

(1)

Perfect Sampling - We will see two methods of generating perfect samples $X \sim \pi$ for given Markov chain P

- i) Strong stationary times
- ii) Coupling from the past

Eg - (Top-to-random shuffle) - Given n cards

- - take top card and insert it u.c.r in position $\{1, 2, \dots, n\}$
- ~~Random~~ Random walk on $S_n \equiv$ Symmetric group
- $\sigma_t \equiv$ Permutation at time t (ie, set of permutations of $[n]$)

• Let $\tau_{\text{top}} = 1 + \min \{t \geq 0 \mid \sigma_t(1) = \sigma_0(n)\}$
(i.e., first time bottom card comes to top plus 1)

• Propn - If at time t , k cards below ^{original} bottom card, then these are perfectly shuffled. Hence $\sigma_{\tau_{\text{top}}} \equiv$ Uniform in S_n

Pf - True at $t=0$. Suppose it's true at time t ; then in σ_{t+1} , either # of cards below ^{original} bottom card is same, or new card inserted at random position below original bottomed.

• Mixing time \equiv Coupon collector

• Stopping time T is a .

i) Stationary time if $P_x[X_T = y] = \pi(y) \forall x, y$

ii) Strong stationary time if $P_x[T=t, X_T=y] = P_x[T=t]\pi(y)$
(i.e., X_T has dist π and $T \perp\!\!\!\perp X_T$) or $P_x[X_T=y|T=t] = \pi(y)$

• Lemma - Let $X_t \sim MC(\Omega, P)$ with stationary dist π . If Z_{st} is a strong stopping time for P , then

$$d(t) = \max_{x \in \Omega} \|P_x^{(t)} - \pi\|_{TV} \leq \max_{x \in \Omega} P_x[Z_{st} > t]$$

Pf. For any $x \in \Omega$, $d_x(t) = \max_{A \subseteq \Omega} \|P_x^{(t)}(A) - \pi(A)\|_{TV}$. Now

$$\begin{aligned} P_x[X_t \in A] &= P_x[X_t \in A, Z_{st} > t] + \sum_{t' \leq t} P_x[X_t \in A, Z_{st} = t'] \\ &= P_x[X_t \in A | Z_{st} > t] P_x[Z_{st} > t] + \pi(A) \sum_{i \leq t} P_x[Z_{st} = i] \\ &= P_x[X_t \in A | Z_{st} > t] P_x[Z_{st} > t] + \pi(A) (1 - P_x[Z_{st} > t]) \end{aligned}$$

$$\Rightarrow P_x[X_t \in A] - \pi(A) = P_x[Z_{st} > t] \underbrace{(P_x[X_t \in A | Z_{st} > t] - \pi(A))}_{\in [-1, 1]}$$

This holds $\forall x \in \Omega, A \subseteq \Omega$

$$\Rightarrow d(t) \leq \max_{x \in \Omega} P_x[Z_{st} > t]$$

Note - Independence of I_{st} and $X_{I_{st}}$ is critical!

Eg - For RW on n -cycle, let $\vec{e}_t = \text{Ber}(1/n)$. If $I=0$, then $X_I \stackrel{\Delta}{=} 0$; else, run RW from 0 and let $\tau = \text{cover time}$.

Eg - For RW on hypercube $\{0,1\}^n$, consider following chain

- Pick I_t var from $\{0,1,2,\dots,n\}$
- Flip $X_t(I_t)$ w.p $1/2$

Claim - $\tau_{st} = \text{First time } \{I_1, I_2, \dots, I_t\} = \{1, 2, \dots, n\}$

- Thus $\tau_{st} \equiv \text{Coupon collector time}$
- $\mathbb{P}_z[\tau_{st} > n \ln n + cn] \leq e^{-c}$
- $\Rightarrow t_{\text{mix}}(\epsilon) \leq n(\ln n + \ln(1/\epsilon))$
- Identical to coupling time of random walks.

Eg - Transposition Shuffle (Pick L_t, R_t var with replacement & swap)

- Earlier we showed $t_{\text{mix}}(\epsilon) \leq \frac{6n^2}{\pi^2 \epsilon}$ via mixing arguments
- SS time (Breden) - Start with no marked card
- (Sec 8.2 in LPU) - (In round t , mark R_t if unmarked) and (L_t marked OR $L_t = R_t$)
- $\mathbb{E}[\tau_{st}] = \sum_{k=0}^{n-1} \frac{n^2}{(k+1)(n-k)} = 2n(\ln n + O(1))$, $\text{Var}(\tau_{st}) = O(n^2)$
- By Chebyshev - $t_{\text{mix}}(\epsilon) \leq n \ln n (2 + \sqrt{\epsilon})$

Random Mapping Representation of MC

(5)

- Given MC (Ω, P) , and an n -valued ^{independent} random variable Z ~~satisfying~~, a random mapping representation is a fn $f: \Omega \times \Lambda \rightarrow \Omega$ s.t.

$$P[f(x|Z) = y] = P(x, y)$$

- Eg - i) RW on n -cycle - $Z \sim \pm 1$ w.p. $1/2$
 $f(x|Z) = (x + Z) \bmod n$
- ii) Lazy RW on hypercube - $Z \sim \text{Unif}(\{1, 2, \dots, n\})$, $\text{Ber}(1/2)$
 $f(x|Z) \equiv \text{flip } x(I) \text{ if } F=1, \text{ else no change}$

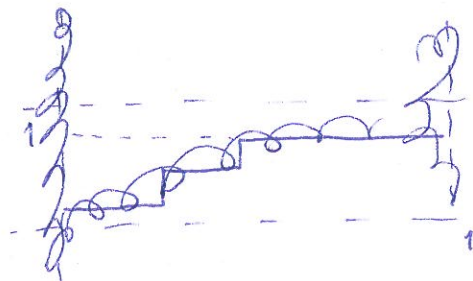
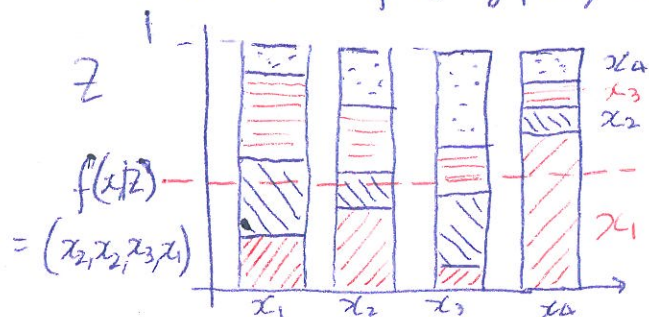
Lemma: Every P on finite Ω has a random mapping representation

Pf - - Choose arbitrary ordering $\Omega = \{x_1, x_2, \dots, x_n\}$

- Generate $Z \sim \text{Unif}([0, 1])$

- Define $F_{j,k} = \sum_{i=1}^k P(x_j, x_i)$

- Set $f(x_j|Z) = x_k$ when $F_{j,k-1} < Z \leq F_{j,k}$



• Some obs on random mapping representations

- Not unique

- $\{f(x|z)\}_x$ forms a grand coupling (i.e., $(f(x|z), f(y|z))$ is a coupling for A)

- If $|\Omega|=n$, then sufficient to consider a discrete r.v. Z with $\leq n^2$ values

(Corresponding to breakpoints $F_{j,k} \forall (j,k) \in \Omega^2$)

- If $(X_t, Y_t) \sim (f(X_{t-1}|Z_t), f(Y_{t-1}|Z_t))$, then

T_{couple} related to $\max_{z,y} \|f(x|z) - f(y|z)\|_{TV}$

(or more generally $\|f(f(\dots f(x|z_1)|z_2)\dots|z_t) - f(f(\dots f(y|z_1)|z_2)\dots|z_t)\|_r$)

- Henceforth write $f \circ f(x) = f(f(x|z_1)|z_2)$
composition of random fns

- Can obtain a SS time from a random function representation (with $Z \in \Lambda, |\Lambda| \leq n^2$) by

No →

sampling $z_1, z_2, \dots, z_{T_{\text{st}}}$ until we hit every value in Λ (i.e., coupon collector on Λ)

- Problem - Difficult to specify Λ in general

Eg - MC (Ω, P) s.t. $\alpha = \sum_y \min_x (P(x, y)) > 0$ (7)

(i.e., a sum over states of min transition prob)

- Related to the strong Doeblin condition

- Can write $P = \alpha I^T \Theta + (1-\alpha) Q$,

where $\Theta \equiv$ distn over Ω and $Q \equiv$ stochastic matrix

- Natural induced coupling - w.p. α , ^{set} $X_t = Y_t \sim \Theta$
- else, advance (X_t, Y_t) using Q

This is a random mapping representation!

- $P[\tau_{\text{couple}} > t] \leq (1-\alpha)^t$

$\Rightarrow t_{\text{mix}}(\epsilon) \leq \frac{1}{\epsilon} \ln(1-\alpha)$

- To generate a perfect sample from Π

- Sample $X_0 \sim \Theta$, $\tau \sim \text{Geom}(\alpha)$

- Return $X_{(\tau-1)}$

Pf - $\Pi P = \alpha \underbrace{\Pi I^T}_{\Theta} \Theta + (1-\alpha) \Pi Q = \Pi$

$\Rightarrow \Pi = (I - (1-\alpha)Q)^{-1} \alpha \Theta = \sum_{k=0}^{\infty} (1-\alpha)^k \alpha Q^k \Theta$
 $= E[\Theta^{\tau-1} \Theta]$