

# Theory exam

Mars Gao

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## 1 Sufficient Statistics

**Theorem 1** (Factorization Theorem).  $T \equiv T(X)$  is sufficient for  $\mathcal{P}$  if and only if  $f_\theta(x)$  factors as follows:

$$f_\theta(x) = g_\theta(T(x)) \cdot h(x),$$

where  $g_\theta(T(x))$  depends on  $x$  only through  $T(x)$  and  $h(x)$  does not depend on  $\theta$ .

**Theorem 2** (Lehmann-Scheffe). Suppose that  $X$  has pdf  $f_\theta(x), \theta \in \Omega$  and that  $T^*(X)$  satisfies the following:

$$\frac{f_\theta(y)}{f_\theta(x)} \text{ is } \theta \text{ free} \iff T^*(x) = T^*(y).$$

## 2 Theory of UMVUE

**Theorem 3** (Rao-Blackwell-Lehmann-Scheffe (RBLS)). Let  $T = T(X)$  be complete and sufficient for  $\theta$ . If there exists at least one unbiased estimator  $\tilde{\tau} \equiv \tilde{\tau}(X)$  for  $\tau(\theta)$ , then there exists a unique UMVUE  $\hat{\tau} \equiv \hat{\tau}(T)$  for  $\tau(\theta)$ , namely,

$$\hat{\tau}(T) \equiv \mathbb{E}[\tilde{\tau}(X)|T].$$

**Corollary 1** (The UMVUE Supermarket). Let  $T \equiv T(X)$  be complete and sufficient for  $\theta$ . Then any function  $\phi(T)$  is the UMVUE of its expectation  $\mathbb{E}_\theta[\phi(T)] \equiv \tau(\theta)$ .

## 3 The Information Inequality

**Theorem 4** (The Cramer-Rao-Frechet Lower Bound in 1d). Let  $T \equiv T(X)$  be complete and suffic Assume that  $I_X(\theta) > 0$ . Then,

$$\text{Var}_\theta[T(X)] \geq \frac{\left(\frac{d}{d\theta} \mathbb{E}_\theta[T(X)]\right)^2}{I_X(\theta)},$$

where

$$I_X(\theta) = \mathbb{E}_\theta \left[ \left[ \frac{d \log f_\theta(X)}{d\theta} \right]^2 \right] = -\mathbb{E}_\theta \left[ \frac{d^2 \log f_\theta(X)}{d\theta^2} \right].$$

Equality holds iff  $\{f_\theta(x)\}$  is a 1-parameter exponential family of the form

$$f_\theta(x) = e^{A(\theta)} e^{B(\theta)T(x)} e^{c(x)}.$$

**Corollary 2** (Unbiased estimator CR-bound). *Suppose that  $T(X)$  is an unbiased estimator of  $\tau(\theta)$ , a smooth function of  $\theta$ . Then,*

$$\text{Var}_\theta(T(X)) \geq \frac{[\tau'(\theta)]^2}{I_X(\theta)}.$$

## 4 Hypothesis Testing

**Definition 1** (NP criterion). *Fix  $0 < \alpha < 1$ . A test  $\phi$  is level  $\alpha$  (size  $\alpha$ ) for  $H_0$  if*

$$\pi_\phi(\theta_0) \equiv \mathbb{E}_{\theta_0} [\phi(X)] \leq \alpha.$$

*A level  $\alpha$  test for  $H_0 : \theta \in \Omega_0$  is most powerful level  $\alpha$  for  $H_1$  if*

$$\pi_\phi(\theta_1) \equiv \mathbb{E}_{\theta_1} [\phi(X)] = \sup_{\phi'} \pi_{\phi'}(\theta_1).$$

**Theorem 5** (NP theorem). *Let  $\phi$  be a LR test of the form in previous definition with  $c < \infty$  and set  $\alpha = \mathbb{E}_{\theta_0} [\phi(X)]$ . Then  $\phi$  is a MP level  $\alpha$  test for testing  $H_0$  vs  $H_1$ . That is, if  $\phi'$  is any other test such that*

$$\mathbb{E}_{\theta_0} [\phi'(X)] \leq \alpha,$$

*then*

$$\mathbb{E}_{\theta_1} [\phi'(X)] \leq \mathbb{E}_{\theta_1} [\phi(X)].$$

**Theorem 6** (Wilks theorem). *Let  $X_1, \dots, X_n$  be i.i.d. samples from a regular family that satisfies Fisher-Cramer and Wald conditions. The MLE  $\hat{\theta}$  is CANE for  $\theta$ . Let  $\lambda_n$  be the Likelihood ratio test statistics for testing. Then, under  $H_0$ ,*

$$-2 \log \lambda_n \xrightarrow{d} \chi_{d-d_0}^2.$$

## 5 Checklist of examples

- 12.14 UMVUE of Uniform.
- 14.3 MLE of Uniform.
- 14.4 MLE of Gaussian  $\mu, \sigma$ .
- 12.12 UMVUE of Gaussian  $\mu, \sigma$ .
- 14.5 MLE of Cauchy.
- 14.25 Multinomial Multivariate MLE.
- 14.34 General location-scale family MLE.
- 13.7 UMVUE of Binomial.
- 12.13 UMVUE of Poisson.
- 11.11 Minimal sufficient statistics of exponential/truncation family.
- Proposition 12.1 and 12.2 for exponential family completeness.
- Theorem 18.6 NP hypothesis testing.
- 2019.1 Confidence interval asymptotic.
- 2019.3 Hypothesis testing LRT, p-value, posterior testing.
- 2019.1 MLE of Poisson.
- 2019.2 Unbiased estimation of Poisson.
- 2019.6 Uniform  $[-\theta, \theta^2]$ .
- 2020.2 Definition of consistent and unbiased.
- 2020.2 Bernoulli unbiased estimator.
- 2020.3 Multinomial. MLE, MSS.
- 2020.4 Sample median and MLE of Uniform (0, 1).
- 2020.5 2d unit disk variable transformation.
- 2020.6 Bayesian for what??
- 2018.1 Mixture of normal.
- 2018.2 Proof of Chebyshev.
- 2018.3 Poisson UMVUE.
- 2018.4 Logit model.

- 2018.5 Bernoulli MLE. Bayes estimator with MSE.
- 2018.6 MLE of Beta. Information lower bound.
- 2017.1 MGF of Normal and Normal mean.
- 2017.2 All hypothesis testing for exponential.
- 2017.3 Convergence in distribution and probability.
- 2017.4 Variable transformation.
- 2017.5 Exponential (change) MLE, information lower bound. Existence of MLE.
- 2017.6 Mixture of two distribution.
- 2016.1 Uniform k-th moment UMVUE.
- 2016.2 Binomial mean. Binomial cannot find unbiased (polynomial proof).
- 2016.3 Distribution Chi-square and testing.
- 2016.4 Linear model and lowest MSE.
- 2016.5 Multi-nomial, complete statistics, conditional distribution and UMVUE.
- 2016.6 Poisson with 0 events. Unbiased estimator.