

Discussion 5

Fall 2017

1. Convergence of Exponentials

Let X_1, X_2, \dots be i.i.d. $\text{Exponential}(\lambda)$ random variables. Show that

$$\frac{X_n}{\ln n} \rightarrow 0 \quad \text{in probability as } n \rightarrow \infty.$$

Solution:

Fix $\varepsilon > 0$.

$$\mathbb{P}\left(\frac{X_n}{\ln n} \geq \varepsilon\right) = \mathbb{P}(X_n \geq \varepsilon \ln n) = \exp(-\lambda \varepsilon \ln n) = n^{-\lambda \varepsilon} \rightarrow 0$$

as $n \rightarrow \infty$.

2. Exponential Bounds

Let $X \sim \text{Exponential}(\lambda)$. For $x > \lambda^{-1}$, calculate bounds on $\mathbb{P}(X \geq x)$ using Markov's Inequality, Chebyshev's Inequality, and the Chernoff Bound.

Solution:

Since $\mathbb{E}[X] = \lambda^{-1}$, Markov's Inequality gives

$$\mathbb{P}(X \geq x) \leq \frac{\mathbb{E}[X]}{x} = \frac{1}{\lambda x},$$

and from $\text{var } X = \lambda^{-2}$, Chebyshev's Inequality gives

$$\begin{aligned} \mathbb{P}(X \geq x) &= \mathbb{P}(X - \lambda^{-1} \geq x - \lambda^{-1}) \leq \mathbb{P}(|X - \lambda^{-1}| \geq x - \lambda^{-1}) \\ &\leq \frac{\text{var } X}{(x - \lambda^{-1})^2} = \frac{1}{(\lambda x - 1)^2}. \end{aligned}$$

For the Chernoff Bound, for any $s > 0$,

$$\mathbb{P}(X \geq x) = \mathbb{P}(\exp(sX) \geq \exp(sx)) \leq \frac{M_X(s)}{\exp(sx)} = \frac{\lambda}{(\lambda - s) \exp(sx)}.$$

We wish to optimize this bound over $s > 0$. It suffices to maximize the denominator $(\lambda - s) \exp(sx)$. Taking derivatives,

$$-\exp(sx) + x(\lambda - s) \exp(sx) = 0,$$

so $1 = x(\lambda - s)$, that is, $s = \lambda - x^{-1}$. Thus,

$$\begin{aligned} \mathbb{P}(X \geq x) &\leq \frac{\lambda}{(\lambda - (\lambda - x^{-1})) \exp((\lambda - x^{-1})x)} = \frac{\lambda}{x^{-1} \exp(\lambda x - 1)} \\ &= \lambda x \exp(-(\lambda x - 1)). \end{aligned}$$

Observe that the Chernoff Bound is the only one which decreases exponentially with x , which is the true behavior: $\mathbb{P}(X \geq x) = \exp(-\lambda x)$.