CSCE 689: Special Topics in Modern Algorithms for Data Science Fall 2023

Lecture 23 — October 25, 2023

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Recall

Theorem 1 (Bernstein's Inequality). Let $y_1, ..., y_n \in [-M, M]$ be independent random variables and let $y = y_1 + ... + y_n$ have mean μ and variance σ^2 . Then for any $t \ge 0$:

$$Pr[|y - \mu| \ge t] \le 2e^{-\frac{t^2}{2\sigma^2 + \frac{4}{3}Mt}}$$

Sampling for Sum Estimation

Goal: Approximate the original sum by the sum of the rescaled sampled numbers. Consider a fixed set $X = \{x_1, ..., x_n\}$ of *n* numbers, and suppose we sample each point x_i with some probability p_i and rescale by $\frac{1}{p_i}$, also we let y_i be the contribution of the sample corresponding to x_i , we have:

- $y_i = 0$ with probability $1 p_i$
- $y_i = \frac{1}{p_i} \cdot x_i$ with probability p_i
- $\mathbb{E}[y_i] = x_i$
- Expected sum: $\mathbb{E}[y_1 + ... + y_n] = x_1 + ... + x_n$

Uniform Sampling

Suppose $p_i = p$ for all $i \in [n]$, given Theorem 1, if $x_1 = \dots = X_n = 1$, set $M = \frac{1}{p}$, $t = \frac{n}{2}$, and $\sigma^2 = \frac{n}{p}$, then:

$$\Pr[|y - \mu| \ge \frac{n}{2}] \le 2\exp(-\frac{(n/2)^2}{2(n/p) + (4/3)(n/2p)})$$

We require $2(\frac{n}{p}) \approx (\frac{n}{2})^2$, so we need $p = \Theta(\frac{1}{n})$. If $x_1, ..., x_n \in [1, 2]$, set $M = \frac{2}{p}$, $t = \frac{x}{2}$ and $\sigma^2 \approx \frac{4n}{p}$, then:

$$\Pr[|y - \mu| \ge \frac{x}{2}] \le 2\exp(-\frac{(x/2)^2}{2(4n/p) + (4/3)(x/p)})$$

We require $2(\frac{4n}{p}) \approx \frac{x^2}{2}$ and x can be as small as n, so $p \approx \frac{2}{n}$. If $x_1, ..., x_n \in [1, 100]$, set $M = \frac{100}{p}$, $t = \frac{x}{2}$ and $\sigma^2 \approx \frac{10000n}{p}$, then:

$$\Pr[|y - \mu| \ge \frac{x}{2}] \le 2\exp(-\frac{(x/2)^2}{2(10000n/p) + (4/3)(100x/p)})$$

We require $2(\frac{10000n}{p}) \approx \frac{x^2}{2}$ and x can be as small as n, so $p \approx \frac{80000}{n}$. If $x_1, ..., x_n \in [1, n]$, set $M = \frac{n}{p}$, $t = \frac{x}{2}$ and $\sigma^2 \approx \frac{n^2}{p}$, then:

$$\Pr[|y - \mu| \ge \frac{x}{2}] \le 2\exp(-\frac{(x/2)^2}{2(n^2/p) + (4/3)(nx/2p)})$$

We require $2(\frac{n^2}{p}) \approx \frac{x^2}{2}$ and x can be as small as n, so we need $p \approx 1$. To sum up:

| Domain of $x_1,, x_n$ | Conditions of p for 2-approximation | Expected samples amount |
|------------------------------|---------------------------------------|-------------------------|
| $x_1 = \dots = x_n = 1$ | $p = \Theta(\frac{1}{n})$ | $np = \Theta(1)$ |
| $x_1,, x_n \in [1, 2]$ | $p \approx \frac{2}{n}$ | slightly larger np |
| $x_1,, x_n \in [1, 100]$ | $p \approx \frac{80000}{n}$ | way larger np |
| $x_1, \dots, x_n \in [1, n]$ | $p \approx 1$ | n |

Table 1: Uniform Sampling Summary

Importance Sampling

Suppose $x = x_1 + \ldots + x_n$, since p_i is chosen proportionally to x_i , let $p_i = \frac{x_i}{x}$, we have $y_i \leq \frac{1}{p} \cdot x_i = \frac{x_i}{x_i} \cdot x_i = x$, thus:

- $\operatorname{Var}[y_i] \leq \frac{1}{p_i} x_i^2 \leq x_i \cdot x$
- $\operatorname{Var}[y] = \operatorname{Var}[y_1] + \ldots + \operatorname{Var}[y_n] \le x \cdot (x_1 + \ldots + x_n) = x^2$

Given Theorem 1, set M = x, $t = \frac{x}{2}$, and $\sigma^2 \approx x^2$. Then

$$\Pr[|y - \mu| \ge \frac{x}{2}] \le 2\exp{-\frac{(x/2)^2}{2x^2 + (4/3)(x^2/2)}}$$

Suppose $x_1, ..., x_n \in [1, n]$, we can get a 2-approximation and based on $\frac{x_1}{x} + ... + \frac{x_n}{x} = 1$, we expect a constant number of samples.