

STAT 5010 Tutorial 2*

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Definitions

1. A parametric family $\{P_\theta : \theta \in \Theta\}$ dominated by a σ -finite measure ν on (Ω, \mathcal{F}) is called an *exponential family* if and only if

$$\frac{dP_\theta}{d\nu}(\omega) = \exp\{[\eta(\theta)]^\tau T(\omega) - \xi(\theta)\} h(\omega), \quad \omega \in \Omega,$$

where $\exp\{x\} = e^x$, T is a random p -vector with a fixed positive integer p , η is a function from Θ to \mathcal{R}^p , h is a nonnegative Borel function on (Ω, \mathcal{F}) , and $\xi(\theta) = \log \left\{ \int_{\Omega} \exp\{[\eta(\theta)]^\tau T(\omega)\} h(\omega) d\nu(\omega) \right\}$

2. In an exponential family, consider the reparameterization $\eta = \eta(\theta)$ and

$$f_\eta(\omega) = \exp\{\eta^\tau T(\omega) - \zeta(\eta)\} h(\omega), \quad \omega \in \Omega,$$

where $\zeta(\eta) = \log \left\{ \int_{\Omega} \exp\{\eta^\tau T(\omega)\} h(\omega) d\nu(\omega) \right\}$. This is the *canonical form* for the family, which is not unique. The new parameter η is called the *natural parameter*. The new parameter space $\Xi = \{\eta(\theta) : \theta \in \Theta\}$, a subset of \mathcal{R}^p , is called the *natural parameter space*. An exponential family in canonical form is called a *natural exponential family*. If there is an open set contained in the natural parameter space of an exponential family, then the family is said to be of *full rank*.

3. A measurable function of $X, T(X)$, is called a *statistic* if $T(X)$ is a known value whenever X is known, i.e., the function T is a known function.
4. Let X be a sample from an unknown population $P \in \mathcal{P}$, where \mathcal{P} is a family of populations. A statistic $T(X)$ is said to be *sufficient* for $P \in \mathcal{P}$ (or for $\theta \in \Theta$ when $\mathcal{P} = \{P_\theta : \theta \in \Theta\}$ is a parametric family) if and only if the conditional distribution of X given T is known (does not depend on P or θ).
5. Let T be a sufficient statistic for $P \in \mathcal{P}$. T is called a *minimal sufficient* statistic if and only if, for any other statistic S sufficient for $P \in \mathcal{P}$, there is a measurable function ψ such that $T = \psi(S)$ a.s. \mathcal{P}
6. A statistic $V(X)$ is said to be *ancillary* if its distribution does not depend on the population P and *first-order ancillary* if $E[V(X)]$ is independent of P .
7. A statistic $T(X)$ is said to be *complete* for $P \in \mathcal{P}$ if and only if, for any Borel f , $E[f(T)] = 0$ for all $P \in \mathcal{P}$ implies $f(T) = 0$ a.s. \mathcal{P} . T is said to be *boundedly complete* if and only if the previous statement holds for any bounded Borel f .

Propositions and Theorems

1. If η_0 is an interior point of the natural parameter space, then the m.g.f. ψ_{η_0} of $P_{\eta_0} \circ T^{-1}$ is finite in a neighborhood of 0 and is given by

$$\psi_{\eta_0}(t) = \exp\{\zeta(\eta_0 + t) - \zeta(\eta_0)\}$$

2. (The factorization theorem) Suppose that X is a sample from $P \in \mathcal{P}$ and \mathcal{P} is a family of probability measures on $(\mathcal{R}^n, \mathcal{B}^n)$ dominated by a σ -finite measure ν . Then $T(X)$ is sufficient for $P \in \mathcal{P}$ if and only if there are nonnegative Borel functions h (which does not depend on P) on $(\mathcal{R}^n, \mathcal{B}^n)$ and g_P (which depends on P) on the range of T such that

$$\frac{dP}{d\nu}(x) = g_P(T(x))h(x).$$

3. Suppose that \mathcal{P} contains p.d.f.'s f_P w.r.t. a σ -finite measure and that there exists a sufficient statistic $T(X)$ such that, for any possible values x and y of X , $f_P(x) = f_P(y)\phi(x, y)$ for all P implies $T(x) = T(y)$, where ϕ is a measurable function. Then $T(X)$ is minimal sufficient for $P \in \mathcal{P}$.
4. If P is in an exponential family of full rank with p.d.f.'s given as in Definition. 2, then $T(X)$ is complete and sufficient for $\eta \in \Xi$.

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5. (Basu's theorem) Let V and T be two statistics of X from a population $P \in \mathcal{P}$. If V is ancillary and T is boundedly complete and sufficient for $P \in \mathcal{P}$, then V and T are independent w.r.t. any $P \in \mathcal{P}$.
6. A complete and sufficient statistic is also minimal sufficient. However, a minimal sufficient statistic is not necessarily complete.

Question 1

1. Let X and Y be two random variables such that Y has the binomial distribution with size N and probability π and, given $Y = y$, X has the binomial distribution with size y and probability p . Suppose that $p \in (0, 1)$ and $\pi \in (0, 1)$ are unknown and N is known. Show that (X, Y) is minimal sufficient for (p, π) .

Solution:

Let $A = \{(x, y) : x = 0, 1, \dots, y, y = 0, 1, \dots, N\}$. The joint probability density of (X, Y) with respect to the counting measure is

$$\begin{aligned} & \binom{N}{y} \pi^y (1 - \pi)^{N-y} \binom{y}{x} p^x (1 - p)^{y-x} I_A \\ &= \exp \left\{ x \log \frac{p}{1-p} + y \log \frac{\pi(1-p)}{1-\pi} + N \log(1 - \pi) \right\} \binom{N}{y} \binom{y}{x} I_A. \end{aligned}$$

Hence, (X, Y) has a distribution from an exponential family of full rank ($0 < p < 1$ and $0 < \pi < 1$). This implies that (X, Y) is minimal sufficient for (p, π) .

Question 2

1. Let X_1, \dots, X_n be i.i.d. random variables from P_θ , the uniform distribution $U(\theta, \theta + 1)$, $\theta \in \mathcal{R}$. Prove that $T = (X_{(1)}, X_{(n)})$ is minimal sufficient.

Solution: Suppose that $n > 1$. The joint Lebesgue p.d.f. of (X_1, \dots, X_n) is

$$f_\theta(x) = \prod_{i=1}^n I_{(\theta, \theta+1)}(x_i) = I_{(\underline{x_{(n)}}-1, \underline{x_{(1)}})}(\theta), \quad x = (x_1, \dots, x_n) \in \mathcal{R}^n,$$

where $x_{(i)}$ denotes the i th smallest value of x_1, \dots, x_n . By the factorization theorem, $T = (X_{(1)}, X_{(n)})$ is sufficient for θ . Note that

$$x_{(1)} = \sup \{\theta : f_\theta(x) > 0\} \quad \text{and} \quad x_{(n)} = 1 + \inf \{\theta : f_\theta(x) > 0\}.$$

If $S(X)$ is a statistic sufficient for θ , then by the factorization theorem, there are Borel functions h and g_θ such that $f_\theta(x) = g_\theta(S(x))h(x)$. For x with $h(x) > 0$, $x_{(1)} = \sup \{\theta : g_\theta(S(x)) > 0\}$ and $x_{(n)} = 1 + \inf \{\theta : g_\theta(S(x)) > 0\}$. Hence, there is a measurable function ψ such that $T(x) = \psi(S(x))$ when $h(x) > 0$. Since $h > 0$ a.s. \mathcal{P} , we conclude that T is minimal sufficient.

Proofs for Some propositions

- Prop. 3.

Proof. From Bahadur (1957), there exists a minimal sufficient statistic $S(X)$. The result follows if we can show that $T(X) = \psi(S(X))$ a.s. \mathcal{P} for a measurable function ψ . By the factorization theorem, there are Borel functions g_P and h such that $f_P(x) = g_P(S(x))h(x)$ for all P . Let $A = \{x : h(x) = 0\}$. Then $P(A) = 0$ for all P . For x and y such that $S(x) = S(y)$, $x \notin A$ and $y \notin A$,

$$\begin{aligned} f_P(x) &= g_P(S(x))h(x) \\ &= g_P(S(y))h(x)h(y)/h(y) \\ &= f_P(y)h(x)/h(y) \end{aligned}$$

for all P . Hence $T(x) = T(y)$. This shows that there is a function ψ such that $T(x) = \psi(S(x))$ except for $x \in A$. It remains to show that ψ is measurable. Since S is minimal sufficient, $g(T(X)) = S(X)$ a.s. \mathcal{P} for a measurable function g . Hence g is one-to-one and $\psi = g^{-1}$. The measurability of ψ follows from Theorem 3.9 in Parthasarathy (1967). ■

- Prop. 4.

Proof. Obviously, T is sufficient. Suppose that there is a function f such that $E[f(T)] = 0$ for all $\eta \in \Xi$. Then,

$$\int f(t) \exp \{ \eta^\tau t - \zeta(\eta) \} d\lambda = 0 \quad \text{for all } \eta \in \Xi,$$

where λ is a measure on $(\mathcal{R}^p, \mathcal{B}^p)$. Let η_0 be an interior point of Ξ . Then

$$\int f_+(t) e^{\eta^\tau t} d\lambda = \int f_-(t) e^{\eta^\tau t} d\lambda \quad \text{for all } \eta \in N(\eta_0),$$

where $N(\eta_0) = \{ \eta \in \mathcal{R}^p : \|\eta - \eta_0\| < \epsilon \}$ for some $\epsilon > 0$. In particular,

$$\int f_+(t) e^{\eta_0^\tau t} d\lambda = \int f_-(t) e^{\eta_0^\tau t} d\lambda = c.$$

If $c = 0$, then $f = 0$ a.e. λ . If $c > 0$, then $c^{-1}f_+(t)e^{\eta_0^\tau t}$ and $c^{-1}f_-(t)e^{\eta_0^\tau t}$ are p.d.f.'s w.r.t. λ and this implies that their m.g.f.'s are the same in a neighborhood of 0. Thus, $c^{-1}f_+(t)e^{\eta_0^\tau t} = c^{-1}f_-(t)e^{\eta_0^\tau t}$, i.e., $f = f_+ - f_- = 0$ a.e. λ . Hence T is complete. ■

Prop 1. $\eta_0 \longleftrightarrow P_{\eta_0}$. $\boxed{T = T(X)}$ pdf.

$$\psi_{\eta_0}(t) = E \exp\{t^T T\} = \int \exp\{t^T T\} \exp\{\eta_0^T T(w) - \xi(\eta_0)\} h(w) dw.$$

$$= \int \exp\{(t + \eta_0)^T T(w) - \xi(\eta_0)\} h(w) dw.$$

$$= \frac{\int \exp\{(t + \eta_0)^T T(w)\} h(w) dw}{\exp(\xi(\eta_0))} \quad \checkmark$$

$$\exp(\xi(\eta)) = \int_{\Omega} \exp\{\eta^T T(w)\} h(w) dw. \quad = \frac{\exp\{\xi(\eta_0 + t)\}}{\exp\{\xi(\eta_0)\}}.$$

$$\exp(\xi(\eta + t)) = \int_{\Omega} \exp\{(\eta + t)^T T(w)\} h(w) dw. \quad = \exp\{\xi(\eta_0 + t) - \xi(\eta_0)\}.$$

Prop 6. Suppose T c.s. \boxed{S} minimal suff. $S = g(T)$.

Complete & sufficient stat. may not exist.

$$E[X|Y] = f(Y). \quad E[X|Y=y] = f(y).$$

$$\text{consider: } T - E[T|S] = T - \underbrace{E[T|g(T)]} \\ = T - \psi(g(T)).$$

$$\underbrace{E[T - E[T|S]]} = E[T] - \underbrace{E[E[T|S]]} \\ = E[T] - E[T] = 0.$$

$$\Rightarrow P(T - E[T|S] = 0) = 1. \quad \text{by completeness of } T.$$

$$\Rightarrow T = \underbrace{E[T|S]}_{\triangle} = f(S). \quad \Rightarrow T \text{ is min. suff.}$$

$$\underline{S = g(T). \quad E[f(S)] = E[f(g(T))] = 0. \quad f(g(T)) = 0. \Leftrightarrow f(S) = 0.}$$