

2-1-22 Sets & Probs

- If S has n elementaries \Rightarrow # event = 2^n
- Set properties: ① Commutative / ② Associative / ③ Distributive
④ $(A \cup B)^c = A^c \cap B^c$, $(A \cap B)^c = A^c \cup B^c$
- Axioms of Probability:
 - ① $P(A) \geq 0$
 - ② $P(S) = 1$
 - ③ If $A \cap B = \emptyset$, then $P(A \cup B) = P(A) + P(B)$
- Extensions from Axiom:
 - ① $P(A^c) = 1 - P(A)$
 - ② $0 \leq P(A) \leq 1$
 - ③ $P(\emptyset) = 0$
 - ④ $P(A \cup B) = P(A) + P(B) - P(A \cap B)$
 - ⑤ three events: $P(A \cup B \cup C) = P(A) + P(B) + P(C)$
 $- P(A \cap B) - P(A \cap C) - P(B \cap C) + P(A \cap B \cap C)$
 - ⑥ $P(A \cup B) \leq P(A) + P(B)$
 - ⑦ If $A \subseteq B$ then $P(A) \leq P(B)$

2-3 Counting Methods

- Sampling WITH Replacement, WITH Ordering
 $\# N$ elements total, select $k = n^k$
- NO Replacement, WITH ordering [P]
 $\# N$ total, select out $k = \frac{N!}{(N-k)!}$
 ie. $N \times (N-1) \times (N-2) \times \dots \times (N-k+1)$
 \rightarrow Full Permutation = $N!$
- No Replacement, NO Ordering (C)
 $\# N$ total, select $k = \binom{N}{k} = \frac{N!}{k!(N-k)!}$
- When split $\binom{n}{k}$ into multiple groups:
 Ex. n balls in bucket, split into categories:
 n_1 belong to A^{class} , $\# B=n_2$, $\# C=n_3 \dots \# k=n_k$
 $\text{and } n_1+n_2+n_3+\dots+n_k = n$

different ways to PUT BALLS IN A ROW
 $= \binom{n}{n_1, n_2, \dots, n_k} = \frac{n!}{n_1! n_2! \dots n_k!} \xrightarrow{\text{divided by}} \frac{1}{\text{RBRBB 相概率}}$

2-4 Conditional

- Conditional Def: $P(A|B) = \frac{P(A \cap B)}{P(B)}$ for $P(B) > 0$
- Th. Total Probability: $P(A) = \sum_{i=1}^n P(A|B_i) P(B_i)$
- Baye's Rule:

$$P(B_i|A) = \frac{P(A|B_i)}{P(A)} = \frac{P(A|B_i) \cdot P(B_i)}{\sum_{k=1}^n P(A|B_k) \cdot P(B_k)}$$

2.5 Independent

- Independent Def: $P(A \cap B) = P(A) P(B)$
 means: $P(A|B) = P(A)$, $P(B|A) = P(B)$
- Pairwise Independent: if A, B, C are pairwise IND:
 $P(A \cap B) = P(A) P(B)$. $A \overset{\text{ind}}{\leftrightarrow} B$, $B \overset{\text{ind}}{\leftrightarrow} C$, $A \overset{\text{ind}}{\leftrightarrow} C$
- Jointly Independent:
 based on previous example, also need:
 $P(A \cap B \cap C) = P(A) P(B) P(C)$

2.6 Sequential Experiments

- Bernoulli: single trial, success or fail
 $P(\text{success}) = p$
- Binomial: n trials, within them, k success
 $P(X=k) = \binom{n}{k} p^k (1-p)^{n-k}$
- Multinomial l : n trials, every trial has M classes, result B_j with probability p_j
 probability that every class happen time (k_1, k_2, \dots, k_m):
 $P(k_1, \dots, k_m) = \frac{n!}{k_1! k_2! \dots k_m!} p_1^{k_1} p_2^{k_2} \dots p_m^{k_m}$
 \rightarrow Ex: 把飞镖, 9次, 落在区域 1, 2, 3 概率为 $0.2, 0.3, 0.5$
 每个区域正好落 3 次的概率: $\frac{9!}{3!3!3!} (0.2)^3 (0.3)^3 (0.5)^3$
- Geometric: repeat trials until first success
 $P(\text{first success when } m^{\text{th}} \text{ trial happens})$:
 $P(m) = (1-p)^{m-1} p$, $m=1, 2, \dots$

$$P(m) = (1-p)^{m-1} p, \quad m=1, 2, \dots$$

$$P(\text{超过 } k \text{ 次才成功}) = P(m > k) = (1-p)^k$$

$$P(\text{Dependent Experiments})$$

Ex: two urn experiment

$$P(s_0, s_1, \dots, s_n) = P(s_0) P(s_1 | s_0) P(s_2 | s_1, s_0) \dots P(s_n | s_{n-1}, \dots, s_0)$$

$$\bullet \sum_{k=0}^{\infty} q^k = \frac{1}{1-q}, |q| < 1 \quad \bullet \sum_{k=1}^{\infty} k q^{k-1} = \frac{1}{(1-q)^2}$$

3.1-3.2 Random Variable, Discrete, and pmf

- Probability Mass Function (PMF): $p_x(x) = P(X=x)$ ① ② ③ Used for pmf valid check
- PMF Props: ① $p_x(x) \geq 0$ ② $\sum_{x \in S_x} p_x(x) = 1$ ③ $P(X=B) = \sum_{x \in S_B} p_x(x)$
- Valid RV: 对所有 $B \subseteq \mathbb{R}$, 事件 $\{z : X(z) \in B\}$ 都属于 \mathcal{F} . 小X在原本S的事件中存在 $p_x(x) = 0$ for all $x \notin S_x$

3.3&3.5 Expectation & Variance, and Important RVs

- Expected Value:
 $E[X] = \mu_x = \sum_{x \in S_x} x p_x(x)$
- Expected RV: if $Z = g(x)$ then we have:
 $E[Z] = E[g(x)] = \sum_k g(x_k) p_x(x_k)$
- E Formulas:
 if $Z = a g(x) + b h(x) + c$, then: $E[Z] = aE[g(x)] + bE[h(x)] + c$
 - $\rightarrow E[c] = c$
 - $\rightarrow E[aX] = aE[X]$
 - $\rightarrow E[X+c] = E[X]+c = E[g(x)] + E[h(x)]$

$$\bullet \text{Variance: } \text{VAR}[X] = E[(X - \mu_x)^2] = \sum_x (x - \mu_x)^2 p_x(x)$$

$$\bullet \text{standard dev: } \sigma_x = \sqrt{\text{VAR}(X)} \quad \text{VAR}[X] = E[X^2] - (E[X])^2 \leftarrow$$

$$\bullet \text{VAR props: } \text{VAR}[X+C] = \text{VAR}[X] \quad \text{VAR}[cX] = c^2 \text{VAR}[X]$$

分布	pmf	$E[X]$	$\text{Var}[X]$	Poisson
Bernoulli(p)	$p^x (1-p)^{1-x}, x=0,1$	p	$p(1-p)$	$P(X=k)$
Binomial(n, p)	$\binom{n}{k} p^k (1-p)^{n-k}$	np	$np(1-p)$	
Geometric(p)	$(1-p)^{k-1} p$	$1/p$	$(1-p)/p^2$	
Poisson(λ)	$e^{-\lambda} \lambda^k / k!$	λ	λ	$\lambda \cdot \text{平均次数}$
Uniform($1 \dots L$)	$1/L$	$(L+1)/2$	$(L^2-1)/12$	$K = \text{实际次数}$
Zipf(L)	$(1/c_L) (1/k)$	L/c_L	$L(L+1)/(2c_L) - (L/c_L)^2$	

3.4 Conditional RVs

- Def: $p_{X|C}(x) = P(X=x | C) = \frac{P(\{X=x\} \cap C)}{P(C)}$
- Conditional Expectation: $E[X|B] = \sum_x x p_{X|B}(x)$
- Conditional VAR: $\text{VAR}[X|B] = E[(X - E[X|B])^2 | B] = \sum_x (x - E[X|B])^2 p_{X|B}(x)$
- Law of Total Expectation: $E[X] = \sum_i E[X|B_i] P(B_i)$
- $\sum_{j=1}^{n-1} r^j = \frac{1-r^n}{1-r}, |r| < 1$

若 C 本身是“关于 X 的事件”，例如 $C = \{a \leq X \leq b\}$,
 则条件 pmf 可直接由 X 的 pmf 算出:

$$p_{X|C}(x) = \begin{cases} \frac{p_X(x)}{P(C)}, & x \in C \\ 0, & x \notin C \end{cases}$$