Stationary distributions

Sunday, October 15, 2017

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A Markov chain with transition matrix
Pxs has stutionary distribution TT if,

 $\pi P = \tau$

To interpret this if $X_0 \sim V$, that is $|P(X_0 = h) = Y_0$ then $|P(X_1 = j)| = \sum_{k} |P(X_1 = j \mid X_0 = k) \cdot P(X_0 = k)$

 $= \sum_{n} V_{n} \cdot P_{nj} = (r P)_{j}$

So X, 2 VP. In general X, ~ VP".

So if IT is stationary then Xn ~ TI for all n if Xo NTI.

Example: RW on a graph.

'4) $T_i = \frac{d_i}{2|E|}$ where d_i is degree of i.

 $(\Pi P)_{j} = \sum_{i} \frac{d_{i}}{2|E|} \cdot P_{ij} = \sum_{i} \frac{d_{i}}{2|E|} \cdot \frac{1}{d_{i}} I(in_{j})$ $= \frac{1}{2E} \sum_{i} I(in_{j}) = \frac{d_{i}}{2|E|}$

· Random to top shuffle

Let $g_n = (12...k)$ and G_n uniform on $\{g_n: 1 \leq h \leq n\}$

Then Xn = Gn Xn-, is a top to random shuffle

Let T(0) = n: be the uniform permutation.

Questions: ls T unique? Does X > T. How Fast?

· Example: RW on disconneted graph

A Markov chain is irreducible if for all is; $\exists n \quad \text{such that} \quad (P^n)_{ij}, \forall 0, \text{ that is}$ $P[X_n=j|X_0=i] \neq 0.$

Perron - Frobenius Theorem

It P is a stochastic matrix

then it has a left eigenvector M

with MP=M and Z.M:=1. The entries
of M are positive. If P is irreducible then

M is unique.

Proof Linear Algebra.

Probabilistic Existence proof:

Let $M = \frac{1}{n} \sum_{i=1}^{n} M P^{i}$. $M_n P - M_n = \frac{1}{n} M(P^{n+1} - P) \rightarrow 0$ ME {VE[0,1] : \(\frac{1}{2} \) \(V;=1 \) Compact set so I no such that Mr. -> T. Since Mn. (P-I) 70, M(P-I)=0 =7 m is stationary. For any is In such that (Pn): >0.

Positivity: We must have M; 70 For some i. $M_{ij} = (M_{ij}^{n})_{ij} \times M_{ij} P_{ij}^{n} > 0.$

Uniqueness: Let S = inf{n}1: X=i? Then $u_i = (E[S]X_{o}=i])^{-1}$ Suppose XoNV and V is Stationary.

led Tre be hith visit to i. Then Tr - Tr-, IID

 $= 7 \quad \frac{T_n}{n} \rightarrow \mathbb{E} S = \frac{1}{n!} \quad a. s.$

If No = * { 15 ES n: X, = i} then

 $\frac{N_n}{N} \rightarrow M$: a.s.

So
$$E(N_n)/n \rightarrow M$$
:

But $E(N_n) = \sum_{k=1}^n |P(X_k = i)| = n V_i$
 $= 7 M_i = V_i$

Periodicity:
$$T(x) = \frac{1}{3} \quad uniform$$

$$P(X_{3n} = 1 \mid X_0 = 13 = 1)$$

 $P(X_{3n+1} = 2 \mid X_0 = 1) = 1$ So $X_n \stackrel{4}{\longleftrightarrow} \pi$.

A state x in a Markov (hain is aperiodic if G(O(S)=1) where $S=\{n\neq 1: |P(X_n=x|X_0=x_0^2)>0\}$.

Claim Closel under Addition: If $n,m \in S$ than $n+m \in S$ $P_{xx}^{n+m} = \sum_{y} P_{xy}^{n} P_{yx}^{y} \neq P_{xx}^{n} > 0$.

Fact: If G(D(A)=1) and A closed under addition then $|N \setminus A| < \infty$, i.e In such that $\forall n \mid n$, $n \in A$.

Detn A Markov chain is <u>ergodic</u> if it is irreducible and aperiodic.

Claim: If Xn is ergodiz then IN such that $\forall x, y, n \ge M$ then $P_{xy}^n \ne 0$.

Proof: Suppose Z is aperiodic so V m > M Pzz > O. Non for some k, l Pxz 70, Pzy >0.

Yn > h+l+M, Pxy 7 Pxz Pzz Pzy 70.

Theorem: If Xn is ergodic with stationary
distribution To then Xn => TO For any initial Xo.

Coupling: If X and Y are two R.V. a coupling (X', Y') is a joint distribution

defined on the same probability space such that $X \stackrel{d}{=} X'$, $Y \stackrel{d}{=} Y'$.

We often define a coupling with one of two goals

- a) X' = Y' stochastic domination
- b) minimize P(X'+Y') to compare X = Y.

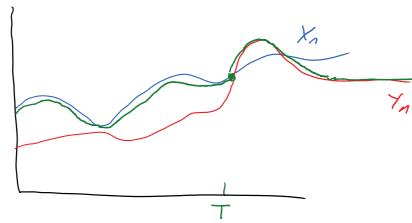
Example: Xn Bin (n,p), Yn Bin (m, p) for m>n.

Show that IP(X > X) < IP(Y > X)

IP[Y=x] = IP[Y'=x] = IP[X'=x]= IP[X=x].

$|||(Y \neq x)| = |||(Y \neq x)| \neq ||P(X \neq x)| = ||P(X \neq x)|.$

Let $X_0 = x_0$, we will prove that $X_n \xrightarrow{d} \pi$. Let Y_n be an independent copy of the chain, $Y_0 \sim \pi$. Let $T = \min \{ n \ge 0 : X_n = Y_n \}$ Let $Z_n = \begin{cases} X_n & T \le n \\ Y_n & T > n \end{cases}$



Then Z_n is a Markon chain with the same distribution as X_n and $IP(X_n = x) = IP(Z_n = x)$.

For some large M, min $P_{xy} = \alpha > 0$.

We can check every M steps to see if T has happened

Then IPC T > (R+1)M | T > RM]

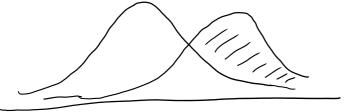
< max IP[Xe+11M = Ye+11M | Xem=x, Yen=9]

<1- min IP[Xexim = Yexim = x | Ken =x, Yen=y]

$$T_{oful} \quad variation \quad Distance$$

$$d_{\tau \nu} (M, V) = \max_{A} |M(A) - V(A)|$$

$$= \sum_{n} \frac{1}{2} |M_n - V_n|$$



Optimal coupling of Xnn, Ynr

IPCX' = Y'3 = dry (n, r)

Proof: For any compling

IP(X' = Y') > IP(X=A) - IP(Y=A)

= dr. (M,Y) for some A

Let
$$p = 1 - d_{TV}(M, r)$$
, $Z \sim Ber(p)$

$$\theta_1 = \frac{M \wedge r}{p}$$

$$\theta_2 = \frac{M - M \wedge r}{1 - p}$$
Probability measures.

$$\theta_3 = \frac{r - M \wedge r}{1 - p}$$
Let $W_1 \sim \theta_1$. Then set

$$X' = Z W_1 + (1 - Z) W_2$$

$$Y' = Z W_1 + (1 - Z) W_3$$

$$|P(X' = Y'] \geq |P(Z = 1] = 1 - d_{TV}(M, r)$$
So $|P(X' \neq Y'] \leq d_{TV}(M, r)$.

Need to check $X' \sim M$, $Y' \sim r$

$$Cuse 1 M \sim T \sim T$$

$$|P(X' = k) = |P(Z = 1) \cdot |P(W_1 = k) + |P(Z = 0) \cdot |P(W_2 = k)$$

$$= P \frac{M \approx N r_k}{p} + (1 - p) \frac{M_k - M_k \wedge N r_k}{1 - p} = M_k \wedge r$$