

CS 331, Fall 2025
lecture 20 (11/5)

Today: - Basic boosting
- Chebyshev's inequality
- Mean/median boosting
- Morris's counter

Basic boosting (Part VII, Section 3)

Today: improving the failure probability

$$\Pr[A \text{ does } \dots] \leq \delta \quad \leftarrow$$

Basic setting: you have A , succeeds

@ outputting "good" $\times$ w.p. $\geq p$

e.g. Find pivot

$$p = \frac{1}{2}$$

Contation resolution

$$p = \frac{1}{n}$$

What if p small?

If we can check whether x good...

- Run A T times
- Check all of them, return any good

We'll get good x except w.p.

$$(1-p)^T$$

Key fact: $\forall p \in (0, \frac{1}{2})$,

$$\frac{1}{4} \leq (1-p)^{\frac{1}{p}} \leq \frac{1}{e}$$

$$\text{Thus, } T \geq \frac{1}{p} \log\left(\frac{1}{\delta}\right)$$

$$\Rightarrow (1-p)^T \leq \delta$$

What if we can't verify??

Motivation: running experiment to estimate x^*
design algo to output x , $\mathbb{E}[x] = x^*$

but...

- We care about $|x - x^*|$
- Can't check (don't know x^*)

Today: how to prove things like

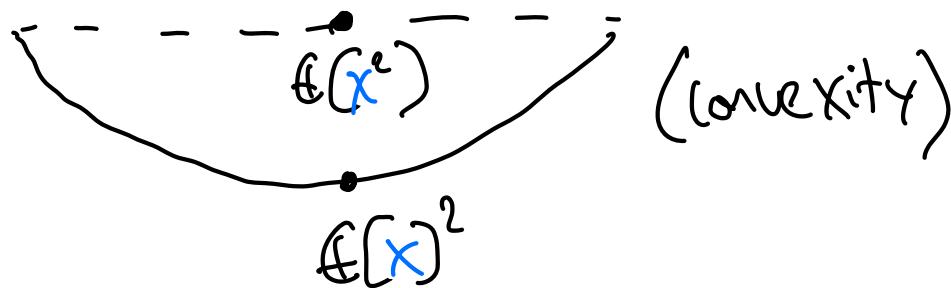
$$\Pr \left[x \notin \underbrace{[x^* - R, x^* + R]}_{\text{"confidence interval"}} \right] \leq \delta$$

- Mean boosting: improve R
- Median boosting: improve δ

Chebyshev's inequality (Part VII, Section 3.1)

Key players:

$$\text{Var}[X] := \mathbb{E}[X^2] - \mathbb{E}[X]^2 \geq 0$$



$$\text{Stdev}[X] := \sqrt{\text{Var}[X]} \quad \text{"spread"}$$

Aside Variance cash course

Reminder about $\mathbb{E}$: (no crests!!!)

$$\textcircled{1} \mathbb{E}[cX]$$

$$= c \mathbb{E}[X]$$

$$\textcircled{2} \mathbb{E}[X + Y]$$

$$= \mathbb{E}[X] + \mathbb{E}[Y]$$

$$E[(X - E[X])^2] = E[X^2]$$

$$- 2E[X \cdot \underbrace{E[X]}_{\text{constant, use ①}}] + E[X]^2$$

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

Also, we have

$$\textcircled{1} \text{Var}[cX] = c^2 \text{Var}[X]$$

$$\textcircled{2} \text{Var}[X + Y] = \text{Var}[X] + \text{Var}[Y]$$

only if X, Y independent!

(Compare to E , ① & ② both always true)

Our intuition: for reasonable δ ,
pick $R \approx \text{stddev}(X)$

Markov's inequality

let r.v. $X \geq 0$, then $\forall \delta \in (0, 1)$,

$$\Pr\left[X \geq \frac{E[X]}{\delta}\right] \leq \delta$$

e.g. $X < 10 E[X]$ except w.p. $\frac{1}{10}$

Proof: if not, let $\tau = \frac{E[X]}{\delta}$,

$$\begin{aligned} E[X] &= \Pr[X \geq \tau] \cdot \tau + \Pr[X < \tau] \cdot 0 \\ &> \delta \cdot \tau > E[X] \end{aligned}$$

Chebyshev's inequality

Apply Markov's with $X \leftarrow (x - E[X])^2$

$$\Pr \left[(x - E[X])^2 \geq \frac{\text{Var}[X]}{\delta} \right] \leq \delta$$

$$\Leftrightarrow |x - E[X]| \geq \frac{\text{stddev}[X]}{\sqrt{\delta}}$$

Thus we have shown except w.p. δ ,

$$x \in \left[E[X] - \frac{\text{stddev}[X]}{\sqrt{\delta}}, E[X] + \frac{\text{stddev}[X]}{\sqrt{\delta}} \right]$$

e.g. $\delta = \frac{1}{4}$, confidence interval

$$\left[E[X] - \underbrace{2\text{stddev}[X]}_{\text{radius } R}, E[X] + 2\text{stddev}[X] \right]$$

Mean / median boosting (Part VII, Section 3.2)

Idea: Improve R (mean boosting)

Recall that $R \propto \text{stdv}(X)$

How to halve R ? Decrease Var !

Basic fact:

Let $X_1, X_2, \dots, X_k$ independent copies of X ,

$$\bar{X} = \frac{1}{k} \sum_{i \in [k]} X_i$$

$$\text{Var}(\bar{X}) = \frac{1}{k^2} \text{Var}\left(\sum_{i \in [k]} X_i\right)$$

$$= \frac{k}{k^2} \text{Var}(X) = \frac{1}{k} \text{Var}(X)$$

Takeaway: if $k \approx \frac{1}{\epsilon^2}$,

then our confidence gets ϵ better

$$\text{Var}(\bar{X}) = \epsilon^2 \text{Var}(X), \sqrt{\epsilon^2} \times \text{for stdv.}$$

Idea 2: Improve δ (median boosting)

Suppose r.v. X has

$$\text{Pr}(X \in I) \geq \frac{3}{4} \text{ for interval } I \subset \mathbb{R}$$

Claim: let $\hat{X} = \text{median}(X_1, X_2, \dots, X_k)$

$$k \geq 12 \log\left(\frac{1}{\delta}\right)$$

Then, $\hat{X} \in I$ except w.p. δ

How to prove?

Step 1: it's enough to show $\geq \frac{k}{2}$ copies $\in I$

e.g. if $\hat{x} \notin I$ to the right, clearly $< \frac{k}{2} \in I$



Our goal: we have coin that H w.p. $\geq \frac{3}{4}$
toss k coins, $\geq \frac{k}{2}$ are H.

Step 2: "binomial concentration"

Suppose $i \geq \frac{k}{2}$ copies miss I



$$\begin{aligned} \Pr(\geq \frac{k}{2} \text{ are } T) &= \sum_{i=\frac{k}{2}}^k \binom{k}{i} (1-p)^i p^{k-i} \\ &\leq \sum_{i=\frac{k}{2}}^k \binom{k}{i} \left(\frac{1}{4}\right)^i \left(\frac{3}{4}\right)^{k-i} \end{aligned}$$

Observe if $i \leftarrow i+1$, $\underbrace{\binom{k}{i}}_{\text{decreases}} \underbrace{\left(\frac{1}{4}\right)^i \left(\frac{3}{4}\right)^{k-i}}_{\text{decreases } 3x}$

Thus geom sequence. Enough to get first term

$$\binom{k}{k/2} \left(\frac{1}{4}\right)^{k/2} \left(\frac{3}{4}\right)^{k/2} = O(\delta)$$

$$\leq 2^k \cdot \left(\frac{3}{16}\right)^{k/2} = \left(\frac{3}{4}\right)^{k/2}$$

Good enough if $k = O\left(\log\left(\frac{1}{\delta}\right)\right)!$

- 3-stage plan:
- Basic interval via $V_S(X)$
 $R = 2 \text{ stdev}(X)$, $\delta = 1/4$
 - Mean boost $\times \frac{4}{\epsilon^2}$
 $R = \epsilon \text{ stdev}(X)$, $\delta = 1/4$
 - Median boost $\times 12 \log\left(\frac{1}{\delta}\right)$
 $R = \epsilon \text{ stdev}(X)$, $\delta = \delta$

Morris's Counter (Part VII, Section 3.3)

Goal: build data structure to store counter $i \in \mathbb{O}$

- $\text{Count}() : i \leftarrow i + 1$
- $\text{Report}() : \text{return } i$

How much space needed to $\text{Count}()$ n times?

Trivial: $O(\log(n))$ space (store i)

Today: $O(\log \log(n))$ space (randomized)

Motivation:

- Count many large things
- Constant factors help a lot!
- Web crawler, search engine, etc.
- Morris: spellchecker, needed 26^3 trigram counters

Algo: $X \leftarrow 0,$

Count(): w.p. $2^{-X}, X \leftarrow X+1$

Report(): return $2^X - 1$

It works ????

Intuition: $X \approx \log(i+1)$

Increase X once in each bucket

$i \in (3, 4], [5, 8], [9, 16], \dots$

Let $X_k =$ value of X after $i = k$

Claim: $E[2^{X_k}] = k+1$

$\forall k \in (n)$

$\text{Var}[2^{X_k}] \leq 2k^2$

3-step plan: gives estimate in

$$[(1-\epsilon)n, (1+\epsilon)n] \text{ w.p. } \geq 1-\delta$$

$$(NY22): \text{ space } O\left(\log \log \frac{n}{\delta} + \log \frac{1}{\epsilon}\right)$$

Proof of $\mathbb{E}$: true when $k=0$. Induct!

$$\mathbb{E}(2^{X_{k+1}}) = 2^j + 1$$

$$= \sum_{j=0}^{\infty} \Pr(X_k = j) \cdot \left(2^j \left(1 - \frac{1}{2^j}\right) + 2^{j+1} \left(\frac{1}{2^j}\right) \right)$$

$$= \sum_{j=0}^{\infty} \left(\Pr(X_k = j) 2^j + \Pr(X_k = j) \right)$$

$$= \mathbb{E}(2^{X_k}) + 1. \quad \square \quad (\text{Var: see notes})$$