

Skills

一. 多变量问题, 则只抓两个!

题欲证明: " $f(p_1, p_2, \dots, p_n)$ 当且仅当 $p_1 = p_2 = \dots = p_n = ?$ 有 min"

这种结论时, 只抓 p_1 & p_2 , 证明在 $p_1 = p_2 = \frac{p_1 + p_2}{2}$ 下有较小值

i.e., $f\left(\frac{p_1 + p_2}{2}, \frac{p_1 + p_2}{2}, p_3, \dots\right) \leq f(p_1, p_2, p_3, \dots)$

其中 p_3, \dots, p_n 为常数, p_1, p_2 为变量

这样, 欲 min, 则 $p_1 = p_2, p_2 = p_3, \dots, p_1 = \dots = p_n$ 下有 min

二. 用概率与期望证明存在性

关键 好分数原则: 设 X 为随机选择的某个对象的得分

思想: 如果 $E(X) \geq C$, 则存在一个得分不小于 C 的对象

可能性原则: 设 A 为一个事件表示在集合中随机选择的某个对象具有某种特定的性质。若 $P(A) > 0$, 则存在这种性质的对象

例: 100个人, 分到15个规模为20人的组中, 每个人服务于3个组。

证明: 存在两个组, 它们至少有3个公共的人。

证: X 表示为任取两个组, 组内公共人的数量

令 $X = I_1 + I_2 + \dots + I_{100}$, I_j 代表第 j 个人同时在两个组里

$$\therefore E(X) = 100 E(I_j)$$

$$\text{而 } E(I_j) = P(\text{第 } j \text{ 个人同时在两个组中}) = \frac{C_3^2}{C_{15}^2} = \frac{1}{35}$$

$$\therefore E(X) = \frac{100}{35} = \frac{20}{7}$$

\therefore 至少有两个组, 有3个公共人

三. 灵活利用 Adam's Law

$E[E(Y|X)] = E(Y)$, 可以让 Y 与 X 缠上关系, 且可以凑题目欲求的项!



Eg. 2022 Fall T5 (a).

欲证 $E[(\theta - E[\theta|X])h(X)] = 0$ for any function $h(\cdot)$

Proof: $\theta - E[\theta|X]$ 与 $h(X)$ 不独立, 但在 X 条件下便独立

$$E(E[(\theta - E[\theta|X])h(X)|X]).$$

$$= h(X) E[(\theta - E[\theta|X])|X].$$

$$= h(X) \cdot [E(\theta|X) - E[E[\theta|X]|X]] = 0.$$

Eg. 2022 Fall T7 (a).

$$[P(T_1 = T_2) < 1]$$

$T_2 = E(T_1|R)$. 欲证 $E[(T_2 - \theta)^2] < E[(T_1 - \theta)^2]$.

右式如何凑 $T_1|R$ 呢? Conditioning on R !

$$\text{LHS} = E[(E[T_1|R] - \theta)^2]$$

$$= E[E[(T_1 - \theta)^2|R]]$$

$$\text{RHS} \stackrel{\text{Adam}}{=} E[E[(T_1 - \theta)^2|R]]$$

Jensen with Condition: $E(g(X)|R) \leq g(E(X|R))$ \square

$E[g(X)|R] \geq g(E[X|R])$, 凸 \Rightarrow 即可证毕

四. MMSE \Rightarrow "后验" α 找分布

$$f_{Y|X} = \frac{f_{X|Y} \cdot f_Y}{\sum \leftarrow \text{定}} \propto f_{X|Y}(\alpha|y) f_Y(y), \text{ 往往它好求!}$$

找分布, 然后 $E(f_{X|Y}(\alpha))$ 就是分布的 mean (公式)

Eg. 2022 Fall T5 (b) 求 MMSE $E[\theta|X]$.

$$\text{Solution: } f_{\theta|X}(\theta|x) \propto f_{\theta}(\theta) f_{X|\theta}(x|\theta) = \theta^\alpha (1-\theta)^{n-\alpha}$$

$$\text{则 } \theta|X=x \sim \text{Beta}(x+1, n-x+1) \text{ 故 } E[\theta|X=x] = \frac{x+1}{n+2}$$

$$\Delta: \text{Beta}(a, b) \propto x^{a-1} (1-x)^{b-1}, E(x) = \frac{a}{a+b}$$



Eq2. Fall 2022 T6. (a) 令 $X = \sum_{i=1}^n X_i$

$$J_{p|X}(p|X=k) \propto J_{X|p}(X=k|p) \cdot J_p(p) = p^k (1-p)^{n-k}$$

$\therefore \sim \text{Beta}(k+1, n-k+1)$, 其分布仅依赖于 k, n .

(b) $P(X_{n+1}=1 | X=k)$ 实质: $X=k \Rightarrow p \Rightarrow X_{n+1}$, $p \in [0, 1]$.

$$= \int_0^1 P(X_{n+1}|p) J_{p|X}(p|X=k) dp$$

$$= \int_0^1 p \cdot J_{p|X}(p|X=k) dp = E(\text{Beta}(k+1, n-k+1)) = \frac{k+1}{n+2}$$

为什么可以不管不顾, 只需 Bayes 找好算的, 然后发现分布上“类似的部分”就断言它是 XX 分布?

这实际上是归一化的手法; Bayes Net 中使用了相同的做法

五. Conjugate Prior

二项分布的 PMF 可视为 Beta 分布, 这是一种巧合吗?

其实这样一类的分布可称为一个 distribution family

若对 p 从 Beta 先验分布, 而 data 是 p 下的二项分布, 则从先验推后验, 我们没有离开 Beta 分布 family。We say that Beta is the conjugate prior of the Binomial.

Conjugate (Likelihood) Model: Discrete

Sample Space	Sampling Dist	Conjugate Prior	Posterior
$X = \{0, 1\}$	Bern(θ)	Beta(α, β)	Beta($\alpha+n\bar{x}, \beta+n\bar{x}$)
$X = Z_+$	Pois(λ)	Gamma(α, β)	Gamma($\alpha+n\bar{x}, \beta+n$)
$X = Z_{++}$	Geom(θ)	Gamma(α, β)	Gamma($\alpha+n, \beta+n\bar{x}$)

Continuous:

	Expo(θ)	Gamma(α, β)	Gamma($\alpha+n, \beta+n\bar{x}$)
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六. MAP, MMSE & LLSE

A coin, heads up: θ . 令先验: PDF $f_{\theta} \sim \text{Unif}(0,1)$
 n tosses, k heads up. (Denote X : num of heads up)

$$\textcircled{1} \text{ MAP: } f_{\theta|X=k}(\theta) = \frac{f_{X=k|\theta}(\theta) \cdot f_{\theta}(\theta)}{\sum_{\theta} \rightarrow \text{constant}}$$

$$\propto f_{X=k|\theta}(\theta) = \binom{n}{k} \theta^k (1-\theta)^{n-k} \propto \theta^k (1-\theta)^{n-k} = g(\theta)$$

$$\text{欲 } g(\theta) \text{ max} \Rightarrow \log g(\theta) \text{ max} \Rightarrow k \ln \theta + n-k \ln(1-\theta) = \psi(\theta) \text{ max}$$

$$\psi'(\theta) = \frac{k}{\theta} - \frac{n-k}{1-\theta} = 0, \quad \theta = \frac{k}{n}$$

$$\text{i.e., } \hat{\theta} = \arg \max_{\theta} f_{X=k|\theta} = \frac{X}{n} \quad (X: \text{r.v., } \theta = \frac{k}{n} \text{ when } X=k)$$

$$\textcircled{2} \text{ MMSE: } f_{\theta|k} \propto \theta^k (1-\theta)^{n-k} \Rightarrow f_{\theta|k} \sim \text{Beta}(k+1, n-k+1)$$

$$\therefore E[f_{\theta|k}] = \frac{k+1}{n+2}$$

$\textcircled{3}$ LLSE:

$$L[\theta|X] = E[\theta] + \frac{\text{Cov}(\theta, X)}{\text{Var}(X)} [X - E(X)]$$

$$E[\theta] = \frac{1}{2}, \quad \text{Cov}(\theta, X) = E[\theta X] - E[\theta]E[X]$$

$$E[\theta X] = E[E[\theta X|\theta]] = E[\theta \cdot E[X|\theta]] = nE[\theta^2] = n \int_0^1 \theta^2 d\theta = \frac{n}{3}$$

$$E[X] = E[E(X|\theta)] = E[n\theta] = \frac{1}{2}n$$

$$\therefore \text{Cov}(\theta, X) = \frac{n}{3} - \frac{1}{2} \cdot \frac{n}{2} = \frac{n}{12}$$

$$\text{又 } \text{Var}(X) = E[\text{Var}(X|\theta)] + \text{Var}[E[X|\theta]] \quad (\text{Eve's Law})$$

$$= E(n\theta(1-\theta)) + \text{Var}(n\theta) = n \cdot \left(\frac{1}{2} - \frac{1}{3}\right) + n^2 \cdot \frac{1}{12}$$

$$= \frac{n}{12}(n+2)$$

$$\therefore L[\theta|X] = \frac{1}{2} + \frac{\frac{n}{12}}{\frac{n}{12}(n+2)} \left(X - \frac{n}{2}\right) = \frac{X+1}{n+2}$$



七. 矩母用途

① 切比诺夫: $P(X \geq a) \leq \frac{E(e^{tX})}{e^{ta}}$

$E(e^{tX})$ 就是矩母; t 灵活取值可以用来凑欲证不等式 (2023 T9, 2022 T8)

② 应对 $E(\sin X)$ $E(\cos X)$

Eg: $X, Y \stackrel{i.i.d.}{\sim} N(0, 1)$, $Z_1 = \sin(X+Y)$ $Z_2 = \cos(X+Y)$ (2021 T9)
求 $E(Z_1)$ $E(Z_2)$ $\text{Var}(Z_1)$ $\text{Var}(Z_2)$

Solution: (X, Y) 是 BVN , 则 $X+Y \sim (0, 2) = Z$.

Δ : 对于 $aW+b$, $W \sim (\mu, \sigma^2)$, 欲求 $E[\cos(aW+b)]$ $E[\sin(aW+b)]$
 $\cos \theta = \frac{e^{i\theta} + e^{-i\theta}}{2}$, $E[\cos(aW+b)] = \frac{1}{2} (E[e^{i(aW+b)}] + E[e^{-i(aW+b)}])$
 $= \frac{1}{2} \cdot [e^{ib} E[e^{iaw}] + e^{-ib} E[e^{-iaw}]]$ 矩阵代 $t = \dots$

$E(e^{tW}) = e^{t\mu + \frac{1}{2}\sigma^2 t^2}$, 则 $E[e^{iaw}] = e^{i\mu a - \frac{1}{2}\sigma^2 a^2}$
 $E[e^{-iaw}] = e^{-i\mu a - \frac{1}{2}\sigma^2 a^2}$

$\therefore E = \frac{1}{2} \cdot e^{-\frac{1}{2}\sigma^2 a^2} \cdot (e^{i(\mu a + b)} + e^{-i(\mu a + b)})$
 $= e^{-\frac{1}{2}\sigma^2 a^2} \cdot \cos(a\mu + b)$

同理: $E[\sin(aW+b)] = e^{-\frac{1}{2}\sigma^2 a^2} \sin(a\mu + b)$

$\therefore E(Z_1) = 0$ $E(Z_2) = e^{-1}$

$E(Z_1^2) = \frac{1}{2} - \frac{1}{2} E[\cos(2W)] = \frac{1}{2} - \frac{1}{2} e^{-4}$

$E(Z_2^2) = \frac{1}{2} + \frac{1}{2} E[\cos(2W)] = \frac{1}{2} + \frac{1}{2} e^{-4}$

$\therefore \text{Var}(Z_1) = \frac{1}{2} - \frac{1}{2} e^{-4}$

$\text{Var}(Z_2) = \frac{1}{2} + \frac{1}{2} e^{-4} - e^{-2}$

八. min. max 换元

$\min\{X, Y\} = M$, $\max\{X, Y\} = N$, 则 $X+Y = M+N$, $-M+N = |X-Y|$

由其: $P(N \leq t) = P(X \leq t) P(Y \leq t) = F_X(t) F_Y(t)$

则 $P(N=t) = (F_X(t) F_Y(t))'$



(b) $X, Y \stackrel{i.i.d.}{\sim} N(0,1)$, $\neq \swarrow$

例: (a) $X, Y \stackrel{i.i.d.}{\sim} \text{Unif}(0,1)$, 求 $[\text{Corr}(\max(X, Y), \min(X, Y))]$

Solution: $M = \max(X, Y)$, $N = \min(X, Y)$

$$\text{Corr}(M, N) = \frac{E(MN) - E(M)E(N)}{\sqrt{\text{Var}(M)\text{Var}(N)}}$$

$$E(MN) = E(XY) = E(X)E(Y) = \frac{1}{4}$$

$$E(M) = \int_0^1 t \cdot p(M=t) dt, \quad p(M=t) = t^2 \therefore p(M=t) = 2t$$

$$= \int_0^1 2t^2 dt = \frac{2}{3} \quad \text{同理 } E(N) = \frac{1}{3}^*$$

$$E(M^2) = \int_0^1 t^2 \cdot 2t dt = \frac{1}{2} \therefore \text{Var}(M) = E(M^2) - E(M)^2 = \frac{1}{2} - \frac{4}{9} = \frac{1}{18}$$

$$*: p(N \geq t) = (1-t)^2, \quad p(N=t) = 2(1-t)$$

$$E(N) = \int_0^1 2t(1-t) dt = \left(t^2 - \frac{2}{3}t^3 \right) \Big|_0^1 = \frac{1}{3}$$

$$E(N^2) = \int_0^1 2(1-t)t^2 = \left(\frac{2}{3}t^3 - \frac{1}{2}t^4 \right) \Big|_0^1 = \frac{1}{6}$$

$$\text{Var}(N) = \frac{1}{6} - \frac{1}{9} = \frac{1}{18} \therefore \text{Corr}(MN) = 18 \cdot \left(\frac{1}{4} - \frac{2}{9} \right) = 18 \cdot \frac{1}{36} = \frac{1}{2}$$

(b) $M = \max(X, Y)$, $N = \min(X, Y)$, $E(MN) = E(XY) = 0$

但此处 $E(M)$ 在用 CDF \Rightarrow PDF 不再好用! 因为正态分布 CDF 未知!

$$E(M) = E(X|X>Y) + E(Y|Y>X) = 2E(X|X>Y) \text{ (symmetric)}$$

欣赏: 令 $W = X+Y$, $Z = X-Y$, 则 $X = \frac{W+Z}{2}$, $E(X|X>Y) = \frac{1}{2}E(W+Z|Z>0)$

$$= \frac{1}{2}E(W|Z>0) + \frac{1}{2}E(Z|Z>0), \text{ 那么 } W \text{ 与 } Z \text{ 独立吗?}$$

$$\text{Cov}(X+Y, X-Y) = \text{Cov}(X, X) - \text{Cov}(Y, Y) + \text{Cov}(Y, X) - \text{Cov}(X, Y)$$

$$= \text{Var}(X) - \text{Var}(Y) = 0, \text{ 一般 } \text{Cov}=0 \text{ 为 } i.i.d., \text{ 但是有定理:}$$

$BVN(X, Y)$, if $\text{Corr}(X, Y) = \text{Cov}(X, Y) = 0 \Rightarrow X, Y$ 独立

$$\therefore W \sim (0, 2), \quad Z \sim (0, 2), \quad E(W|Z>0) = 0$$

$$E(Z|Z>0) = \int_0^{\infty} z f(z) dz = \int_0^{\infty} z \frac{1}{\sqrt{2\pi} \cdot \sigma} e^{-\frac{z^2}{2\sigma^2}} dz$$

$$\text{令 } t = \frac{z}{\sigma}, \quad z = \sigma t, \quad dz = \sigma dt, \quad \text{则:}$$

$$I = \int_0^{\infty} \sigma t \cdot \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{t^2}{2}} \sigma dt = \frac{\sigma}{\sqrt{2\pi}} \int_0^{\infty} t e^{-\frac{t^2}{2}} dt = \sigma \sqrt{\frac{2}{\pi}}$$

$$\sigma = \sqrt{2}, \quad \text{则 } \frac{1}{2}E(Z|Z>0) = \sqrt{\frac{2}{\pi}} = E(M), \text{ 同理, } E(N) = -\frac{1}{\sqrt{\pi}}$$

$$\text{而: } M^2 + N^2 = X^2 + Y^2, \quad E(M^2) = E(N^2) = 1 \therefore \text{Var}(M) = \text{Var}(N) = 1 - \left(\pm \sqrt{\frac{2}{\pi}} \right)^2 = 1 - \frac{2}{\pi}$$

$$\text{Campus: } \therefore \text{Corr}(M, N) = \frac{\frac{1}{\sqrt{\pi}}}{1 - \frac{2}{\pi}} = \frac{1}{\pi - 2}$$

