第六周 常见随机变量的期望与方差和应用实例

6.2 几何分布的期望与方差

几何分布
$$X \sim Ge(p)$$
, $0 , $P(X = k) = p \cdot (1 - p)^{k-1}$, $k = 1, 2, \cdots$$

$$E(X) = \sum_{k=1}^{\infty} k \cdot P(X = k) = \sum_{k=1}^{\infty} k \cdot p(1-p)^{k-1} = \sum_{k=1}^{\infty} [(k-1)+1] \cdot p(1-p)^{k-1}$$

$$= \sum_{k=1}^{\infty} (k-1) \cdot p(1-p)^{k-1} + \sum_{k=1}^{\infty} p(1-p)^{k-1} = (1-p) \sum_{k=2}^{\infty} (k-1) \cdot p(1-p)^{k-2} + 1$$

$$= (1-p) \sum_{j=1}^{\infty} j \cdot p(1-p)^{j-1} + 1 = (1-p) E(X) + 1$$

$$E(X) = (1-p)E(X)+1 \Rightarrow E(X) = \frac{1}{p}$$

$$E(X^{2}) = \sum_{k=1}^{\infty} k^{2} P(X = k) = \sum_{k=1}^{\infty} k^{2} p (1-p)^{k-1} = \sum_{k=1}^{\infty} \left[(k-1)^{2} + 2(k-1) + 1 \right] \cdot p (1-p)^{k-1}$$

$$= \sum_{k=1}^{\infty} \left[(k-1)^{2} + 2k - 1 \right] \cdot p (1-p)^{k-1}$$

$$= \sum_{k=1}^{\infty} (k-1)^{2} \cdot p (1-p)^{k-1} + 2 \sum_{k=1}^{\infty} k \cdot p (1-p)^{k-1} - \sum_{k=1}^{\infty} p (1-p)^{k-1}$$

$$= (1-p) \sum_{j=1}^{\infty} j^{2} \cdot p (1-p)^{j-1} + 2 E(X) - 1 = (1-p) E(X^{2}) + 2 E(X) - 1$$

$$E(X^{2}) = (1-p)E(X^{2}) + 2E(X) - 1 \implies E(X^{2}) = \frac{2}{p^{2}} - \frac{1}{p}$$

$$Var(X) = E(X^{2}) - E(X)^{2} = \frac{1 - p}{p^{2}}$$

备注:

计算泊松分布和几何分布随机变量的期望过程中,做了诸如将k 拆分为(k-1)+1,以及 k^2 拆分为k(k-1)+k 或 $(k-1)^2+2(k-1)+1$ 等等的等价变形处理,这一类的拆分是概率统计计算中常用的处理方法,目的是为了凑出随机变量的分布列求和或期望等的求和式,利用求和式的概率意义和已知的概率结果往往可以简化计算。
