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Markov Inequality

X is a random variable, $X \in \{0, 1, 2, \dots\}$

$$E(X) < \infty$$

For any $t \geq 1$

$$P_r(X \geq t) \leq \frac{E(X)}{t}$$

$$\begin{aligned}
 E(X) &= \underbrace{E(X) \int_{-\infty}^{\infty} P_r(X < t) dt}_{\text{a.c.}} \\
 &\quad + E(X) \int_{-\infty}^{\infty} P_r(X > t) dt \\
 &= E(X) \int_{-\infty}^{\infty} P_r(X < t) dt \\
 &\quad + \int_{-\infty}^{\infty} P_r(X > t) dt
 \end{aligned}$$

Ex. sample: If $t = 1$

$$P_r(X \geq 1) \leq E(X)$$

First
Moment
Method

m balls (distinguishable)

$\rightarrow n$ boxes

$Z = \#$ of empty boxes

$$m \geq (1+\epsilon) n \log_e n$$

$$\mathbb{E}(Z) = n \left(1 - \frac{1}{n}\right)^m$$

$$1+x \leq e^x$$

$$\leq n e^{-m/n}$$

$$\leq n e^{-(1+\epsilon) \log n}$$

$$\leq$$

$$n$$

$$e^{-1-\epsilon}$$

$$\rightarrow$$

$$n^{-1-\epsilon} \rightarrow 0$$

$$P(Z \geq 1) \leq n^{-\epsilon}$$

$\mathcal{F}$ empty

box

$$\underline{\text{Variance}} \quad E(Z) = \mu$$

$$\text{Var}(Z) = E(|Z - \mu|^2)$$

$$\begin{aligned} &= \dots \\ &= E(Z^2) - \mu^2 \end{aligned}$$

$$Z = B_{n,p} \quad \mu = np \quad k(k-1) + k$$

$$\text{Var}(B_{n,p}) = \sum_{k=1}^n k^2 \binom{n}{k} p^k (1-p)^{n-k} - (np)^2$$

$\in \mathbb{Z}^2$

$$= \sum_{k=1}^n k(k-1) \binom{n}{k} p^k (1-p)^{n-k} + np - (np)^2$$

$$\frac{n!}{(k-2)!(n-k)!} = n(n-1) \binom{n-2}{k-2}$$

$$= n(n-1)p^2 \sum_{k=2}^n \binom{n-2}{k-2} p^{k-2} (1-p)^{n-k} + np - (np)^2$$

$$(1-p+p)^{n-2}$$

$$= n(n-1)p^2 + np - n^2p^2$$

$$np(1-p)$$

Chebyshev Inequality

$$\sigma = \sqrt{\text{Variance}}$$

Standard
deviation

$$P(|Z - \mu| \geq k\sigma) \leq \frac{1}{k^2}$$

$\uparrow$

$$P(Z - \mu)^2 \geq k^2 \sigma^2$$
$$\leq \frac{E((Z - \mu)^2)}{k^2 \sigma^2} = \frac{\sigma^2}{k^2 \sigma^2} = \frac{1}{k^2}$$

$\swarrow$

Markov Inequality

$$P\left[|B_{n,p} - np| \geq \epsilon np\right]$$

$$\leq \frac{\text{Var}(B_{n,p})}{\epsilon^2 n^2 p^2}$$

$$= \frac{np(1-p)}{\epsilon^2 n^2 p^2} \leq \frac{1-p}{\epsilon^2 np}$$

Hoeffding Inequality

(Simple Case: Chernoff bounds)

$X_1, X_2, \dots, X_n$ are independent

$$P_i(X_i = 1) = \frac{1}{2}$$

$$P_i(X_i = -1) = \frac{1}{2}$$

$$X = X_1 + X_2 + \dots + X_n$$

$$E(X) = 0$$

$$P_r(|X| \geq t) \leq 2e^{-t^2/2n}$$

For $\lambda > 0$ (choose k later)

$$P_r(X \geq t) = P_r(e^{\lambda X} \geq e^{\lambda t})$$

$$\xrightarrow{\text{Markov inequality}} \leq e^{-\lambda t} E(e^{\lambda X})$$

$$E(e^{\lambda X}) = E\left(\prod_{i=1}^n e^{\lambda x_i}\right)$$

x_i independent $\rightarrow$

$$= \prod_{i=1}^n E(e^{\lambda x_i})$$

$E(x_i) = E(x_1) = \dots = E(x_n) = 0$

$$E(e^{\lambda X_n}) = \frac{e^{-\lambda} + e^{\lambda}}{2}$$

$$= 1 + \frac{\lambda^2}{2!} + \frac{\lambda^4}{4!} + \dots$$

$$< e^{\lambda^2/2}$$

$$P_r(X \geq t) \leq e^{-\lambda t} \cdot (e^{\lambda^2/2})^n$$

$$= e^{\frac{-\lambda t + \lambda^2 n/2}{}}$$

choose λ to minimize this.

$-\lambda t + \lambda^2 n = 0$

$$\begin{aligned}
 P_1(X \geq t) &\leq e^{-\lambda t} \cdot (e^{\lambda^2/2})^n \\
 &= e^{\frac{-\lambda t + \lambda^2 n/2}{\text{choose } \lambda \text{ to minimize}}} \\
 &\quad \boxed{-t + \lambda n = 0}
 \end{aligned}$$

thus.

$$P_{\lambda t} \lambda = t/n$$

$$P_1(X \geq t) \leq e^{-t^2/2n}$$