

Today:

- Amplifying the probability of success of the AMS algorithm for F_2
- The Chernoff bound

Review of what we know already:

- We will refer to copies of this algorithm as Z_i
- Algorithm that outputs a multiplicative $(1+\epsilon)$ -approximation to F_2 w.p. $\geq \frac{8}{9}$
 - How: - Maintain $k = \lceil 18/\epsilon^2 \rceil$ sketches

$$Y_i = \sum_{x \in X} h_i(x) f(x)$$
 - Estimate: $Z = \frac{\sum_{i=1}^k Y_i^2}{k}$
 - Space: $O(1/\epsilon^2)$

Challenge from last time:

Improve probability of outputting multiplicative $(1+\epsilon)$ -approximation to $1-S$ for $S \in (0, 1/9)$.

Attempt 1 (last time):

- Take the mean of more estimators Y_i^2

- Outcome via Chebyshev's inequality:
needed $k = \Theta\left(\frac{1}{s^2 \epsilon^2}\right)$

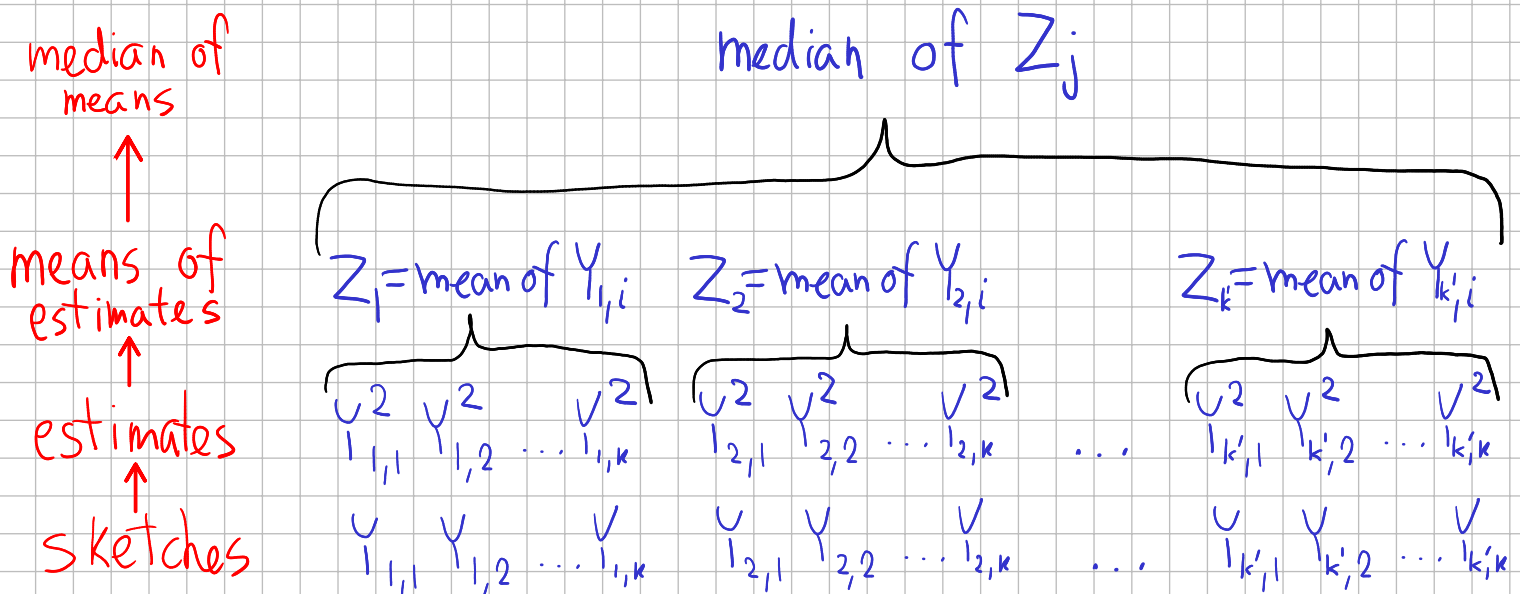
- So total space $O\left(\frac{1}{s^2 \epsilon^2}\right)$

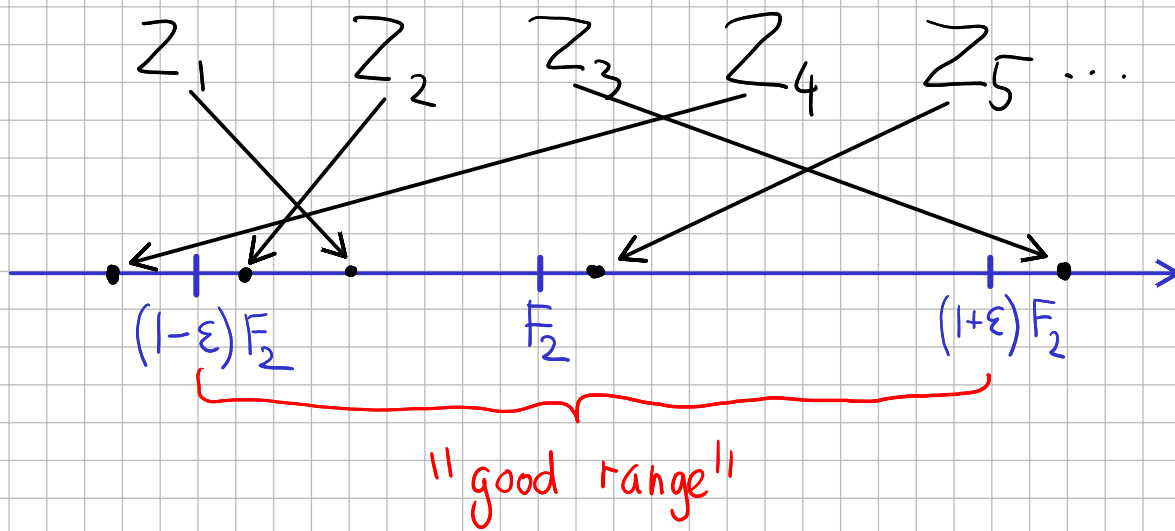
Bigger overhead
than Count-Min
Sketch, which
required $O(\log(1/s))$

Attempt 2 (now): Run k' copies of Z_i
& return the median of
estimates

to be set later

Pictorially:





each Z_i in this range w.p. $\geq \frac{8}{9}$

The median of Z_i 's outside of the "good range" only if

① at least half of Z_i 's $> (1+\varepsilon)F_2$

or

at least half of Z_i 's $< (1-\varepsilon)F_2$

We'll show that probability of

⊛ \equiv (at least half of Z_i 's outside of the "good range")

is small. If ⊛ does not occur, then neither

do ① and ②, and the median of Z_i 's is

a multiplicative $(1+\varepsilon)$ -approximation.

The Chernoff bound

Setup: X_1, X_2, \dots, X_n - independent random variables that take on values in $[0, 1]$

$$X \stackrel{\text{def}}{=} \sum_{i=1}^n X_i \quad \& \quad \mu \stackrel{\text{def}}{=} \mathbb{E}[X]$$

Claim: For $\varepsilon \in [0, 1]$,

$$\Pr[X \leq (1-\varepsilon)\mu] \leq e^{-\varepsilon^2 \mu / 2}$$

and

$$\Pr[X \geq (1+\varepsilon)\mu] \leq e^{-\varepsilon^2 \mu / 3}$$

For $\varepsilon \geq 1$,

$$\Pr[X \geq (1+\varepsilon)\mu] \leq e^{-\varepsilon \mu / 4}$$

Bounding the probability of \odot :

For $i \in [k']$, $X_i = \begin{cases} 1 & \text{if } Z_i \text{ in "good range"} \\ 0 & \text{otherwise} \end{cases}$

*independent
random variables*

Define $X \stackrel{\text{def}}{=} \sum_{i=1}^{k'} X_i$. Then $\mathbb{E}[X] \geq \frac{8}{9} k'$.

$$\begin{aligned} \Pr[\odot] &= \Pr\left[X \leq \frac{k'}{2}\right] = \Pr\left[X \leq \frac{9}{16} \cdot \frac{8}{9} k'\right] \\ &\leq \Pr\left[X \leq \left(1 - \frac{7}{16}\right) \mathbb{E}[X]\right] \end{aligned}$$

Chernoff bound \rightarrow

$$\leq e^{-\frac{(7)^2 \mathbb{E}[X]}{16}} / 3 \leq e^{-\frac{(7)^2}{16} \cdot \frac{8}{9} k' / 3} = e^{-\frac{49}{864} k'}$$

By setting $k' = \lceil \frac{864}{49} \ln(1/\delta) \rceil$, we get

$$\Pr[\circledast] \leq e^{-\ln(1/\delta)} = \delta.$$

Total space: $O(k' \cdot k) = O\left(\frac{1}{\epsilon^2} \log(1/\delta)\right)$

↑
overhead of $O(\log(1/\delta))$
(asymptotically better than $O(1/\delta)$)

Note: Nothing specific about F_2 here. The same strategy, taking the median of several estimators, would work for any estimator as long as it gives a "good" estimate with probability p , where p is a constant greater than $1/2$. The number of estimators needed is $O(\log(1/\delta))$.