

E2 207: Concentration Inequalities

Naman Mishra

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Contents

Lectures

1	Tue, August 5	2
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The course

Resources:

(1)

Lecture 1.

Tuesday

August 5

Evaluation:

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Question .1. Toss a fair coin 10000 times. How many heads do you expect to see?

Our expectation is 5000.

Theorem .2 (Weak law of large numbers). If X_1, X_2, \dots are iid random variables with finite mean and variance, then $\frac{1}{n} \sum_{i=1}^n X_n \xrightarrow{P} \mathbf{E}[X_1]$. That is, for any $\epsilon > 0$,

$$\lim_{n \rightarrow \infty} \mathbf{P} \left\{ \left| \frac{1}{n} \sum_{i=1}^n X_i - \mathbf{E}[X_1] \right| > \epsilon \right\} = 0.$$

The finite variance assumption is not necessary.

For a fixed ϵ , the error bound on S_n is $O(n)$.

Theorem .3 (Central limit theorem). If X_1, X_2, \dots are iid random variables with finite mean μ and variance σ , then

$$\frac{1}{\sqrt{n}} \left(\sum_{i=1}^n (X_i - \mu) \right) \xrightarrow{d} N(0, \sigma^2).$$

Using this, once more, we have that the number of heads is $5000 \pm \sqrt{10000(\frac{1}{4})Q^{-1}(0.005)}$ with probability at least 0.99, if we assume that $10000 = \infty$, where $Q(x) = \mathbf{P}(Z \geq x)$ for $Z \sim N(0, 1)$. $Q^{-1}(0.005) \approx \sqrt{(-\log 0.005)}$. The uncertainty here happens to be around 115 tosses.

We will eventually see the Berry-Esseen theorem.

Theorem .4 (Berry-Esseen inequality). *If X_1, X_2, \dots are iid with zero mean and finite **third** moment, then there exists a constant C such that for each n and ε ,*

$$\left| \mathbf{P} \left\{ \frac{1}{\sigma\sqrt{n}} \sum_{i=1}^n X_i > \varepsilon \right\} - Q(\varepsilon) \right| \leq \frac{C}{\sqrt{n}},$$

The constant here is known to be between 0.4 and 0.5.

Theorem .5 (Large deviations). *For every interval $A \subseteq \mathbb{R}$,*

$$\lim_{n \rightarrow \infty} -\frac{1}{n} \log \mathbf{P} \left\{ \frac{1}{n} \sum_{i=1}^n X_i \in A \right\} = \inf_{x \in A} I(x),$$

where $I(\cdot)$ is the rate function.

Thus assuming $10000 = \infty$, we can say that for $x > \frac{1}{2}$,

$$\mathbf{P}(S_n > nx) = \exp(-nI(x)).$$

We have written $I(x)$ in place of $\inf_{y>x} I(y)$ because I happens to be decreasing in this case.

These were largely asymptotic results. The two simplest concentration inequalities are non-asymptotic.

Theorem .6 (Markov's inequality). *If X is a non-negative random variable, then for any $t > 0$,*

$$\mathbf{P}(X \geq t) \leq \frac{\mathbf{E}[X]}{t}.$$

Theorem .7 (Chebyshev's inequality). *If X is a random variable with finite variance, then for any $\delta > 0$,*

$$\mathbf{P}(|X - \mu| \geq \delta) \leq \frac{\sigma^2}{\delta^2}.$$

Chebyshev's inequality gives that the number of heads is $5000 \pm \sqrt{\frac{2500}{0.01}}$ with probability at least 0.99. (The uncertainty is 500 tosses.)

In general, CLT and Chebyshev's inequality give results

$$n \mathbf{E} X_1 \pm \sqrt{n \text{Var } X_1} \sqrt{\log \frac{2}{\delta}},$$

$$\text{and } n \mathbf{E} X_1 \pm \sqrt{n \text{Var } X_1} \sqrt{\frac{1}{\delta}}$$

with probability at least $1 - \delta$, respectively.

The Chernoff bound, for iid Bernoulli random variables, gives

$$n \mathbf{E} X_1 \pm C \sqrt{n \operatorname{Var} X_1} \sqrt{\log \frac{1}{\delta}}$$

for some constant C .

Roadmap

We'll study Chernoff and Cramér bounds, which lead to Hoeffding's, Bennett's and Bernstein's inequalities.

We'll look at the coupon collector's problem.

Another question with applications in bioinformatics and `git diffing` is the following.

Question .8. *Given two iid finite sequences (X_1, \dots, X_n) and (Y_1, \dots, Y_n) , what is the length of a longest common subsequence?*

A common subsequence of length k is a sequence of length k that is a (not necessarily contiguous) subsequence of both X and Y .

Theorem .9 (Talagrand's principle). *A random variable that depends smoothly on a large number of random variables satisfies Chernoff-type bounds.*

Methods

- Chernoff-type bounds
- Tensorization techniques to break a function of random variables into its constituents using the (1) martingale, (2) Effron-Stein, and (3) entropy methods.
- Applying isoperimetric inequalities to probability spaces.
- Transportation method

Applications

- Johnson-Lindenstrauss lemma
- Hypercontractivity
- Bounds on the performance of LASSO
- Blowing-up lemmas
- Bin-packing, coupon collector, birthday paradox, LCS, et cetera.