MATH 285: Stochastic Processes

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Today: Periodic, aperiodic, reducible, irreducible Markov chains with finite state space

Homework 2 is due on Friday, January 21 11:59 PM

Prop. 7.1 Let (Xn) be a MC with finite state space S. Suppose that there exists no EN s.t [P]; >0 for all i,jes Then for each i, Ti(j) is equal to the asymptotic expected fraction of time the chain spends in state j, i.e., $\lim_{n\to\infty} \mathbb{E} \left[\frac{1}{n+1} \sum_{k=0}^{\infty} \mathbb{1}_{\{X_k=j\}} \right] = \pi(j)$ Proot. $\mathbb{E}\left[\frac{1}{n+1}\sum_{k=0}^{n}\mathbb{1}_{\left\{X_{k}=j\right\}}\right] = \frac{1}{n+1}\sum_{k=0}^{n}\mathbb{P}\left[X_{k}=j\right] = \frac{1}{n+1}\sum_{k=0}^{n}\mathbb{E}\left[X_{k}=j\right] \mathbb{P}\left[X_{o}=i\right]$ $=\frac{1}{n+1}\sum_{k=0}^{\infty}\left[\pi_{o}P^{k}\right].$ By Cor. 6.6. $[\Pi_0 P^K]_j \to \Pi(j)$, $k \to \infty$, for all jes and Π_0 .

Therefore, $\frac{1}{n+1} \sum_{k=0}^{n} [\Pi_0 P^K]_j \to \Pi(j)$ [if $a_n \to a$, $n \to \infty$, then $\frac{1}{n} \sum_{k=1}^{n} a_k \to a$]

Stationary distribution and long-run behavior

Stationary distribution and expected return times

Recall that Tik denotes the time of the k-th visit to state i.

Then by the strong Markov property

$$\{Y_{k}\}_{k=1}^{\infty}$$
 is a collection of i.i.d. random variables

 $Y_{k} \sim T_{i} = T_{i,2}$. Notice that $\sum_{k=1}^{m} Y_{k} = \sum_{k=1}^{m} T_{i,k+1} - T_{i,k} = T_{i,m+1}$
 $\lim_{k \to \infty} T_{i,m+1} = \lim_{k \to \infty} \sum_{k=1}^{m} Y_{k} \rightarrow \mathbb{E}[T_{i}]$, $m \to \infty$, so $T_{i,m+1} \approx m \cdot \mathbb{E}[T_{i}]$

Take m large, and let $n = m \cdot \mathbb{E}[T_{i}]$. Then $T_{i,m+1} \approx n$

Take m large, and let n=m E(Ti). Then Ti, m+, ≈n So $\sum_{k=0}^{n} \mathbb{1}_{\{X_k=i\}} = m+1 \approx n \, \mathbb{T}(i)$. Then $\frac{m+1}{n} \approx \mathbb{T}(i) \approx \frac{1}{\mathbb{E}[T_i]}$

Periodic and aperiodic chains Let (Xn) be a MC with state space 5 and transition probability p(i,j). Def For ies, denote Ji = {nzo: p(i,i)>0 } . We call d(i) := greatest common divisor of Ji (period of i) pe(0,1) J, = { 0 } J,={0,1,2,..} $J_1 = \{0, 2, 4, ... \}$ $J_1 = \{0, 2, 3, 4, 5, ... \}$ d(1)=1 no period d(1) = 2 d(1) = 1 Def If d(i)=1 for all i ∈ S, then (Xn) is called aperiodic

Periodic and aperiodic chains Lemma 7.2 If P is the transition matrix for an irreducible Markov chain then d(i)=d(j) for all states i.j. Proof. Fix ies. (1) If mine Ji, then mine Ji (2) Let d=d(i). Then J; c{0,d,2d,...} (definition of d(i)) (3) Pirreducible => 3 m,n s.t. pm (i,j)>0, pn (j,i)>0. ⇒ Pm+n (i,i)>0 => m+n ∈ Ji ⇒ 3 k∈N: m+n = kd (4) If le J; then pe(j,j)>0 and thus pm+e+n (i,i)>0 ⇒ m+l+n e Ji ⇒ 3 k': l = k'd ⇒ l divisible by d $\Rightarrow d \text{ is a common divisor of } J_j \Rightarrow J_{q,e} N \text{ s.t. } d(j) = q, d(i)$ (5) Swap i and j: $J_{q_2} \in N \text{ s.t. } d(i) = q_2 d(j) \Rightarrow d(i) = d(j)$

RW on bipartite graphs

Example 7.3 Let G=(V,E) be finite connected graph.

- · SRW on G is irreducible (all vertices have the same period) - we call the common period the period of MC
- For any i~; P(i,j)>0, P(j,i)>0, so P2(i,i)>0, 2 € Ji

$$V = V_1 \coprod V_2$$
, $E \subset (V_1 \times V_2 \cup V_2 \times V_1)$

⇒ d(i) ≤ 2

$$V = H$$
, $V_1 = \text{even numbers}$
 $V_2 = \text{odd numbers}$

Irreducible aperiodic Markov chains Theorem 7.4 Let P be a transition matrix for a finite-state, irreducible, aperiodic Markov chain. Then there exists a unique stationary distribution II, II = IT P, and for any initial probability distribution) $\lim_{n\to\infty} \mathcal{P}^n = 11$ Proof. (1) By PF theorem, enough to show that there exists no>o s.t. V iij [P"] ij>o . Fix iijes (2) d(i) = 1 (aperiodic) => = Mi s.t. Ji contains all n=Mi (3) irreducible => 3 mij s.t. pmij (i,j)>0 (2)+(3): Yn 2 Mir mij Pn(i,j)>0 Take $n_0 = \max_{i,j} (M_i + m_{ij}) \Rightarrow \forall i,j \in S pr_0(i,j) > 0$

Reducible Markov chains Not irreducible MC = reducible MC Def 7.5 Let (Xn) be a MC with state space 5. We say that states i and i communicate, denoted i is if there exist nime Nulois.t. pn (iij)>0 and pm (jii)>0 2 Pe(0,1) Lemma 7.6 Relation \leftrightarrow on S is an equivalence relation. (reflexivity, i +i) po(i,i) = 1, so i +i (symmetry, i + j > j + i) Follows from Def 7.5 (transitivity, i +) j + k => i + k) i + j: pn(i,j)>0, pm (j,i)>0 j +> k: Pn(j,k)>0, Pm'(kij)>0. Then Pn+n'(c,k)>0 Pm'+m (k,i) >0

Communication classes Equivalence relation \(\rightarrow \) splits the state space into communication classes (sets of states that communicate with each other). 1 pe(011) 1 P MC is irreducible iff it consists of one communication class Class properties: [proof as in Prop 4.8, Prop. 7.2] - transience or recurrence: either all states in one class are transient (class) or all are recurrent (class) - periodicity: all states in one class have the same period

Communication classes Suppose i and j belong to different classes.

• If p(i,j) > 0, then $p_n(j,i) = 0$ for all $n \in \mathbb{N}$ (otherwise $i \leftrightarrow j$.

i $\leftrightarrow j$.

If p(i,j) > 0 and p(j,i) = 0 for all $n \in \mathbb{N}$, then $P_i[X_n = i \text{ for infinitely many } n] \leq 1 - p(i,j) < 1$,

and thus i is transient

Therefore, if i and j belong to different classes

and i is recurrent, then p(i,j) = 0 (once in a

recurrent class, MC stays there forever)

If we split the state space into communication classes,

with Re denoting recurrent classes, then the transition matrix has the following form

General form of transition matrix with finite S Pe submatrix for the recurrent class Re Pe is a stochastic matrix, we can consider it as a Markov chain on Re T { S, Q [SIQ] transition probabilities starting from transient • If Pe is aperiodic, then Pe → [π(), n→∞ · What about transient states? · What if Pe is not aperiodic?