Moments and Deviations

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Recap Tail bounds

Tail bounds Markov's inequality Chebyshev's Inequality



Main points

Probability spaces, Random Variable,



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Probability spaces, Random Variable, Distribution of a random variable (pmf),

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Probability spaces, Random Variable, Distribution of a random variable (pmf), Expected Value,

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Probability spaces, Random Variable, Distribution of a random variable (pmf), Expected Value, Variance,

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Main points

Probability spaces, Random Variable, Distribution of a random variable (pmf), Expected Value, Variance, Samples of randomized algorithms.

Tail bounds

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Note

The tail bounds of a random variable X are concerned with the probability that it deviates significantly from its expected value E[X] on a run of the experiment.

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Example

Consider the experiment of tossing a fair coin *n* times. What is the probability that the number of heads exceeds $\frac{3}{4} \cdot n$?

Markov's inequality

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Let X be a non-negative random variable and let c > 0 be a positive constant. Then,

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$$\sum_{0 \le x < c} x \cdot P(X = x) + \sum_{x \ge c} x \cdot P(X = x)$$

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$$\ge \sum_{x \ge c} x \cdot P(X = x)$$

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$$\begin{aligned} F[X] &= \sum_{x} x \cdot P(X = x) \\ &= \sum_{0 \le x < c} x \cdot P(X = x) + \sum_{x \ge c} x \cdot P(X = x) \\ &\ge \sum_{x \ge c} x \cdot P(X = x) \\ &\ge \sum_{x \ge c} c \cdot P(X = x) \end{aligned}$$

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Let X be a non-negative random variable and let c>0 be a positive constant. Then, $P(X \geq c) \leq \frac{E[X]}{c}.$

Proof.

 \Rightarrow

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$$= c \cdot P(X \ge c)$$

$$\cdot P(X \ge c) \le \frac{E[X]}{c}$$

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Markov's inequality Chebyshev's Inequality

Markov's Inequality (contd.)

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Let X be a non-negative random variable and let c > 0 be a positive constant. Then, $P(X \ge c \cdot E[X]) \le \frac{1}{c}$.

$$P(X \ge \frac{3n}{4}) = P(X \ge \frac{3}{2} \cdot \frac{n}{2})$$

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$$\le \frac{1}{\frac{3}{2}}$$

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$$P(|X - E[X]| \ge a) = P(|X - E[X]|^2 \ge a^2)$$

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Let X be a random variable (not necessarily positive). Then, $P(|X - E[X]| \ge a) \le \frac{Var[X]}{a^2}$.

Proof.

$$\begin{array}{ll} \mathcal{P}(|X - E[X]| \geq a) & = & \mathcal{P}(|X - E[X]|^2 \geq a^2) \\ & \leq & \frac{E[(X - E[X])^2]}{a^2} \end{bmatrix}, \, \text{Markov's inequality} \end{array}$$

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Proof.

$$P(|X - E[X]| \ge a) = P(|X - E[X]|^2 \ge a^2)$$

$$\le \frac{E[(X - E[X])^2]}{a^2}], \text{ Markov's inequality}$$

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Chebyshev's theorem is alternatively stated as:

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Note

 $\begin{array}{l} \mbox{Chebyshev's theorem is alternatively stated as:} \\ P(|X - E[X]| \geq a \cdot E[X]) \leq \frac{\text{Var}[X]}{(a \cdot E[X])^2}. \end{array}$

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$$P(X \ge \frac{3n}{4}) = P(X - \frac{n}{2} \ge \frac{n}{4})$$
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Chebyshev's inequality (contd.)

$$P(X \ge \frac{3n}{4}) = P(X - \frac{n}{2} \ge \frac{n}{4})$$
$$\leq P(|X - \frac{n}{2}| \ge \frac{n}{4})$$
$$= P(|X - E[X]| \ge \frac{1}{2}E[X]$$

Chebyshev's inequality (contd.)

$$\begin{array}{lll} P(X \geq \frac{3n}{4}) &=& P(X - \frac{n}{2} \geq \frac{n}{4}) \\ &\leq& P(|X - \frac{n}{2}| \geq \frac{n}{4}) \\ &=& P(|X - E[X]| \geq \frac{1}{2}E[X] \\ &\leq& \frac{\frac{n}{4}}{(\frac{1}{2})^2 \cdot (\frac{n}{2})^2} \end{array}$$

Chebyshev's inequality (contd.)

$$P(X \ge \frac{3n}{4}) = P(X - \frac{n}{2} \ge \frac{n}{4})$$

$$\le P(|X - \frac{n}{2}| \ge \frac{n}{4})$$

$$= P(|X - E[X]| \ge \frac{1}{2}E[X])$$

$$\le \frac{\frac{n}{4}}{(\frac{1}{2})^2 \cdot (\frac{n}{2})^2}$$

$$= \frac{4}{n}$$

The coupon collecting problem

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Remark

Let X denote the number of coupons to be collected in order to ensure that we have one coupon of each type.

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Remark

Let X denote the number of coupons to be collected in order to ensure that we have one coupon of each type. We have shown that $E[X] = n \cdot H_n$, where H_n is the n^{th} harmonic number.

Tail bounds for coupon collecting

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Markov

 $P(X \ge 2 \cdot n \cdot H_n)$

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Chebyshev

What do we need?

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What do we need? Var[X].

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What do we need? **Var**[X]. Observe that **Var**[X] = $\sum_{i=1}^{n}$ **Var**[X_i], where X_i is the random variable which counts the number of coupons to be drawn assuming that (*i* - 1) distinct types have already been drawn, in order to draw a coupon of a new type.

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What do we need? **Var**[X]. Observe that **Var**[X] = $\sum_{i=1}^{n}$ **Var**[X_i], where X_i is the random variable which counts the number of coupons to be drawn assuming that (i - 1) distinct types have already been drawn, in order to draw a coupon of a new type. For a geometric variable X_i with parameter p, we know that **Var**[X_i] = $\frac{1-p_i}{p_i^2} \le \frac{1}{p_i^2}$.

But recall that $p_i = \frac{n-i+1}{n}$.

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But recall that $p_i = \frac{n-i+1}{n}$. Therefore, $\frac{1}{p_i} = \frac{n}{n-i+1}$.

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$$P(X \ge 2 \cdot n \cdot H_n) \le \frac{1}{2}$$

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But recall that $p_i = \frac{n-i+1}{n}$. Therefore, $\frac{1}{p_i} = \frac{n}{n-i+1}$. Hence,

$$\operatorname{Var}[X] = \sum_{i=1}^{n} \operatorname{Var}[X_i]$$

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But recall that $p_i = \frac{n-i+1}{n}$. Therefore, $\frac{1}{p_i} = \frac{n}{n-i+1}$. Hence,

$$Var[X] = \sum_{i=1}^{n} Var[X_i]$$
$$\leq \sum_{i=1}^{n} \frac{1}{p_i^2}$$

Tail bounds for coupon collecting (contd.)

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$$\operatorname{Var}[X] \leq \sum_{i=1}^{n} (\frac{n}{n-i+1})^2$$

Tail bounds for coupon collecting (contd.)

$$Par[X] \leq \sum_{i=1}^{n} (\frac{n}{n-i+1})^2$$

= $n^2 \cdot \sum_{i=1}^{n} (\frac{1}{n-i+1})^2$

Tail bounds for coupon collecting (contd.)

$$\begin{aligned} \mathsf{Var}[X] &\leq \sum_{i=1}^{n} (\frac{n}{n-i+1})^2 \\ &= n^2 \cdot \sum_{i=1}^{n} (\frac{1}{n-i+1})^2 \\ &= n^2 \cdot \sum_{i=1}^{n} \frac{1}{i^2} \end{aligned}$$

Tail bounds for coupon collecting (contd.)

Chebyshev (contd.)

$$Var[X] \leq \sum_{i=1}^{n} \left(\frac{n}{n-i+1}\right)^{2}$$
$$= n^{2} \cdot \sum_{i=1}^{n} \left(\frac{1}{n-i+1}\right)^{2}$$
$$= n^{2} \cdot \sum_{i=1}^{n} \frac{1}{i^{2}}$$
$$\leq n^{2} \cdot \frac{\pi^{2}}{6}$$

Tail bounds for coupon collecting (contd.)

Analysis (contd.)

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$$P(X \ge 2 \cdot n \cdot H_n) =$$

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Tail bounds for coupon collecting (contd.)

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$$P(X \ge 2 \cdot n \cdot H_n) = P((X - n \cdot H_n) \ge n \cdot H_n)$$

$$\leq P(|X - n \cdot H_n| \ge n \cdot H_n)$$

Tail bounds for coupon collecting (contd.)

Analysis (contd.)

$$\begin{array}{lll} \mathsf{P}(X \geq 2 \cdot n \cdot H_n) & = & \mathsf{P}((X - n \cdot H_n) \geq n \cdot H_n) \\ & \leq & \mathsf{P}(|X - n \cdot H_n| \geq n \cdot H_n) \\ & \leq & \frac{n^2 \cdot \frac{\pi^2}{6}}{(n \cdot H_n)^2} \end{array}$$

Tail bounds for coupon collecting (contd.)

Analysis (contd.)

$$P(X \ge 2 \cdot n \cdot H_n) = P((X - n \cdot H_n) \ge n \cdot H_n)$$

$$\leq P(|X - n \cdot H_n| \ge n \cdot H_n)$$

$$\leq \frac{n^2 \cdot \frac{\pi^2}{6}}{(n \cdot H_n)^2}$$

$$\in O(\frac{1}{\ln^2 n})$$

Tail bounds (first principles)

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First principles

Focus on a coupon of type *i*.

Tail bounds (first principles)

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$$(1-\frac{1}{n})^{n\cdot\ln n+c\cdot n} =$$

Tail bounds (first principles)

First principles

$$\left(1-\frac{1}{n}\right)^{n\cdot\ln n+c\cdot n} = \left(1-\frac{1}{n}\right)^{n\cdot\left(\ln n+c\right)}$$

Tail bounds (first principles)

First principles

$$(1-\frac{1}{n})^{n\cdot\ln n+c\cdot n} = (1-\frac{1}{n})^{n\cdot(\ln n+c)}$$

$$< e^{-1\cdot(\ln n+c)}$$

Tail bounds (first principles)

First principles

$$(1-\frac{1}{n})^{n\cdot\ln n+c\cdot n} = (1-\frac{1}{n})^{n\cdot(\ln n+c)}$$
$$\leq e^{-1\cdot(\ln n+c)}$$
$$= \frac{1}{e^c \cdot n}$$

Tail bounds (first principles)

First principles

Focus on a coupon of type *i*. What is the probability that a coupon of type *i* has not been drawn after $n \cdot \ln n + c \cdot n$ trials?

$$(1-\frac{1}{n})^{n\cdot\ln n+c\cdot n} = (1-\frac{1}{n})^{n\cdot(\ln n+c)}$$
$$\leq e^{-1\cdot(\ln n+c)}$$
$$= \frac{1}{e^c \cdot n}$$

What is the probability that a coupon of any type has not been drawn after $n \cdot \ln n + c \cdot n$ trials?

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What is the probability that a coupon of any type has not been drawn after $n \cdot \ln n + c \cdot n$ trials? At most e^{-c} .

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What is the probability that a coupon of any type has not been drawn after $n \cdot \ln n + c \cdot n$ trials? At most e^{-c} . Hence, the probability that a coupon of some type is not picked after $2 \cdot n \cdot \ln n$ trials is at most

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Focus on a coupon of type *i*. What is the probability that a coupon of type *i* has not been drawn after $n \cdot \ln n + c \cdot n$ trials?

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What is the probability that a coupon of any type has not been drawn after $n \cdot \ln n + c \cdot n$ trials? At most e^{-c} . Hence, the probability that a coupon of some type is not picked after $2 \cdot n \cdot \ln n$ trials is at most $e^{-\ln n} = \frac{1}{n}$.

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What is the probability that a coupon of any type has not been drawn after $n \cdot \ln n + c \cdot n$ trials? At most e^{-c} . Hence, the probability that a coupon of some type is not picked after $2 \cdot n \cdot \ln n$ trials is at most $e^{-\ln n} = \frac{1}{n}$. Moral of the story:

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What is the probability that a coupon of any type has not been drawn after $n \cdot \ln n + c \cdot n$ trials? At most e^{-c} . Hence, the probability that a coupon of some type is not picked after $2 \cdot n \cdot \ln n$ trials is at most $e^{-\ln n} = \frac{1}{n}$. Moral of the story: First principle bounds are always better than cookie cutter bounds.