

# Lecture 4: Card Shuffling and Covertime

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CAMBRIDGE

## Shuffling and Strong Stationary Times

Covertime

$s - t$  Connectivity

2-Sat



## Card Shuffling

A *Permutation*  $\sigma$  of  $[n] = \{1, \dots, n\}$  is a bijection  $\sigma : [n] \rightarrow [n]$ .

Let  $\Sigma_n$  be the set of all  $n!$  permutations of  $[n]$ .

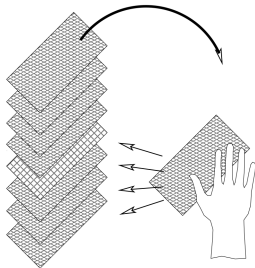
### Sampling from uniform.

Given an ordered set  $[n]$  we wish to sample a permutation of  $[n]$  uniformly.

### Top-to-Random (T-to-R) Shuffling

Given a deck of  $n$  cards take the top card and place it at random position in the deck.

- Markov chain on  $\Sigma_n$  with  $\pi$  uniform.



## Strong Stationary Time

A *Strong Stationary Time* for a Markov Chain  $(X_t)$  with stationary distribution  $\pi$  is a stopping time  $\tau$ , possibly depending on the starting state  $x$ , such that

$$\mathbf{P}_x[t = \tau, X_\tau = y] = \mathbf{P}_x[t = \tau] \pi_y.$$

- Thus  $X_\tau$  has distribution  $\pi$  and is independent of  $\tau$ .

— Mixing from Strong Stationary Times —

If  $\tau$  is a strong stationary time then for any  $x \in \mathcal{I}$ ,

$$\left\| P_x^t - \pi \right\|_{tv} \leq \mathbf{P}[\tau > t \mid X_0 = x].$$

**Proof:** For any  $A \subseteq \mathcal{I}$  the difference  $\mathbf{P}_x[X_t \in A] - \pi(A)$  is equal to

$$\begin{aligned} & \mathbf{P}_x[X_t \in A \mid \tau > t] \mathbf{P}_x[\tau > t] + \mathbf{P}_x[X_t \in A \mid \tau \leq t] (1 - \mathbf{P}_x[\tau > t]) - \pi(A) \\ &= (\mathbf{P}_x[X_t \in A \mid \tau > t] - \pi(A)) \mathbf{P}_x[\tau > t]. \end{aligned}$$

Then since  $-1 \leq \mathbf{P}_x[X_t \in A \mid \tau > t] - \pi(A) \leq 1$  we have

$$|\mathbf{P}_x[X_t \in A] - \pi(A)| = |\mathbf{P}_x[X_t \in A \mid \tau > t] - \pi(A)| \mathbf{P}_x[\tau > t] \leq \mathbf{P}_x[\tau > t],$$

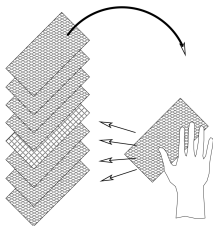
for any  $A \subset \mathcal{I}$ . We can take  $\sup_{A \subset \mathcal{I}}$  to complete the result.  $\square$



## Strong Stationary time for Top-to-Random Shuffling

- Let  $B$  be the card at the bottom of the deck at  $t = 0$ .
- Let  $\tau_{top}$  be one step after the first time when  $B$  is on top of the deck.

Strong Stationary time for T-to-R  
 $\tau_{top}$  is a Strong Stationary time for the T-to-R chain.



**Proof:** At any  $t \geq 0$  all arrangements of the cards under  $B$  are equally likely.

**Induction:** When  $t = 0$ , there are no cards under  $B$ . Suppose that the claim holds at time  $t \geq 0$  with  $k \geq 0$  cards under  $B$ . Two cases for time  $t + 1$ : either the top card is placed under  $B$ , or it is placed above  $B$ .

**Case 1** Hypothesis: all orderings of the  $k$  cards already under  $B$  are equiprobable. Top card is equally likely to be added to any of the  $k + 1$  possible locations under  $B$ , so each of the  $(k + 1)!$  arrangements is equiprobable.

**Case 2** New card goes above  $B$ , so cards under  $B$  remain in random order.  $\square$

Thus at time  $\tau_{top} - 1$   $B$  sits on the top of a uniform permutation of  $[n] \setminus \{B\}$ , then we place  $B$  in at random so  $\mathbf{P}[X_{\tau_{top}} \mid \tau_{top} = t] = 1/n!$ .  $\square$



## Top-to-Random Shuffle

Mixing of Top-to-Random Shuffle

Let  $\epsilon > 0$  then for the top to random shuffle,  $\tau(\epsilon) \leq n \ln n + O(n)$ .

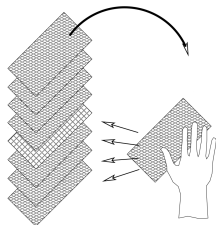
**Proof:** For  $1 \leq k \leq n - 1$  the time between the  $(k - 1)^{\text{th}}$  and  $k^{\text{th}}$  cards going under  $B$  is distributed  $\text{Geo}(k/n)$ . This means that  $\tau_{\text{top}}$  is distributed the same as the number of balls thrown until no bin is empty in “*Balls and Bins*”. Thus

$$\mathbf{P}[\tau > n \ln n + Cn] \leq \mathbf{P}[\exists \text{ empty bin after } n \ln n + Cn \text{ balls}] \stackrel{\text{Lecture 1}}{\leq} e^{-C}.$$

Taking  $C$  large enough such that  $e^{-C} \leq \epsilon$  yields the result.  $\square$

- Since the state space  $\Sigma_n$  has size  $n!$ , we have

$$t_{\text{mix}} \approx \ln(|\Sigma_n|).$$



## Realistic Shuffling - Riffle Shuffle

### Riffle Shuffle

- Split the deck into two piles  $L, R$  where  $L$  is the first  $\text{Bin}(n, 1/2)$  cards and  $R$  is the rest.
- Form a new pile iteratively by adding a card from  $L$  with probability  $\ell/(r + \ell)$ , where  $\ell, r$  sizes of  $L, R$  at that time, or otherwise from  $R$  with probability  $r/(\ell + r)$ .

Riffle is fast

For the Riffle shuffle  $t_{mix} \leq 2 \log_2(4n/3)$ .

- Same state space  $\Sigma_n$  as T-to-R however this time

$$t_{mix} \approx \ln \ln (|\Sigma_n|).$$

- May have heard "7 riffle shuffles is enough".

$t$	$\leq 4$	5	6	7	8	9
$\Delta(t)$	1.00	.92	.61	.33	.17.	.09



# Outline

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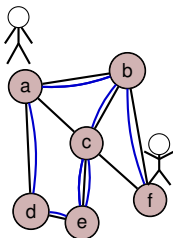
## Covetime

The *Cover time*  $t_{\text{cov}}(G)$  of a graph  $G = (V, E)$  is given by

$$t_{\text{cov}}(G) = \max_{v \in V} \mathbf{E}_v[\tau_{\text{cov}}] \quad \text{where} \quad \tau_{\text{cov}} := \inf \left\{ t : \cup_{i=0}^t \{X_i\} = V \right\}.$$

- Expected time for a walk to visit the whole graph from worst case start.

Example:



$$\begin{aligned} |V| &= 6 \\ \tau_{\text{cov}}(G) &= 9. \end{aligned}$$



Let  $P$  be the SRW on a connected graph  $G$ , then  $\pi_x = d(x) / 2|E|$ .

**Proof:** Note that  $\sum_{x \in V} \pi = 1$  and that for any  $x \in V$

$$(\pi P)_x = \sum_{y \in V} \pi_y P_{y,x} = \sum_{y \in d(x)} \frac{d(y)}{2|E|} \frac{1}{d(y)} = \frac{d(x)}{2|E|}. \quad \square$$

Let  $xy \in E(G)$  where  $G$  is any finite connected graph then  $h_{x,y} \leq 2|E|$ .

**Proof:** Since the SRW on any connected finite graph is irreducible we know

$$\mathbf{E}_y[\tau_y^+] = \frac{1}{\pi_y} = \frac{2|E|}{d(y)}.$$

By the Markov property we have

$$\frac{2|E|}{d(y)} = \mathbf{E}_y[\tau_y^+] = 1 + \sum_{z \sim y} \frac{h_{z,y}}{d(y)}.$$

It follows that  $\sum_{z \sim y} h_{z,y} \leq d(y) (\mathbf{E}_y[\tau_y^+] - 1)$  and thus

$$h_{x,y} \leq \sum_{z \sim y} h_{z,y} \leq d(y) \cdot \left( \frac{2|E|}{d(y)} - 1 \right) \leq 2|E|. \quad \square$$



For any connected graph  $t_{cov}(G) \leq 4n|E| \leq 2n^3$ .

**Proof:** Any connected graph has a spanning tree  $T$  with  $n - 1$  edges.

Choose any root  $v_0$  for  $T$  and fix a tour  $v_0, \dots, v_{2n-2}$  on  $T$  which visits every vertex and returns to the root.

The Covertime of  $G$  is at most the expected length of this tour (from worst case start vertex). Thus

$$t_{cov}(G) \leq \sum_{i=0}^{2n-3} h_{v_i, v_{i+1}} = \sum_{xy \in E(T)} (h_{xy} + h_{yx}) \leq 2 \sum_{xy \in E(T)} |E| \leq 4n|E|,$$

since for any  $xy \in E$  we have  $h_{x,y} \leq 2|E|$ . □

For any graph  $G$  we have

$$t_{cov}(G) \leq \left( \sum_{m=1}^{n-1} \frac{1}{m} \right) \cdot \max_{x,y \in V} h_{x,y} \approx (\ln n) \cdot \max_{x,y \in V} h_{x,y}.$$



## Random Walk on a path

The  $n$ -path  $P_n$  is the graph with  $V(P_n) = [n]$  and  $E(P_n) = \{ij : j = i + 1\}$ .

Proposition

For the SRW on  $P_n$  we have  $h_{k,n} = n^2 - k^2$ , for any  $0 \leq k \leq n$ .

**Proof:** Let  $f_k = h_{k,n}$  and observe that  $f_n = 0$ . By the Markov property

$$f_0 = 1 + f_1 \quad \text{and} \quad f_k = 1 + \frac{f_{k-1}}{2} + \frac{f_{k+1}}{2} \quad \text{for } 1 \leq k \leq n-1.$$

System of  $n$  independent equations in  $n$  unknowns so has a unique solution.

Thus it suffices to check that  $f_k = n^2 - k^2$  satisfies the above. Indeed

$$f_n = n^2 - n^2 = 0, \quad f_0 = 1 + f_1 = 1 + n^2 - 1^2 = n^2,$$

and for any  $1 \leq k \leq n-1$  we have,

$$f_k = 1 + \frac{n^2 - (k-1)^2}{2} + \frac{n^2 - (k+1)^2}{2} = n^2 - k^2. \quad \square$$

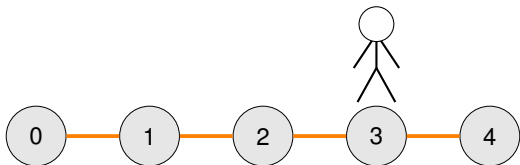


For the path  $P_n$  on  $n$  vertices we have

$$n^2 \leq t_{\text{cov}}(P_n) \leq 2n^2.$$

**Proof:** For the lower bound, take the random walk from the left hand end point (vertex 0). To cover the path we must at reach the righthand end point (vertex  $n$ ), this takes time  $n^2$  in expectation.

For the upper bound the max time to reach one end point from any start point is at most  $n^2$ . Now from this end point if we reach the opposite end point we must have visited every vertex, this takes an additional  $n^2$  expected time.  $\square$



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## $s - t$ Connectivity

### $s - t$ Connectivity Problem

- **Given:** Undirected graph  $G = (V, E)$  and  $s, t \in V$
- **Goal:** Determine if  $s$  is connected by a path to  $t$ .

### $s - t$ Connectivity Algorithm

- Start a random walk from  $s$ .
- If the walk hits  $t$  within  $4n^3$  steps, return **True**. O/W return **False**.

### Proposition

The  $s - t$  Connectivity Algorithm runs in time  $4n^3$  and returns the correct answer w.p. at least  $1/2$  and never returns **True** incorrectly.

**Proof:** By Markov inequality if there is a path to  $t$  we will find it w.p.  $\geq 1/2$ .  $\square$

- Running this  $T$  times gives the correct answer with probability  $\geq 1 - 1/2^T$ .
- Only uses logspace.



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## SAT Problems

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A *Satisfiability (SAT)* formula is a logical expression that's the conjunction (AND) of a set of *Clauses*, where a clause is the disjunction (OR) of *Literals*.

A *Solution* to a SAT formula is an assignment of the variables to the values True and False so that all the clauses are satisfied.

Example:

$$\text{SAT: } (x_1 \vee \overline{x_2} \vee \overline{x_3}) \wedge (\overline{x_1} \vee \overline{x_3}) \wedge (x_1 \vee x_2 \vee x_4) \wedge (x_4 \vee \overline{x_3}) \wedge (x_4 \vee \overline{x_1})$$

**Solution:**  $x_1 = \text{True}$ ,  $x_2 = \text{False}$ ,  $x_3 = \text{False}$  and  $x_4 = \text{True}$ .

- If each clause has  $k$  literals we call the problem *k-SAT*.
- In general, determining if a SAT formula has a solution is NP-hard
- In practice solvers are fast and used to great effect
- A huge amount of problems can be posed as a SAT:
  - Model Checking and hardware/software verification
  - Design of experiments
  - Classical planning
  - ...



## RAND 2-SAT Algorithm

- (1) Start with an arbitrary truth assignment.
- (2) Repeat up to  $2n^2$  times, terminating if all clauses are satisfied:
  - (a) Choose an arbitrary clause that is not satisfied
  - (b) Choose one of its literals UAR and switch the variables value.
- (3) If a valid solution is found return it. O/W return **unsatisfiable**

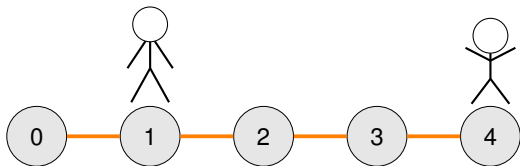
- Call each loop of (2) a *Step*. Let  $A_i$  be the variable assignment at step  $i$ .
- Let  $S$  be any solution and  $X_i = |\text{variable values shared by } A_i \text{ and } S|$ .

## Example 1 : Solution Found

$$(x_1 \vee \bar{x}_2) \wedge (\bar{x}_1 \vee \bar{x}_3) \wedge (x_1 \vee x_2) \wedge (x_4 \vee \bar{x}_3) \wedge (x_4 \vee \bar{x}_1)$$

T   F   F   T   T   T   T   T   T   F

$$S = (T, T, F, T).$$



$t$	$x_1$	$x_2$	$x_3$	$x_4$
0	F	F	F	F
1	F	T	F	F
2	T	T	F	F
3	T	T	F	T



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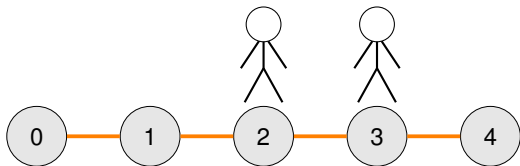
- Call each loop of (2) a *Step*. Let  $A_i$  be the variable assignment at step  $i$ .
- Let  $S$  be any solution and  $X_i = |\text{variable values shared by } A_i \text{ and } S|$ .

## Example 2 : Solution Found

$$(x_1 \vee \bar{x}_2) \wedge (\bar{x}_1 \vee \bar{x}_3) \wedge (x_1 \vee x_2) \wedge (x_4 \vee x_3) \wedge (x_4 \vee \bar{x}_1)$$

T   F   F   T   T   T   T   F   T   F

$$S = (T, F, F, T).$$



$t$	$x_1$	$x_2$	$x_3$	$x_4$
0	F	F	F	F
1	F	F	F	T
2	F	T	F	T
3	T	T	F	T



## 2-SAT and the SRW on the path

Expected iterations of (2) in RAND 2-SAT

If a valid solution  $S$  exists then the expected number of iterations of loop (2) before RAND 2-SAT outputs a valid solution is at most  $n^2$ .

**Proof:** Fix any solution  $S$ , then for any  $i \geq 0$  and  $1 \leq k \leq n - 1$ ,

- (i)  $\mathbf{P}[X_{i+1} = 1 \mid X_i = 0] = 1$
- (ii)  $\mathbf{P}[X_{i+1} = k + 1 \mid X_i = k] \geq 1/2$
- (iii)  $\mathbf{P}[X_{i+1} = k - 1 \mid X_i = k] \leq 1/2$ .

Notice that if  $X_i = n$  then  $A_i = S$  thus solution found (may find another first).

Assume (pessimistically) that  $X_0 = 0$  (we get non of our initial guesses right).

The stochastic process  $X_i$  is complicated to describe in full however by

(i) – (iii) we can couple it with  $Y_i$ - the SRW on the  $n$ -path from 0. This gives

$$\mathbf{E}[\text{time to find } S] \leq \mathbf{E}_0[\inf\{t : X_t = n\}] \leq \mathbf{E}_0[\inf\{t : Y_t = n\}] = h_{0,n} = n^2. \quad \square$$

Proposition

Provided a solution exists the **RAND 2-SAT Algorithm** will return a valid solution in time  $2n^2$  with probability at least  $1/2$ .

