

Reducing Randomness Via Random Walks  
on special graphs

# Reducing Randomness

For decision problem  $L$ ,

Let  $A$  be algorithm st. 1)  $\forall x \in L \quad \Pr[A(x)=1] \geq 99/100$  almost always correct  
 2)  $\forall x \notin L \quad \Pr[A(x)=0] = 1$  always correct

To get error  $< 2^{-k}$ :

## Method:

## # random bits used

- |   |            |
|---|------------|
| 1) run $k$ times & output "x ∈ L" if ever see "x ∈ L" else output "x ∉ L" | $O(kr)$    |
| 2) use p.i. random bits   | $O(k+r)$   |
| 3) today: use random walk on graph to choose random bits                  | $r + O(k)$ |

## Plan:

- associate all (random) strings in  $\{0,1\}^n$  with nodes of a graph  $G$
- problem of picking a random string is now equivalent to problem of picking a random node ← easier?
- picking several random strings  $\Rightarrow$  picking several nodes ← easier?
- picking several strings, one of which is "good"  $\Rightarrow$  picking several nodes, one of which is "good" ← "easier"!

The graph  $G$  :  $\swarrow$  we get to pick  $G$ !

- constant degree  $d$ -regular, connected, nonbipartite
- transition matrix  $P$  for r.w. on  $G$  has  $|\lambda_2| \leq \frac{1}{10}$   
stationary dist  $\pi$  uniform since  $d$ -reg
- # nodes =  $2^r \sim r$  random bits

### The Algorithm:

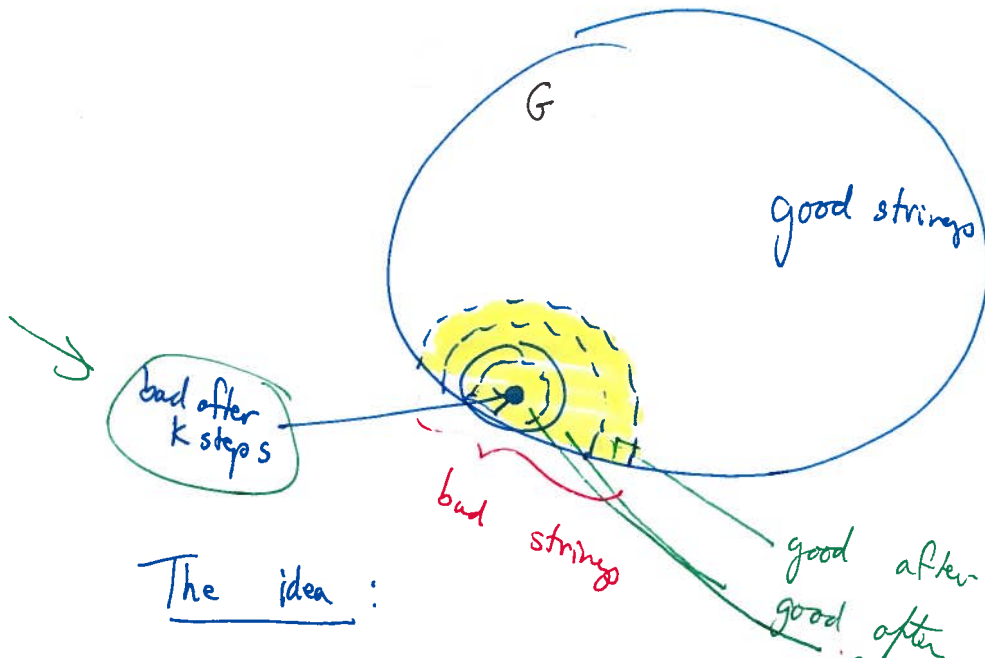
- pick random start node  $w \in \{0,1\}^r$  r bits
- Repeat  $K$  times:
  - $w \leftarrow$  random neighbor of  $w$   $O(1)$  bits  $\times K$
  - run  $A(x)$  with  $w$  as random bits
  - if  $A$  outputs " $x \in L$ " then output " $x \in L$ " + halt
  - else continue

• Output " $x \notin L$ "

total :  $r + O(K)$   
random bits

Claim: error of new algorithm  $\leq (\frac{1}{5})^K$  for  $x \in L$   
(still 0-error for  $x \notin L$ )

very few of these

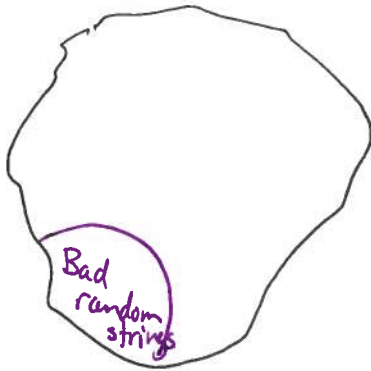


The idea :

very unlikely to pick a start location that is bad after  $k$  steps.

Behavior:

idea:



bad case - walk only on "bad" random strings + never get out to "good" random strings

why would this not work on arbitrary  $G$ ?  
e.g.  $G = \text{line}$

if  $X \notin L$ : algorithm never errs (there are no bad strings)

if  $X \in L$ :

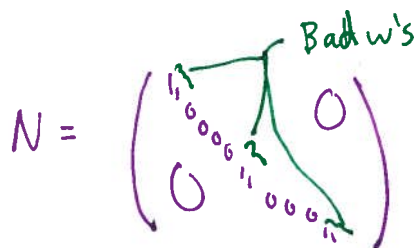
most random bits say  $X \in L$ :  $\geq \frac{99}{100} \cdot 2^r$

define  $B \leftarrow \{w \mid A(x) \text{ with random bits } w \text{ is incorrect ie. says } X \notin L\}$   
"Bad w's"

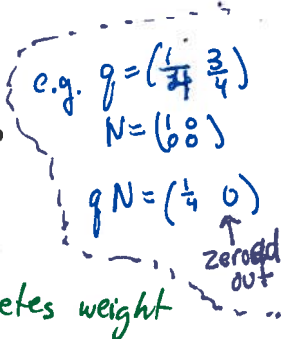
$$|B| \leq \frac{2^r}{100}$$

Want linear algebraic way of describing walks that stay in badset:  
define  $N$  diagonal matrix such that

$$N_w = \begin{cases} 1 & \text{if } w \in B \leftarrow \text{incorrect} \\ 0 & \text{o.w.} \leftarrow \text{correct} \end{cases}$$



$q$  any probability distribution,  $q \cdot N$  is ??



$\|qN\|_1 = \Pr_{w \in q} [w \text{ is bad}]$  i.e.  $qN$  deletes weight that finds a witness to  $x \in L$

can compose:

$\|q \cdot PN\|_1 = \Pr_{w \in q} [\text{start at } q, \text{ take a step + land on "bad"}]$

⋮

$\|q \cdot (PN)^k\|_1 = \Pr_{w \in q} [\text{start at } q, \text{ take } k \text{ steps + each is "bad"}]$

ignores whether start node is bad, this just hurts us so it is ok to ignore

Lemma  $\forall \pi \quad \|\pi PN\|_2 \leq \frac{1}{5} \|\pi\|_2$

First: How do we use the lemma?

If always see bad w's, then answer incorrect

$\Rightarrow \Pr[\text{incorrect}] \leq \|p_0 \cdot (PN)^k\|_1$

$\leq \sqrt{2^r} \|p_0 (PN)^k\|_2$

since  $\|p\|_1 \leq \sqrt{\text{domain size}} \cdot \|p\|_2$

$\leq \sqrt{2^r} \cdot \|p_0\|_2 \left(\frac{1}{5}\right)^k$

apply lemma  $k$  times

$\frac{1}{\sqrt{2^r}}$  since start at uniform +  $L_2$  norm of uniform =  $\sqrt{\frac{1}{2^r}}$  =  $\frac{1}{\sqrt{2^r}}$

$= \left(\frac{1}{5}\right)^k$

Proof of lemma

let  $V_1, \dots, V_{2^r}$  be e-vecs of  $P$ , +  $V_1$  is st.  $\|V_1\|_2 = 1$   
 note,  $V_1 = (\frac{1}{\sqrt{2^r}}, \dots, \frac{1}{\sqrt{2^r}})$

then  $\Pi = \sum_{i=1}^{2^r} \alpha_i V_i$

Note: 1)  $\|\Pi\|_2 = \sqrt{\sum \alpha_i^2}$  (from before)

2)  $\forall w \quad \|wN\|_2 = \sqrt{\sum_{i \in B} w_i^2} \leq \sqrt{\sum_i w_i^2} = \|w\|_2$

So:

$$\|\Pi P N\|_2 = \left\| \sum_{i=1}^{2^r} \alpha_i V_i P N \right\|_2$$

$$= \left\| \sum_{i=1}^{2^r} \alpha_i \lambda_i V_i N \right\|_2$$

$$\leq \underbrace{\|\alpha_1 \lambda_1 V_1 N\|_2}_A + \underbrace{\left\| \sum_{i=2}^{2^r} \alpha_i \lambda_i V_i N \right\|_2}_B$$

Cauchy-Schwarz

bounding:  $\|\alpha_1 \lambda_1 V_1 N\|_2 = \|\alpha_1 V_1 N\|_2$  since  $\lambda_1 = 1$

$$= |\alpha_1| \sqrt{\sum_{i \in B} \left(\frac{1}{\sqrt{2^r}}\right)^2}$$

since  $V_1 = (\frac{1}{\sqrt{2^r}}, \dots, \frac{1}{\sqrt{2^r}})$

$$= |\alpha_1| \sqrt{\frac{|B|}{2^r}}$$

$$\leq \frac{|\alpha_1|}{10}$$

since  $\frac{|B|}{2^r} \leq \frac{1}{100}$

$$\leq \frac{\|\Pi\|_2}{10}$$

since  $\|\Pi\|_2 = \sqrt{\sum \alpha_i^2}$

use that uniform is unlikely to be on bad string

Bounding (B) :  $\left\| \sum_{i=2}^{2^r} \alpha_i \lambda_i v_i N \right\|_2 \leq \left\| \sum_{i=2}^{2^r} \alpha_i \lambda_i v_i \right\|_2$  from note

use "mixing"

$$= \sqrt{\sum (\alpha_i \lambda_i)^2}$$

$$\leq \sqrt{\sum \alpha_i^2 \left(\frac{1}{10}\right)^2}$$

$$\lambda_i \leq 1/10$$

$$\leq \frac{1}{10} \|\pi\|_2$$

So:  $\|\pi P N\|_2 \leq \frac{\|\pi\|_2}{5}$  ■