

## • One-time Pad :

→ encrypting a message  $m \in \{0,1\}^n$  using one-time pad  $k \in \{0,1\}^n$  picked at random

$$m \xrightarrow{\text{Enc}} m \oplus k = e \xrightarrow{\text{Dec}} m = e \oplus k$$

provably secure if  $k$  is used only once.

→ inefficient

want:  $g: \{0,1\}^k \rightarrow \{0,1\}^n$

pick  $s \in \{0,1\}^k$  at random

$$m \xrightarrow{\text{Enc}} m \oplus g(s) \xrightarrow{\text{Dec}} m = e \oplus g(s)$$

$g$ ... "pseudorandom generator"

Q: when is it secure?

•  $f: \{0,1\}^n \rightarrow \{0,1\}$  ... test

$g: \{0,1\}^k \rightarrow \{0,1\}^n$  ... PRG

Def:  $g$  fools  $f$  if

$$\epsilon = \frac{1}{n} \quad \epsilon = \frac{1}{n^{\omega(1)}}$$

$$\left| \Pr_{z \in \{0,1\}^n} [f(z) = 1] - \Pr_{s \in \{0,1\}^k} [f(g(s)) = 1] \right| \leq \epsilon$$

- Want  $g$  which fools all  $F$  from some class of fun's,  
e.g. all polynomially computable  $F$ :

- $F$  computable probabilistically in time  $m(n) = n^{O(1)}$

$$l = O(\lg n)$$

$$g: \{0,1\}^{l(n)} \rightarrow \{0,1\}^{m(n)}$$

$$\text{s.t. } \forall n \forall x \in \{0,1\}^n \left| \Pr_{z \in \{0,1\}^{m(n)}} [F(x,z) = 1] - \Pr_{s \in \{0,1\}^{l(n)}} [F(x,g(s)) = 1] \right| \leq \frac{1}{4}$$

→ derandomization, combinatorial constructions, ...

→ cryptography

- PRG is good  $\approx$  one cannot predict next bit of its output.

$$g: \{0,1\}^l \rightarrow \{0,1\}^{l+1}$$

- PRG's fooling BPP computers exist

$\Leftrightarrow \exists f \in E$  which requires ch's of size  $2^{\delta n}$   
for some  $\delta > 0$ .

Ex: Randomized alg. for stconn (reachability)

Input: undirected graph  $G$ ,  $s, t \in V(G)$

Output: Is there a path from  $s$  to  $t$ ?

Output: Is there a path from  $s$  to  $t$  :

Alg: for  $16n^2$  steps repeat:  
if  $s = t$  then output YES.  
 $s \leftarrow$  random neighbor of  $s$   
output NO.

If  $s \rightsquigarrow t$  then  $\Pr[\text{Alg}(G, s, t) = \text{YES}] \geq \frac{2}{3}$

If  $s \not\rightsquigarrow t$  then  $\Pr[\text{Alg}(G, s, t) = \text{NO}] = 1.$

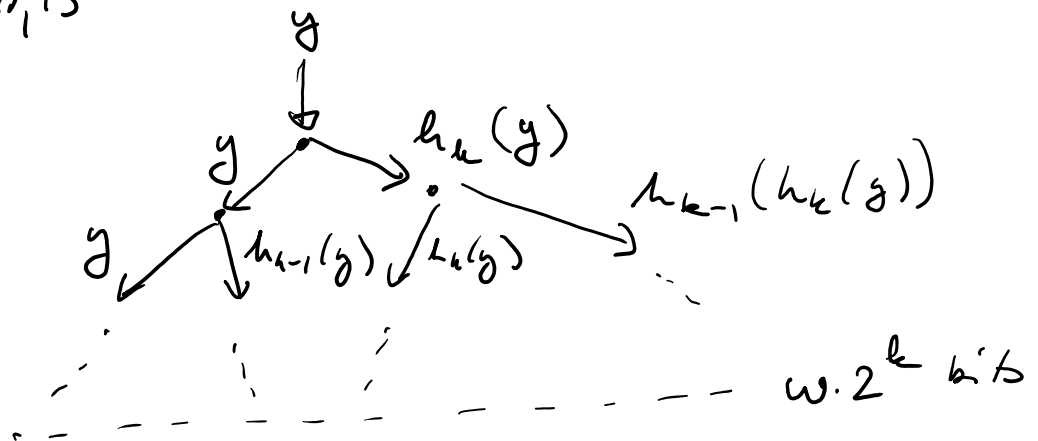
Algorithm runs in randomized logarithmic space = RL.

Nisan's PRG for RL :

$h_1, h_2, h_3, \dots, h_k : \{0, 1\}^w \rightarrow \{0, 1\}^w$

pair-wise int.  
hash fun's.

$y \in \{0, 1\}^w$



$$G_0(y) = y$$

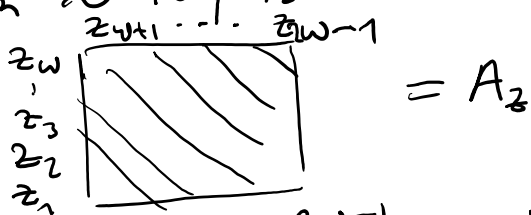
$$G_k(y, h_1, h_2, \dots, h_k) = G_{k-1}(y, h_1, h_2, \dots, h_{k-1}) \circ G_{k-1}(h_k(y), h_{k-1})$$

$h: \{0,1\}^w \rightarrow \{0,1\}^w \dots$  pair-wise ind.

e.g. 1)  $h_{A,b}(x) = Ax + b$

$w \times w$   
 $A \in \{0,1\}^{w \times w}$   
 $b \in \{0,1\}^w$   
 random

2) convolution  $\approx$  Toeplitz matrix



pick random  $z \in \{0,1\}^{2w-1}$   $b \in \{0,1\}^w$

$h_{z,b}(x) = A_z x + b$



$(x, z_{i \dots i+w-1}) + b_i$

$w = O(\lg n)$

$k = O(\lg n)$

$A, B \subseteq \{0,1\}^w$

$\alpha = \frac{|A|}{2^w}$

$\beta = \frac{|B|}{2^w}$

$C_h = \{x \in \{0,1\}^w; x \in A \& h(x) \in B\}$

$C_h \approx A \times B$

$\gamma_h = \frac{|C_h|}{2^w}$

want:  $\gamma_h \approx \alpha \cdot \beta$

claim: 1)  $\mathbb{E}_h[\gamma_h] = \alpha \cdot \beta$

(\*\*\*\*) 2)  $\Pr[|\gamma_h - \alpha \cdot \beta| > \epsilon] \leq \frac{\alpha \beta}{\epsilon^2 \cdot 2^w} \leq \frac{1}{\epsilon^2 \cdot 2^w}$

$$(\text{***}) \quad 2) \quad \Pr_L [ |g_h - \alpha \cdot \beta| > \varepsilon ] = \frac{1}{\varepsilon^2 \cdot 2^w} = \varepsilon^2 \cdot 2^w$$

Pf: 1)  $\forall x \in \{0,1\}^w$  r.v.  $Y_x = \begin{cases} 1 & L(x) \in B \\ 0 & \text{else} \end{cases}$

$$|c_h| = \sum_{x \in A} Y_x$$

$$\begin{aligned} \mathbb{E}_h [g_h] &= \mathbb{E}_h \left[ \frac{|c_h|}{2^w} \right] = \frac{1}{2^w} \sum_{x \in A} \mathbb{E}[Y_x] \\ &= \frac{|A|}{2^w} \cdot \beta = \alpha \cdot \beta \end{aligned}$$

$$\begin{aligned} 2) \quad \mathbb{E} [ (g_h - \alpha \cdot \beta)^2 ] &= \mathbb{E} \left[ \left( \frac{1}{2^w} \sum_{x \in A} Y_x - \alpha \cdot \beta \right)^2 \right] \\ &= \mathbb{E} \left[ \frac{1}{(2^w)^2} \left( \sum_{x \in A} Y_x \right)^2 - \frac{2}{2^w} \alpha \cdot \beta \cdot \sum_{x \in A} Y_x + (\alpha \beta)^2 \right] \\ &= \frac{1}{2^{2w}} \sum_{x,y \in A} \mathbb{E}[Y_x \cdot Y_y] - 2 \cdot \alpha \cdot \beta \cdot \underbrace{\mathbb{E}[g_h]}_{\alpha \cdot \beta} + (\alpha \beta)^2 \\ &= \frac{1}{2^{2w}} \sum_{x,y \in A} \mathbb{E}[Y_x \cdot Y_y] - (\alpha \beta)^2 \\ &= \frac{1}{2^{2w}} \left( \sum_{\substack{x,y \in A \\ x \neq y}} \mathbb{E}[Y_x \cdot Y_y] + \sum_{x \in A} \mathbb{E}[Y_x] \right) - (\alpha \beta)^2 \\ &= \frac{1}{2^{2w}} \left( |A| \cdot |A| - |A| \right) \cdot \beta^2 + \frac{\alpha \beta}{2^w} - (\alpha \beta)^2 \\ &\leq \frac{\alpha \beta}{2^w} \end{aligned}$$

$$\Pr_L [ |g_h - \alpha \beta| \geq \varepsilon ] = \Pr_L [ (g_h - \alpha \beta)^2 \geq \varepsilon^2 ]$$

$$P_h^n [ |y_h - \alpha \beta| \geq \epsilon ] = P_h^n [ (y_h - \alpha \beta) = \epsilon ]$$

$$\leq \frac{\alpha \beta}{2^w} \cdot \frac{1}{\epsilon^2}$$

marker

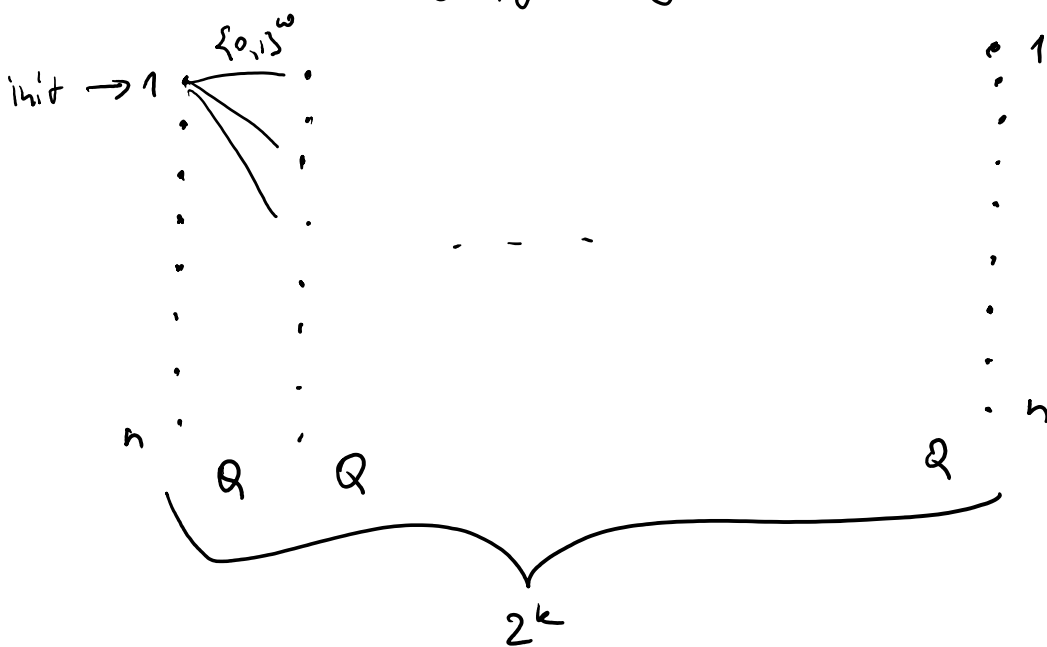
$$P_h [ X \geq k E(X) ] \leq \frac{1}{k}$$

$$k \cdot E(X) = \epsilon^2 \rightarrow \frac{1}{k} = \frac{E(X)}{\epsilon^2}$$

□

• We say  $h$  is  $(\epsilon, \alpha, \beta)$ -independent if  $|y_h - \alpha \beta| < \epsilon$ .

• consider "branching pgm" / graph:



$Q$  represent the transitions of a log-space machine between different configurations (configs  $\{1, \dots, n\}$ )

want:

$$P_{z \in \{0,1\}^{w \cdot 2^k}} [ Q^{2^k}(z) = 1 ] \approx P_{\substack{y \in \{0,1\}^w \\ h_1, h_2, \dots, h_k}} [ Q^{2^k}(G_k(y, h_1, h_2, \dots, h_k)) = 1 ]$$

For distribution  $D$  on  $\{0,1\}^{w2^k}$

$Q(D)$  is the  $n \times n$  matrix of transition probabilities, i.e.

$$Q(D)_{i,j} = \Pr[D \text{ takes us from } i \text{ to } j]$$

For matrix  $M$  def.  $\|M\| = \max_i \sum_j |M_{i,j}|$   
 ... the  $l_1$ -norm of largest row

claim:  $\|M+N\| \leq \|M\| + \|N\|$

$$\|M \cdot N\| \leq \|M\| \cdot \|N\|$$

Pf: (exc).  $\square$

$(h_1, \dots, h_k)$  are  $(\epsilon, Q)$ -good if

$$\|Q(G_{\epsilon}(U_w, h_1, \dots, h_k)) - Q(U_{\{0,1\}^{w2^k}})\| \leq \epsilon$$

$\uparrow$  uniform on  $\{0,1\}^w$        $\uparrow$  uniform on  $\{0,1\}^{w2^k}$

Thm:  $\forall Q: [n] \times \{0,1\}^w \rightarrow [n]$

$$\Pr_{h_1, h_2, \dots, h_k} \left[ (h_1, h_2, \dots, h_k) \text{ is not } ((2^k - 1)\epsilon, Q)\text{-good} \right] \leq \frac{n^7 \cdot k}{\epsilon^2 \cdot 2^w}$$

Pf: induction on  $k$ :

$$\left[ \begin{array}{l} k=0 \quad \dots \text{trivial} \\ k=1 \quad \dots \text{by claim (****)} \end{array} \leq \frac{n^2}{\epsilon^2 \cdot 2^w} \right]$$

$$\underline{k > 0}$$

$h_1, h_2, \dots, h_k$  picked at random

def.  $B_{i,j}^{h_1, \dots, h_{k-1}} = \{x \in \{0,1\}^w, G_{k-1}(x, h_1, h_2, \dots, h_{k-1}) \text{ takes } i \xrightarrow{Q} j\}$

Consider event:

1)  $(h_1, h_2, \dots, h_{k-1})$  is  $((2^{k-1}-1)\epsilon, Q)$ -good

2)  $\forall i, l, j$ :  $h_k$  is  $(\frac{\epsilon}{n^2}, B_{i,l}^{h_1, \dots, h_{k-1}}, B_{l,j}^{h_1, \dots, h_{k-1}})$ -independent

Claim: 1) & 2)  $\Rightarrow (h_1, \dots, h_k)$  is  $((2^k-1)\epsilon, Q)$ -good

$$\|Q(G_k(U_w, h_1, \dots, h_k)) - Q(U_{w,2^k})\| \leq \epsilon$$

$$\begin{aligned} &\leq \|Q(G_k(U_w, h_1, \dots, h_k)) - (Q(G_{k-1}(U_w, h_1, \dots, h_{k-1})))\|^2 \quad (*) \\ &+ \| (Q(G_{k-1}(U_w, h_1, \dots, h_{k-1}))) - Q(U_{w,2^k}) \|^2 \quad (**) \end{aligned}$$

$\leq (2^{k-2})\epsilon$

triangle  
ineq.

to bound (\*) consider some fixed  $i$  & arbitrary  $j$ :

$$Q(G_k(U_w, h_1, \dots, h_k))_{i,j} = \sum_x \Pr_x [x \in B_{i,l}^{h_1, \dots, h_{k-1}} \& h_k(x) \in B_{l,j}^{h_1, \dots, h_{k-1}}]$$

$$\left( Q(G_{k-1}(U_w, h_1, \dots, h_{k-1})) \right)_{i,j}^2 = \sum_x \Pr_x [x \in B_{i,l}^{h_1, \dots, h_{k-1}}] \cdot \Pr_y [y \in B_{l,j}^{h_1, \dots, h_{k-1}}]$$

by 2)  $\left| \Pr_x [x \in B_{i,l}^{h_1, \dots, h_{k-1}} \& h_k(x) \in B_{l,j}^{h_1, \dots, h_{k-1}}] - \Pr_x [x \in B_{i,l}^{h_1, \dots, h_{k-1}}] \cdot \Pr_y [y \in B_{l,j}^{h_1, \dots, h_{k-1}}] \right| \leq \frac{\epsilon}{n^2}$

$$\left| \Pr_x [x \in B_{i,l}^{h_1, \dots, h_{k-1}} \& h_k(x) \in B_{l,j}^{h_1, \dots, h_{k-1}}] - \Pr_x [x \in B_{i,l}^{h_1, \dots, h_{k-1}}] \cdot \Pr_y [y \in B_{l,j}^{h_1, \dots, h_{k-1}}] \right| \leq \frac{\epsilon}{n^2}$$

$$\Rightarrow \left| \Pr_x [x \in B_{i,l}^{h_1, \dots, h_{k-1}} \& h_k(x) \in B_{l,j}^{h_1, \dots, h_{k-1}}] - \Pr_x [x \in B_{i,l}^{h_1, \dots, h_{k-1}}] \cdot \Pr_y [y \in B_{l,j}^{h_1, \dots, h_{k-1}}] \right| \leq \frac{\epsilon}{n^2}$$



$$\Rightarrow \left| Q(G_k(u_w, h_1, \dots, h_k))_{i,j} - \left[ Q(G_{k-1}(u_w, h_1, \dots, h_{k-1})) \right]_{i,j}^2 \right| \leq \frac{\epsilon}{n}$$

$$\Rightarrow (*) \leq \epsilon$$

to bound (\*\*)

$$Q(u_{w \cdot 2^k}) = \left( Q(u_{w \cdot 2^{k-1}}) \right)^2$$

$$\text{Hence, } (**) = \left\| \underbrace{Q(G_{k-1}(u_w, h_1, \dots, h_{k-1}))}_M - \underbrace{\left( Q(u_{w \cdot 2^{k-1}}) \right)^2}_N \right\|$$

$$= \|M^2 - N^2\| = \|M^2 - MN + MN - N^2\|$$

$$\leq \|M\| \cdot \|M - N\| + \|M - N\| \cdot \|N\|$$

$$\text{by 1) } \|M - N\| \leq (2^{k-1} - 1) \epsilon$$

$$\|M\|, \|N\| \leq 1 \quad \text{as } M \text{ \& } N \text{ are matrices of probabilities}$$

$$\Rightarrow (**) \leq 2 \cdot (2^{k-1} - 1) \epsilon = (2^k - 2) \epsilon.$$

Hence, it remains to bound the probability of 1) & 2) not happening. By i.H. prob of 1) is  $\leq \frac{n^7 \cdot (k-1)}{\epsilon^2 \cdot 2^w}$ .

For fixed  $h_1, \dots, h_{k-1}$ ,  
By "Mixty" Claim:

$$\Pr[2] \leq n^3 \cdot \frac{n^4}{\epsilon^2 \cdot 2^w} = \frac{n^7}{\epsilon^2 \cdot 2^w}$$

The claim follows. □

Corollary:  $RL \subseteq DTISP(n^{O(1)}, \log^2 n)$

Pf: Pick the hash fcn's  $h_1, h_2, \dots, h_k$  one by one while always checking that the next one is satisfied  $\&$  from the above proof.

Given  $h_1, \dots, h_k$  we can calculate  $\forall i, j$   
 $O(G_e(u_{ij}, h_1, \dots, h_k))_{i,j}$   
in  $O(\log n)$  space so checking the condition<sup>2)</sup> can be done in  $\log$ -space.  $\square$

Thm: (Saks-Zhou '96):  $RL \subseteq DSPACE(\log^{3/2} n)$

idea: ... pick  $\sqrt{\log n}$  hash fcn & reuse them  $\sqrt{\log n}$  times.

Application:

(D. Sivakumar '02)

Discrepancy: Given  $A_1, \dots, A_m \subseteq [n]$

Find  $D \subseteq [n]$  s.t.

$$\forall i: |(A_i \cap D) - (A_i \setminus D)| \leq \sqrt{|A_i| \cdot \log m}.$$

$\rightarrow$  random bits will give a good set  $D$  w.h.p.

$\rightarrow$  use Nisan's PRG to generate  $D$ .

2. It takes  $O(m \lg n)$  space to check the condition for all  $i=1, \dots, m$  if  $D$  is given "on-the-fly".

Solution: Design  $m$   $O(\lg n)$ -space tests for each  $i$  individually. Nisan's PRG will work for each if their w.p.  $\geq 1 - \frac{1}{m^2}$  so it will work for all of them at once w.p.  $\geq 1 - \frac{1}{m}$ .

→ find the appropriate hash functions deterministically like in  $RL \subseteq DTISP(n^{O(1)}, \lg^2 n)$ . □

→ Johnson-Lindenstrauss lemma, ... using similar technique.

streaming, ...