

Power enhancement for dimension detection of Gaussian signals

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S1. Further simulations

In the present section, our objective is to provide Monte-Carlo simulation results to corroborate the conclusions drawn from Proposition 1 and in Section 4. In the first simulation exercise, the objective is to illustrate Proposition 1. We generated $M = 10,000$ independent samples of i.i.d. observations $\mathbf{X}_1^{(b)}, \dots, \mathbf{X}_{10,000}^{(b)}$, for $b = 0, \frac{1}{4}, \frac{1}{2}, 1$. The $\mathbf{X}_i^{(b)}$'s are i.i.d. with a common ($p = 8$)-dimensional Gaussian distribution with mean zero and covariance matrix

$$\boldsymbol{\Sigma}(b) = \text{diag}((1 + n^{-b})\mathbf{1}_q, \mathbf{1}_{p-q}).$$

For various values of q , we computed the value of $T_q^{(n)}$ and performed the test that rejects the null hypothesis $\mathcal{H}_{0q}^{(n)}$ when $T_q^{(n)} > \chi_{d(p,q);.95}^2$. In Figure 1, we provide histograms of the distribution of the values of $T_q^{(n)}$ (obtained

from the $M = 10,000$ replications). The histograms have to be compared with (i) the red line which is the chi-square density function with $d(p, q)$ degrees of freedom and (ii) the grey line which is an approximation of the density of $T_q^{(n)}$ obtained in Proposition 1; the approximation has been obtained by computing a kernel density estimator based on 100,000 replications of the random variable in (3.3). In Figure 2 we provide the empirical rejection frequencies (out of the $M = 10,000$ replications) of the tests rejecting $\mathcal{H}_{0q}^{(n)}$ when $T_q^{(n)} > \chi_{d(p,q);.95}^2$. Inspection of Figures 1 and 2 clearly reveals that the conclusions drawn from Proposition 1 are correct. Provided that $n^{1/2}r_q^{(n)} \rightarrow \infty$, the weak limit of $T_q^{(n)}$ is chi-square with $d(p, q)$ degrees of freedom. Now if $n^{1/2}r_q^{(n)}$ does not diverge to ∞ , the weak limit of $T_q^{(n)}$ is not chi-square and the test $\phi^{(n)}$ is such that $\lim_{n \rightarrow \infty} \mathbb{E}[\phi^{(n)}]$ is far below the asymptotic nominal level α .

The second simulation study illustrates the results obtained in Section 4. We generated $M = 1,000$ independent samples of i.i.d. observations

$$\mathbf{X}_1^{(b,\tau)}, \dots, \mathbf{X}_{10,000}^{(b,\tau)},$$

for $\tau = 0, 1, 2, 4, 6, 8$ and $b = 0, \frac{1}{8}, \frac{1}{4}, \frac{1}{2}, 1, 2$. The $\mathbf{X}_i^{(b,\tau)}$'s are i.i.d. with a common $(p = 5)$ -dimensional Gaussian distribution with mean zero and

S1. FURTHER SIMULATIONS

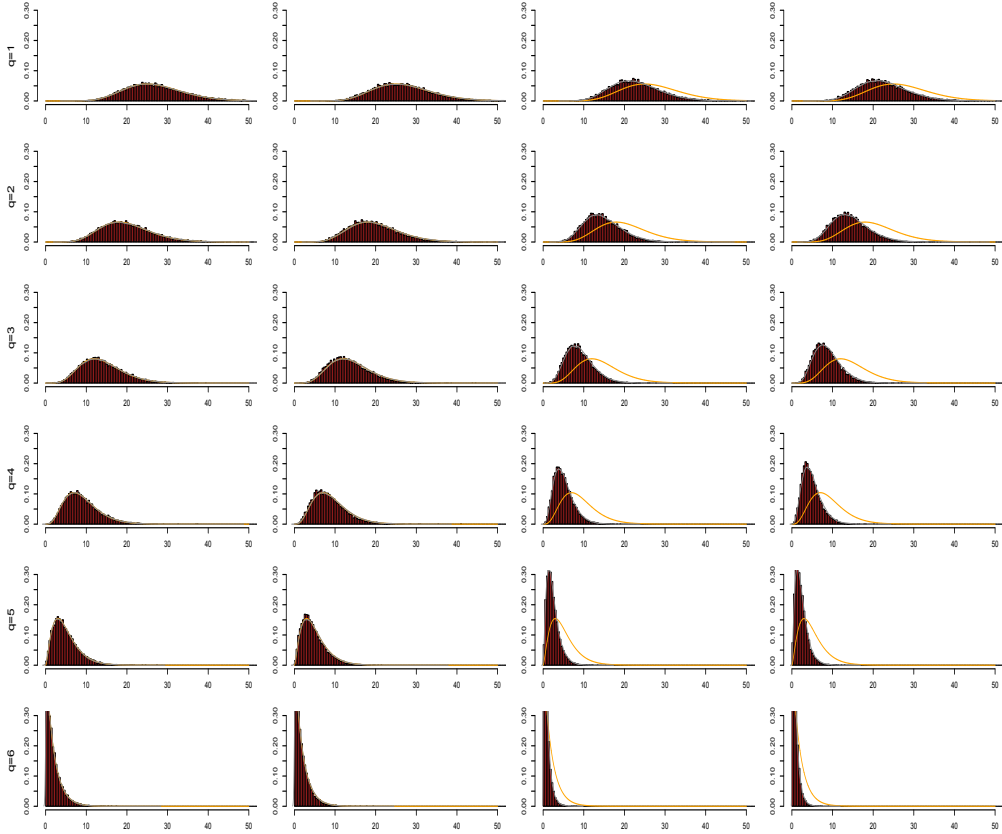


Figure 1: In red, histograms of the distribution of the values of $T_q^{(n)}$ (obtained from the $M = 10,000$ replications) for various values of q and b . The histograms have to be compared with (i) the orange line which is the chi-square density function with $d(p, q)$ degrees of freedom and (ii) the grey line which is an approximation of the density of $T_q^{(n)}$ obtained in Proposition 1.

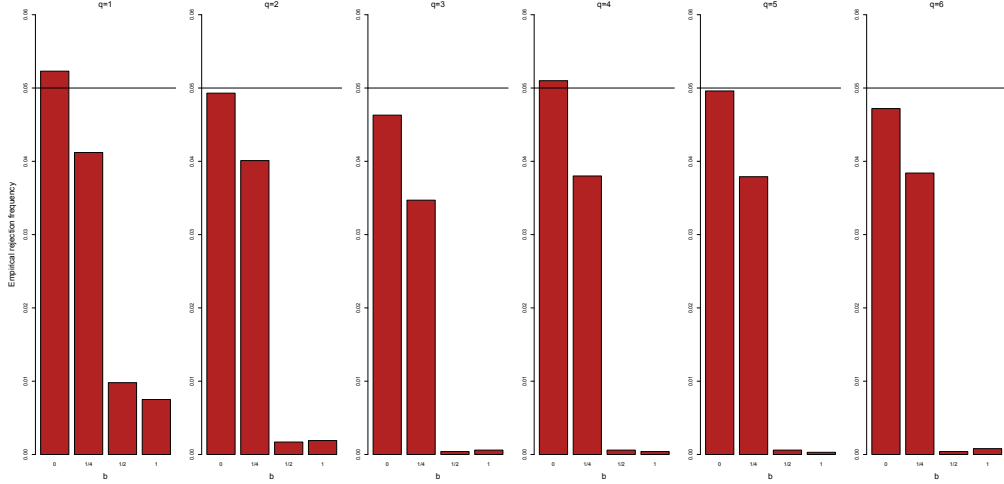


Figure 2: Empirical rejection frequencies of the tests that reject the null hypothesis $\mathcal{H}_{0q}^{(n)}$ when $T_q^{(n)} > \chi_{d(p,q);.95}^2$ for various values of q and various values of b .

covariance matrix

$$\Sigma(b, \tau) = \text{diag}(3, 1 + n^{-b}, 1 + n^{-b}, 1, 1 - \frac{\tau}{n^{1/2}}).$$

The values $\tau = 0$ and $b < 1/2$ provide data generating processes that belong to the null hypothesis $\lambda_3 > \lambda_4 = \lambda_5$ while for $\tau = 1, 2, 4, 6, 8$, the corresponding distributions are increasingly under the alternative. The value $b = 0$ provides data generating processes with three eigenvalues (virtually) in block 1 of (3.2), the values $b = \frac{1}{8}, \frac{1}{4}$ provide data generating processes with one eigenvalue (virtually) in block 1 and two eigenvalues (virtually) in block 2 of (3.2), the value $b = \frac{1}{2}$ provides data generating processes with one

eigenvalue (virtually) in block 1 and two eigenvalues (virtually) in block 3 of (3.2) while the values $b = 1, 2$ provide data generating processes with one eigenvalue (virtually) in block 1 and two eigenvalues (virtually) in block 4 of (3.2). For each scenario, we performed at each replication the three tests $\phi_{\beta}^{(n)}$, $\phi_{\text{LRT}}^{(n)}$ and $\phi^{(n)}$ at the nominal level $\alpha = .05$. In Figure 3, we provide the empirical power curves of the various tests as functions of τ . Inspection of Figure 3 clearly confirms our theoretical findings: the three tests $\phi^{(n)}$, $\phi_{\text{LRT}}^{(n)}$ and $\phi_{\beta}^{(n)}$ are asymptotically equivalent provided that $n^{1/2}r_q^{(n)}$ diverges to infinity as $n \rightarrow \infty$ and the tests $\phi^{(n)}$ and $\phi_{\text{LRT}}^{(n)}$ (that are asymptotically equivalent in all regimes as stated in Lemma 1) are such that $\lim_{n \rightarrow \infty} \mathbb{E}[\phi^{(n)}]$ is far below the asymptotic nominal level α when $n^{1/2}r_q^{(n)}$ does not diverge to infinity as $n \rightarrow \infty$.

Now, we conclude this Section by providing some further simulation results for a test similar to $\phi_{\text{new}}^{(n)}$ based on the pseudo-Gaussian tests mentioned in the real data application section. Letting $\hat{\kappa}^{(n)}$ be a consistent estimator of the underlying kurtosis coefficient, define

$$\tilde{T}_{q,q+1}^{(n)} := (1 + \hat{\kappa}^{(n)})^{-1} T_{q,q+1}^{(n)},$$

and

$$\tilde{T}_q^{(n)} := (1 + \hat{\kappa}^{(n)})^{-1} T_q^{(n)}.$$

Our pseudo-Gaussian test $\tilde{\phi}_{\text{new}}^{(n)}$ rejects $\mathcal{H}_{0q}^{(n)}$ at asymptotic confidence level

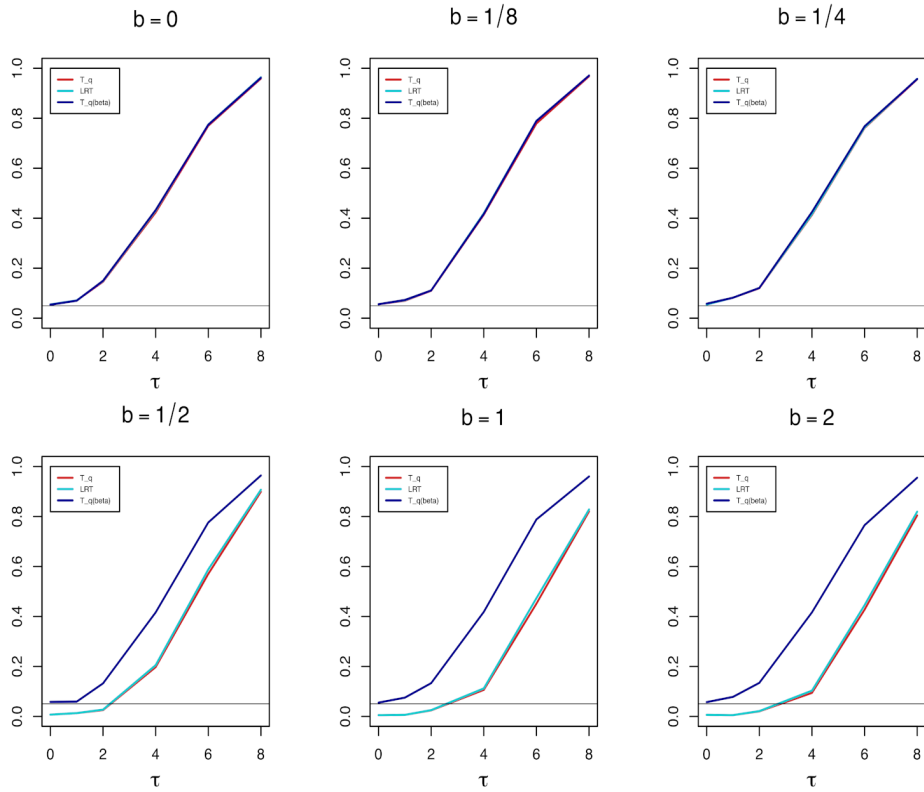


Figure 3: Empirical rejection frequencies of the tests $\phi_{\text{LRT}}^{(n)}$ (based on $L_q^{(n)}$), $\phi^{(n)}$ (based on $T_q^{(n)}$) and $\phi_{\beta}^{(n)}$ (based on $T_q^{(n)}(\beta)$) performed at the nominal level $\alpha = .05$.

α when

$$\tilde{\phi}_{\text{new}}^{(n)} := \mathbb{I}[\tilde{T}_q^{(n)} > \chi_{d(p,q);1-\alpha}^2] \mathbb{I}[\tilde{T}_{q,q+1}^{(n)} > \chi_{2;1-\gamma}^2] + \mathbb{I}[\tilde{T}_{q,q+1}^{(n)} \leq \chi_{2;1-\gamma}^2] = 1 \quad (\text{S1.1})$$

Letting $\tilde{\phi}^{(n)}$ be the pseudo-Gaussian version of $\phi^{(n)}$, we compare here $\tilde{\phi}_{\text{new}}^{(n)}$ and $\tilde{\phi}^{(n)}$ through simulations.

To do so, we generated $M = 2,000$ independent samples of i.i.d. observations

$$\mathbf{X}_1^{(b,\tau)}, \dots, \mathbf{X}_n^{(b,\tau)},$$

for $\tau = 0, 1, 2, 4, 6, 8$ and $b = 0, \frac{1}{8}, \frac{1}{4}, \frac{1}{2}, 1, 2$. The $\mathbf{X}_i^{(b,\tau)}$'s are i.i.d. having ($p = 5$)-dimensional multivariate student distribution with zero mean, 5 degrees of freedom and scatter matrix

$$\Sigma(b, \tau) = \text{diag}(3, 1 + n^{-b}, 1 + n^{-b}, 1, 1 - \frac{\tau}{n^{1/2}}).$$

We performed the classical test $\phi^{(n)}$, three versions of the $\phi_{\text{new}}^{(n)}$ test ($\gamma = .9$, $\gamma = .5$ and $\gamma = .05$) for $\mathcal{H}_{03}^{(n)}$ ($q = 3$). We also performed the pseudo-Gaussian versions $\tilde{\phi}^{(n)}$ and $\tilde{\phi}_{\text{new}}^{(n)}$ of these tests, with same choices of γ . In Figures 4 and 5, we provide empirical rejection frequencies of the Gaussian tests as functions of τ for sample sizes $n = 500$ and $n = 10,000$. Clearly, the lack of robustness to non-Gaussian assumptions is shown in Figures 4 and 5.

In Figures 6 and 7, the same empirical rejection frequencies are displayed for the pseudo-Gaussian tests. Pseudo-Gaussian procedures are clearly robust to non-Gaussianity.

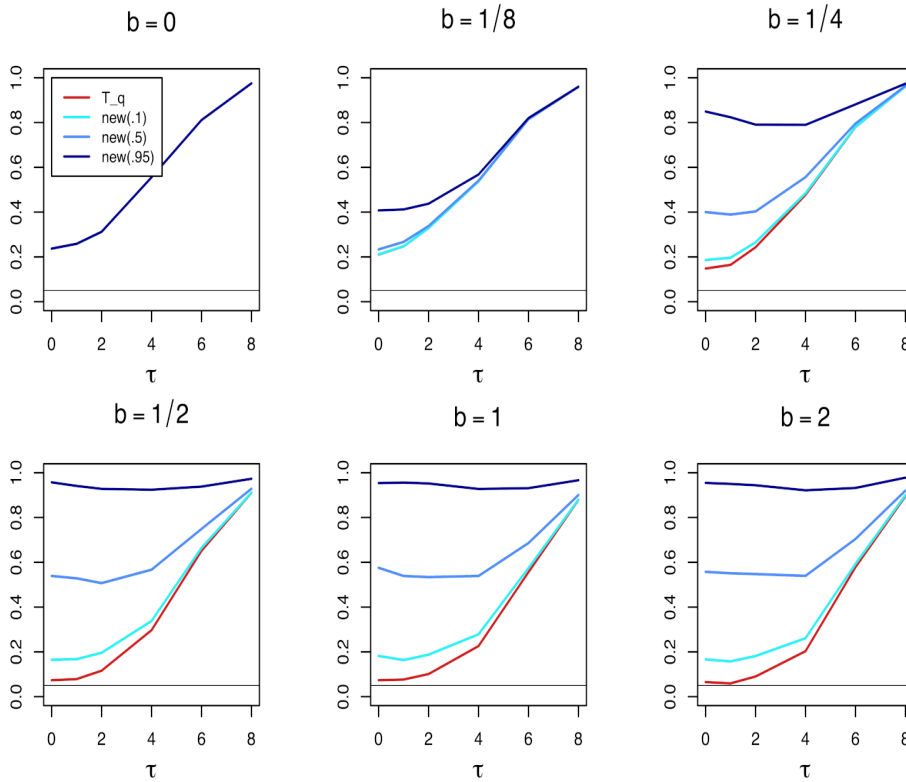


Figure 4: Empirical rejection frequencies of the classical gaussian test $\phi^{(n)}$ performed at the asymptotic nominal level .05 and three versions of the $\phi_{new}^{(n)}$ test (all with $\alpha = .05$) based on three different choices of γ : $\gamma = .9$ (denoted as $new(.1)$), $\gamma = .5$ (denoted as $new(.5)$) and $\gamma = .05$ (denoted as $new(.95)$). The sample size is $n = 500$.

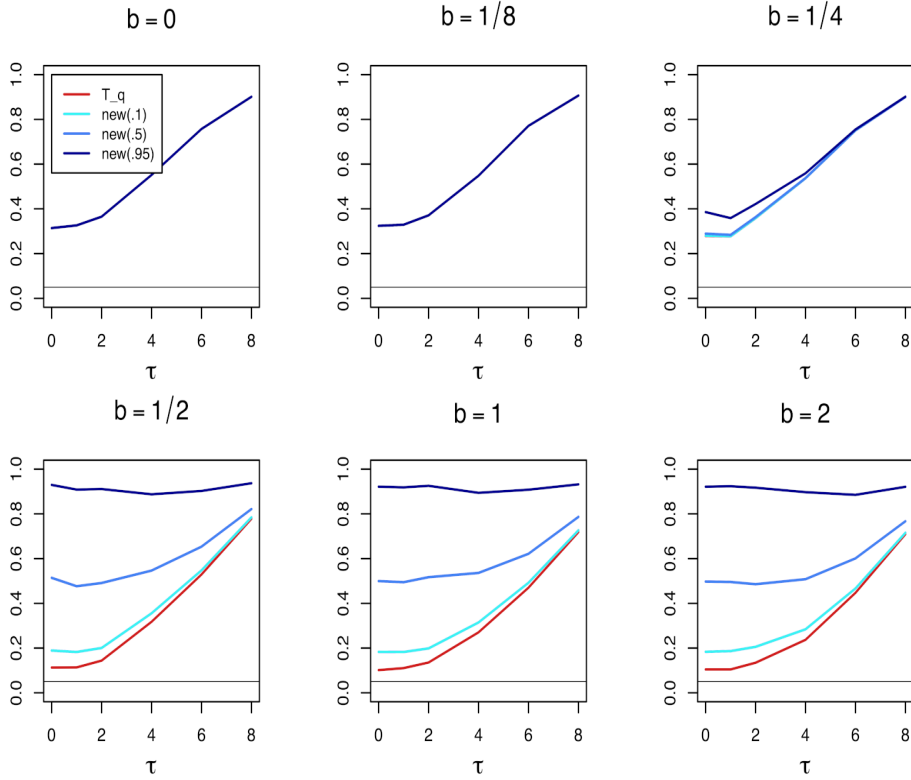


Figure 5: Empirical rejection frequencies of the classical gaussian test $\phi^{(n)}$ performed at the asymptotic nominal level .05 and three versions of the $\phi_{\text{new}}^{(n)}$ test (all with $\alpha = .05$) based on three different choices of γ : $\gamma = .9$ (denoted as new(.1)), $\gamma = .5$ (denoted as new(.5)) and $\gamma = .05$ (denoted as new(.95)). The sample size is $n = 10,000$.

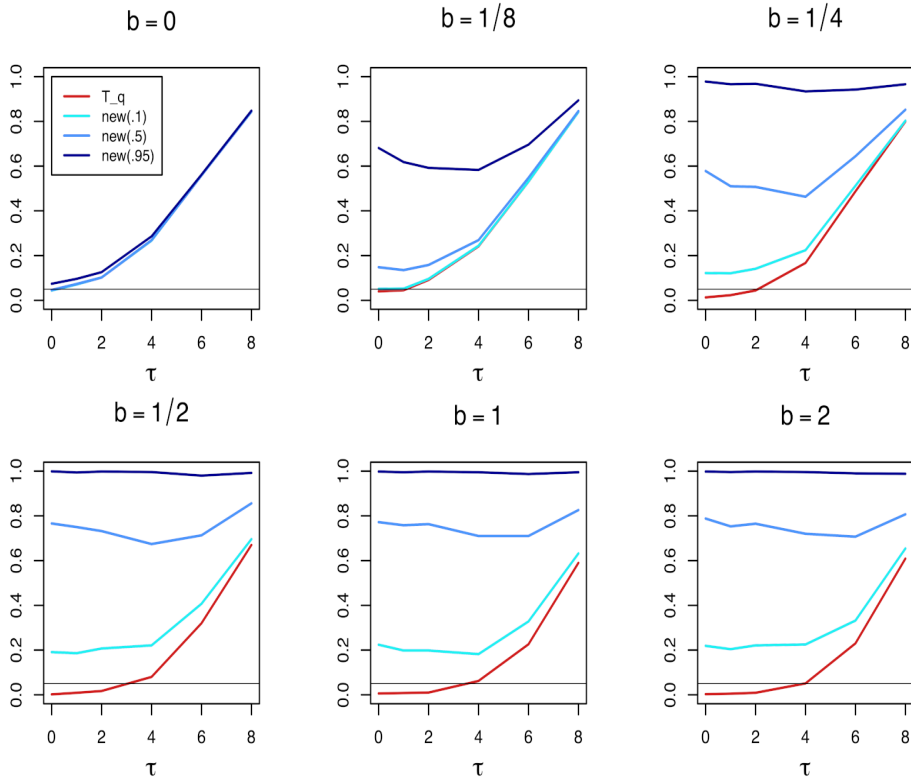


Figure 6: Empirical rejection frequencies of the classical pseudo-gaussian $\tilde{\phi}^{(n)}$ performed at the asymptotic nominal level .05 and three versions of the $\tilde{\phi}_{\text{new}}^{(n)}$ test (all with $\alpha = .05$) based on three different choices of γ : $\gamma = .9$ (denoted as $\text{new}(.1)$), $\gamma = .5$ (denoted as $\text{new}(.5)$) and $\gamma = .05$ (denoted as $\text{new}(.95)$). The sample size is $n = 500$.

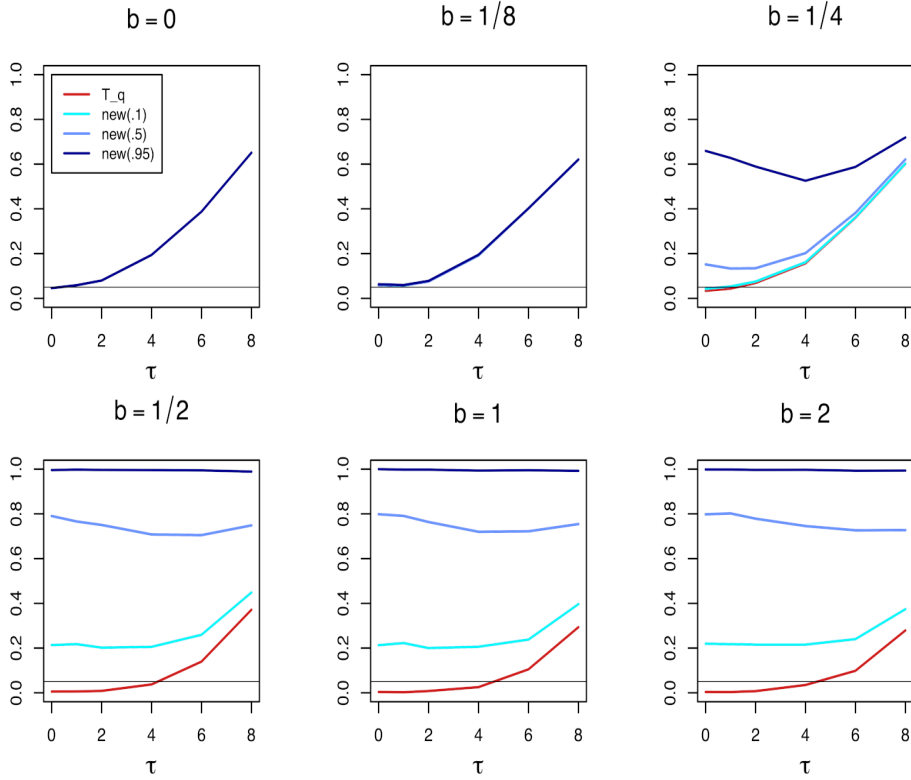


Figure 7: Empirical rejection frequencies of the classical pseudo-gaussian $\tilde{\phi}^{(n)}$ performed at the asymptotic nominal level .05 and three versions of the $\tilde{\phi}_{\text{new}}^{(n)}$ test (all with $\alpha = .05$) based on three different choices of γ : $\gamma = .9$ (denoted as $\text{new}(.1)$), $\gamma = .5$ (denoted as $\text{new}(.5)$) and $\gamma = .05$ (denoted as $\text{new}(.95)$). The sample size is $n = 10,000$.

S2. Proofs of the various results.

Proof of Proposition 1. The proof of Proposition 1 directly follows from Proposition 1 below. \square

Proposition 6. *Let $\mathbf{r}^{(n)}$ and \mathbf{v} be such that (3.1) holds and such that for $0 \leq s_1 \leq s_2 \leq s_3 \leq q$, (i) $r_j^{(n)} \equiv 1$ for each $1 \leq j \leq s_1$, (ii) $r_j^{(n)} = o(1)$ with $n^{1/2}r_j^{(n)} \rightarrow \infty$, for each $s_1 < j \leq s_2$, (iii) $r_j^{(n)} = n^{-1/2}$, for each $s_2 < j \leq s_3$ and (iv) $r_j^{(n)} = o(n^{-1/2})$, for each $s_3 < j \leq q$. Let*

$$\mathbf{Z}(v_1, \dots, v_{s_1}) = \begin{pmatrix} \mathbf{Z}_{11} & \mathbf{Z}'_{21} \\ \mathbf{Z}_{21} & \mathbf{Z}_{22} \end{pmatrix}$$

where \mathbf{Z}_{11} is the $s_2 \times s_2$ upper left block of $\mathbf{Z}(v_1, \dots, v_{s_1})$, \mathbf{Z}_{22} is the $p - s_2 \times p - s_2$ lower right block of $\mathbf{Z}(v_1, \dots, v_{s_1})$, etc, be such that

$$\text{vec}(\mathbf{Z}(v_1, \dots, v_{s_1})) \sim \mathcal{N}_{p^2}(\mathbf{0}, (\mathbf{I}_{p^2} + \mathbf{K}_p)(\text{diag}(1 + v_1, \dots, 1 + v_{s_1}, \mathbf{1}'_{p-s_1}))^{\otimes 2}).$$

Then as $n \rightarrow \infty$ under $\mathbb{P}_{\boldsymbol{\beta}, \boldsymbol{\lambda}^{(n)}}^{(n)}$,

$$\boldsymbol{\ell}_{p-q}^{(n)} = (\ell_{q+1}^{(n)}, \dots, \ell_p^{(n)})' := n^{1/2}(\hat{\lambda}_{q+1} - 1, \dots, \hat{\lambda}_p - 1)' \quad (\text{S2.1})$$

converges weakly to the $(p-q)$ smallest roots of $\mathbf{Z}_{22} + \text{diag}(v_{s_2+1}, \dots, v_{s_3}, \mathbf{0}'_{q-s_3}, \mathbf{0}'_{p-q})$.

Proof of Proposition 1. Throughout the proof, we put

$$\mathbf{Z}^{(n)} := n^{1/2} \boldsymbol{\beta}' (\mathbf{S}^{(n)} - \boldsymbol{\Sigma}^{(n)}) \boldsymbol{\beta} = \begin{pmatrix} \mathbf{Z}_{11}^{(n)} & (\mathbf{Z}_{21}^{(n)})' \\ \mathbf{Z}_{21}^{(n)} & \mathbf{Z}_{22}^{(n)} \end{pmatrix},$$

where $\mathbf{Z}_{11}^{(n)}$ is the $s_2 \times s_2$ upper left block of $\mathbf{Z}^{(n)}$, $\mathbf{Z}_{22}^{(n)}$ the $p - s_2 \times p - s_2$ lower right block, etc. It follows along the same lines as in Lemma 2.1 of Paindaveine et al. (2020) that $\mathbf{Z}^{(n)}$ converges weakly to $\mathbf{Z}(v_1, \dots, v_{s_1})$ with

$$\text{vec}(\mathbf{Z}(v_1, \dots, v_{s_1})) \sim \mathcal{N}_{p^2}(\mathbf{0}, (\mathbf{I}_{p^2} + \mathbf{K}_p)(\text{diag}(1 + v_1, \dots, 1 + v_{s_1}, \mathbf{1}'_{p-s_1}))^{\otimes 2})$$

under $P_{\beta, \lambda}^{(n)}$. First note that for every $q + 1 \leq j \leq p$, $\ell_j^{(n)}$ is the j th largest root of $(\mathbf{\Lambda}^{(n)} = \beta' \Sigma^{(n)} \beta)$

$$\begin{aligned} P_{q+1, n}(h) &= \det(n^{1/2}(\mathbf{S}^{(n)} - \mathbf{I}_p) - h\mathbf{I}_p) \\ &= \det(\mathbf{Z}^{(n)} + n^{1/2}(\mathbf{\Lambda}^{(n)} - \mathbf{I}_p) - h\mathbf{I}_p) = \det(\mathbf{Z}^{(n)} + n^{1/2} \text{diag}((\text{diag}((\mathbf{r}^{(n)})')\mathbf{v})', \mathbf{0}'_{p-q}) - h\mathbf{I}_p). \end{aligned}$$

It follows that $(\ell_{q+1}^{(n)}, \dots, \ell_p^{(n)})$ are the smallest roots of

$$\begin{aligned} \frac{P_{q+1, n}(h)}{(n^{1/2}r_{s_2}^{(n)}v_{s_2})^{s_2}} &= \frac{1}{(n^{1/2}r_{s_2}^{(n)}v_{s_2})^{s_2}} \det(\mathbf{Z}^{(n)} + n^{1/2} \text{diag}((\text{diag}((\mathbf{r}^{(n)})')\mathbf{v})', \mathbf{0}'_{p-q})) - h\mathbf{I}_p) \\ &= \det \left(\begin{array}{cc} (n^{1/2}r_{s_2}^{(n)}v_{s_2})^{-1}(\mathbf{Z}_{11}^{(n)} + \text{diag}(n^{1/2}r_1^{(n)}v_1 - h, \dots, n^{1/2}r_{s_2}^{(n)}v_{s_2} - h)) & (\mathbf{Z}_{21}^{(n)})' \\ & (n^{1/2}r_{s_2}^{(n)}v_{s_2})^{-1}(\mathbf{Z}_{21}^{(n)}) & \mathbf{Z}_{22}^{(n)} \end{array} \right) \\ &\quad + \text{diag}(\mathbf{0}'_{s_2}, v_{s_2+1} - h, \dots, v_{s_3} - h, n^{1/2}r_{s_3+1}^{(n)}v_{s_3+1} - h, \dots, n^{1/2}r_q^{(n)}v_q - h, -h\mathbf{1}'_{p-q}). \end{aligned}$$

Since $(n^{1/2}r_{s_2}^{(n)}v_{s_2})^{-1} = o(1)$, $(n^{1/2}r_{s_2}^{(n)}v_{s_2})^{-1}\mathbf{Z}_{21}^{(n)} = o_P(1)$ and $(n^{1/2}r_{s_2}^{(n)}v_{s_2})^{-1}\mathbf{Z}_{11}^{(n)} = o_P(1)$ as $n \rightarrow \infty$. It follows that the s_2 largest roots of $(n^{1/2}r_{s_2}^{(n)}v_{s_2})^{-s_2}P_{q+1, n}(h)$ converge to $+\infty$ in probability while the $p - s_2$ smallest roots converge weakly to the roots (in the decreasing order) of the weak limit of the poly-

nomial

$$\det(\mathbf{Z}_{22}^{(n)} - h\mathbf{I}_{p-s_2} + \text{diag}(v_{s_2+1}, \dots, v_{s_3}, n^{1/2}r_{s_3+1}^{(n)}v_{s_3+1}, \dots, n^{1/2}r_q^{(n)}v_q, \mathbf{0}'_{p-q}));$$

that is the vector of the ordered eigenvalues of

$$\mathbf{Z}_{22} + \text{diag}(v_{s_2+1}, \dots, v_{s_3}, \mathbf{0}'_{q-s_3}, \mathbf{0}'_{p-q}).$$

This implies that the vector $(\ell_{q+1}^{(n)}, \dots, \ell_p^{(n)})$ converges weakly to the $(p-q)$ smallest eigenvalues of

$$\mathbf{Z}_{22} + \text{diag}(v_{s_2+1}, \dots, v_{s_3}, \mathbf{0}'_{q-m}, \mathbf{0}'_{p-q})$$

which is the desired result. □

In the proof of Proposition 2, we will use the following preliminary Lemma that follows along the same lines as Lemma S.1.1 in Paindaveine et al. (2020).

Lemma 1. *Let \mathbf{A} be a $p \times p$ matrix. Assume that λ is an eigenvalue of \mathbf{A} and that the corresponding eigenspace V_λ has dimension one. Denoting as $C = (C_{ij})$ the cofactor matrix of $\mathbf{A} - \lambda\mathbf{I}_p$, assume that for $1 \leq j \leq p$ $\mathbf{v} := (C_{j1}, \dots, C_{jp})' \neq \mathbf{0}$. Then $V_\lambda = \{t\mathbf{v} : t \in \mathbb{R}\}$.*

Proof of Proposition 2. First, note that using the fact that

$$\mathbf{E}^{(n)} = \begin{pmatrix} \mathbf{E}_{11}^{(n)} & \mathbf{E}_{12}^{(n)} \\ \mathbf{E}_{21}^{(n)} & \mathbf{E}_{22}^{(n)} \end{pmatrix}$$

is an orthogonal matrix, we have that

$$\mathbf{E}_{21}^{(n)}(\mathbf{E}_{11}^{(n)})' = -\mathbf{E}_{22}^{(n)}(\mathbf{E}_{12}^{(n)})' \quad \text{and} \quad \mathbf{E}_{22}^{(n)}(\mathbf{E}_{22}^{(n)})' = \mathbf{I}_{p-q} - \mathbf{E}_{21}^{(n)}(\mathbf{E}_{21}^{(n)})'. \quad (\text{S2.2})$$

We first will prove (a). Following the notations of the proof of Proposition 1, we put

$$\mathbf{Z}^{(n)} := n^{1/2}\boldsymbol{\beta}'(\mathbf{S}^{(n)} - \boldsymbol{\Sigma}^{(n)})\boldsymbol{\beta} = \begin{pmatrix} \mathbf{Z}_{11}^{(n)} & (\mathbf{Z}_{21}^{(n)})' \\ \mathbf{Z}_{21}^{(n)} & \mathbf{Z}_{22}^{(n)} \end{pmatrix},$$

where $\mathbf{Z}_{11}^{(n)}$ the $q \times q$ upper-left block and $\mathbf{Z}_{22}^{(n)}$ the $p - q \times p - q$ lower-right block of $\mathbf{Z}^{(n)}$. For $1 \leq j \leq q$, we put

$$\hat{\boldsymbol{\theta}}_j^{(n)} := \boldsymbol{\beta}'\hat{\boldsymbol{\beta}}_j = (\mathbf{b}_j^{(n)}, \mathbf{e}_j^{(n)})',$$

where $\mathbf{b}_j^{(n)}$ is the j th line of $\mathbf{E}_{11}^{(n)}$ and $\mathbf{e}_j^{(n)}$ the j th line of $\mathbf{E}_{12}^{(n)}$. We also use the notation

$$\boldsymbol{\ell}_q^{(n)} = (\ell_1^{(n)}, \dots, \ell_q^{(n)}) := n^{1/2}(\hat{\lambda}_1 - (1 + r_1^{(n)}v_1), \dots, \hat{\lambda}_q - (1 + r_q^{(n)}v_q)) \quad (\text{S2.3})$$

as in Proposition 1. For any $1 \leq j \leq q$, it is easy to verify that $\ell_j^{(n)}$ is the j th (largest) eigenvalue of

$$n^{1/2}\boldsymbol{\beta}'\mathbf{S}^{(n)}\boldsymbol{\beta} - n^{1/2}(1 + r_j^{(n)}v_j)\mathbf{I}_p = \mathbf{Z}^{(n)} - \text{diag}(k_{j1}^{(n)}, \dots, k_{jq}^{(n)}, n^{1/2}r_j^{(n)}v_j\mathbf{1}_{p-q}'),$$

where $k_{j\ell}^{(n)} := n^{1/2}(r_j^{(n)}v_j - r_\ell^{(n)}v_\ell)$; obviously $k_{jj}^{(n)} = 0$. The eigenvector associated with $\ell_j^{(n)}$ is $\hat{\boldsymbol{\theta}}_j^{(n)}$. Therefore, using Lemma 1, we have that $\hat{\boldsymbol{\theta}}_j^{(n)}$ is

proportional to the vector of the cofactors obtained with respect to the j th row of $(\mathbf{K}^{(n)} := \text{diag}(k_{j1}^{(n)}, \dots, k_{jq}^{(n)}))$

$$\begin{pmatrix} \mathbf{Z}_{11}^{(n)} - \mathbf{K}^{(n)} - \ell_j^{(n)} \mathbf{I}_q & (\mathbf{Z}_{21}^{(n)})' \\ \mathbf{Z}_{21}^{(n)} & \mathbf{Z}_{22}^{(n)} - n^{1/2} r_j^{(n)} v_j \mathbf{I}_{p-q} - \ell_j^{(n)} \mathbf{I}_{p-q} \end{pmatrix},$$

or equivalently, proportional to the vector of the cofactors obtained with respect to the j th row of

$$\left(\prod_{\ell=1}^q c_{j\ell}^{(n)} (n^{1/2} r_j^{(n)} v_j)^{-(p-q)} \begin{pmatrix} \mathbf{Z}_{11}^{(n)} - \mathbf{K}^{(n)} - \ell_j^{(n)} \mathbf{I}_q & (\mathbf{Z}_{21}^{(n)})' \\ \mathbf{Z}_{21}^{(n)} & \mathbf{Z}_{22}^{(n)} - n^{1/2} r_j^{(n)} v_j \mathbf{I}_{p-q} - \ell_j^{(n)} \mathbf{I}_{p-q} \end{pmatrix} \right), \quad (\text{S2.4})$$

where

$$c_{j\ell}^{(n)} := \begin{cases} (k_{j\ell}^{(n)})^{-1} & \text{if } \lim_{n \rightarrow \infty} k_{j\ell}^{(n)} \neq 0 \\ 1 & \text{else.} \end{cases}$$

Note that since $n^{1/2} r_q^{(n)}$ diverges to ∞ , provided that $\lim_{n \rightarrow \infty} r_j^{(n)} v_j - r_\ell^{(n)} v_\ell \neq 0$, $(k_{j\ell}^{(n)})^{-1} = c_{j\ell}^{(n)}$ converges to zero as $n \rightarrow \infty$ while obviously $c_{jj}^{(n)} = 1$. Following (S2.4), letting $\mathbf{C}^{(n)} := \text{diag}(c_{j1}^{(n)}, \dots, c_{jq}^{(n)})$, classical algebra yields to the fact that $\hat{\boldsymbol{\theta}}_j^{(n)}$ is also proportional to the vector of the cofactors obtained with respect to the j th row of

$$\begin{pmatrix} (\mathbf{C}^{(n)} (\mathbf{Z}_{11}^{(n)} - \ell_j^{(n)} \mathbf{I}_q) - \mathbf{C}^{(n)} \mathbf{K}^{(n)}) & \mathbf{C}^{(n)} (\mathbf{Z}_{21}^{(n)})' \\ (n^{1/2} r_j^{(n)} v_j)^{-1} \mathbf{Z}_{21}^{(n)} & (n^{1/2} r_j^{(n)} v_j)^{-1} \mathbf{Z}_{22}^{(n)} - \mathbf{I}_{p-q} - \frac{\ell_j^{(n)}}{n^{1/2} r_j^{(n)} v_j} \mathbf{I}_{p-q} \end{pmatrix}. \quad (\text{S2.5})$$

It is easy to check that $(\mathbf{b}_j^{(n)})' = O_P(1)$ (and not $o_P(1)$) while the cofactors associated with the j th line and the ℓ th column of the above matrix with $q < \ell \leq p$ are all $O_P((n^{1/2}r_j^{(n)})^{-1})$; indeed these cofactors are obtained by computing the determinants of matrices containing a line with only $O_P((n^{1/2}r_j^{(n)})^{-1})$ elements or zeros while the other entries are $O_P(1)$ as $n \rightarrow \infty$. It follows that $\mathbf{e}_j^{(n)} = O_P((n^{1/2}r_j^{(n)})^{-1})$. Point (a) then follows easily from (S2.2). We now turn to point (b). For point (b), we will show the result for the q th line of $\mathbf{E}_{12}^{(n)}$ which is obviously enough. First note that since $n^{1/2}r_q^{(n)} \rightarrow c < \infty$,

$$n^{1/2}(\hat{\lambda}_q - 1) = \ell_q^{(n)} + n^{1/2}r_q^{(n)}v_q$$

is (i) $O_P(1)$ and (ii) the q th (largest) eigenvalue of

$$\mathbf{Z}^{(n)} + \text{diag}(n^{1/2}r_1^{(n)}v_1, \dots, n^{1/2}r_q^{(n)}v_q, \mathbf{0}'_{p-q}) = \mathbf{Z}^{(n)} + \text{diag}(n^{1/2}(\text{diag}((\mathbf{r}^{(n)})'\mathbf{v})'), \mathbf{0}'_{p-q}).$$

The associated eigenvector is $\hat{\boldsymbol{\theta}}_q^{(n)} = \boldsymbol{\beta}'\hat{\boldsymbol{\beta}}_q = (\mathbf{b}_q^{(n)}, \mathbf{e}_q^{(n)})'$, using the same notations as in point (a) above. Using again the same technique as for point (a) above, Lemma 1 entails that $\hat{\boldsymbol{\theta}}_q^{(n)}$ is proportional to the vector of the cofactors obtained with respect to the first row of

$$\left(\prod_{\ell=1}^{s^*} (n^{1/2}r_\ell^{(n)}v_\ell)^{-1} \begin{pmatrix} \mathbf{Z}_{11}^{(n)} + n^{1/2}\text{diag}((\text{diag}((\mathbf{r}^{(n)})'\mathbf{v})') - n^{1/2}(\hat{\lambda}_q - 1)\mathbf{I}_q) & (\mathbf{Z}_{21}^{(n)})' \\ \mathbf{Z}_{21}^{(n)} & \mathbf{Z}_{22}^{(n)} - n^{1/2}(\hat{\lambda}_q - 1)\mathbf{I}_{p-q} \end{pmatrix} \right), \quad (\text{S2.6})$$

where $s^* = \max\{j \in \{1, \dots, q\}, n^{1/2}r_j^{(n)} \rightarrow \infty\}$ ($s^* \equiv 0$ if the maximum does not exist). Letting

$$\mathbf{R}^{(n)} := \text{diag}((n^{1/2}r_1^{(n)}v_1)^{-1}, \dots, (n^{1/2}r_{s^*}^{(n)}v_{s^*})^{-1}, \mathbf{1}'_{q-s^*}),$$

it follows from (S2.6) that $\hat{\boldsymbol{\theta}}_q^{(n)}$ is proportional to the vector of the cofactors obtained with respect to the q th row of

$$\mathbf{M}^{(n)} := \begin{pmatrix} \mathbf{R}^{(n)}\mathbf{Z}_{11}^{(n)} + n^{1/2}\mathbf{R}^{(n)}\text{diag}((\text{diag}((\mathbf{r}^{(n)})'\mathbf{v}))' - n^{1/2}(\hat{\lambda}_q - 1)\mathbf{R}^{(n)} & \mathbf{R}^{(n)}(\mathbf{Z}_{21}^{(n)})' \\ \mathbf{Z}_{21}^{(n)} & \mathbf{Z}_{22}^{(n)} - n^{1/2}(\hat{\lambda}_q - 1)\mathbf{I}_{p-q} \end{pmatrix}.$$

Since $\mathbf{Z}^{(n)}$ and $n^{1/2}(\hat{\lambda}_q - 1)$ are both $O_P(1)$ (and not $o_P(1)$) as $n \rightarrow \infty$, we readily have that

$$\mathbf{M}^{(n)} := \begin{pmatrix} \mathbf{I}_{s^*} & \mathbf{0} \\ \mathbf{Y}_{11}^{(n)} & \mathbf{Y}_{12}^{(n)} \\ \mathbf{Z}_{21}^{(n)} & \mathbf{Z}_{22}^{(n)} - n^{1/2}(\hat{\lambda}_q - 1)\mathbf{I}_{p-q} \end{pmatrix} + o_P(1)$$

as $n \rightarrow \infty$, where $(\mathbf{Y}_{11}^{(n)} : \mathbf{Y}_{12}^{(n)})$ is a $(q - s^*) \times p$ matrix whose elements are $O_P(1)$ (and not $o_P(1)$). It is then easy to see that the cofactor associated with the q th line and the $(q + 1)$ th column of the above matrix is not $o_P(1)$.

Point (b) follows. □

Before providing the proof of Proposition 3, the following Lemma can be obtained along the same lines as Proposition 1.

Lemma 2. *Let $\mathbf{r}^{(n)}$, \mathbf{v} and \mathbf{Z} be as in Proposition 1. Then we have that as $n \rightarrow \infty$ under $P_{\beta, \lambda^{(n)}}^{(n)}$ with $\lambda^{(n)}$ as in (3.1), $((\boldsymbol{\ell}_q^{(n)})', (\boldsymbol{\ell}_{p-q}^{(n)})')$ defined through (S2.1) and (S2.3) is $O_P(1)$ as $n \rightarrow \infty$.*

Proof of Proposition 3. In this proof, we put $\hat{\mathbf{\Lambda}}^{(n)} = \text{diag}(\hat{\mathbf{\Lambda}}_q^{(n)}, \hat{\mathbf{\Lambda}}_{p-q}^{(n)}) = \hat{\boldsymbol{\beta}}' \mathbf{S}^{(n)} \hat{\boldsymbol{\beta}}$, where $\hat{\mathbf{\Lambda}}_q^{(n)} := \text{diag}(\hat{\lambda}_1, \dots, \hat{\lambda}_q)$ and $\hat{\mathbf{\Lambda}}_{p-q}^{(n)} := \text{diag}(\hat{\lambda}_{q+1}, \dots, \hat{\lambda}_p)$ and $\mathbf{R}^{(n)} := \text{diag}((\mathbf{r}^{(n)})')$. We have that

$$\mathbf{S}_{\mathbf{Y}}^{(n)} = (\mathbf{0}_{(p-q) \times q} \vdots \mathbf{I}_{p-q}) (\mathbf{E}^{(n)})' \hat{\mathbf{\Lambda}}^{(n)} \mathbf{E}^{(n)} (\mathbf{0}_{(p-q) \times q} \vdots \mathbf{I}_{p-q})' = ((\mathbf{E}_{12}^{(n)})' \vdots (\mathbf{E}_{22}^{(n)})') \hat{\mathbf{\Lambda}}^{(n)} ((\mathbf{E}_{12}^{(n)})' \vdots (\mathbf{E}_{22}^{(n)})')'. \quad (\text{S2.7})$$

This entails that

$$\begin{aligned} n^{1/2}(\mathbf{S}_{\mathbf{Y}}^{(n)} - \mathbf{I}_{p-q}) &= n^{1/2}(((\mathbf{E}_{12}^{(n)})' \vdots (\mathbf{E}_{22}^{(n)})') \hat{\mathbf{\Lambda}}^{(n)} ((\mathbf{E}_{12}^{(n)})' \vdots (\mathbf{E}_{22}^{(n)})')' - \mathbf{I}_{p-q}) \\ &= n^{1/2}((\mathbf{E}_{12}^{(n)})' \hat{\mathbf{\Lambda}}_q^{(n)} \vdots (\mathbf{E}_{22}^{(n)})' \hat{\mathbf{\Lambda}}_{p-q}^{(n)}) ((\mathbf{E}_{12}^{(n)})' \vdots (\mathbf{E}_{22}^{(n)})')' - \mathbf{I}_{p-q}). \end{aligned}$$

Now, since $\mathbf{E}^{(n)}$ is orthogonal, $(\mathbf{E}_{12}^{(n)})' \mathbf{E}_{12}^{(n)} + (\mathbf{E}_{22}^{(n)})' \mathbf{E}_{22}^{(n)} = \mathbf{I}_{p-q}$, which implies, that letting

$$\begin{aligned} n^{1/2}(\mathbf{S}_{\mathbf{Y}}^{(n)} - \mathbf{I}_{p-q}) &= n^{1/2}((\mathbf{E}_{12}^{(n)})' (\hat{\mathbf{\Lambda}}_q^{(n)} - \mathbf{I}_q) \mathbf{E}_{12}^{(n)} + (\mathbf{E}_{22}^{(n)})' (\hat{\mathbf{\Lambda}}_{p-q}^{(n)} - \mathbf{I}_{p-q}) \mathbf{E}_{22}^{(n)}) \\ &= (\mathbf{E}_{12}^{(n)})' (\mathbf{R}^{(n)})^{-1} (\hat{\mathbf{\Lambda}}_q^{(n)} - \mathbf{I}_q) n^{1/2} \mathbf{R}^{(n)} \mathbf{E}_{12}^{(n)} + (\mathbf{E}_{22}^{(n)})' (\hat{\mathbf{\Lambda}}_{p-q}^{(n)} - \mathbf{I}_{p-q}) \mathbf{E}_{22}^{(n)} \end{aligned}$$

Note that for each $1 \leq j \leq q$, $(r_j^{(n)})^{-1} (\hat{\lambda}_j - 1) = (r_j^{(n)})^{-1} (\hat{\lambda}_j - 1 - r_j^{(n)} v_j) + v_j$.

Since $n^{1/2} r_j^{(n)} \rightarrow \infty$, Lemma 2 entails that

$$(r_j^{(n)})^{-1} (\hat{\lambda}_j - 1 - r_j^{(n)} v_j) = (n^{1/2} r_j^{(n)})^{-1} n^{1/2} (\hat{\lambda}_j - 1 - r_j^{(n)} v_j)$$

is $o_{\mathbb{P}}(1)$ as $n \rightarrow \infty$ so that since from Proposition 2 (point (a)), $n^{1/2}\mathbf{R}^{(n)}\mathbf{E}_{12}^{(n)}$ is $O_{\mathbb{P}}(1)$ as $n \rightarrow \infty$, we have that

$$n^{1/2}(\mathbf{S}_{\mathbf{Y}}^{(n)} - \mathbf{I}_{p-q}) = n^{1/2}(\mathbf{E}_{22}^{(n)})'(\hat{\boldsymbol{\Lambda}}_{p-q}^{(n)} - \mathbf{I}_{p-q})\mathbf{E}_{22}^{(n)} + o_{\mathbb{P}}(1) \quad (\text{S2.8})$$

as $n \rightarrow \infty$. Now, using (S2.8),

$$\begin{aligned} T_q^{(n)}(\boldsymbol{\beta}) &= \frac{n}{2} \left(\frac{(p-q)}{\text{tr}(\mathbf{S}_{\mathbf{Y}}^{(n)})} \right)^2 \left(\text{tr}((\mathbf{S}_{\mathbf{Y}}^{(n)})^2) - \frac{1}{p-q} \text{tr}^2(\mathbf{S}_{\mathbf{Y}}^{(n)}) \right) \\ &= \frac{n}{2} \left(\frac{(p-q)}{\text{tr}(\mathbf{S}_{\mathbf{Y}}^{(n)})} \right)^2 \left(\text{tr}((\mathbf{S}_{\mathbf{Y}}^{(n)} - \mathbf{I}_{p-q})^2) - \frac{1}{p-q} \text{tr}^2(\mathbf{S}_{\mathbf{Y}}^{(n)} - \mathbf{I}_{p-q}) \right) \\ &= \frac{1}{2} \left(\frac{(p-q)}{\sum_{j=q+1}^p \hat{\lambda}_j} \right)^2 \left(\text{tr}((n^{1/2}(\mathbf{E}_{22}^{(n)})'(\hat{\boldsymbol{\Lambda}}_{p-q}^{(n)} - \mathbf{I}_{p-q})\mathbf{E}_{22}^{(n)})^2) \right. \\ &\quad \left. - \frac{1}{p-q} \text{tr}^2(n^{1/2}(\mathbf{E}_{22}^{(n)})'(\hat{\boldsymbol{\Lambda}}_{p-q}^{(n)} - \mathbf{I}_{p-q})\mathbf{E}_{22}^{(n)}) \right) + o_{\mathbb{P}}(1), \quad (\text{S2.9}) \end{aligned}$$

as $n \rightarrow \infty$. Using the fact that Proposition 2 yields $\mathbf{E}_{22}^{(n)}(\mathbf{E}_{22}^{(n)})' = \mathbf{I}_{p-q} + o_{\mathbb{P}}(1)$, (S2.9) and the Slutsky Lemma yield

$$\begin{aligned} T_q^{(n)}(\boldsymbol{\beta}) &= \frac{1}{2} \left(\frac{(p-q)}{\sum_{j=q+1}^p \hat{\lambda}_j} \right)^2 \left(\sum_{j=q+1}^p (n^{1/2}(\hat{\lambda}_j - 1))^2 - \frac{1}{p-q} \left(\sum_{j=q+1}^p n^{1/2}(\hat{\lambda}_j - 1) \right)^2 \right) + o_{\mathbb{P}}(1) \\ &= T_q^{(n)} + o_{\mathbb{P}}(1) \end{aligned}$$

under $\mathbb{P}_{\boldsymbol{\beta}, \boldsymbol{\lambda}^{(n)}}^{(n)}$ as $n \rightarrow \infty$, which ends the proof. \square

Proof of Proposition 4. Fix $0 < q \leq p-2$ (the case $q = 0$ is trivial). Letting

$$\boldsymbol{\ell}_{2,q}^{(n)} := (\ell_q^{(n)}, \ell_{q+1}^{(n)}) = n^{1/2}((\hat{\lambda}_q - (1 + r_q^{(n)}v_q), (\hat{\lambda}_{q+1} - 1)),$$

we have that (with $\mathbf{e}_1 = (1, 0) \in \mathbb{R}^2$)

$$\begin{aligned}
 T_{q,q+1}^{(n)} &= \frac{n(\sum_{j=q}^{q+1} \hat{\lambda}_j^2 - \frac{1}{2}(\sum_{j=q}^{q+1} \hat{\lambda}_j)^2)}{\frac{1}{2}(\sum_{j=q}^{q+1} \hat{\lambda}_j)^2} \\
 &= \frac{n(\sum_{j=q}^{q+1} (\hat{\lambda}_j - 1)^2 - \frac{1}{2}(\sum_{j=q}^{q+1} (\hat{\lambda}_j - 1))^2)}{\frac{1}{2}(\sum_{j=q}^{q+1} \hat{\lambda}_j)^2} \\
 &= \frac{(\boldsymbol{\ell}_{2,q}^{(n)} + n^{1/2}r_q^{(n)}v_q\mathbf{e}_1)'(\mathbf{I}_2 - \mathbf{1}_2(\mathbf{1}'_2\mathbf{1}_2)^{-1}\mathbf{1}'_2)(\boldsymbol{\ell}_{2,q}^{(n)} + n^{1/2}r_q^{(n)}v_q\mathbf{e}_1)}{\frac{1}{2}(\sum_{j=q}^{q+1} \hat{\lambda}_j)^2}
 \end{aligned}$$

so that since from Lemma 2, $\boldsymbol{\ell}_{2,q}^{(n)}$ is $O_P(1)$ as $n \rightarrow \infty$ under $\mathbb{P}_{\boldsymbol{\beta}, \boldsymbol{\lambda}^{(n)}}^{(n)}$,

$$T_{q,q+1}^{(n)} = \frac{(\boldsymbol{\ell}_{2,q}^{(n)} + n^{1/2}r_q^{(n)}v_q\mathbf{e}_1)'(\mathbf{I}_2 - \mathbf{1}_2(\mathbf{1}'_2\mathbf{1}_2)^{-1}\mathbf{1}'_2)(\boldsymbol{\ell}_{2,q}^{(n)} + n^{1/2}r_q^{(n)}v_q\mathbf{e}_1)}{\frac{1}{2}(\sum_{j=q}^{q+1} \hat{\lambda}_j)^2} \tag{S2.10}$$

converges to $+\infty$ in probability when $n^{1/2}r_q^{(n)} \rightarrow \infty$. We therefore have

that, for any $\gamma \in (0, 1)$,

$$\begin{aligned}
 \mathbb{E}[|\phi_{\text{new}}^{(n)} - \phi^{(n)}|] &= \mathbb{P}[(T_q^{(n)} < \chi_{d(p,q);1-\alpha}^2) \cap (T_{q,q+1}^{(n)} \leq \chi_{2;1-\gamma}^2)] \\
 &\leq \mathbb{P}[T_{q,q+1}^{(n)} \leq \chi_{2;1-\gamma}^2],
 \end{aligned}$$

so that $\mathbb{E}[|\phi_{\text{new}}^{(n)} - \phi^{(n)}|]$ converges to zero when $n^{1/2}r_q^{(n)} \rightarrow \infty$. □

S3. Consistency of \hat{k}_{new}

In this Section, we provide the consistency (under some conditions) of the estimator \hat{k}_{new} .

Proposition 7. *Let $c^{(n)}$ and $b_0^{(n)}, \dots, b_{p-2}^{(n)}$ be positive sequences that diverge to ∞ and are such that (i) $c^{(n)} = o(n)$ and $b_q^{(n)} = o(n)$ as $n \rightarrow \infty$ for $q = 0, \dots, p-2$ and (ii) $(\max(c^{(n)}, b_0^{(n)}, \dots, b_{p-2}^{(n)}))^{-1/2} n^{1/2} (\lambda_k^{(n)} - \lambda_{k+1}^{(n)})$ diverges to ∞ as $n \rightarrow \infty$. Then $\lim_{n \rightarrow \infty} \mathbb{P}(\hat{k}_{\text{new}} = k) = 1$.*

To simplify the interpretation of Proposition 2, assume that the sequences $c^{(n)}$ and $b_0^{(n)} = \dots = b_{p-2}^{(n)} \equiv c^{(n)}$ are all the same. Proposition 2 then shows that provided that $c^{(n)}$ does not diverge too quickly to ∞ in the sense that $c^{(n)} = o(n)$ and $(c^{(n)})^{-1/2} n^{1/2} (\lambda_k^{(n)} - \lambda_{k+1}^{(n)})$ diverges to ∞ as $n \rightarrow \infty$, the resulting estimator \hat{k}_{new} is consistent.

Proof of Proposition 2. Fix $k < p-1$ (the case $k = p-1$ is considered at the end of the proof). In the proof, we assume without loss of generality that $\boldsymbol{\lambda}^{(n)}$ is as in (3.1) and work under $\mathbb{P}_{\boldsymbol{\beta}, \boldsymbol{\lambda}^{(n)}}^{(n)}$ with $\boldsymbol{\lambda}^{(n)}$ such that $\mathcal{H}_{0k}^{(n)}$ holds. Therefore, since

$$(\max(c^{(n)}, b_0^{(n)}, \dots, b_{p-2}^{(n)}))^{-1/2} n^{1/2} (\lambda_k^{(n)} - \lambda_{k+1}^{(n)}) \rightarrow \infty$$

as $n \rightarrow \infty$, we have that $(c^{(n)})^{-1/2} n^{1/2} (\lambda_k^{(n)} - \lambda_{k+1}^{(n)}) \rightarrow \infty$ as $n \rightarrow \infty$. We then have using the same notations as in (S2.10) that

$$\frac{T_{k,k+1}^{(n)}}{c^{(n)}} = \frac{(c^{(n)})^{-1/2} (\boldsymbol{\ell}_{2,k}^{(n)} + n^{1/2} r_k^{(n)} v_k \mathbf{e}_1)' (\mathbf{I}_2 - \mathbf{1}_2 (\mathbf{1}'_2 \mathbf{1}_2)^{-1} \mathbf{1}'_2) (c^{(n)})^{-1/2} (\boldsymbol{\ell}_{2,k}^{(n)} + n^{1/2} r_k^{(n)} v_k \mathbf{e}_1)}{\frac{1}{2} (\sum_{q=k}^{k+1} \hat{\lambda}_q)^2} \quad (\text{S3.11})$$

converges to $+\infty$ in probability for $k > 0$. When $k = 0$, $T_{0,1}^{(n)}$ is arbitrarily

taken such that $T_{0,1}^{(n)} > c^{(n)}$ ($T_{0,1}^{(n)}$ can be arbitrarily fixed). Now, since $\mathcal{H}_{0k}^{(n)}$ holds, we also have by (1) that

$$(\ell_{k+1}^{(n)}, \dots, \ell_p^{(n)}) = n^{1/2}(\hat{\lambda}_{k+1} - 1, \dots, \hat{\lambda}_p - 1) = O_{\mathbb{P}}(1)$$

as $n \rightarrow \infty$. Following the same lines as in S3.11 then yields that

$$(b_k^{(n)})^{-1}T_k^{(n)} = o_{\mathbb{P}}(1) \tag{S3.12}$$

as $n \rightarrow \infty$. Combining S3.11 and S3.12, we obtain that under $\mathbb{P}_{\boldsymbol{\beta}, \boldsymbol{\lambda}^{(n)}}^{(n)}$ with $\boldsymbol{\lambda}^{(n)}$ such that $\mathcal{H}_{0k}^{(n)}$ holds,

$$\lim_{n \rightarrow \infty} \mathbb{P}(\mathbb{I}[T_k^{(n)} > b_k^{(n)}] \mathbb{I}[T_{k,k+1}^{(n)} > c^{(n)}] + \mathbb{I}[T_{k,k+1}^{(n)} \leq c^{(n)}] = 0) = 1. \tag{S3.13}$$

Now, for $0 \leq j < k$, since $(b_j^{(n)})^{-1/2}n^{1/2}(\lambda_k^{(n)} - \lambda_{k+1}^{(n)}) \rightarrow \infty$, working along the same line as S3.11, we obtain that $T_j^{(n)}/b_j^{(n)}$ converges to $+\infty$ in probability.

Therefore,

$$\lim_{n \rightarrow \infty} \mathbb{P}(\mathbb{I}[T_j^{(n)} > b_j^{(n)}] \mathbb{I}[T_{j,j+1}^{(n)} > c^{(n)}] + \mathbb{I}[T_{j,j+1}^{(n)} \leq c^{(n)}] = 1) = 1. \tag{S3.14}$$

as $n \rightarrow \infty$ under $\mathbb{P}_{\boldsymbol{\beta}, \boldsymbol{\lambda}^{(n)}}^{(n)}$ with $\boldsymbol{\lambda}^{(n)}$ such that $\mathcal{H}_{0k}^{(n)}$ holds. Since

$$\hat{k}_{\text{new}} = \min_{j \in \{0, \dots, p-2\}} \{\mathbb{I}[T_j^{(n)} > b_j^{(n)}] \mathbb{I}[T_{j,j+1}^{(n)} > c^{(n)}] + \mathbb{I}[T_{j,j+1}^{(n)} \leq c^{(n)}] = 0\},$$

it follows from S3.13 and S3.14 that $\hat{k}_{\text{new}} - k$ is $o_{\mathbb{P}}(1)$ as $n \rightarrow \infty$. Finally,

for $k = p - 1$, since

$$(\max(c^{(n)}, b_0^{(n)}, \dots, b_{p-2}^{(n)}))^{-1/2}n^{1/2}(\lambda_{p-1}^{(n)} - \lambda_p^{(n)}) \rightarrow \infty$$

as $n \rightarrow \infty$ we have that

$$\lim_{n \rightarrow \infty} \mathbb{P}(T_j^{(n)} \geq b_j^{(n)}) = 1 \quad (\text{S3.15})$$

for all $0 \leq j \leq p - 2$. The result follows. \square

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