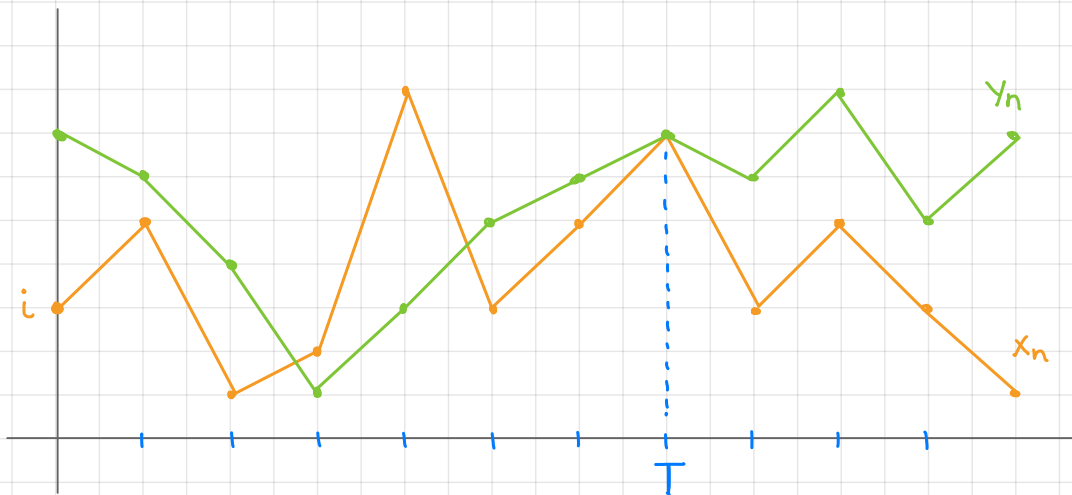
# Convergence theorem

## Proof of Thm 12.1

- Idea: couple two independent MCs, one starting from  $i$ , another with initial distribution  $\pi$ , wait until they collide.

$(X_n)$ : starting from  $i$

$(Y_n)$ : initial distr.  $\pi$



$(Z_n)$ :

## Convergence theorem

Let  $X_0 = i$ . Let  $(Y_n)$  be a MC with initial distribution  $\pi$ , transition probabilities  $p(i, j)$  (same as  $(X_n)$ ), and independent of  $(X_n)$ .

Take  $b \in S$  and define

(i)

Consider the process  $W_n :=$

$(W_n)$  is a MC with  $\mathbb{P}[W_0 = (k, e)] =$  and

transition probabilities  $\tilde{p}((k, e), (s, t)) =$

•  $(X_n)$  is aperiodic  $\Rightarrow \exists n$  s.t.  $p_n(k, s) > 0, p_n(e, t) > 0 \forall k, s, e, t$

$\Rightarrow \forall (k, e), (s, t) \in S \times S \tilde{p}_n((k, e), (s, t)) =$

$\Rightarrow (W_n)$  is



## Convergence theorem

- $\tilde{\pi}(k, \ell) := \pi(k)\pi(\ell)$  is the stationary distribution for  $(W_n)$   
$$\sum_{s, t \in S} \tilde{\pi}(s, t) p((s, t), (k, \ell)) = \sum_{s, t \in S} \pi(s)\pi(t) p(s, k) p(t, \ell) = \pi(k)\pi(\ell)$$
- Corollary 11.1  $\Rightarrow (W_n)$  is
- $T = \min\{n \geq 1 : W_n = (b, b)\} \stackrel{\text{HW 3, P1}}{\Rightarrow} \mathbb{P}[T < \infty] = 1$

Define  $Z_n = \begin{cases} X_n & \text{if } n \leq T \\ Y_n & \text{if } n > T \end{cases}$        $Z'_n = \begin{cases} Y_n & \text{if } n \leq T \\ X_n & \text{if } n > T \end{cases}$

(ii)  $(Z_n)$  is a MC starting from  $i$  with transition probabilities  $p(i, j)$

- $T$  is the stopping time for  $(W_n)$

## Convergence theorem

- By SMP  $(X_{T+n}, Y_{T+n})_{n \geq 0}$  is MC starting from  $(b, b)$  with transition probabilities independent of
- By symmetry  $(Y_{T+n}, X_{T+n})_{n \geq 0}$  is also a MC starting from  $(b, b)$  with transition probabilities independent of
- Therefore,  $Z_n$  has the same initial distribution and transition probabilities as  $Z_n$ . In particular,  $Z_n$  is a MC starting from  $i$  with trans. prob

## Convergence theorem

$$(iii) \quad |P[X_n = j] - \pi(j)| \leq$$

- $P[Z_n = j] =$

- By (ii)

- $P[Y_n = j] =$

- $|P[X_n = j] - \pi(j)| =$

=

$$(iv) \quad \text{By (i)}$$