

Recall:

We were proving Reed-Solomon, and had the following expected values:

$$\text{keep} := \left(1 - \frac{p}{2}\right)^{\Delta_L(G)} \approx e^{-\frac{p\Delta}{2}} \quad (= \Pr(c \in L'(u) | c \in L(u)))$$

$$\mathbb{E}[|L'(u)|] = \text{keep} \cdot |L|$$

$$\mathbb{E}[d_{L',G}(u,c)] = d_{L,G}(u,c) \left((1-p) \cdot \text{keep} + p \left(\text{keep} \cdot \frac{1}{|L|} + \left(1 - \frac{1}{|L|}\right) \left(1 - \frac{2}{|L|}\right)^{\Delta} \right) \right)$$

\uparrow not activated
 \uparrow keeps c
 \uparrow if $d(u)=c$ and keeps c
 \nwarrow $d(u) \geq c$ and keeps c
 \uparrow keeps $\neq c$ and c

But ~~the~~ it turns out that

this variable doesn't concentrate well, so we'll use another.

Picking a better variable:

$$d_{L',G}(u,c) = |N_{L,G}(u,c)| = |\{v \in N_{L,G}(u,c) \cap V(G') : c \in L(v)\}|$$

$\underbrace{\qquad\qquad\qquad}_{G' \rightarrow G'} \qquad \qquad \underbrace{\qquad\qquad\qquad}_{L' \rightarrow L}$

But, it turns out we don't need to care about the list, so instead let's look at $|N_{L,G}(u,c) \cap V(G')|$.

Clearly

$$d_{L',G}(u,c) \leq |N_{L,G}(u,c) \cap V(G')|.$$

$$\mathbb{E}[|N_{L,G}(u,c) \cap V(G')|]$$

$$= \sum_{u \in N_{L,G}(u,c)} \Pr[u \in V(G')] \quad (\text{line of exp.})$$

$$= \sum_{u \in N_{L,G}(u,c)} 1 - \Pr[u \notin V(G')]$$

These events independent.

$$= \sum_{u \in N_{L,G}(u,c)} 1 - \Pr[u \in A \text{ and } \phi(u) \text{ is kept}] = d_{L,G}(u,c) (1 - p \cdot \text{keep})$$

On to some iterative calculations:

- Now, if we could show that with "high enough" probability every $|L'(G)|$ and $|N_{L(G)}(G, c) \cap V(G')|$ are close enough to their expectations to apply the LLT, then we'll be happy with the following calculations:

$$\frac{|L'|}{\Delta_L(G)} \approx \frac{\mathbb{E}[|L'(G)|]}{\mathbb{E}[|N_{L(G)}(G, c) \cap V(G')|]} = \frac{\text{keep} \cdot |L|}{(1 - p \cdot \text{keep}) \cdot \Delta_L(G)}$$

Some ~~extra~~ factor \rightarrow we will argue this is close to 1

$$\geq \frac{|L|}{\Delta_L(G)} \cdot \frac{1 - \frac{p\Delta}{14}}{1 - p(1 - \frac{p\Delta}{14})} \quad \left\{ \begin{array}{l} \text{using that } (1 + \frac{x}{n})^n \geq 1 + x \text{ for } nx \geq 1 \text{ and } |x| \leq n \end{array} \right.$$

Note: $\frac{\Delta}{14} = \frac{1}{1+\epsilon}$ (see assumption we made)

$$= \frac{|L|}{\Delta_L(G)} \cdot \frac{1 - p/(1+\epsilon)}{1 - p(1 - p/(1+\epsilon))}$$

$$= 1 + \frac{p\epsilon}{1+\epsilon} - \frac{\epsilon^2}{1+\epsilon}$$

Now, if we choose $p \geq \epsilon/2$, then we get

$$= 1 + \frac{\epsilon^2}{4(1+\epsilon)} \quad \left\{ \begin{array}{l} \text{so, we get some improvement in each iteration} \end{array} \right.$$

How to show a variable is close to its expectation:

Concentration Inequalities:

Markov's Inequality:

If $X \geq 0$ is a random variable,

$$\Pr[X \geq k \mathbb{E}[X]] \leq 1/k. \quad \forall k \geq 1$$

Chebyshev's Inequality:

The variance of a random variable X , denoted $\text{Var}[X]$ is

$$\text{Var}[X] = \mathbb{E}[(X - \mathbb{E}[X])^2]$$

The standard deviation σ of X is $\sqrt{\text{Var}[X]}$.

~~Then~~, let $\mu = \mathbb{E}[X]$, then

$$\Pr[X \geq \mu + k\sigma] \leq 1/k^2.$$

Proof: Use Markov's inequality.

Chernoff Bounds:

Let $X = \sum_{i=1}^n X_i$ where X_i is a Bernoulli r.v. (i.e. only takes values of 0 and 1), and all independent:

$$\Pr[X \leq (1-\delta)\mu] \leq e^{-\frac{\delta^2 \mu}{2}} \quad (0 \leq \delta \leq 1)$$

$$\Pr[X \geq (1+\delta)\mu] \leq e^{-\frac{\delta^2 \mu}{3}} \quad \text{if } 0 \leq \delta \leq 1$$

$$e^{-\frac{\delta^2 \mu}{3}} \quad \text{if } \delta \geq 1$$

(i.e. exponentially small if δ constant and gives something meaningful if $\delta \gg \frac{1}{\sqrt{\mu}}$).

Example:

$$\mathbb{E}[|A \cap N_{L,c}(u,c)|] = p \Delta_L(u)$$

$$\Pr[|A \cap N_{L,c}(u,c)| \geq p \Delta_L(u)(1+\delta)] \leq e^{-\frac{\delta^2 (p \Delta_L(u))^2}{3}} \quad (\text{By Chernoff}).$$

$$\text{i.e. } |A \cap N_{L,c}(u,c)| = p \Delta_L(u) \pm \sqrt{\Delta_L(u) \log \Delta_L(u)}$$

with prob. $1 - \frac{1}{\Delta_L(u)}$ for any u .

So, by LL, this holds for all u with some positive prob.

But, we need to concentrate:

$$|L(u)| \quad \text{and} \quad |N_{L,c}(u,c) \cap V(G')|$$

Depends on color $c \in L(u)$
being in $L(u)$ which
depends on activation flips
and color-assignment for
 $N_{L,c}(u,c)$ (But not independent!)

→ Even worse!
(Stronger interaction with
neighbors + 2nd neighbors)

"Simple" Concentration Bound (Basis of Hoeffding & Freedman)

Let X be a r.v. that depends only on the outcome
of a set of independent trials T_1, \dots, T_n . Suppose that
changing the outcome of any one trial changes X by
at most C (C constant). (This is called C -Lipschitz).

Then,

$$\Pr[|X - \mathbb{E}[X]| \geq t + 8C \sqrt{\mathbb{E}[X]}] \leq e^{-\frac{t^2}{32n}}$$

Remark: Note that the denominator has the # of trials.
(Just exponential)

Jan 21st

Concentrating $|L'(v)|$:

Trials: activating flips & colourings of $X_L(G, c)$ & $\text{col}(v)$

- 1-lipschitz: Since changing any one trial (activation or colouring) changes $|L'(v)|$ by at most 1.

Recall: $\mathbb{E}[|L'(v)|] = k_{\text{avg}} \cdot \Delta_L(G) = \Theta(\Delta_L(G))$

What is n ?

- If not doing colour degree, then 2Δ (Good!)

- But, w/ colour degree it's:

$$2\Delta_L(G) \cdot |L| \quad (\text{If every colour the neighbours holds})$$

$$= 2(1 + \epsilon \Delta_L(G))^2$$

(This is Bad! The simple concentration bound is only meaningful for $t \geq \frac{\Delta_L(G)}{\epsilon}$.)

Talagrand's Inequality (over)

CA)

Combinatorial version: (There is also a probability one)

Let $X \geq 0$ depend on independent trials T_1, \dots, T_n . If X is C -lipschitz and r -verifiable, then for any $t \geq 9C\sqrt{rc^2 \mathbb{E}[X]} + 28rc^2$,

then:

$$\Pr[|X - \mathbb{E}[X]| > t] \leq 4e^{-\frac{t^2}{8rc^2(4\mathbb{E}[X] + t)}}$$

Remark: If r, c constant, then get exponentially small in t , if $t = \Theta(\mathbb{E}[X])$ and still meaningful for $t \geq \sqrt{\mathbb{E}[X]}$.

③

r-Verifiable: For every $\delta > 0$, if $X \geq S$, then there exists a set Z of at most $r\delta$ trials that "verify" that $X \geq S$, i.e. Changing any single outcome of Z still results in $X \geq S$.

So, of course, counting # of heads ^{in coin flips} is 1-verifiable.

Similarly, $|A \cap N_{G,C}(u, c)|$ is ~~1~~ 1-verifiable.

(We just exhibit the heads on the activated set).

What about $|L(u)|$?

Example: How do we show $|L(u)| \geq 1$ (i.e. One column kept).

For all $u \in N_{G,C}(u, c)$ show either $u \in A$ or $\phi(u) \neq c$.

i.e. Need $\Delta_{G,C}(u, c)$ trials to verify

i.e. Need $r \geq \Delta_{G,C}(u, c)$ (This is bad!)

Pick a better variable:

$|L(u)| - |L'(u)| = \# \text{ of columns lost}$

$$\mathbb{E}[|L(u)| - |L'(u)|] = |L| - \text{keep} |L|$$

$$= |L| (1 - \text{keep})$$

$$\geq (1 - \epsilon) \Delta_{G,C}(u, c) (1 - \text{keep}) = \Theta(\Delta_{G,C}(u, c))$$

Obviously, this variable is 1-Lipschitz.

What do we need to verify a column is lost?

→ Need a neighbor $v \in N_{G,C}(u, c)$ to be activated and $\phi(v) = c$

i.e. Need 2 trials

More generally, if $|L(u)| - |L'(u)| \geq S$, need $2S$, so $r = 2$ works.