

## 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  $1 - \frac{1}{k^2}$ .