

## Lecture Notes for STAT-F404

# 6 Hypothesis Testing: UMPU Tests

## 6.1 Unbiasedness and similarity

Let  $\mathcal{H}_0$  and  $\mathcal{H}_1$  characterize a testing problem in a parametric family indexed by  $\boldsymbol{\theta} \in \Theta \subset \mathbb{R}^k$ . Recall that a test  $\phi$  is an  $\alpha$ -level test if  $E_{\boldsymbol{\theta}}[\phi] \leq \alpha$  for any  $\boldsymbol{\theta} \in \mathcal{H}_0$ . It is not unnatural to require that a test rejects more often under  $\mathcal{H}_1$  than under  $\mathcal{H}_0$ , which leads to the concept of *unbiasedness*.

**Definition 1.** A test  $\phi$  for  $\mathcal{H}_0$  against  $\mathcal{H}_1$  is unbiased at level  $\alpha$  if (i)  $\phi$  is an  $\alpha$ -level test and (ii)  $E_{\boldsymbol{\theta}}[\phi] \geq \alpha$  for any  $\boldsymbol{\theta} \in \mathcal{H}_1$ .

The unbiasedness principle then consists in restricting to tests that are unbiased at level  $\alpha$ : a test  $\phi^*$  will be said to be *uniformly most powerful unbiased* (UMPU) at level  $\alpha$  if (a)  $\phi^*$  is unbiased at level  $\alpha$ , and (b) for any  $\phi$  that is unbiased at level  $\alpha$ , one has  $E_{\boldsymbol{\theta}}[\phi^*] \geq E_{\boldsymbol{\theta}}[\phi]$  for any  $\boldsymbol{\theta} \in \mathcal{H}_1$ .

Now, if the size/power function  $\boldsymbol{\theta} \mapsto E_{\boldsymbol{\theta}}[\phi]$  is continuous (as it is for any test in the framework of exponential families), then a necessary condition for unbiasedness is *similarity*.

**Definition 2.** A test  $\phi$  for  $\mathcal{H}_0$  against  $\mathcal{H}_1$  is similar at level  $\alpha$  (or  $\alpha$ -similar) if  $E_{\boldsymbol{\theta}}[\phi] = \alpha$  for any  $\boldsymbol{\theta} \in \bar{\mathcal{H}}$ , where  $\bar{\mathcal{H}} := \text{adh}(\mathcal{H}_0) \cap \text{adh}(\mathcal{H}_1)$  is the boundary between  $\mathcal{H}_0$  and  $\mathcal{H}_1$ .

Parallel as above, we say that a test  $\phi^*$  is uniformly most powerful in the class of  $\alpha$ -similar tests if (a)  $\phi^*$  is similar at level  $\alpha$ , and (b) for any  $\phi$  that is  $\alpha$ -similar, one has  $E_{\boldsymbol{\theta}}[\phi^*] \geq E_{\boldsymbol{\theta}}[\phi]$  for any  $\boldsymbol{\theta} \in \mathcal{H}_1$ . As an exercise, the reader can prove that the following statement holds as soon as  $\boldsymbol{\theta} \mapsto E_{\boldsymbol{\theta}}[\phi]$  is continuous: if  $\phi^*$  is uniformly most powerful in the class of  $\alpha$ -similar tests and if  $\phi^*$  is an  $\alpha$ -level test, then  $\phi^*$  is UMPU at level  $\alpha$ .

## 6.2 Two-sided testing problems in exponential families

Let  $(\mathcal{X}, \mathcal{A}, \mathcal{P} = \{P_\theta : \theta \in \Theta \subset \mathbb{R}\})$  be an exponential model indexed by a scalar parameter  $\theta$  (here,  $\Theta$  is an interval of  $\mathbb{R}$ , or  $\mathbb{R}$  itself). Recall that, after appropriately choosing the  $\sigma$ -finite dominating measure  $\mu$  at hand, this implies that the corresponding densities take the form

$$f_\theta(x) = C(\theta) \exp(\theta T(x)),$$

and that, when  $\mathbf{X} \sim P_\theta$ , the distribution of  $T$  admits the density

$$f_\theta^T(t) = C(\theta) \exp(\theta t)$$

with respect to the induced dominating measure  $\mu^T$ . In this framework, consider the two-sided problem

$$H_0 : \theta = \theta_0 \quad \text{vs} \quad H_1 : \theta \neq \theta_0,$$

where  $\theta_0$  is a fixed value in the interior of  $\Theta$ . Since a UMP test at level  $\alpha$  cannot exist for this problem (why?), we are after a UMPU test at level  $\alpha$ .

**Theorem 1.** *In the exponential model above, fix  $\alpha \in (0, 1)$  and  $\theta_0 \in \text{int}(\Theta)$ . Then, (i) there exist  $\gamma_{1,\alpha}, \gamma_{2,\alpha} \in [0, 1]$  and  $t_{1,\alpha}, t_{2,\alpha} \in \mathbb{R}$  with  $t_{1,\alpha} \leq t_{2,\alpha}$  such that the test defined by*

$$\phi_\alpha^*(\mathbf{x}) = \begin{cases} 1 & \text{if } T(\mathbf{x}) \notin [t_{1,\alpha}, t_{2,\alpha}] \\ \gamma_{1,\alpha} & \text{if } T(\mathbf{x}) = t_{1,\alpha} \\ \gamma_{2,\alpha} & \text{if } T(\mathbf{x}) = t_{2,\alpha} \\ 0 & \text{if } T(\mathbf{x}) \in (t_{1,\alpha}, t_{2,\alpha}) \end{cases}$$

*satisfies  $E_{\theta_0}[\phi_\alpha^*] = \alpha$  and  $E_{\theta_0}[\phi_\alpha^* T] = \alpha E_{\theta_0}[T]$ . (ii) The test  $\phi_\alpha^*$  is UMPU at level  $\alpha$  for the problem of testing  $H_0 : \theta = \theta_0$  against  $H_1 : \theta \neq \theta_0$ .*

We now give an interpretation for the constraint  $E_{\theta_0}[\phi_\alpha^* T] = \alpha E_{\theta_0}[T]$ . Note that we have

$$\begin{aligned} \frac{d}{d\theta} E_\theta[\phi] &= \int_{\mathcal{X}} \phi(x) \frac{d}{d\theta} \left( C(\theta) \exp(\theta T(x)) \right) d\mu(x) \\ &= \frac{C'(\theta)}{C(\theta)} E_\theta[\phi] + E_\theta[\phi T], \end{aligned} \quad (6.1)$$

which, for  $\phi \equiv 1$ , yields  $E_\theta[T] = -C'(\theta)/C(\theta)$ . Using this in (6.1) provides

$$\frac{d}{d\theta} E_\theta[\phi] = E_\theta[\phi T] - E_\theta[\phi] E_\theta[T] \quad (= \text{Cov}_\theta[\phi, T]).$$

The constraint  $E_{\theta_0}[\phi_\alpha^* T] = \alpha E_{\theta_0}[T]$  in the theorem above may then be interpreted as

$$\frac{d}{d\theta} E_\theta[\phi_\alpha^*] |_{\theta=\theta_0} = 0.$$

In the exponential family considered (where the size/power function of any test is smooth), the constraints  $E_{\theta_0}[\phi_\alpha^*] = \alpha$  and  $E_{\theta_0}[\phi_\alpha^* T] = \alpha E_{\theta_0}[T]$  thus clearly are necessary conditions for the test  $\phi_\alpha^*$  to be unbiased at level  $\alpha$ . This will play a role in the proof of the theorem.

*Proof.* (i) Consider the induced model  $(\mathcal{T}, \mathcal{B}, \mathcal{P}^T = \{P_\theta^T : \theta \in \Theta\})$ , which, as recalled above, is dominated by the induced measure  $\mu^T$ , leading to the corresponding densities  $f_\theta^T(t) = C(\theta) \exp(\theta t)$ . In this induced model, fix  $\theta_1 > \theta_0$  arbitrarily, and consider the problem of testing  $\mathcal{H}_0 : \theta = \theta_0$  against  $\mathcal{H}_1 : \theta = \theta_1$  in the class of tests  $\varphi = \varphi(T)$  satisfying

$$\begin{aligned} \int_{\mathcal{T}} \varphi(t) f_{\theta_0}^T(t) d\mu^T(x) &= E_{\theta_0}[\varphi] \\ &= \alpha \end{aligned} \quad (6.2)$$

and

$$\begin{aligned} \int_{\mathcal{T}} \varphi(t) t f_{\theta_0}^T(t) d\mu^T(x) &= E_{\theta_0}[\varphi T] \\ &= \alpha E_{\theta_0}[T]. \end{aligned} \quad (6.3)$$

For this problem, consider

$$\mathcal{M} = \{(\mathbb{E}_{\theta_0}[\varphi], \mathbb{E}_{\theta_0}[\varphi T]) : \varphi \text{ a test}\}.$$

For any  $u$  in a neighborhood  $[\alpha - \varepsilon, \alpha + \varepsilon]$  of  $\alpha$ , the UMP test,  $\varphi_u^+$  say, for  $\mathcal{H}_0 : \theta \leq \theta_0$  against  $\mathcal{H}_1 : \theta > \theta_0$  at level  $u$  provides  $(\mathbb{E}_{\theta_0}[\varphi_u^+], \mathbb{E}_{\theta_0}[\varphi_u^+ T]) = (u, c_u^+)$ , with  $c_u^+ > u\mathbb{E}_{\theta_0}[T]$  (since  $\frac{d}{d\theta}\mathbb{E}_{\theta}[\varphi_u^+]|_{\theta=\theta_0} = \mathbb{E}_{\theta_0}[\varphi_u^+ T] - u\mathbb{E}_{\theta_0}[T] > 0$ ; see the remark at the end of Section 5.3). Similarly, for any  $u \in [\alpha - \varepsilon, \alpha + \varepsilon]$ , the UMP test,  $\varphi_u^-$  say, for  $\mathcal{H}_0 : \theta \geq \theta_0$  against  $\mathcal{H}_1 : \theta < \theta_0$  at level  $u$  provides  $(\mathbb{E}_{\theta_0}[\varphi_u^-], \mathbb{E}_{\theta_0}[\varphi_u^- T]) = (u, c_u^-)$ , with  $c_u^- < u\mathbb{E}_{\theta_0}[T]$ . Jointly with the convexity of  $\mathcal{M}$ , this implies that  $(\alpha, \alpha\mathbb{E}_{\theta_0}[T])$  is an interior point of  $\mathcal{M}$ . Part (iv) of the second version of the generalized Neyman-Pearson lemma thus implies that there exist  $k_1, k_2 \in \mathbb{R}$  and a measurable function  $\gamma(\cdot)$  such that the test  $\varphi$  defined by

$$\varphi(t) = \begin{cases} 1 & \text{if } f_{\theta_1}^T(t) > k_1 f_{\theta_0}^T(t) + k_2 t f_{\theta_0}^T(t) \\ \gamma(t) & \text{if } f_{\theta_1}^T(t) = k_1 f_{\theta_0}^T(t) + k_2 t f_{\theta_0}^T(t) \\ 0 & \text{if } f_{\theta_1}^T(t) < k_1 f_{\theta_0}^T(t) + k_2 t f_{\theta_0}^T(t) \end{cases}$$

satisfies (6.2)–(6.3). Using the explicit expression of  $f_{\theta_1}^T$ , this test rewrites

$$\varphi(t) = \begin{cases} 1 & \text{if } \exp((\theta_1 - \theta_0)t) > l_1 + l_2 t \\ \gamma(t) & \text{if } \exp((\theta_1 - \theta_0)t) = l_1 + l_2 t \\ 0 & \text{if } \exp((\theta_1 - \theta_0)t) < l_1 + l_2 t, \end{cases}$$

where we let  $l_i := k_i C(\theta_0)/C(\theta_1)$ ,  $i = 1, 2$ . Since we cannot have  $\varphi(t) = 1$  for any  $t$  (this would provide  $\mathbb{E}_{\theta_0}[\varphi] = 1 \neq \alpha$ , which would contradict (6.2)), nor either of

$$\varphi(t) = \begin{cases} 1 & \text{if } t > t_0 \\ \gamma & \text{if } t = t_0 \\ 0 & \text{if } t < t_0 \end{cases} \quad \text{or} \quad \varphi(t) = \begin{cases} 1 & \text{if } t < t_0 \\ \gamma & \text{if } t = t_0 \\ 0 & \text{if } t > t_0 \end{cases}$$

(this would provide  $\frac{d}{d\theta}\mathbb{E}_{\theta}[\varphi]|_{\theta=\theta_0} = \mathbb{E}_{\theta_0}[\varphi T] - \alpha\mathbb{E}_{\theta_0}[T] > 0$  or  $< 0$ , respectively, which would

contradict (6.3)), we must then have

$$\varphi(t) = \begin{cases} 1 & \text{if } t \notin [t_1, t_2] \\ \gamma_1 & \text{if } t = t_1 \\ \gamma_2 & \text{if } t = t_2 \\ 0 & \text{if } t \in (t_1, t_2), \end{cases}$$

for some  $t_1, t_2$  such that  $t_1 \leq t_2$ . It follows that the test  $\phi$  defined by

$$\phi(\mathbf{x}) = \varphi(T(\mathbf{x})) = \begin{cases} 1 & \text{if } T(\mathbf{x}) \notin [t_1, t_2] \\ \gamma_1 & \text{if } T(\mathbf{x}) = t_1 \\ \gamma_2 & \text{if } T(\mathbf{x}) = t_2 \\ 0 & \text{if } T(\mathbf{x}) \in (t_1, t_2) \end{cases}$$

satisfies  $E_{\theta_0}[\phi] = \alpha$  and  $E_{\theta_0}[\phi T] = \alpha E_{\theta_0}[T]$ , which establishes the result.

(ii) Let  $\phi_\alpha^*$  be the test described in Part (i) of the theorem. Fix  $\theta_1 \neq \theta_0$  arbitrarily. Then, it is possible (why?<sup>1</sup>) to find  $\ell_1, \ell_2 \in \mathbb{R}$  such that

$$\phi_\alpha^*(\mathbf{x}) = \begin{cases} 1 & \text{if } \exp((\theta_1 - \theta_0)T(\mathbf{x})) > \ell_1 + \ell_2 T(\mathbf{x}) \\ \gamma_1 & \text{if } T(\mathbf{x}) = t_1 \\ \gamma_2 & \text{if } T(\mathbf{x}) = t_2 \\ 0 & \text{if } \exp((\theta_1 - \theta_0)T(\mathbf{x})) < \ell_1 + \ell_2 T(\mathbf{x}). \end{cases}$$

Thus,

$$\phi_\alpha^*(\mathbf{x}) = \begin{cases} 1 & \text{if } f_{\theta_1}(T(\mathbf{x})) > k_1 f_{\theta_0}(T(\mathbf{x})) + k_2 T(\mathbf{x}) f_{\theta_0}(T(\mathbf{x})) \\ \gamma_1 & \text{if } T(\mathbf{x}) = t_1 \\ \gamma_2 & \text{if } T(\mathbf{x}) = t_2 \\ 0 & \text{if } f_{\theta_1}(T(\mathbf{x})) < k_1 f_{\theta_0}(T(\mathbf{x})) + k_2 T(\mathbf{x}) f_{\theta_0}(T(\mathbf{x})) \end{cases}$$

for some  $k_1, k_2 \in \mathbb{R}$ . Part (ii) of the second version of the generalized Neyman-Pearson lemma thus entails that  $\phi_\alpha^*$  is most powerful for the problem of testing  $H_0 : \theta = \theta_0$  against  $H_1 : \theta = \theta_1$

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<sup>1</sup>When answering this question, do not forget the case  $t_{1,\alpha} = t_{2,\alpha}$

in the class of tests satisfying  $E_{\theta_0}[\phi] = \alpha$  and  $E_{\theta_0}[\phi T] = \alpha E_{\theta_0}[T]$ . Since  $\phi_\alpha^*$  does not depend on  $\theta_1$ , it follows that  $\phi_\alpha^*$  is uniformly most powerful for the problem of testing  $H_0 : \theta = \theta_0$  against  $H_1 : \theta \neq \theta_0$  in the class of tests satisfying  $E_{\theta_0}[\phi] = \alpha$  and  $E_{\theta_0}[\phi T] = \alpha E_{\theta_0}[T]$ . Recalling that any unbiased test at level  $\alpha$  must satisfy these constraints,  $\phi_\alpha^*$  is then UMPU at level  $\alpha$  for the problem of testing  $H_0 : \theta = \theta_0$  against  $H_1 : \theta \neq \theta_0$  (unbiasedness of  $\phi_\alpha^*$  at level  $\alpha$  follows by comparing its power function to that of the trivial test defined by  $\phi(\mathbf{x}) = \alpha$  for any  $\mathbf{x}$ ).  $\square$

If the distribution of  $T$  under  $P_{\theta_0}$  is symmetric (automatically about  $E_{\theta_0}[T]$ ), then it is always possible to find  $\gamma_\alpha \in [0, 1]$  and  $h_\alpha \in \mathbb{R}^+$  such that the test  $\phi_\alpha^*$  defined by

$$\phi_\alpha^*(\mathbf{x}) = \begin{cases} 1 & \text{if } T(\mathbf{x}) \notin [E_{\theta_0}[T] - h_\alpha, E_{\theta_0}[T] + h_\alpha] \\ \gamma_\alpha & \text{if } T(\mathbf{x}) \in \{E_{\theta_0}[T] - h_\alpha, E_{\theta_0}[T] + h_\alpha\} \\ 0 & \text{if } T(\mathbf{x}) \in (E_{\theta_0}[T] - h_\alpha, E_{\theta_0}[T] + h_\alpha) \end{cases} \quad (6.4)$$

satisfies  $E_{\theta_0}[\phi_\alpha^*] = \alpha$  (this can be checked by rewriting this test as

$$\phi_\alpha^*(\mathbf{x}) = \begin{cases} 1 & \text{if } |T(\mathbf{x}) - E_{\theta_0}[T]| > h_\alpha \\ \gamma_\alpha & \text{if } |T(\mathbf{x}) - E_{\theta_0}[T]| = h_\alpha \\ 0 & \text{if } |T(\mathbf{x}) - E_{\theta_0}[T]| < h_\alpha \end{cases}$$

and by proceeding as in the proof of Theorem 1(i) from the previous chapter). For this test, note that we have

$$\begin{aligned} E_{\theta_0}[\phi_\alpha^* T] &= E_{\theta_0}[\phi_\alpha^*(T - E_{\theta_0}[T])] + E_{\theta_0}[\phi_\alpha^*] E_{\theta_0}[T] \\ &= 0 + \alpha E_{\theta_0}[T] \\ &= \alpha E_{\theta_0}[T], \end{aligned}$$

so that the second constraint is then automatically satisfied, which implies that this test is UMPU at level  $\alpha$ . A typical example is obtained when testing  $\mathcal{H}_0 : \mu = \mu_0$  against  $\mathcal{H}_1 : \mu \neq \mu_0$  at level  $\alpha$  on the basis of  $\mathbf{X} = (X_1, \dots, X_n)$ , where the  $X_i$ 's are i.i.d. Gaussian with

unknown mean  $\mu$  and fixed variance  $\sigma_0^2$ . If the distribution of  $T$  under  $P_{\theta_0}$  is not symmetric, then both constraints need to be imposed (which, in practice, sometimes can only be achieved numerically); an example is obtained when testing  $\mathcal{H}_0 : \sigma^2 = \sigma_0^2$  against  $\mathcal{H}_1 : \sigma^2 \neq \sigma_0^2$  at level  $\alpha$  on the basis of  $\mathbf{X} = (X_1, \dots, X_n)$ , where the  $X_i$ 's are i.i.d. Gaussian with fixed mean  $\mu_0$  and unknown variance  $\sigma^2$ .

### 6.3 Problems in exponential families with nuisance parameters

Let  $(\mathcal{X}, \mathcal{A}, \mathcal{P} = \{P_{\tau, \boldsymbol{\lambda}}\})$  be an exponential model with densities (with respect to a  $\sigma$ -finite dominating measure  $\mu$ ) of the form

$$f_{\tau, \boldsymbol{\lambda}}(x) = C(\tau, \boldsymbol{\lambda}) \exp(\tau T(x) + \boldsymbol{\lambda}' \mathbf{S}(x)),$$

with  $\tau \in \mathbb{R}$  and  $\boldsymbol{\lambda} \in \mathbb{R}^s$ . In this framework, consider the one-sided problem

$$H_0 = \{P_{\tau, \boldsymbol{\lambda}} : \tau \leq \tau_0\} \quad \text{vs} \quad H_1 = \{P_{\tau, \boldsymbol{\lambda}} : \tau > \tau_0\} \quad (6.5)$$

at level  $\alpha$ , where  $\tau_0$  is a fixed value and  $\boldsymbol{\lambda}$  remains unspecified, hence plays the role of a nuisance parameter. Using the notation from Section 6.1, we have

$$\bar{\mathcal{H}} = \{P_{\tau, \boldsymbol{\lambda}} : \tau = \tau_0\},$$

which is an exponential subfamily indexed by  $\boldsymbol{\lambda}$  and with natural statistic  $\mathbf{S}$ . For the corresponding submodel,  $\mathbf{S}$  is thus sufficient and complete.

**Theorem 2.** *Fix  $\alpha \in (0, 1)$ . Then, we have the following in the framework above: (i) there exists a test of the form*

$$\phi_\alpha^*(\mathbf{x}) = \begin{cases} 1 & \text{if } T(\mathbf{x}) > t(\mathbf{S}(x)) \\ \gamma(\mathbf{S}(x)) & \text{if } T(\mathbf{x}) = t(\mathbf{S}(x)) \\ 0 & \text{if } T(\mathbf{x}) < t(\mathbf{S}(x)) \end{cases}$$

satisfying

$$E_{\tau_0, \boldsymbol{\lambda}}[\phi_\alpha^* | \mathbf{S}] = \alpha \quad P_{\tau_0, \boldsymbol{\lambda}}\text{-a.s. for any } \boldsymbol{\lambda}. \quad (6.6)$$

(ii) This test is UMPU at level  $\alpha$  for the testing problem (6.5).

A test that satisfies (6.6) is said to have *Neyman  $\alpha$ -structure with respect to  $\mathbf{S}$*  for the testing problem (6.5). Note that a test has Neyman  $\alpha$ -structure with respect to  $\mathbf{S}$  if and only if it is  $\alpha$ -similar for the same problem (the necessary condition is trivial, whereas the sufficient one follows from sufficiency and completeness of  $\mathbf{S}$  in the  $\bar{\mathcal{H}}$ -submodel). A corollary is that one may restrict to tests having Neyman  $\alpha$ -structure with respect to  $\mathbf{S}$  when looking for a UMPU test at level  $\alpha$ , which makes condition (6.6) natural.

As an exercise, one can apply Theorem 2 to show that, when  $X_1, \dots, X_n$  are i.i.d. Gaussian with mean  $\mu$  and variance  $\sigma^2$ , then

$$\phi_\alpha^*(\mathbf{x}) = \begin{cases} 1 & \text{if } ns^2/\sigma_0^2 > \chi_{n-1, 1-\alpha}^2 \\ 0 & \text{if } ns^2/\sigma_0^2 \leq \chi_{n-1, 1-\alpha}^2, \end{cases}$$

where  $s^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2$  and  $\chi_{n-1, 1-\alpha}^2$  is the  $(1 - \alpha)$ -quantile of the  $\chi_{n-1}^2$  distribution, is UMPU at level  $\alpha$  for the problem of testing  $\mathcal{H}_0 : \sigma^2 \leq \sigma_0^2$  against  $\mathcal{H}_1 : \sigma^2 > \sigma_0^2$ , and that, in the same model,

$$\phi_\alpha^*(\mathbf{x}) = \begin{cases} 1 & \text{if } \sqrt{n-1}(\bar{X} - \mu_0)/s > t_{n-1, 1-\alpha} \\ 0 & \text{if } \sqrt{n-1}(\bar{X} - \mu_0)/s \leq t_{n-1, 1-\alpha}, \end{cases}$$

where  $t_{n-1, 1-\alpha}$  is the  $(1 - \alpha)$ -quantile of the  $t_{n-1}$  distribution, is UMPU at level  $\alpha$  for the problem of testing  $\mathcal{H}_0 : \mu \leq \mu_0$  against  $\mathcal{H}_1 : \mu > \mu_0$  (the exercise is actually more complicated for this second testing problem<sup>2</sup>).

In the same general exponential framework as in the beginning of this section, one may also consider the two-sided problem

$$H_0 = \{P_{\tau, \boldsymbol{\lambda}} : \tau = \tau_0\} \quad \text{vs} \quad H_1 = \{P_{\tau, \boldsymbol{\lambda}} : \tau \neq \tau_0\} \quad (6.7)$$

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<sup>2</sup>Do the change of variables  $X_i \rightsquigarrow X_i - \mu_0$ , which makes the null hypothesis become  $\mathcal{H}_0 : \mu \leq 0$ , and use the fact that  $\bar{X}/s = \bar{X}/(\frac{1}{n} \sum_i X_i^2 - \bar{X}^2)^{1/2}$  is an increasing function of  $\bar{X}$  for fixed  $\sum_i X_i^2$ .

at level  $\alpha$ , where  $\tau_0$  is still a fixed value and  $\boldsymbol{\lambda}$  remains unspecified—which yields again  $\bar{\mathcal{H}} = \{P_{\tau, \boldsymbol{\lambda}} : \tau = \tau_0\}$ . We have the following analog of Theorem 2.

**Theorem 3.** *Fix  $\alpha \in (0, 1)$ . Then, we have the following in the framework above: (i) there exists a test of the form*

$$\phi_\alpha^*(\mathbf{x}) = \begin{cases} 1 & \text{if } T(\mathbf{x}) \notin [t_1(\mathbf{S}(x)), t_2(\mathbf{S}(x))] \\ \gamma_1(\mathbf{S}(x)) & \text{if } T(\mathbf{x}) = t_1(\mathbf{S}(x)) \\ \gamma_2(\mathbf{S}(x)) & \text{if } T(\mathbf{x}) = t_2(\mathbf{S}(x)) \\ 0 & \text{if } T(\mathbf{x}) \in (t_1(\mathbf{S}(x)), t_2(\mathbf{S}(x))) \end{cases}$$

satisfying both

$$E_{\tau_0, \boldsymbol{\lambda}}[\phi_\alpha^* | \mathbf{S}] = \alpha \quad P_{\tau_0, \boldsymbol{\lambda}}\text{-a.s. for any } \boldsymbol{\lambda}$$

and

$$E_{\tau_0, \boldsymbol{\lambda}}[\phi_\alpha^* T | \mathbf{S}] = \alpha E_{\tau_0, \boldsymbol{\lambda}}[\phi_\alpha^* | \mathbf{S}] \quad P_{\tau_0, \boldsymbol{\lambda}}\text{-a.s. for any } \boldsymbol{\lambda}.$$

(ii) *This test is UMPU at level  $\alpha$  for the testing problem (6.7).*

Coming back to the Gaussian model considered above, one may then show in particular that the test defined by

$$\phi_\alpha^*(\mathbf{x}) = \begin{cases} 1 & \text{if } |\sqrt{n-1}(\bar{X} - \mu_0)/s| > t_{n-1, 1-(\alpha/2)} \\ 0 & \text{if } |\sqrt{n-1}(\bar{X} - \mu_0)/s| \leq t_{n-1, 1-(\alpha/2)} \end{cases}$$

is UMPU at level  $\alpha$  for the problem of testing  $\mathcal{H}_0 : \mu = \mu_0$  against  $\mathcal{H}_1 : \mu \neq \mu_0$ .