

## Lecture 13

- finish saving random bits via random walks
- Linearity testing intro

From last time:

## Linear Algebra Review

def  $v$  is an eigenvector of  $A$  with corresponding eigenvalue  $\lambda$  iff

$$vA = \lambda v$$

def  $\ell_2$ -norm of  $v = (v_1 \dots v_n) = \sqrt{\sum_{i=1}^n v_i^2} = v \cdot v$

def  $v^{(1)} \dots v^{(m)}$  orthonormal if

$$\underbrace{v^{(i)} \cdot v^{(j)}}_{\text{inner product}} = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$$
$$= \sum_l v_l^{(i)} \cdot v_l^{(j)}$$

Thm Transition matrix  $P$  real + symmetric

$\Rightarrow \exists$  e-vecs  $v^{(1)} \dots v^{(n)}$

forming orthonormal basis with corresponding e-values  $| = \lambda_1 \geq |\lambda_2| \geq \dots \geq |\lambda_n|$

$$+ v^{(1)} = \frac{1}{\sqrt{n}} (1 \dots 1)$$

$\curvearrowleft$  chosen so that  $\|v^{(1)}\|_2 = 1$

$\Rightarrow$  any vector  $w$  is expressible as linear

Combination of  $v^{(i)}$ 's

$$w = \sum \alpha_i \cdot v^{(i)}$$

+  $L_2$ -norm of  $w$  is  $\sqrt{\sum \alpha_i^2}$

From last time:

## Useful Facts:

Assume  $P$  has all positive entries & evecs  $v^{(1)} \dots v^{(n)}$  with

### Facts

corresponding e-vals  $\lambda_1 \dots \lambda_n$

(1)  $\alpha P$  has e-vecs  $v^{(1)} \dots v^{(n)}$  with corresponding evals  $\alpha \lambda_1 \dots \alpha \lambda_n$

(2)  $P + I$  " " " " "  $\lambda_1 + 1, \dots, \lambda_n + 1$

(3)  $P^k$  " " " " "  $\lambda_1^k, \dots, \lambda_n^k$

(4)  $P$  stochastic  $\Rightarrow |\lambda_i| \leq 1 \quad \forall i$

← useful today

Note: add self-loops:  $\frac{P+I}{2}$  = "stay put with prob  $\frac{1}{2}$  & walk with prob  $\frac{1}{2}$ "  
 $\Rightarrow$  new eigenvalues  $\frac{\lambda_1 + 1}{2}, \dots, \frac{\lambda_n + 1}{2}$

Thm  $P$  is transition matrix of undirected,

→ non-k-partite, d-reg connected graph

can put self loop on each node

$\pi_0$  is start dist.

$\pi$  is stationary dist  $= (\frac{1}{n}, \dots, \frac{1}{n})$

(so  $\pi P = \pi$ )

Then  $\|\pi_0 P^t - \pi\|_2 \leq |\lambda_2|^t$

## Reducing Randomness via Random Walks:

For language  $L$ ,

let  $A$  be algorithm s.t.

$$(1) \forall x \in L \quad \Pr_{\text{d's coins}}[A(x)=1] \geq 99/100 \quad \text{usually correct}$$

$$(2) \forall x \notin L \quad \Pr_{\text{d's coins}}[A(x)=0] = 1 \quad \text{always correct}$$

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

### Method

- |  | <u># random bits used</u> |
|--|---------------------------|
| 1) run $k$ times & output " $x \in L$ " if see 0<br>else output " $x \notin L$ " | $k \cdot r$               |
| 2) use pairwise ind random bits  | $O(k+r)$                  |
| 3) today: use random walks to choose bits  | $r + O(k)$                |

The graph  $G$ : ← we get to pick  $G$ !!!

- constant degree  $d$ -regular, connected, nonbipartite

- transition matrix  $P$  for r.w. on  $G$

$$\text{has } |\lambda_2| \leq \frac{1}{10}$$

$d$ -reg  $\Rightarrow$  stat dist  $\pi$  is uniform

- # nodes =  $2^r$

Corresponds to all  
possible choices of  $r$   
random bits

# The Algorithm

# Random bits

- Pick random start node  $w \in \{0,1\}^r$

r

- Repeat K times:

$w \leftarrow$  random nbr of w

run  $A(x)$  with w as random bits.

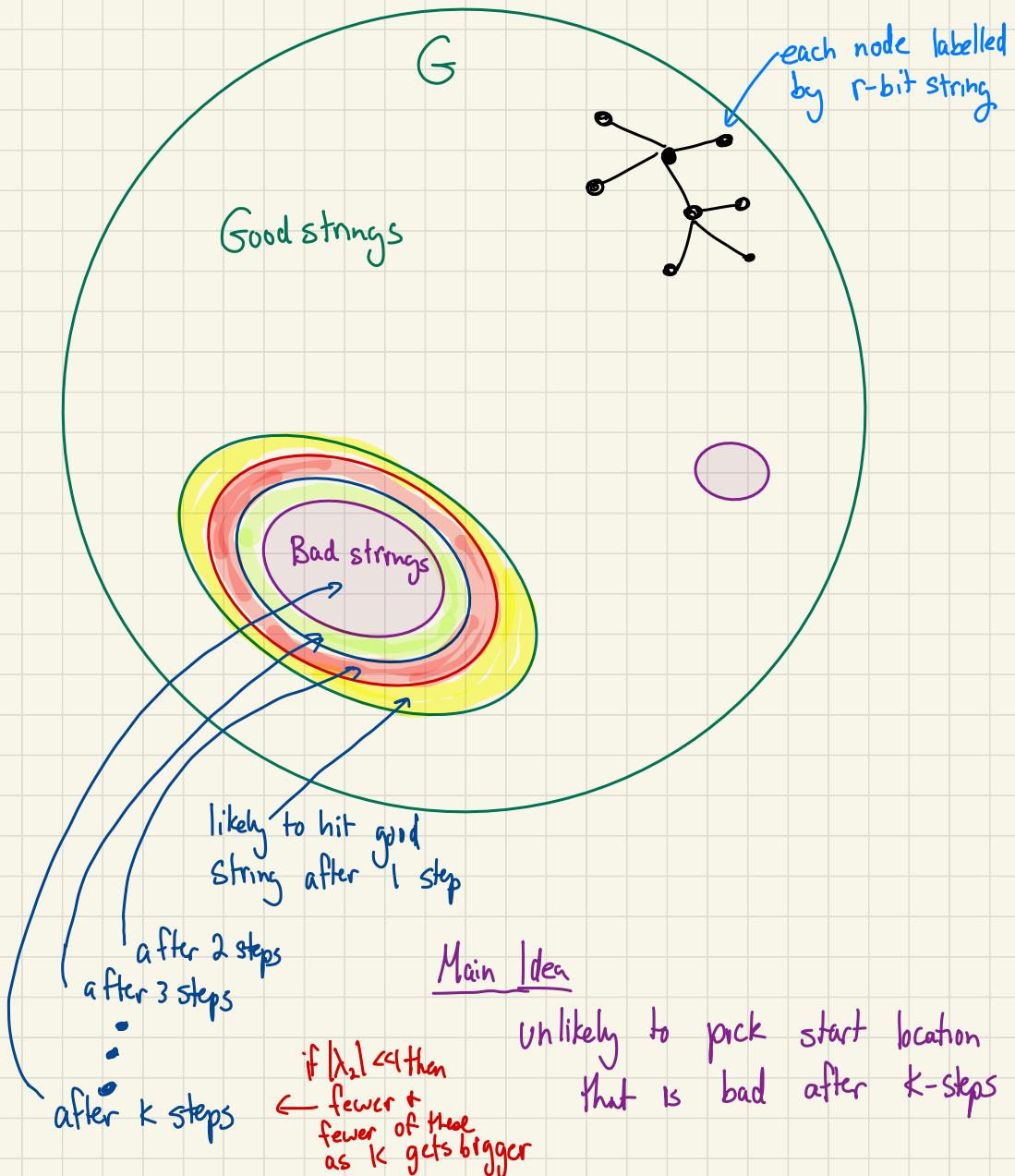
If  $A(x)$  outputs " $x \in L$ ", output " $x \in L$ " & halt  
else continue

$O(d) \times k$   
↑  
d  
is  
const  
↑  
# loops

- Output " $x \notin L$ "

total:  $r + O(k)$

Behavior: Claim: error of new algorithm is  $\leq (\frac{1}{s})^k$  for  $x \notin L$   
(still 0 error for  $x \in L$ )



bad case: walk only on "bad strings" & never reach good strings

why is this possible if  $f$  arbitrary? e.g. line

$\uparrow \lambda_2$  is close to 1

## Proof of Claim

$x \notin L$ : algorithm never errs (no bad strings)

$x \in L$ :

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

define  $B = \{w \mid A(x) \text{ with random bits } w\}$   
is incorrect.  
i.e. says  $x \notin L$

"bad w's"

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

need lin. alg. way of describing walks that  
stay in bad set:

define  $N$  diagonal matrix

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

← incorrect

← correct

$$N = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

Bad w's

For  $q$  any probability dist'.

$q \cdot N$  is ??

example:

$$q = \left( \frac{1}{4} \frac{3}{4} \right) \quad N = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$q \cdot N = \left( \frac{1}{4} 0 \right)$$

this  
0-entry  
zeroed out  
the  $\frac{3}{4}$

$q \cdot N$  deletes weight that finds  
a witness to  $x \in L$

$$\|q \cdot N\|_1 = \Pr_{w \in q} [w \text{ is bad}]$$

Can compose:

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

.

.

.

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

ignores whether start node  
bad. this just hurts vs,  
so ok to ignore.

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

First: how do we use lemma?

answer incorrect only if always see bad w/3

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

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

Since  $\|\pi\|_1 \leq \sqrt{\text{domain size}} \cdot \|\pi\|_2$

$$\begin{aligned}
 &\leq \cancel{\sqrt{2^r}} \|\rho_0\|_2 \left(\frac{1}{5}\right)^k \quad \text{apply lemma } k \text{ times} \\
 &= \cancel{\frac{1}{\sqrt{2^r}}} \quad \text{since start at uniform} \\
 &\quad + \|_{L_2} \text{ norm of uniform} \\
 &= \sqrt{\sum \left(\frac{1}{2^r}\right)^2} = \sqrt{\frac{1}{2^r}} \\
 &= \left(\frac{1}{5}\right)^k
 \end{aligned}$$

### Proof of lemma:

let  $V_1 \dots V_{2^r}$  be e-vecs of  $P$

+  $V_1$  is st.  $\|V_1\|_2 = 1$  (so  $V_1 = \left(\frac{1}{\sqrt{2^r}}, \dots, \frac{1}{\sqrt{2^r}}\right)$ )

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

$$\text{note: 1) } \|\Pi\|_2 = \sqrt{\alpha_i^2} \quad \text{by (x) proved previously}$$

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

So:

$$\|\Pi PN\|_2 = \left\| \sum_{i=1}^{2^n} \alpha_i v_i P N \right\|_2$$

since any  $\Pi$   
is lin comb of  
basis vectors

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

$$\leq \underbrace{\left\| \alpha, \lambda, v, N \right\|_2}_{(A)} + \underbrace{\left\| \sum_{i=2}^{2^n} \alpha_i \lambda_i v_i N \right\|_2}_{(B)}$$

Cauchy-Schwarz

bound (A):

$$\left\| \alpha, \lambda, v, N \right\|_2 = \left\| \alpha, v, N \right\|_2 \quad \text{since } \lambda_i = 1$$

$$= |\alpha_1| \cdot \sqrt{\sum_{i \in B} \left(\frac{1}{2^r}\right)^2} \quad \text{since } v_i = \left(\frac{1}{2^r}, \dots, \frac{1}{2^r}\right)$$

uses that  
uniform dist  
is unlikely  
to be on  
a bad string

$$+ N = \begin{pmatrix} \dots & 0 \\ 0 & \dots & \dots & 0 \end{pmatrix}$$

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

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

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

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

$$\text{since } \|\Pi\|_2 = \sqrt{\sum_{i=1}^{2^n} \alpha_i^2}$$

bound ③:

$$\left\| \sum_{i=2}^{2^r} \alpha_i \lambda_i v_i N \right\|_2 \leq \left\| \sum_{i=2}^{2^r} \alpha_i \lambda_i v_i \right\|_2 \quad \text{from note}$$

uses

"mixing"  
of  
 $v_i$ 's  
for  $i > 2$ .

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

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

$$\leq \frac{1}{10} \cdot \|\pi\|_2 \quad (*)$$

( $v_i$  could have lots of weight in bad areas, but "expansion" of graph causes it to step out of bad area)

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

New topic

## Linearity Testing

$$f: G \rightarrow H$$

$G$  is finite group  
 $H$  " " "

def.  $f$  is "linear" if  
(homomorphism)

$$\forall x, y \in G \quad f(x) +_H f(y) = f(x +_G y)$$

e.g.  $f(x) = x$

$$f(x) = ax \bmod p \quad \text{for } f = \mathbb{Z}_p$$

$$f_{\tilde{a}}(x) = \sum a_i x_i \bmod 2$$

def  $f$  is " $\varepsilon$ -linear" if  $\exists$  linear  $g$

s.t.  $f + g$  agree on  $\geq 1 - \varepsilon$  fraction  
of inputs.

Notation note that the following are equivalent statements:

- $f + g$  agree on  $\geq 1 - \varepsilon$  fraction of inputs
- $\frac{|\{x \mid f(x) = g(x), x \in G\}|}{|G|} \geq 1 - \varepsilon$
- $\Pr_{x \in G} [f(x) = g(x)] \geq 1 - \varepsilon$

How hard is it to test linearity?

do we need to try all  $x, y, xy$  tuples?

if domain is size  $n$ , this requires  $n^2$  tests  
of  $f(x) + f(y) = f(xy)$

Proposed test: Pick random  $x, y$   
Test  $f(x) + f(y) = f(xy)$

repeat  
how  
many  
times?