

## Lec 9&10

### Independence of r.v.

R.V.  $X, Y$  are said to be independent if:

$$P(X \leq x, Y \leq y) = P(X \leq x) P(Y \leq y) \text{ for } \forall x, y \in \mathbb{R}$$

# Def: R.V.  $X_1, \dots, X_n$  are independent if

$$P(X_1 \leq x_1, \dots, X_n \leq x_n) = P(X_1 \leq x_1) \cdots P(X_n \leq x_n)$$

for  $\forall x_1, \dots, x_n \in \mathbb{R}$



I.I.D.  $\Rightarrow$  independent and identically distributed

关于是否独立 & 是否相同分布, 则  $X, Y$  关系就有四种

Binomial Distribution = 二项分布

Theorem: if  $X \sim \text{Bin}(n, p)$ , 则可视为  $n$  个独立 Bernoulli 实验, 则  $X = X_1 + \dots + X_n$ ,  $X_i$  是 i.i.d.  $\text{Bern}(p)$

Theorem: If  $X \sim \text{Bin}(n, p)$ ,  $Y \sim \text{Bin}(m, p)$ ,  $X$  与  $Y$  独立, 则  $X+Y \sim \text{Bin}(m+n, p)$  \* 因为  $X, Y$  独立

$$\begin{aligned} \text{Proof: } P(X+Y=k) &= \sum_{j=0}^k P(X+Y=k | X=j) P(X=j) \\ &= \sum_{j=0}^k P(Y=k-j) P(X=j) = \sum_{j=0}^k \binom{m}{k-j} p^{k-j} (1-p)^{m-k+j} \binom{n}{j} p^j (1-p)^{n-j} \\ &= \sum_{j=0}^k \binom{m}{k-j} \binom{n}{j} p^k (1-p)^{m+n-k} \\ &= \binom{m+n}{k} p^k (1-p)^{m+n-k} \end{aligned}$$

\* story proof:  $m$  男  $n$  女, 共取  $k$  人, 可视为男取  $k-j$  人, 女取  $j$  人,  $j \in [0, k]$

Conditional Independence of R.V.s

Def: R.V.  $X, Y$  are conditionally independent given an r.v.  $Z$  if for all  $x, y \in \mathbb{R}$  and all  $z$  in the support of  $Z$ ,

$$P(X \leq x, Y \leq y | Z=z) = P(X \leq x | Z=z) P(Y \leq y | Z=z)$$

Conditional PMF

Def: For any discrete r.v.s  $X$  &  $Z$ , function  $P(X=x | Z=z)$ , when considered as a function of  $x$  for fixed  $z$ , is called the conditional PMF of  $X$  given  $Z=z$  ( $z$  是 constant,  $x$  在变)



## Binomial & Hypergeometric

Theorem: If  $X \sim \text{Bin}(n, p)$ ,  $Y \sim \text{Bin}(m, p)$ ,  $X, Y$  间独立, 则  $X+Y=r$  的条件分布为  $\text{HGeom}(n, m, r)$

Theorem: If  $X \sim \text{HGeom}(w, b, n)$  and  $N = w+b \rightarrow \infty$  s.t.  $p = w/(w+b)$  remains fixed, 则  $X$  的 PMF 收敛至  $\text{Bin}(n, p)$  PMF

Proof 1:  $X+Y \sim \text{Bin}(n+m, p)$  \* 为什么看这个 PMF? Story: HGeom:  $n+m$  中取  $r$ ,  $r$  中有  $x$  个  $X$  的分布  $\Rightarrow X+Y=r$  下,  $X=x$  的条件 PMF

$$P(X=x | X+Y=r)^* = \frac{P(X=x, X+Y=r)}{P(X+Y=r)}$$

$$= \frac{P(X=x, Y=r-x)}{P(X+Y=r)} = \frac{P(X=x) P(Y=r-x)}{P(X+Y=r)}$$

$$= \frac{\binom{n}{x} p^x (1-p)^{n-x} \binom{m}{r-x} p^{r-x} (1-p)^{m-r+x}}{\binom{n+m}{r} p^r (1-p)^{m+n-r}} = \frac{\binom{n}{x} \binom{m}{r-x}}{\binom{n+m}{r}}$$

## Lec 9 Expectation

Def: The expected value of a discrete r.v.  $X$

$$E(X) = \sum_{j=1}^{\infty} x_j P(X=x_j)$$

Theorem: If  $X$  &  $Y$  是有相同分布的离散 r.v.s, 则  $E(X) = E(Y)$

Theorem: Linearity:  $E(X+Y) = E(X) + E(Y)$   
 $E(cX) = cE(X)$

Expectation via Survival Function:

Let  $X$  be a nonnegative integer-valued r.v. Let  $F$  be the CDF of  $X$ , and  $G(x) = 1 - F(x) = P(X > x)$ .

The  $G$  is called the survival function of  $X$ . 则:

$$E(X) = \sum_{n=0}^{\infty} G(n)$$

KOKUYO





这条性质很灵性:  $x=1$  只覆盖一次,  $x=2$  二次, ...,  $x=i$ ,  $i$  次

Law of the unconscious statistician (LOTUS)

$$E(g(x)) = \sum_x g(x) P(X=x)$$

Variance & Standard Deviation

$$\text{Var}(X) = E(X^2) - [E(X)]^2$$

$$\text{SD}(X) = \sqrt{\text{Var}(X)}$$

Geometric Distribution

考虑一系列 Bernoulli 试验, 概率为  $p \in (0, 1)$ , 试验一直做直至一次成功。X 是在成功前失败的次数, 则认为  $X \sim \text{Geom}(p)$

$$\text{PMF: } P(X=k) = q^n p$$

Property: Memory less

$$P(X \geq n+k | X \geq k) = P(X \geq n)$$

vice versa: 若离散 r.v. X 满足  $P(X \geq n+k | X \geq k) = P(X \geq n)$

则  $X \sim \text{Geom}(p)$

Proof: ①  $k=0$   $\vee$  ②  $k \geq 1$ :  $P(X \geq n+k | X \geq k)$

$$= \frac{P(X \geq n+k, X \geq k)}{P(X \geq k)} = \frac{P(X \geq n+k)}{P(X \geq k)} = P(X \geq n)$$

令  $G(n) = P(X \geq n)$ ,  $G(0) = 1$ , 则欲证:  $G(n+k) = G(n)G(k)$

$$G(n) = 1 - \sum_{i=0}^{n-1} P(X=i) = 1 - P(q^0 + q^1 + \dots + q^{n-1})$$

$$= 1 - p \cdot \frac{1 - q^n}{1 - q} = q^n$$

$$\therefore q^{n+k} = q^n \cdot q^k \text{ 显然成立}$$





Properties of Indicator R.V.  $I\{A\} = \begin{cases} 1, & \text{若 } A \text{ 发生} \\ 0, & \text{若 } A \text{ 不发生} \end{cases}$

$$(IA)^k = IA \quad I_{A^c} = 1 - IA \quad I_{A \cap B} = IAIB \quad I_{A \cup B} = IA + IB - IAIB$$

Theorem:  $P(A) = E(IA)$ , 架起了 Probability & Expectation 桥梁  
但左右两者计算量不同! 在多变量  $\cap \cup$  下, 利用 I.r.v 的 property 与期望线性性,  $E(I)$  计算更方便!

Eg: 证: Boole's Inequality:

$$P\left(\bigcup_{i=1}^n A_i\right) \leq \sum_{i=1}^n P(A_i)$$

欲证:  $I(A_1 + \dots + A_n) \leq I(A_1) + \dots + I(A_n)$

① if LHS = 0, ✓

② if LHS = 1, 说明  $A_1, \dots, A_n$  间至少有一个  $A_i$ ,  $I(A_i) = 1$   
 $\therefore$  RHS  $\geq 1$ , RHS  $\geq$  LHS

## Lec 11

Moments and Indicators:

设 events  $A_1, \dots, A_n$ , indicators  $I_j, j = 1, \dots, n$ .

$X = \sum_{j=1}^n I_j$  代表 events 发生数量

$\binom{X}{2} = \sum_{i < j} I_i I_j$  代表 发生的不同事件对的数量

$$\text{则 } E\left(\binom{X}{2}\right) = \sum_{i < j} P(A_i \cap A_j)$$

$$E(X^2) = 2 \sum_{i < j} P(A_i \cap A_j) + E(X)$$

$$\text{Var}(X) = 2 \sum_{i < j} P(A_i \cap A_j) + E(X) - (E(X))^2$$

$$\Delta: E\left(\binom{X}{k}\right) = \binom{n}{k} p^k$$

Poisson Distribution:  $e^{-\lambda} \lambda^k$ .

$$\text{PMF: } P(X=k) = \frac{e^{-\lambda} \lambda^k}{k!}, \quad X \sim \text{Pois}(\lambda).$$





核心:  $\sum_{k=0}^{\infty} \frac{e^{-\lambda} \lambda^k}{k!} = 1$  . 有点无穷级数味道

Theorem: ①  $E(X) = \lambda$   

$$E(X) = \sum_{k=0}^{\infty} k \cdot \frac{e^{-\lambda} \lambda^k}{k!} = e^{-\lambda} \cdot \lambda \sum_{k=1}^{\infty} \frac{\lambda^{k-1}}{(k-1)!}$$

$$= e^{-\lambda} \cdot \lambda \cdot e^{\lambda} = \lambda$$

②  $Var(X) = \lambda$

Poisson 近似: Let  $A_1, \dots, A_n$  是  $p_j = P(A_j)$ . 当  $n$  很大,  $p_j$  很小,  $A_j$  独立时;

$$X = \sum_{j=1}^n I(A_j) \text{ 近似于 } \text{Pois}(\lambda).$$
 其中  $\lambda = \sum_{j=1}^n p_j$

Theorem: if  $X \sim \text{Pois}(\lambda_1)$   $Y \sim \text{Pois}(\lambda_2)$ .  $X$  与  $Y$  独立,  
 则  $X+Y \sim \text{Pois}(\lambda_1 + \lambda_2)$

且:  $X+Y = n$  的  $X$  条件分布为  $\text{Bin}(n, \lambda_1 / (\lambda_1 + \lambda_2))$

Theorem: If  $X \sim \text{Bin}(n, p)$ , 并令  $n \rightarrow \infty, p \rightarrow 0$ , s.t.  $\lambda = np$  remains fixed, 则  $X$  的 PMF 收敛至  $\text{Pois}(\lambda)$  PMF

### Distance between Two Probability Distributions

Def:  $\mu, \nu$  两个分布间的 total variation distance:

$$d_{TV}(\mu, \nu) = \|\mu - \nu\|_{TV}$$

$$= \frac{1}{2} \sum_{x \in \Omega} |\mu(x) - \nu(x)|$$

### The Law of Rare Events.

有独立 r.v.  $Y_1, \dots, Y_n$  s.t.  $1 \leq m \leq n, P(Y_m=1) = p_m$   
 &  $P(Y_m=0) = 1 - p_m, S_n = \sum_{i=1}^n Y_i$ . 设:  

$$\sum_{m=1}^n p_m \rightarrow \lambda \in (0, \infty) \text{ as } n \rightarrow \infty$$



No.

Date

$$A: \max_{1 \leq m \leq n} P_{m \rightarrow 0} \quad \text{as } n \rightarrow \infty$$

$$\text{则 } d_{TV}(S_n, \text{Poi}(\lambda)) \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

Probability Generating Function.  
The PGF of r.v.  $X$  with PMF  $p_k = P(X=k)$  is the generating function of PMF:

$$E(t^X) = \sum_{k=0}^{\infty} p_k t^k$$

