

Chebyshev's Inequality

$$P(|X - E(X)| \geq k) \leq \frac{\text{Var}(X)}{k^2}, \text{ for } k > 0$$

Why? Use Markov's Inequality with $(X - E(X))^2$ instead of X and k^2 instead of a .

Markov's Inequality says:

$$P(\underbrace{(X - E(X))^2}_{\text{both positive}} \geq k^2) \leq \frac{E((X - E(X))^2)}{k^2}$$

$$\boxed{P(|X - E(X)| \geq k) \leq \frac{\text{Var}(X)}{k^2}}$$

another version, use $k\sigma_X$ instead of k
 $k^2 \text{Var}(X)$ instead of k^2 , we get:

$$P(|X - E(X)| \geq k\sigma_X) \leq \frac{\text{Var}(X)}{k^2 \text{Var}(X)}$$

$$P(|X - E(X)| \geq k\sigma_X) \leq \frac{1}{k^2}.$$

i.e. The probability that X is more than k standard deviations away from its mean is no larger than $\frac{1}{k^2}$.

Equivalently,

$$P(|X - E(X)| \leq k\sigma_X) \geq 1 - \frac{1}{k^2}$$

i.e. the probability X is within k standard deviations of its mean is at least $1 - \frac{1}{k^2}$.