

PSRGs

Why?

1. Random bits are sort of rare.
2. So can re-run algorithm.
3. Analyzable - understand how much randomness is necessary.

examples: eigenvector/value computation.

primality testing

How?

Good enough - standard in most languages.

example: random d -regular is expander.

But flawed proved need good prng.

Cryptographic

* For specific sorts of algorithms.

Today: re-running an algorithm to boost confidence.

Say is right 99% of the time, but want more.

(maybe outputs yes/no)

Re-run, use majority answer

Fix input, and now view random bits $r \in \{0,1\}^n$ as input

Or, view as repeating an experiment

input $r \in \{0,1\}^n$, and output will be correct 99% of the

time.

$$X = \{r \in \{0,1\}^n \text{ or which wrong}\} \quad |X| \leq \frac{2^n}{100}$$

$$Y = \{0,1\}^n - X. \quad |Y| \geq \frac{99}{100} 2^n$$

To run k times, will generate $\tau_0, \tau_1, \tau_2, \dots, \tau_k$
each in $\{0,1\}^n$.

$$\text{Want } \Pr[\text{most } \tau_i \in X] \leq \epsilon^k$$

Naive approach: use $n(k+1)$ bits.

Today: only need $n + 9k$ bits, for $\epsilon = \frac{2}{\sqrt{5}}$

Let G be a d -regular $\frac{1}{10}$ -expander with
vertex set $U = \{0,1\}^n$.

Recall \Rightarrow all eivals of adjacency $\leq \frac{d}{10}$

For w_1, w_2, \dots, w_n eivals of $W = M D^{-1}$

$$\text{satisfy } |w_i| \leq \frac{1}{10}, \quad i \geq 2$$

Can find with $d = 400$, $\log_2 d \leq 9$, is why 9 bits.

Choose τ_0 uniform in $\{0,1\}^n$.

For each i , let τ_i be random neighbor of τ_{i-1} .

Use random walk.

Will prove: $\Pr[\text{rand walk in } X \text{ most of } k+1 \text{ steps}] \leq \left(\frac{2}{\sqrt{5}}\right)^{k+1}$

To write using matrices,

$$D_X = \text{diagonal}(1_X) \quad D_Y = \text{diagonal}(1_Y)$$

Let $S \subseteq \{0, \dots, k\}$

$\Pr[\tau_i \in X \text{ for } i \in S \text{ and } \tau_i \in Y \text{ for } i \notin S] \quad (*)$

will prove $\leq \left(\frac{1}{5}\right)^{|S|}$

So, $\Pr[\text{walk in } X \text{ most steps}]$

$$\leq \sum_{|S| > \frac{k}{2}} \left(\frac{1}{5}\right)^{|S|} = \sum_{|S| > \frac{k}{2}} \left(\frac{1}{5}\right)^{\frac{k+1}{2}} \leq 2 \left(\frac{1}{5}\right)^{\frac{k+1}{2}} = \left(\frac{2}{5}\right)^{\frac{k+1}{2}}$$

Define $D_i = D_X \quad i \in S$
 $= D_Y \quad i \notin S$

$$(*) = \mathbb{1}^T D_k W_k \cdots W D_1 W D_0 = \mathbb{1}^T D_k W_k \cdots W D_1 W D_0 W \frac{1}{n}$$

To bound this, use matrix norms.

$$\text{Recall } \|M\| = \max_x \frac{\|Mx\|}{\|x\|}$$

$$\text{Note } \|M_1 M_2\| \leq \|M_1\| \cdot \|M_2\|$$

And, for symmetric M , $\|M\| = \max \text{ abs eigen}$.

Claim 1 $\|D_Y \omega\| \leq 1$

proof $\|\omega\| = 1$, $\|D_Y\| = 1$, so $\|D_Y \omega\| \leq 1$

Will prove claim 2: $\|D_X \omega\| \leq \frac{1}{5}$

$$\Rightarrow \left\| D_k \omega D_{k-1} \omega \dots D_0 \omega \right\| \leq \left(\frac{1}{5}\right)^{|S|}$$

$$\begin{aligned} \Rightarrow \left\| D_k \omega \dots D_0 \omega \cdot \frac{1}{n} \right\| &\leq \left(\frac{1}{5}\right)^{|S|} \cdot \left\| \frac{1}{n} \right\| \\ &= \left(\frac{1}{5}\right)^{|S|} \cdot \frac{1}{\sqrt{n}} \end{aligned}$$

$$\text{and } \mathbf{1}^T D_k \omega \dots D_0 \omega \cdot \frac{1}{n} \leq \left(\frac{1}{5}\right)^{|S|} \cdot \frac{1}{\sqrt{n}} \cdot \|\mathbf{1}\| = \left(\frac{1}{5}\right)^{|S|}$$

Proof of claim 2:

$\|D_X \omega\| \leq \frac{1}{5}$. Will prove for all z ,

$$\|D_X \omega z\| \leq \frac{\|z\|}{5}.$$

let $z = c\mathbf{1} + \gamma$, where $\mathbf{1}^T \gamma = 0$.

$$D_X \omega \mathbf{1} = D_X \mathbf{1} = \mathbf{1}_X, \quad \|\mathbf{1}_X\| \leq \sqrt{\frac{n}{100}} = \frac{\sqrt{n}}{10}$$

$$\text{As } \mathbb{1}^T \gamma = 0, \quad \|\omega_\gamma\| \leq \|\gamma\| \cdot \max(\omega_2, \omega_n) \leq \frac{\|\gamma\|}{10}$$

$$\text{So, } \|\mathbb{D}_x \omega z\| \leq \|\mathbb{D}_x \omega c \mathbb{1}\| + \|\mathbb{D}_x \omega_\gamma\|$$

$$\leq \frac{c\sqrt{n}}{10} + \frac{1}{10}\|\gamma\|$$

$$= \frac{1}{10}\|c\mathbb{1}\| + \frac{1}{10}\|\gamma\|$$

$$\leq \frac{2}{10}\|z\| \quad \text{because } \|c\mathbb{1}\| \leq \|z\| \\ \|\gamma\| \leq \|z\|$$

Note: very odd, because used 2-norms for probabilities

Note: for asymmetric, norm has little to do with eigenvalues.

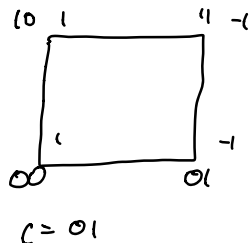
Generalized Hypercube.

$$V = \{0,1\}^d \pmod{2}. \quad (a,b) \in E \iff b = a + \delta_i \pmod{2} \\ \uparrow \\ \text{in pos } i.$$

From product theorem,

have eigenvectors for each $c \in \{0,1\}^d$

$$\psi_c(a) \triangleq (-1)^{c \cdot a} \quad \text{as.}$$



Generalized. Pick $g_1, \dots, g_k \in \{0,1\}^d$, $k \geq d$

edges are $(a, a+g_i) \pmod 2$ $1 \leq i \leq k$

Claim: has same eigenvectors.

lem For $c \in \{0,1\}^d$, ψ_c is an eigvec of M with eigenval

$$\sum_{i=1}^k (-1)^{c^T g_i}$$

$$\text{First } \psi_c(a+b) = (-1)^{c^T(a+b)} = (-1)^{c^T a} (-1)^{c^T b} = \psi_c(a) \psi_c(b)$$

For any vertex a , compute

$$\begin{aligned} (M\psi_c)(a) &= \sum_{i=1}^k \psi_c(a+g_i) \\ &= \sum_{i=1}^k \psi_c(a) \psi_c(g_i) \\ &= \sum_{i=1}^k \psi_c(g_i) \\ &= \sum_{i=1}^k (-1)^{c^T g_i} \end{aligned}$$

need all these small...