

Chebychev's Inequality

For a random variable X with mean value \bar{X} and variance σ_X^2 , it states that

$$P\{|X - \bar{X}| \geq \varepsilon\} \leq \sigma_X^2 / \varepsilon^2 \quad \text{for any } \varepsilon > 0$$

Proof:

$$\begin{aligned} \sigma_X^2 &= \int_{-\infty}^{\infty} (x - \bar{X})^2 f_X(x) dx \geq \int_{|x - \bar{X}| \geq \varepsilon} (x - \bar{X})^2 f_X(x) dx \\ &\geq \varepsilon^2 \int_{|x - \bar{X}| \geq \varepsilon} f_X(x) dx = \varepsilon^2 P\{|X - \bar{X}| \geq \varepsilon\} \end{aligned}$$

Markov's Inequality

For a nonnegative random variable X , and any positive real constant a , we have

$$P\{X \geq a\} \leq E[X] / a \quad a > 0$$

Proof:

$$\begin{aligned} E[X] &= \int_0^{\infty} x f_X(x) dx \geq \int_{x \geq a} x f_X(x) dx \geq \int_{x \geq a} a f_X(x) dx \\ &\geq a \int_{x \geq a} f_X(x) dx = a P\{X \geq a\} \end{aligned}$$

Chernoff's inequality and Bound

Let X be any random variable, nonnegative or not. For any real $v > 0$, it is clear that

$$\exp [v (x - a)] \geq u(x - a) \quad (1)$$

where $u(\cdot)$ is the unit-step function and a is an arbitrary real constant.

$$\text{Since } P\{X \geq a\} = \int_a^{\infty} f_X(x) dx = \int_{-\infty}^{\infty} f_X(x) u(x - a) dx, \quad (2)$$

We have

$$P\{X \geq a\} \leq \int_{-\infty}^{\infty} f_X(x) e^{\nu(x-a)} dx = e^{-\nu a} M_X(\nu)$$

where

$$M_X(\nu) = E[e^{\nu X}] \quad (\text{moment generating function}).$$

For any random variable X , and any real constant a ,

$$P\{X \geq a\} \leq e^{-\nu a} M_X(\nu) \quad (3)$$

Equation (3) is called *Chernoff's inequality*. Because the right side is a function of parameter ν , it can be minimized with respect to this parameter. The minimum value is called *Chernoff's bound*.