

Appendix 0. Proof of equation (16)

The mean square error of the Liu linear shrinkage estimator is as follows:

$$\begin{aligned}
MSE(\hat{\beta}^{LLS}) &= E[(\hat{\beta}^{LLS} - \beta)(\hat{\beta}^{LLS} - \beta)'] \\
&= E[(\delta \hat{\beta}^{RL} + (1 - \delta) \hat{\beta}^{UR} - \beta)(\delta \hat{\beta}^{RL} + (1 - \delta) \hat{\beta}^{UR} - \beta)'] \\
&= E[(\hat{\beta}^{UR} - \beta - \delta(\hat{\beta}^{UR} - \hat{\beta}^{RL}))(\hat{\beta}^{UR} - \beta - \delta(\hat{\beta}^{UR} - \hat{\beta}^{RL}))'] \\
&= E[(\hat{\beta}^{UR} - \beta)(\hat{\beta}^{UR} - \beta)'] + \delta^2 E[(\hat{\beta}^{UR} - \hat{\beta}^{RL})(\hat{\beta}^{UR} - \hat{\beta}^{RL})'] \\
&\quad - 2\delta E[(\hat{\beta}^{UR} - \beta)(\hat{\beta}^{UR} - \hat{\beta}^{RL})']
\end{aligned}$$

We now derive from above equation with respect to δ as follows:

$$\frac{\partial MSE(\hat{\beta}^{LLS})}{\partial \delta} = 2\delta E[(\hat{\beta}^{UR} - \hat{\beta}^{RL})(\hat{\beta}^{UR} - \hat{\beta}^{RL})'] - 2E[(\hat{\beta}^{UR} - \beta)(\hat{\beta}^{UR} - \hat{\beta}^{RL})'] = 0$$

Therefore, the optimum value of δ is as

$$\delta_{optimal} = \frac{\overbrace{E[(\hat{\beta}^{UR} - \beta)(\hat{\beta}^{UR} - \hat{\beta}^{RL})']}^{q_1}}{\underbrace{E[(\hat{\beta}^{UR} - \hat{\beta}^{RL})(\hat{\beta}^{UR} - \hat{\beta}^{RL})']}_{q_2}} \quad (25)$$

Since $\hat{\beta}^{UR} = \mathbf{A}\hat{\beta}$ and $\hat{\beta}^{RL} = \mathbf{A}\hat{\beta}_{RMLE} = \mathbf{A}[\hat{\beta} - \mathbf{B}(\mathbf{H}\hat{\beta} - \mathbf{h})]$, where $\mathbf{B} = (\mathbf{X}'\widehat{\mathbf{W}}\mathbf{X} + \mathbf{I}_p)^{-1}(\mathbf{X}'\widehat{\mathbf{W}}\mathbf{X} + d\mathbf{I}_p)$ and $\mathbf{Z} = \mathcal{I}^{-1}\mathbf{H}'(\mathbf{H}\mathcal{I}^{-1}\mathbf{H}')^{-1}$. Therefore, q_1 can be written as:

$$\begin{aligned}
q_1 &= E[(\hat{\beta}^{UR} - \beta)(\hat{\beta}^{UR} - \hat{\beta}^{RL})'] \\
&= E[(\mathbf{B}\hat{\beta} - \beta)(\mathbf{B}\hat{\beta} - \mathbf{B}\hat{\beta}_{RMLE})'] \\
&= E[(\mathbf{B}\hat{\beta} - \beta)(\mathbf{B}\hat{\beta} - \mathbf{B}[\hat{\beta} - \mathbf{Z}(\mathbf{H}\hat{\beta} - \mathbf{h})])'] \\
&= E[(\mathbf{B}\hat{\beta} - \beta)(\mathbf{A}\mathbf{Z}(\mathbf{H}\hat{\beta} - \mathbf{h}))'] \\
&= \mathbf{B}E[\hat{\beta}\hat{\beta}']\mathbf{H}'\mathbf{Z}'\mathbf{B}' - \mathbf{B}E[\hat{\beta}]\mathbf{h}'\mathbf{Z}'\mathbf{B}' \\
&\quad - \beta E[\hat{\beta}']\mathbf{H}'\mathbf{Z}'\mathbf{B}' + \beta\mathbf{h}'\mathbf{Z}'\mathbf{B}'
\end{aligned}$$

and q_2 is

$$\begin{aligned}
q_2 &= E[(\hat{\beta}^{UR} - \hat{\beta}^{RL})(\hat{\beta}^{UR} - \hat{\beta}^{RL})'] \\
&= E[(\hat{\beta}^{UR} - \beta - (\hat{\beta}^{RL} - \beta))(\hat{\beta}^{UR} - \beta - (\hat{\beta}^{RL} - \beta))'] \\
&= E[(\hat{\beta}^{UR} - \beta)(\hat{\beta}^{UR} - \beta)'] - 2E[(\hat{\beta}^{UR} - \beta)(\hat{\beta}^{RL} - \beta)'] \\
&\quad + E[(\hat{\beta}^{RL} - \beta)(\hat{\beta}^{RL} - \beta)'] \\
&= \underbrace{\Sigma(\hat{\beta}^{UR})}_{q_3} - 2\underbrace{\Sigma(\hat{\beta}^{UR}, \hat{\beta}^{RL})}_{q_4} + \underbrace{\Sigma(\hat{\beta}^{RL})}_{q_5}
\end{aligned}$$

where $\Sigma(\hat{\beta}^{UR})$ and $\Sigma(\hat{\beta}^{RL})$ are variances of $\hat{\beta}^{UR}$ and $\hat{\beta}^{RL}$, respectively, and $\Sigma(\hat{\beta}^{UR}, \hat{\beta}^{RL})$ is the variance-covariance matrix of $\hat{\beta}^{UR}$ and $\hat{\beta}^{RL}$.

Now we can write q_3 as:

$$q_3 = \Sigma(\hat{\beta}^{UR}) = \Sigma(\mathbf{B}\hat{\beta}) = \mathbf{B}\Sigma(\hat{\beta})\mathbf{B}'$$

and q_4 is

$$\begin{aligned} q_4 &= \Sigma(\hat{\beta}^{UR}, \hat{\beta}^{RL}) \\ &= \Sigma(\mathbf{B}\hat{\beta}, \mathbf{B}\hat{\beta}_{RMLE}) \\ &= \Sigma(\mathbf{B}\hat{\beta}, \mathbf{B}[\hat{\beta} - \mathbf{Z}(\mathbf{H}\hat{\beta} - \mathbf{h})]) \\ &= \Sigma(\mathbf{B}\hat{\beta}, \mathbf{B}\hat{\beta}) - \Sigma(\mathbf{B}\hat{\beta}, \mathbf{B}\mathbf{Z}(\mathbf{H}\hat{\beta} - \mathbf{h})) \\ &= \mathbf{B}^2 \Sigma(\hat{\beta}) (\mathbf{B}^2)' - \mathbf{B}^2 \mathbf{Z} \mathbf{H} \Sigma(\hat{\beta}) \mathbf{H}' \mathbf{Z}' (\mathbf{B}^2)' \\ &= \mathbf{B}^2 \Sigma(\hat{\beta}) [\mathbf{I}_p - \mathbf{Z} \mathbf{H} \mathbf{H}' \mathbf{Z}'] (\mathbf{B}^2)' \end{aligned}$$

and q_5 is as follows:

$$\begin{aligned} q_5 &= \Sigma(\hat{\beta}^{RL}) \\ &= \Sigma(\mathbf{B}\hat{\beta}_{RMLE}) \\ &= \mathbf{B} \Sigma(\hat{\beta} - \mathbf{Z}(\mathbf{H}\hat{\beta} - \mathbf{h})) \mathbf{B}' \\ &= \mathbf{B} (\Sigma(\hat{\beta}) - \Sigma(\mathbf{Z} \mathbf{H} \hat{\beta})) \mathbf{B}' \\ &= \mathbf{B} (\Sigma(\hat{\beta}) - \mathbf{Z} \mathbf{H} \Sigma(\hat{\beta}) \mathbf{H}' \mathbf{Z}') \mathbf{B}' \\ &= \mathbf{B} \Sigma(\hat{\beta}) (\mathbf{I}_p - \mathbf{Z} \mathbf{H} \mathbf{H}' \mathbf{Z}') \mathbf{B}' \end{aligned}$$

So, q_2 becomes

$$\begin{aligned} q_2 &= \mathbf{B} \Sigma(\hat{\beta}) \mathbf{B}' + \mathbf{B}^2 \Sigma(\hat{\beta}) [\mathbf{I}_p - \mathbf{Z} \mathbf{H} \mathbf{H}' \mathbf{Z}'] (\mathbf{B}^2)' \\ &\quad - 2 \mathbf{B} \Sigma(\hat{\beta}) [\mathbf{I}_p - \mathbf{Z} \mathbf{H} \mathbf{H}' \mathbf{Z}'] \mathbf{B} \\ &= \mathbf{B} \Sigma(\hat{\beta}) \mathbf{B}' [\mathbf{I}_p - (\mathbf{I}_p - \mathbf{Z} \mathbf{H} \mathbf{H}' \mathbf{Z}') (\mathbf{B} \mathbf{B}' - 2 \mathbf{I}_p)] \end{aligned}$$

Hence, Equation (25) becomes

$$\delta_{optimal} = \frac{\mathbf{B} E[\hat{\beta} \hat{\beta}'] \mathbf{H}' \mathbf{Z}' \mathbf{B}' - \mathbf{B} E[\hat{\beta}] \mathbf{h}' \mathbf{Z}' \mathbf{B}' - \beta E[\hat{\beta}'] \mathbf{H}' \mathbf{Z}' \mathbf{B}' + \beta \mathbf{h}' \mathbf{Z}' \mathbf{B}'}{\mathbf{B} \Sigma(\hat{\beta}) \mathbf{B}' [\mathbf{I}_p - (\mathbf{I}_p - \mathbf{Z} \mathbf{H} \mathbf{H}' \mathbf{Z}') (\mathbf{B} \mathbf{B}' - 2 \mathbf{I}_p)]}$$

Appendix 1. Proof of Lemma 3.1

We present the following lemma which are useful for the proof of the Theorems 3.2 and 3.3.

Lemma .1. Let $\mathbf{y}$ be a p_2 -dimensional random vector distributed as $\mathcal{N}_{p_2}(\boldsymbol{\mu}_y, \boldsymbol{\Sigma}_y)$. Then, for any measurable function φ , we have

$$E[\mathbf{y} \varphi(\mathbf{y}' \mathbf{y})] = \boldsymbol{\mu}_y E[\varphi(\chi_{p_2+2}^2(\Delta^*))], \quad (26)$$

$$E[\mathbf{y}' \mathbf{y} \varphi(\mathbf{y}' \mathbf{y})] = \boldsymbol{\Sigma}_y E[\varphi(\chi_{p_2+2}^2(\Delta^*))] + \boldsymbol{\mu}'_y \boldsymbol{\mu}_y E[\varphi(\chi_{p_2+4}^2(\Delta^*))], \quad (27)$$

where Δ^* is the non-centrality parameter.

Proof: See [Judge and Bock \(1987\)](#).

Under local alternatives given in Equation (21) we show $\boldsymbol{\kappa}_1^n$ is asymptotically normally distributed with mean and variance as follows

$$\begin{aligned} \boldsymbol{\kappa}_1^n &= \sqrt{n}(\widehat{\boldsymbol{\beta}}^{UR} - \boldsymbol{\beta}) \\ &= \sqrt{n}(\mathbf{B}\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}) \end{aligned}$$

As $n \rightarrow \infty$, $\widehat{\boldsymbol{\beta}} \sim \mathcal{N}_p(\mathbf{0}, \boldsymbol{\mathcal{I}}^{-1})$. Therefore, $\boldsymbol{\kappa}_1^n \xrightarrow{D} \boldsymbol{\kappa}_1 \sim \mathcal{N}_p\left(E(\boldsymbol{\kappa}_1), \text{Var}(\boldsymbol{\kappa}_1)\right)$ where

$$\begin{aligned} E(\boldsymbol{\kappa}_1) &= E[\mathbf{B}\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}] \\ &= \mathbf{B}\boldsymbol{\beta} - \boldsymbol{\beta} \\ &= (\mathbf{B} - \mathbf{I}_p)\boldsymbol{\beta} \end{aligned}$$

$$\begin{aligned} \text{Var}(\boldsymbol{\kappa}_1) &= \text{Var}[\mathbf{B}\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta}] \\ &= \mathbf{B} \text{Var}(\widehat{\boldsymbol{\beta}}) \mathbf{B}' \\ &= \mathbf{B} \boldsymbol{\mathcal{I}}^{-1} \mathbf{B}' \end{aligned}$$

We show $\boldsymbol{\kappa}_2^n$ is asymptotically normally distributed with mean and variance as follow

$$\begin{aligned} \boldsymbol{\kappa}_2^n &= \sqrt{n}(\mathbf{H}\widehat{\boldsymbol{\beta}}^{UR} - \mathbf{h}) \\ &= \sqrt{n}(\mathbf{H}\widehat{\boldsymbol{\beta}}^{UR} - \mathbf{H}\boldsymbol{\beta} + \mathbf{H}\boldsymbol{\beta} - \mathbf{h}) \\ &= \mathbf{H} \sqrt{n}(\widehat{\boldsymbol{\beta}}^{UR} - \boldsymbol{\beta}) + \underbrace{\sqrt{n}(\mathbf{H}\boldsymbol{\beta} - \mathbf{h})}_{\boldsymbol{\vartheta}} \\ &= \mathbf{H} \boldsymbol{\kappa}_1^n + \boldsymbol{\vartheta} \end{aligned}$$

as $n \rightarrow \infty$, $\boldsymbol{\kappa}_1^n \xrightarrow{D} \boldsymbol{\kappa}_1 \sim \mathcal{N}_p\left((\mathbf{B} - \mathbf{I}_p)\boldsymbol{\beta}, \mathbf{B} \boldsymbol{\mathcal{I}}^{-1} \mathbf{B}'\right)$. So, $\boldsymbol{\kappa}_2^n \xrightarrow{D} \boldsymbol{\kappa}_2 \sim \mathcal{N}_p\left(E(\boldsymbol{\kappa}_2), \text{Var}(\boldsymbol{\kappa}_2)\right)$ where

$$\begin{aligned} E(\boldsymbol{\kappa}_2) &= \mathbf{H} E(\boldsymbol{\kappa}_1) + \boldsymbol{\vartheta} \\ &= \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} + \boldsymbol{\vartheta} \end{aligned}$$

$$\begin{aligned} Var(\boldsymbol{\kappa}_2) &= \mathbf{H} Var(\boldsymbol{\kappa}_1) \mathbf{H}' \\ &= \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' \mathbf{H}' \end{aligned}$$

Then

$$\begin{aligned} \boldsymbol{\kappa}_3 &= \sqrt{n}(\hat{\boldsymbol{\beta}}^{RL} - \boldsymbol{\beta}) \\ &= \sqrt{n}(\hat{\boldsymbol{\beta}}^{UR} - \mathcal{Z}(\mathbf{H}\hat{\boldsymbol{\beta}}^{UR} - \mathbf{h}) - \boldsymbol{\beta}) \\ &= \sqrt{n}(\hat{\boldsymbol{\beta}}^{UR} - \boldsymbol{\beta}) - \mathcal{Z}\sqrt{n}(\mathbf{H}\hat{\boldsymbol{\beta}}^{UR} - \mathbf{h}) \\ &= \boldsymbol{\kappa}_1^n - \mathcal{Z} \boldsymbol{\kappa}_2^n \\ &= \boldsymbol{\kappa}_1^n - \mathcal{Z} (\mathbf{H} \boldsymbol{\kappa}_1^n + \boldsymbol{\vartheta}) \\ &= (\mathbf{I}_p - \mathcal{Z} \mathbf{H}) \boldsymbol{\kappa}_1^n - \mathcal{Z} \boldsymbol{\vartheta}, \end{aligned}$$

As $n \rightarrow \infty$, $\boldsymbol{\kappa}_1^n \xrightarrow{D} \boldsymbol{\kappa}_1 \sim \mathcal{N}_p\left((\mathbf{B} - \mathbf{I}_p)\boldsymbol{\beta}, \mathbf{B} \mathcal{I}^{-1} \mathbf{B}'\right)$. Thus, $\boldsymbol{\kappa}_3^n \xrightarrow{D} \boldsymbol{\kappa}_3 \sim \mathcal{N}_p\left(E(\boldsymbol{\kappa}_3), Var(\boldsymbol{\kappa}_3)\right)$ where

$$\begin{aligned} E(\boldsymbol{\kappa}_3) &= E[(\mathbf{I}_p - \mathcal{Z} \mathbf{H}) \boldsymbol{\kappa}_1 - \mathcal{Z} \boldsymbol{\vartheta}] \\ &= (\mathbf{I}_p - \mathcal{Z} \mathbf{H}) E[\boldsymbol{\kappa}_1] - \mathcal{Z} \boldsymbol{\vartheta} \\ &= (\mathbf{I}_p - \mathcal{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} - \mathcal{Z} \boldsymbol{\vartheta} \end{aligned}$$

$$\begin{aligned} Var(\boldsymbol{\kappa}_3) &= Var[(\mathbf{I}_p - \mathcal{Z} \mathbf{H}) \boldsymbol{\kappa}_1 - \mathcal{Z} \boldsymbol{\vartheta}] \\ &= Var[(\mathbf{I}_p - \mathcal{Z} \mathbf{H}) \boldsymbol{\kappa}_1] \\ &= (\mathbf{I}_p - \mathcal{Z} \mathbf{H}) \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' (\mathbf{I}_p - \mathcal{Z} \mathbf{H})' \\ &= \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' - \mathcal{Z} \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' \end{aligned}$$

Now we write

$$\begin{aligned} \boldsymbol{\kappa}_4^n &= \sqrt{n}(\hat{\boldsymbol{\beta}}^{UR} - \hat{\boldsymbol{\beta}}^{RL}) \\ &= \sqrt{n}(\hat{\boldsymbol{\beta}}^{UR} - [\hat{\boldsymbol{\beta}}^{UR} - \mathcal{Z}(\mathbf{H}\hat{\boldsymbol{\beta}}^{UR} - \mathbf{h})]) \\ &= \mathcal{Z} \boldsymbol{\kappa}_2^n \\ &= \mathcal{Z} (\mathbf{H} \boldsymbol{\kappa}_1^n + \boldsymbol{\vartheta}) \\ &= \mathcal{Z} \mathbf{H} \boldsymbol{\kappa}_1^n + \mathcal{Z} \boldsymbol{\vartheta} \end{aligned}$$

As $n \rightarrow \infty$, $\boldsymbol{\kappa}_1^n \xrightarrow{D} \boldsymbol{\kappa}_1 \sim \mathcal{N}_p\left((\mathbf{B} - \mathbf{I}_p)\boldsymbol{\beta}, \mathbf{B} \mathcal{I}^{-1} \mathbf{B}'\right)$. Thus, $\boldsymbol{\kappa}_4^n \xrightarrow{D} \boldsymbol{\kappa}_4 \sim \mathcal{N}_p\left(E(\boldsymbol{\kappa}_4), Var(\boldsymbol{\kappa}_4)\right)$ where

$$\begin{aligned}
E(\boldsymbol{\kappa}_4) &= E[\mathbf{Z} \mathbf{H} \boldsymbol{\kappa}_1 + \mathbf{Z} \boldsymbol{\vartheta}] \\
&= \mathbf{Z} \mathbf{H} E[\boldsymbol{\kappa}_1] + \mathbf{Z} \boldsymbol{\vartheta} \\
&= \mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} + \mathbf{Z} \boldsymbol{\vartheta}
\end{aligned}$$

$$\begin{aligned}
Var(\boldsymbol{\kappa}_4) &= Var[\mathbf{Z} \mathbf{H} \boldsymbol{\kappa}_1 + \mathbf{Z} \boldsymbol{\vartheta}] \\
&= (\mathbf{Z} \mathbf{H}) Var[\boldsymbol{\kappa}_1] (\mathbf{Z} \mathbf{H})' \\
&= \mathbf{Z} \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' (\mathbf{Z} \mathbf{H})' \\
&= \mathbf{Z} \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}'.
\end{aligned}$$

The joint distribution of $\boldsymbol{\kappa}_1^n$ and $\boldsymbol{\kappa}_4^n$ is as follows:

$$\begin{aligned}
\begin{pmatrix} \boldsymbol{\kappa}_1^n \\ \boldsymbol{\kappa}_4^n \end{pmatrix} &= \begin{pmatrix} \mathbf{I}_p \boldsymbol{\kappa}_1^n + \mathbf{0} \\ \mathbf{Z} \mathbf{H} \boldsymbol{\kappa}_1^n + \mathbf{Z} \boldsymbol{\vartheta} \end{pmatrix} \\
&= \begin{pmatrix} \mathbf{I}_p \\ \mathbf{Z} \mathbf{H} \end{pmatrix} \boldsymbol{\kappa}_1^n + \begin{pmatrix} \mathbf{0} \\ \mathbf{Z} \boldsymbol{\vartheta} \end{pmatrix},
\end{aligned}$$

$$\text{as } n \rightarrow \infty, \boldsymbol{\kappa}_1^n \xrightarrow{D} \boldsymbol{\kappa}_1 \sim \mathcal{N}_p\left((\mathbf{B} - \mathbf{I}_p)\boldsymbol{\beta}, \mathbf{B} \mathcal{I}^{-1} \mathbf{B}'\right). \text{ So, } \begin{pmatrix} \boldsymbol{\kappa}_1^n \\ \boldsymbol{\kappa}_4^n \end{pmatrix} \xrightarrow{D} \begin{pmatrix} \boldsymbol{\kappa}_1 \\ \boldsymbol{\kappa}_4 \end{pmatrix} \sim \mathcal{N}_{2p}\left(\boldsymbol{\mu}_1, \boldsymbol{\sigma}_1\right)$$

where

$$\begin{aligned}
\boldsymbol{\mu}_1 &= E\left[\begin{pmatrix} \mathbf{I}_p \\ \mathbf{Z} \mathbf{H} \end{pmatrix} \boldsymbol{\kappa}_1 + \begin{pmatrix} \mathbf{0} \\ \mathbf{Z} \boldsymbol{\vartheta} \end{pmatrix}\right] \\
&= \begin{pmatrix} \mathbf{I}_p \\ \mathbf{Z} \mathbf{H} \end{pmatrix} E[\boldsymbol{\kappa}_1] + \begin{pmatrix} \mathbf{0} \\ \mathbf{Z} \boldsymbol{\vartheta} \end{pmatrix} \\
&= \begin{pmatrix} (\mathbf{B} - \mathbf{I}_p)\boldsymbol{\beta} \\ \mathbf{Z} \mathbf{H}(\mathbf{B} - \mathbf{I}_p)\boldsymbol{\beta} + \mathbf{Z} \boldsymbol{\vartheta} \end{pmatrix}
\end{aligned}$$

and

$$\begin{aligned}
\sigma_1 &= Var \left[\begin{pmatrix} \mathbf{I}_p \\ \mathbf{Z} \mathbf{H} \end{pmatrix} \kappa_1 + \begin{pmatrix} \mathbf{0} \\ \mathbf{Z} \vartheta \end{pmatrix} \right] \\
&= Var \left[\begin{pmatrix} \mathbf{I}_p \\ \mathbf{Z} \mathbf{H} \end{pmatrix} \kappa_1 \right] \\
&= \begin{pmatrix} \mathbf{I}_p \\ \mathbf{Z} \mathbf{H} \end{pmatrix} Var[\kappa_1] \begin{pmatrix} \mathbf{I}_p & (\mathbf{Z} \mathbf{H})' \end{pmatrix} \\
&= \begin{pmatrix} \mathbf{I}_p \\ \mathbf{Z} \mathbf{H} \end{pmatrix} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' \begin{pmatrix} \mathbf{I}_p & (\mathbf{Z} \mathbf{H})' \end{pmatrix} \\
&= \begin{pmatrix} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' & \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' (\mathbf{Z} \mathbf{H})' \\ \mathbf{Z} \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' & \mathbf{Z} \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' (\mathbf{Z} \mathbf{H})' \end{pmatrix} \\
&= \begin{pmatrix} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' & \mathbf{Z} \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' \\ \mathbf{Z} \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' & \mathbf{Z} \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' \end{pmatrix}
\end{aligned}$$

In a similar way, we consider

$$\begin{aligned}
\begin{pmatrix} \kappa_3^n \\ \kappa_4^n \end{pmatrix} &= \begin{pmatrix} (\mathbf{I}_p - \mathbf{Z} \mathbf{H}) \kappa_1 - \mathbf{Z} \vartheta \\ \mathbf{Z} \mathbf{H} \kappa_1 + \mathbf{Z} \vartheta \end{pmatrix} \\
&= \begin{pmatrix} \mathbf{I}_p - \mathbf{Z} \mathbf{H} \\ \mathbf{Z} \mathbf{H} \end{pmatrix} \kappa_1 + \begin{pmatrix} -\mathbf{I}_p \\ \mathbf{I}_p \end{pmatrix} \mathbf{Z} \vartheta
\end{aligned}$$

Therefore, as $n \rightarrow \infty$, $\kappa_1^n \xrightarrow{D} \kappa_1 \sim \mathcal{N}_p((\mathbf{B} - \mathbf{I}_p)\beta, \mathbf{B} \mathcal{I}^{-1} \mathbf{B}')$. So, $\begin{pmatrix} \kappa_3^n \\ \kappa_4^n \end{pmatrix} \xrightarrow{D} \begin{pmatrix} \kappa_3 \\ \kappa_4 \end{pmatrix} \sim \mathcal{N}_{2p}(\mu_2, \sigma_2)$ where

$$\begin{aligned}
\mu_2 &= E \left[\begin{pmatrix} \mathbf{I}_p - \mathbf{Z} \mathbf{H} \\ \mathbf{Z} \mathbf{H} \end{pmatrix} \kappa_1 + \begin{pmatrix} -\mathbf{I}_p \\ \mathbf{I}_p \end{pmatrix} \mathbf{Z} \vartheta \right] \\
&= \begin{pmatrix} \mathbf{I}_p - \mathbf{Z} \mathbf{H} \\ \mathbf{Z} \mathbf{H} \end{pmatrix} E[\kappa_1] + \begin{pmatrix} -\mathbf{I}_p \\ \mathbf{I}_p \end{pmatrix} \mathbf{Z} \vartheta \\
&= \begin{pmatrix} (\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \beta - \mathbf{Z} \vartheta \\ \mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \beta + \mathbf{Z} \vartheta \end{pmatrix}
\end{aligned}$$

and

$$\begin{aligned}
\sigma_2 &= Var \left[\begin{pmatrix} \mathbf{I}_p - \mathbf{Z}\mathbf{H} \\ \mathbf{Z}\mathbf{H} \end{pmatrix} \boldsymbol{\kappa}_1 + \begin{pmatrix} -\mathbf{I}_p \\ \mathbf{I}_p \end{pmatrix} \mathbf{Z}\boldsymbol{\vartheta} \right] \\
&= \begin{pmatrix} \mathbf{I}_p - \mathbf{Z}\mathbf{H} \\ \mathbf{Z}\mathbf{H} \end{pmatrix} Var[\boldsymbol{\kappa}_1] \begin{pmatrix} (\mathbf{I}_p - \mathbf{Z}\mathbf{H})' & (\mathbf{Z}\mathbf{H})' \end{pmatrix} \\
&= \begin{pmatrix} \mathbf{I}_p - \mathbf{Z}\mathbf{H} \\ \mathbf{Z}\mathbf{H} \end{pmatrix} \mathbf{B}\mathcal{I}^{-1}\mathbf{B}' \begin{pmatrix} (\mathbf{I}_p - \mathbf{Z}\mathbf{H})' & (\mathbf{Z}\mathbf{H})' \end{pmatrix} \\
&= \begin{pmatrix} (\mathbf{I}_p - \mathbf{Z}\mathbf{H})\mathbf{B}\mathcal{I}^{-1}\mathbf{B}'(\mathbf{I}_p - \mathbf{Z}\mathbf{H})' & (\mathbf{I}_p - \mathbf{Z}\mathbf{H})\mathbf{B}\mathcal{I}^{-1}\mathbf{B}'(\mathbf{Z}\mathbf{H})' \\ \mathbf{Z}\mathbf{H}\mathbf{B}\mathcal{I}^{-1}\mathbf{B}'(\mathbf{I}_p - \mathbf{Z}\mathbf{H})' & \mathbf{Z}\mathbf{H}\mathbf{B}\mathcal{I}^{-1}\mathbf{B}'(\mathbf{Z}\mathbf{H})' \end{pmatrix} \\
&= \begin{pmatrix} \mathbf{B}\mathcal{I}^{-1}\mathbf{B}' - \mathbf{Z}\mathbf{H}\mathbf{B}\mathcal{I}^{-1}\mathbf{B}' & \mathbf{0} \\ \mathbf{0} & \mathbf{Z}\mathbf{H}\mathbf{B}\mathcal{I}^{-1}\mathbf{B}' \end{pmatrix}.
\end{aligned}$$

Appendix 2. Proof of Theorem 3.3

Here, we provide the proof of bias expressions. Based on Lemma 3.1 we have

$$\mathcal{B}(\hat{\beta}^{UR}) = \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\hat{\beta}^{UR} - \beta \right) \right] = E[\kappa_1] = (\mathbf{B} - \mathbf{I}_p) \beta$$

$$\begin{aligned} \mathcal{B}(\hat{\beta}^{RL}) &= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\hat{\beta}^{RL} - \beta \right) \right] \\ &= E[\kappa_3] \\ &= (\mathbf{I}_p - \mathbf{Z} \mathbf{H})(\mathbf{B} - \mathbf{I}_p) \beta - \mathbf{Z} \vartheta \end{aligned}$$

$$\begin{aligned} \mathcal{B}(\hat{\beta}^{LLS}) &= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\hat{\beta}^{LLS} - \beta \right) \right] \\ &= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\delta \hat{\beta}^{RL} + (1 - \delta) \hat{\beta}^{UR} - \beta \right) \right] \\ &= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\hat{\beta}^{UR} - \beta \right) - \delta \sqrt{n} \left(\hat{\beta}^{UR} - \hat{\beta}^{RL} \right) \right] \\ &= E(\kappa_1) - \delta E(\kappa_4) \\ &= \mathcal{B}(\hat{\beta}^{UR}) - \delta \mathbf{Z} [\mathbf{H}(\mathbf{B} - \mathbf{I}_p) \beta + \vartheta] \end{aligned}$$

$$\begin{aligned} \mathcal{B}(\hat{\beta}^{LSPE}) &= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\hat{\beta}^{LSPE} - \beta \right) \right] \\ &= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\hat{\beta}^{UR} - \delta (\hat{\beta}^{UR} - \hat{\beta}^{RL}) I(T_n \leq T_{n,\alpha}) - \beta \right) \right] \\ &= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\hat{\beta}^{UR} - \beta \right) - \delta \sqrt{n} \left(\hat{\beta}^{UR} - \hat{\beta}^{RL} \right) I(T_n \leq T_{n,\alpha}) \right] \\ &= E[\kappa_1] - \delta E \left[\kappa_4 I(T_n \leq T_{n,\alpha}) \right], \end{aligned}$$

based on Equation (26) we can write

$$\begin{aligned} \mathcal{B}(\hat{\beta}^{LSPE}) &= E[\kappa_1] - \delta E[\kappa_4] P(\chi_{p_2+2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2) \\ &= \mathcal{B}(\hat{\beta}^{UR}) - \delta \mathbf{Z} [\mathbf{H}(\mathbf{B} - \mathbf{I}_p) \beta + \vartheta] P(\chi_{p_2+2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2) \\ &= \mathcal{B}(\hat{\beta}^{UR}) - \delta \mathbf{Z} [\mathbf{H}(\mathbf{B} - \mathbf{I}_p) \beta + \vartheta] \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*). \end{aligned}$$

If $\lambda = 1$,

$$\mathcal{B}(\hat{\beta}^{LPT}) = \mathcal{B}(\hat{\beta}^{UR}) - \mathbf{Z} [\mathbf{H}(\mathbf{B} - \mathbf{I}_p) \beta + \vartheta] \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*).$$

In a similar way, we can obtain

$$\begin{aligned}
\mathcal{B}(\hat{\beta}^{LS}) &= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\hat{\beta}^{LS} - \beta \right) \right] \\
&= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\hat{\beta}^{RL} + (1 - (p_2 - 2) T_n^{-1}) (\hat{\beta}^{UR} - \hat{\beta}^{RL}) - \beta \right) \right] \\
&= \lim_{n \rightarrow \infty} \left\{ E \left[\sqrt{n} \left(\hat{\beta}^{UR} - \beta \right) \right] - (p_2 - 2) E \left[T_n^{-1} \sqrt{n} \left(\hat{\beta}^{UR} - \hat{\beta}^{RL} \right) \right] \right\} \\
&= E[\kappa_1] - (p_2 - 2) E[T_n^{-1} \kappa_4] \\
&= \mathcal{B}(\hat{\beta}^{UR}) - (p_2 - 2) E[\kappa_4] E \left[\frac{1}{\chi_{p_2+2}^2(\Delta^*)} \right] \\
&= \mathcal{B}(\hat{\beta}^{UR}) - (p_2 - 2) \mathbf{Z} [\mathbf{H}(\mathbf{B} - \mathbf{I}_p) \beta + \vartheta] E \left[\frac{1}{\chi_{p_2+2}^2(\Delta^*)} \right].
\end{aligned}$$

Also,

$$\begin{aligned}
\mathcal{B}(\hat{\beta}^{LPS}) &= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\hat{\beta}^{LPS} - \beta \right) \right] \\
&= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\mathcal{B}(\hat{\beta}^{LS}) - \beta \right) - (1 - (p_2 - 2) T_n^{-1}) \right. \\
&\quad \left. \times \sqrt{n} \left((\hat{\beta}^{UR} - \hat{\beta}^{RL}) I(T_n < p_2 - 2) \right) \right] \\
&= \mathcal{B}(\hat{\beta}^{LS}) - \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\hat{\beta}^{UR} - \hat{\beta}^{RL} \right) (1 - (p_2 - 2) T_n^{-1}) I(T_n < p_2 - 2) \right] \\
&= \mathcal{B}(\hat{\beta}^{LS}) - E[\kappa_4 (1 - (p_2 - 2) T_n^{-1}) I(T_n < p_2 - 2)] \\
&= \mathcal{B}(\hat{\beta}^{LS}) - E[\kappa_4 I(T_n < p_2 - 2)] + (p_2 - 2) E[\kappa_4 T_n^{-1} I(T_n < p_2 - 2)] \\
&= \mathcal{B}(\hat{\beta}^{LS}) - E[\kappa_4] \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*) + (p_2 - 2) E[\kappa_4] E \left[\frac{I(T_n < p_2 - 2)}{\chi_{p_2+2}^2(\Delta^*)} \right] \\
&= \mathcal{B}(\hat{\beta}^{LS}) - E[\kappa_4] \left\{ \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*) + (p_2 - 2) E \left[\frac{I(T_n < p_2 - 2)}{\chi_{p_2+2}^2(\Delta^*)} \right] \right\} \\
&= \mathcal{B}(\hat{\beta}^{LS}) - \mathbf{Z} [\mathbf{H}(\mathbf{B} - \mathbf{I}_p) \beta + \vartheta] \left\{ \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*) \right. \\
&\quad \left. + (p_2 - 2) E \left[\frac{I(\chi_{p_2+2}^2(\Delta^*) < p_2 - 2)}{\chi_{p_2+2}^2(\Delta^*)} \right] \right\}.
\end{aligned}$$

Appendix 3. Proof of Theorem 3.3

Here, we compute the asymptotic covariance matrices of the estimators

$$\begin{aligned}
\mathcal{V}(\hat{\beta}^{UR}) &= \lim_{n \rightarrow \infty} E\left(\sqrt{n}(\hat{\beta}^{UR} - \beta) \sqrt{n}(\hat{\beta}^{UR} - \beta)'\right) \\
&= \lim_{n \rightarrow \infty} E(\kappa_1^n \kappa_1^{n'}) \\
&= E(\kappa_1 \kappa_1') \\
&= \text{Var}(\kappa_1) + E(\kappa_1 E(\kappa_1')) \\
&= \mathbf{B}\mathcal{I}^{-1}\mathbf{B}' + \left[(\mathbf{B} - \mathbf{I}_p)\beta\right] \left[(\mathbf{B} - \mathbf{I}_p)\beta\right]'
\end{aligned}$$

$$\begin{aligned}
\mathcal{V}(\hat{\beta}^{RL}) &= \lim_{n \rightarrow \infty} E\left(\sqrt{n}(\hat{\beta}^{RL} - \beta) \sqrt{n}(\hat{\beta}^{RL} - \beta)'\right) \\
&= \lim_{n \rightarrow \infty} E(\kappa_3^n \kappa_3^{n'}) \\
&= E(\kappa_3 \kappa_3') \\
&= \text{Var}(\kappa_3) + E(\kappa_3) E(\kappa_3') \\
&= \mathbf{B}\mathcal{I}^{-1}\mathbf{B}' - \mathcal{Z}\mathbf{H}\mathbf{B}\mathcal{I}^{-1}\mathbf{B}' \\
&\quad + \left[(\mathbf{I}_p - \mathcal{Z}\mathbf{H})(\mathbf{B} - \mathbf{I}_p)\beta - \mathcal{Z}\vartheta\right] \left[(\mathbf{I}_p - \mathcal{Z}\mathbf{H})(\mathbf{B} - \mathbf{I}_p)\beta - \mathcal{Z}\vartheta\right]'
\end{aligned}$$

$$\begin{aligned}
\mathcal{V}(\hat{\beta}^{LLS}) &= \lim_{n \rightarrow \infty} E\left(\sqrt{n}(\hat{\beta}^{LLS} - \beta) \sqrt{n}(\hat{\beta}^{LLS} - \delta)'\right) \\
&= \lim_{n \rightarrow \infty} E[(\kappa_1^n - \delta \kappa_4^n) (\kappa_1^n - \delta \kappa_4^n)'] \\
&= E[(\kappa_1 - \delta \kappa_4) (\kappa_1 - \delta \kappa_4)'] \\
&= E[\kappa_1 \kappa_1'] - 2\delta E[\kappa_1 \kappa_4'] + \delta^2 E[\kappa_4 \kappa_4'] \\
&= \mathcal{V}(\hat{\beta}^{UR}) - 2\delta \underbrace{E[\kappa_1 \kappa_4']}_{e_1} + \delta^2 E[\kappa_4 \kappa_4']
\end{aligned}$$

Using the conditional expectation of a multivariate normal distribution, e_1 becomes

$$\begin{aligned}
e_1 &= E[\kappa_1 \kappa_4'] \\
&= E(\kappa_4' E[\kappa_1 | \kappa_4]) \\
&= E(\kappa_4' \{(\mathbf{B} - \mathbf{I}_p)\beta + \kappa_4 - \mathcal{Z}\mathbf{H}(\mathbf{B} - \mathbf{I}_p)\beta - \mathcal{Z}\vartheta\}) \\
&= (\mathbf{B} - \mathbf{I}_p)\beta E(\kappa_4') + E(\kappa_4 \kappa_4') - [\mathcal{Z}\mathbf{H}(\mathbf{B} - \mathbf{I}_p)\beta + \mathcal{Z}\vartheta] E(\kappa_4') \\
&= E(\kappa_4 \kappa_4') + E(\kappa_4') \{(\mathbf{I}_p - \mathcal{Z}\mathbf{H})(\mathbf{B} - \mathbf{I}_p)\beta - \mathcal{Z}\vartheta\}
\end{aligned}$$

Therefore,

$$\begin{aligned}
\mathcal{V}(\widehat{\beta}^{LLS}) &= \mathcal{V}(\widehat{\beta}^{UR}) + \delta^2 E[\kappa_4 \kappa_4'] \\
&\quad - 2\delta \left(E(\kappa_4 \kappa_4') + E(\kappa_4') \{ (\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \beta - \mathbf{Z} \vartheta \} \right) \\
&= \mathcal{V}(\widehat{\beta}^{UR}) - 2\delta \{ (\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \beta - \mathbf{Z} \vartheta \} E(\kappa_4') - \delta (2 - \delta) E[\kappa_4 \kappa_4'] \\
&= \mathcal{V}(\widehat{\beta}^{UR}) - 2\delta \{ (\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \beta - \mathbf{Z} \vartheta \} \{ \mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \beta + \mathbf{Z} \vartheta \}' \\
&\quad - \delta (2 - \delta) \{ \mathbf{Z} \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' + [\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \beta + \mathbf{Z} \vartheta] [\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \beta + \mathbf{Z} \vartheta]' \}
\end{aligned}$$

Next we obtain $\mathcal{V}(\widehat{\beta}^{LSPE})$ as follows

$$\begin{aligned}
\mathcal{V}(\widehat{\beta}^{LSPE}) &= \lim_{n \rightarrow \infty} E \left(\sqrt{n} (\widehat{\beta}^{LSPE} - \beta) \sqrt{n} (\widehat{\beta}^{LSPE} - \beta)' \right) \\
&= \lim_{n \rightarrow \infty} E \left[\{ \kappa_1^n - \delta \kappa_4^n I(T_n \leq T_{n,\alpha}) \} \{ \kappa_1^n - \delta \kappa_4^n I(T_n \leq T_{n,\alpha}) \}' \right] \\
&= \lim_{n \rightarrow \infty} E(\kappa_1^n \kappa_1^{n'}) - 2\delta \lim_{n \rightarrow \infty} E(\kappa_1^n \kappa_4^{n'} I(T_n \leq T_{n,\alpha})) \\
&\quad + \delta^2 \lim_{n \rightarrow \infty} E(\kappa_4^n \kappa_4^{n'} I(T_n \leq T_{n,\alpha})) \\
&= E(\kappa_1 \kappa_1') - 2\delta E(\kappa_1 \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)) + \delta^2 E(\kappa_4 \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)) \\
&= \mathcal{V}(\widehat{\beta}^{UR}) - 2\delta \underbrace{E(\kappa_1 \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2))}_{e_2} + \delta^2 E(\kappa_4 \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)),
\end{aligned}$$

by using conditional expectation, e_2 becomes

$$\begin{aligned}
e_2 &= E(\kappa_1 \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)) \\
&= E[E(\kappa_1 \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2) | \kappa_4)] \\
&= E[E(\kappa_1 | \kappa_4) \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)] \\
&= E[\{ (\mathbf{B} - \mathbf{I}_p) \beta + \kappa_4 - \mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \beta - \mathbf{Z} \vartheta \} \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)] \\
&= (\mathbf{B} - \mathbf{I}_p) \beta E[\kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)] + E[\kappa_4 \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)] \\
&\quad - (\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \beta + \mathbf{Z} \vartheta) E[\kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)] \\
&= \left\{ (\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \beta - \mathbf{Z} \vartheta \right\} E[\kappa_4] \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*) + E[\kappa_4 \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)]
\end{aligned}$$

Therefore,

$$\begin{aligned}
\mathcal{V}(\widehat{\beta}^{LSPE}) &= \mathcal{V}(\widehat{\beta}^{UR}) \\
&\quad - 2\delta \left(\{(\mathbf{I}_p - \mathbf{Z}\mathbf{H})(\mathbf{B} - \mathbf{I}_p)\beta - \mathbf{Z}\vartheta\} E[\kappa_4] \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*) \right) \\
&\quad + E[\kappa_4 \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)] + \delta^2 E(\kappa_4 \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)) \\
&= \mathcal{V}(\widehat{\beta}^{UR}) - 2\delta \left(\{(\mathbf{I}_p - \mathbf{Z}\mathbf{H})(\mathbf{B} - \mathbf{I}_p)\beta - \mathbf{Z}\vartheta\} E[\kappa_4] \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*) \right) \\
&\quad - \delta(2 - \delta) E[\kappa_4 \kappa_4' I(\chi_{p_2}^2(\Delta^*) \leq \chi_{p_2,\alpha}^2)] \\
&= \mathcal{V}(\widehat{\beta}^{UR}) - 2\delta \left(\{(\mathbf{I}_p - \mathbf{Z}\mathbf{H})(\mathbf{B} - \mathbf{I}_p)\beta - \mathbf{Z}\vartheta\} \right. \\
&\quad \times [\mathbf{Z}\mathbf{H}(\mathbf{B} - \mathbf{I}_p)\beta + \mathbf{Z}\vartheta]' \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*) \Big) \\
&\quad - \delta(2 - \delta) \left(\mathbf{Z}\mathbf{H}\mathbf{B}\mathcal{I}^{-1}\mathbf{B}' \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*) + [\mathbf{Z}\mathbf{H}(\mathbf{B} - \mathbf{I}_p)\beta + \mathbf{Z}\vartheta] \right. \\
&\quad \times [\mathbf{Z}\mathbf{H}(\mathbf{B} - \mathbf{I}_p)\beta + \mathbf{Z}\vartheta]' \Psi_{p_2+4}(\chi_{p_2,\alpha}^2; \Delta^*) \Big).
\end{aligned}$$

For $\delta = 1$, $\mathcal{V}(\widehat{\beta}^{LPT})$ reduces to

$$\begin{aligned}
\mathcal{V}(\widehat{\beta}^{LPT}) &= \mathcal{V}(\widehat{\beta}^{UR}) - 2 \left(\left\{ (\mathbf{I}_p - \mathbf{Z}\mathbf{H})(\mathbf{B} - \mathbf{I}_p)\beta - \mathbf{Z}\vartheta \right\} \right. \\
&\quad \times [\mathbf{Z}\mathbf{H}(\mathbf{B} - \mathbf{I}_p)\beta + \mathbf{Z}\vartheta]' \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*) \Big) \\
&\quad - \left(\mathbf{Z}\mathbf{H}\mathbf{B}\mathcal{I}^{-1}\mathbf{B}' \Psi_{p_2+2}(\chi_{p_2,\alpha}^2; \Delta^*) + [\mathbf{Z}\mathbf{H}(\mathbf{B} - \mathbf{I}_p)\beta + \mathbf{Z}\vartheta] \right. \\
&\quad \times [\mathbf{Z}\mathbf{H}(\mathbf{B} - \mathbf{I}_p)\beta + \mathbf{Z}\vartheta]' \Psi_{p_2+4}(\chi_{p_2,\alpha}^2; \Delta^*) \Big).
\end{aligned}$$

Now we obtain $\mathcal{V}(\widehat{\beta}^{LS})$ as follows:

$$\begin{aligned}
\mathcal{V}(\widehat{\beta}^{LS}) &= \lim_{n \rightarrow \infty} E \left(\sqrt{n}(\widehat{\beta}^{LS} - \beta) \sqrt{n}(\widehat{\beta}^{LS} - \beta)' \right) \\
&= \lim_{n \rightarrow \infty} E \left[\sqrt{n} \left(\widehat{\beta}^{RL} + \{1 - (p_2 - 2)T_n^{-1}\} (\widehat{\beta}^{UR} - \widehat{\beta}^{RL}) - \beta \right) \right. \\
&= \lim_{n \rightarrow \infty} E[(\kappa_1^n - (p_2 - 2)T_n^{-1}\kappa_4^n)(\kappa_1^n - (p_2 - 2)T_n^{-1}\kappa_4^n)'] \\
&= E[(\kappa_1 - (p_2 - 2)T_n^{-1}\kappa_4)(\kappa_1 - (p_2 - 2)T_n^{-1}\kappa_4')] \\
&= E[(\kappa_1 \kappa_1' - 2(p_2 - 2) \underbrace{E[\kappa_4 \kappa_1' T_n^{-1}]}_{e_3} + (p_2 - 2)^2 \underbrace{E[\kappa_4 \kappa_4' \mathbf{T}_n^{-2}]}_{e_4})],
\end{aligned}$$

similar to e_1 , we can write e_3 as follows:

$$\begin{aligned}
e_3 &= E[\boldsymbol{\kappa}_4 \boldsymbol{\kappa}'_1 T_n^{-1}] \\
&= E[\boldsymbol{\kappa}_4 \boldsymbol{\kappa}'_4 T_n^{-1}] + [(\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} - \mathbf{Z} \boldsymbol{\vartheta}] E[\boldsymbol{\kappa}_4 T_n^{-1}] \\
&= Var[\boldsymbol{\kappa}_4] E\left[\frac{1}{\chi_{p_2+2}^2(\Delta^*)}\right] + E[\boldsymbol{\kappa}_4] E[\boldsymbol{\kappa}'_1] E\left[\frac{1}{\chi_{p_2+4}^2(\Delta^*)}\right] \\
&\quad + [(\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} - \mathbf{Z} \boldsymbol{\vartheta}] E[\boldsymbol{\kappa}_4 T_n^{-1}],
\end{aligned}$$

and by using Equation (27), e_4 becomes

$$\begin{aligned}
e_4 &= E[\boldsymbol{\kappa}_4 \boldsymbol{\kappa}'_4 T_n^{-2}] \\
&= Var[\boldsymbol{\kappa}_4] E\left[\frac{1}{(\chi_{p_2+2}^2(\Delta^*))^2}\right] + E[\boldsymbol{\kappa}_4] E[\boldsymbol{\kappa}'_1] E\left[\frac{1}{(\chi_{p_2+4}^2(\Delta^*))^2}\right],
\end{aligned}$$

Therefore,

$$\begin{aligned}
\mathcal{V}(\hat{\boldsymbol{\beta}}^{LS}) &= \mathcal{V}(\hat{\boldsymbol{\beta}}^{UR}) \\
&\quad - 2(p_2 - 2) \left([(\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} - \mathbf{Z} \boldsymbol{\vartheta}] [\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} + \mathbf{Z} \boldsymbol{\vartheta}] E\left[\frac{1}{\chi_{p_2+2}^2(\Delta^*)}\right] \right) \\
&\quad + (p_2 - 2)(p_2 - 4) \mathbf{Z} \mathbf{H} \mathbf{B} \mathbf{I}^{-1} \mathbf{B}' \left(E\left[\frac{1}{(\chi_{p_2+2}^2(\Delta^*))^2}\right] - E\left[\frac{1}{\chi_{p_2+2}^2(\Delta^*)}\right] \right) \\
&\quad + (p_2 - 2)(p_2 - 4) [\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} + \mathbf{Z} \boldsymbol{\vartheta}] [\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} + \mathbf{Z} \boldsymbol{\vartheta}]' \\
&\quad \times \left(E\left[\frac{1}{(\chi_{p_2+4}^2(\Delta^*))^2}\right] - E\left[\frac{1}{\chi_{p_2+4}^2(\Delta^*)}\right] \right).
\end{aligned}$$

Finally, we can write $\mathcal{V}(\hat{\boldsymbol{\beta}}^{LPS})$ as follows:

$$\begin{aligned}
\mathcal{V}(\hat{\boldsymbol{\beta}}^{LPS}) &= \lim_{n \rightarrow \infty} E\left(\sqrt{n}(\hat{\boldsymbol{\beta}}^{LPS} - \boldsymbol{\beta}) \sqrt{n}(\hat{\boldsymbol{\beta}}^{LPS} - \boldsymbol{\beta})'\right) \\
&= \lim_{n \rightarrow \infty} E\left[\sqrt{n} \left(\hat{\boldsymbol{\beta}}^{LS} - (1 - (p_2 - 2) T_n^{-1}) I(T_n < p_2 - 2) (\hat{\boldsymbol{\beta}}^{UR} - \hat{\boldsymbol{\beta}}^{RL}) - \boldsymbol{\beta}\right) \right. \\
&\quad \times \left. \sqrt{n} \left(\hat{\boldsymbol{\beta}}^{LS} - (1 - (p_2 - 2) T_n^{-1}) I(T_n < p_2 - 2) (\hat{\boldsymbol{\beta}}^{UR} - \hat{\boldsymbol{\beta}}^{RL}) - \boldsymbol{\beta}\right)'\right] \\
&= \mathcal{V}(\hat{\boldsymbol{\beta}}^{LS}) - 2E[\boldsymbol{\kappa}_4 \boldsymbol{\kappa}'_3 (1 - (p_2 - 2) T_n^{-1}) I(T_n < p_2 - 2)] \\
&\quad - 2E[\boldsymbol{\kappa}_4 \boldsymbol{\kappa}'_4 (1 - (p_2 - 2) T_n^{-1})^2 I(T_n < p_2 - 2)] \\
&\quad + E[\boldsymbol{\kappa}_4 \boldsymbol{\kappa}'_4 (1 - (p_2 - 2) T_n^{-1})^2 I(T_n < p_2 - 2)] \\
&= \mathcal{V}(\hat{\boldsymbol{\beta}}^{LS}) - 2 \underbrace{E[\boldsymbol{\kappa}_4 \boldsymbol{\kappa}'_3 (1 - (p_2 - 2) T_n^{-1}) I(T_n < p_2 - 2)]}_{e_5} \\
&\quad - \underbrace{E[\boldsymbol{\kappa}_4 \boldsymbol{\kappa}'_4 (1 - (p_2 - 2) T_n^{-1})^2 I(T_n < p_2 - 2)]}_{e_6},
\end{aligned}$$

now we obtain e_5

$$\begin{aligned}
e_5 &= E[\kappa_4 \kappa_3' (1 - (p_2 - 2) T_n^{-1}) I(T_n < p_2 - 2)] \\
&= E[\kappa_4 E\{\kappa_3' (1 - (p_2 - 2) T_n^{-1}) I(T_n < p_2 - 2) \mid \kappa_4\}] \\
&= E[\kappa_4 \{(\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} - \mathbf{Z} \boldsymbol{\vartheta}\}' \\
&\quad \times (1 - (p_2 - 2) T_n^{-1}) I(T_n < p_2 - 2)] \\
&= \{(\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} - \mathbf{Z} \boldsymbol{\vartheta}\} E[\kappa_4 (1 - (p_2 - 2) T_n^{-1}) I(T_n < p_2 - 2)] \\
&= \{(\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} - \mathbf{Z} \boldsymbol{\vartheta}\} E[\kappa_4] E\left[\left(1 - \frac{p_2 - 2}{\chi_{p_2+2}^2(\Delta^*)}\right) I(\chi_{p_2+2}^2(\Delta^*) < p_2 - 2)\right] \\
&= \{(\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} - \mathbf{Z} \boldsymbol{\vartheta}\} [\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} - \mathbf{Z} \boldsymbol{\vartheta} + \mathbf{Z} \boldsymbol{\vartheta}] \\
&\quad \times E\left[\left(1 - \frac{p_2 - 2}{\chi_{p_2+2}^2(\Delta^*)}\right) I(\chi_{p_2+2}^2(\Delta^*) < p_2 - 2)\right]
\end{aligned}$$

and based on Equation (27), e_6 becomes

$$\begin{aligned}
e_6 &= E[\kappa_4 \kappa_4' (1 - (p_2 - 2) T_n^{-1})^2 I(T_n < p_2 - 2)] \\
&= Var(\kappa_4) E\left[\left(1 - \frac{p_2 - 2}{\chi_{p_2+2}^2(\Delta^*)}\right)^2 I(\chi_{p_2+2}^2(\Delta^*) < p_2 - 2)\right] \\
&\quad + E(\kappa_4) E(\kappa_4') E\left[\left(1 - \frac{p_2 - 2}{\chi_{p_2+4}^2(\Delta^*)}\right)^2 I(\chi_{p_2+4}^2(\Delta^*) < p_2 - 2)\right] \\
&= \mathbf{Z} \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' E\left[\left(1 - \frac{p_2 - 2}{\chi_{p_2+2}^2(\Delta^*)}\right)^2 I(\chi_{p_2+2}^2(\Delta^*) < p_2 - 2)\right] \\
&\quad + [\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} + \mathbf{Z} \boldsymbol{\vartheta}] [\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} + \mathbf{Z} \boldsymbol{\vartheta}]' \\
&\quad \times E\left[\left(1 - \frac{p_2 - 2}{\chi_{p_2+4}^2(\Delta^*)}\right)^2 I(\chi_{p_2+4}^2(\Delta^*) < p_2 - 2)\right]
\end{aligned}$$

Therefore, $\mathcal{V}(\hat{\boldsymbol{\beta}}^{LPS})$ becomes

$$\begin{aligned}
\mathcal{V}(\hat{\boldsymbol{\beta}}^{LPS}) &= \mathcal{V}(\hat{\boldsymbol{\beta}}^{LS}) \\
&\quad - 2 \left(\{(\mathbf{I}_p - \mathbf{Z} \mathbf{H}) (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} - \mathbf{Z} \boldsymbol{\vartheta}\} [\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} - \mathbf{Z} \boldsymbol{\vartheta} + \mathbf{Z} \boldsymbol{\vartheta}] \right. \\
&\quad \times E\left[\left(1 - \frac{p_2 - 2}{\chi_{p_2+2}^2(\Delta^*)}\right) I(\chi_{p_2+2}^2(\Delta^*) < p_2 - 2)\right] \Big) \\
&\quad - \left(\mathbf{Z} \mathbf{H} \mathbf{B} \mathcal{I}^{-1} \mathbf{B}' E\left[\left(1 - \frac{p_2 - 2}{\chi_{p_2+2}^2(\Delta^*)}\right)^2 I(\chi_{p_2+2}^2(\Delta^*) < p_2 - 2)\right] \right. \\
&\quad + [\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} + \mathbf{Z} \boldsymbol{\vartheta}] [\mathbf{Z} \mathbf{H} (\mathbf{B} - \mathbf{I}_p) \boldsymbol{\beta} + \mathbf{Z} \boldsymbol{\vartheta}]' \\
&\quad \times E\left[\left(1 - \frac{p_2 - 2}{\chi_{p_2+4}^2(\Delta^*)}\right)^2 I(\chi_{p_2+4}^2(\Delta^*) < p_2 - 2)\right] \Big).
\end{aligned}$$

Appendix 4. Bootstrap Standard Errors

Table 6: Coefficients and bootstrapped standard errors of the proposed estimators for econometric data

number	Variables	Coefficients						
		UR	LR	LLS	LPT	LSPE	LS	LPS
1	houseval	0.1035	0.1407	0.1221	0.1092	0.1064	0.1105	0.1105
2	education	-3.5113	-3.5892	-3.5503	-3.5237	-3.5175	-3.5259	-3.5259
3	recreation	-2.9546	-2.6228	-2.7887	-2.9087	-2.9317	-2.8939	-2.894
4	social	-3.3167	-3.4323	-3.3745	-3.3343	-3.3255	-3.3382	-3.3381
5	urbanplanning	-3.6089	-3.6965	-3.6527	-3.6219	-3.6154	-3.6251	-3.6251
6	popdens	-0.0119	-0.0117	-0.0118	-0.0124	-0.0121	-0.0119	-0.0119
7	noleft	-0.0181	-0.021	-0.0195	-0.0200	-0.0191	-0.0191	-0.0191
8	minorityleft	-0.0895	-0.037	-0.0633	-0.0806	-0.0851	-0.0794	-0.0794
9	safety	0.3466	-0.024	0.1613	0.2971	0.3219	0.2792	0.2793
10	tot	-0.0029	-7e-04	-0.0018	-0.0027	-0.0028	-0.0026	-0.0026
Standard Errors								
1	houseval	0.0573	0.0486	0.0515	0.0594	0.0577	0.0558	0.0557
2	education	0.2625	0.2606	0.2589	0.2662	0.2635	0.2618	0.2616
3	recreation	0.5526	0.4194	0.4585	0.5567	0.5463	0.5165	0.5163
4	social	0.1856	0.1475	0.1588	0.1944	0.1877	0.1787	0.1784
5	urbanplanning	0.159	0.1431	0.1465	0.1671	0.1616	0.1569	0.1566
6	popdens	0.0256	0.0055	0.0141	0.022	0.023	0.0203	0.0203
7	noleft	0.0626	0.006	0.0311	0.0555	0.0573	0.0501	0.05
8	minorityleft	0.0589	0.0073	0.0297	0.0583	0.0562	0.0487	0.0485
9	safety	0.445	0.0055	0.223	0.4492	0.433	0.374	0.3737
10	tot	0.0049	0.0013	0.0031	0.0049	0.0048	0.0044	0.0044

Table 7: Coefficients and bootstrapped standard errors of the proposed estimators for educational data

		Coefficients						
	Variables	UR	RR	RLS	RPT	SPE	RS	RPS
1	Father Education	-0.0319	-0.0361	-0.034	-0.0337	-0.0328	-0.0335	-0.0335
2	Age	-0.1026	-0.1087	-0.1056	-0.1058	-0.1042	-0.1051	-0.105
3	Studytime	-0.0291	-0.0252	-0.0271	-0.0269	-0.028	-0.0274	-0.0274
4	Failures	0.0324	0.0267	0.0295	0.0293	0.0308	0.0299	0.03
5	Health	-0.0378	-0.0378	-0.0378	-0.0377	-0.0378	-0.0378	-0.0378
6	First period grade	0.0482	0.0492	0.0487	0.0486	0.0484	0.0485	0.0485
7	second period grade	0.1658	0.166	0.1659	0.166	0.1659	0.166	0.1659
8	Mother education	-0.0044	-4e-04	-0.0024	-0.0028	-0.0036	-0.0029	-0.0029
9	Travel time	-0.0221	-3e-04	-0.0112	-0.0107	-0.0164	-0.0128	-0.0131
10	Absences	-0.0011	-1e-04	-6e-04	-5e-04	-8e-04	-7e-04	-7e-04
11	Weekend Alcohol consumption	-0.0085	-1e-04	-0.0043	-0.0046	-0.0066	-0.0052	-0.0053
12	quality of family relationships	0.0092	-6e-04	0.0043	0.004	0.0066	0.0048	0.0049
13	going out	-0.0202	-3e-04	-0.0102	-0.0096	-0.0149	-0.0117	-0.0119
14	Workday alcohol consumption	4e-04	1e-04	2e-04	5e-04	4e-04	2e-04	2e-04
		Standard Errors						
1	Father Education	0.0199	0.0155	0.0167	0.0178	0.0183	0.0174	0.0173
2	Age	0.0084	0.0055	0.0064	0.0078	0.0077	0.0071	0.0070
3	Studytime	0.0222	0.0211	0.0213	0.0219	0.0218	0.0216	0.0216
4	Failures	0.0307	0.0302	0.0301	0.0308	0.0306	0.0304	0.0303
5	Health	0.0136	0.0132	0.0133	0.0134	0.0134	0.0134	0.0134
6	First period grade	0.0115	0.0111	0.0112	0.0113	0.0113	0.0113	0.0112
7	second period grade	0.0123	0.0116	0.0118	0.0119	0.0120	0.0119	0.0119
8	Mother education	0.0200	0.0004	0.0100	0.0133	0.0152	0.0123	0.0120
9	Travel time	0.0271	0.0005	0.0135	0.0224	0.0224	0.0183	0.0177
10	Absences	0.0021	0.0000	0.0010	0.0015	0.0016	0.0013	0.0012
11	Weekend Alcohol consumption	0.0183	0.0003	0.0091	0.0130	0.0142	0.0116	0.0112
12	quality of family relationships	0.0190	0.0003	0.0095	0.0136	0.0148	0.0122	0.0116
13	going out	0.0176	0.0003	0.0088	0.0166	0.0153	0.0129	0.0123
14	Workday alcohol consumption	0.0245	0.0004	0.0123	0.0163	0.0187	0.0151	0.0147

Appendix 5. Tables of Simulation Results

Table 8: RMSE values for $\rho = 0.6$

		$n = 50$						$n = 100$						$n = 200$							
Δ		RL	LLS	LPT	LSPE	LS	LPS	Δ	RL	LLS	LPT	LSPE	LS	LPS	Δ	RL	LLS	LPT	LSPE	LS	LPS
$p_2 = 10$	0.00	2.5900	1.9060	1.5150	1.3540	1.7340	1.7950	0.00	1.5590	1.3860	1.4110	1.2920	1.3850	1.4380	0.00	1.4790	1.3320	1.3640	1.2590	1.3370	1.3790
	0.20	1.5660	1.5420	1.1490	1.1500	1.3690	1.3930	0.20	1.2490	1.2910	1.0610	1.0730	1.2200	1.2230	0.20	0.9800	1.1410	0.9730	1.0210	1.1010	1.1040
	0.40	1.2270	1.3850	1.0260	1.0620	1.2310	1.2460	0.40	1.0430	1.1990	0.9900	1.0110	1.1290	1.1300	0.40	0.7700	1.0260	0.9720	0.9940	1.0410	1.0410
	0.60	1.0320	1.2730	0.9860	1.0240	1.1650	1.1690	0.60	0.9200	1.1210	0.9900	1.0010	1.0860	1.0860	0.60	0.6520	0.9430	0.9940	0.9980	1.0140	1.0140
	0.80	0.8960	1.1750	0.9650	1.0050	1.1060	1.1080	0.80	0.8410	1.0650	0.9920	0.9990	1.0570	1.0570	0.80	0.5870	0.8910	0.9980	0.9990	1.0030	1.0030
	1.00	0.8050	1.1100	0.9570	0.9960	1.0720	1.0730	1.00	0.7800	1.0210	0.9970	0.9990	1.0400	1.0400	1.00	0.5360	0.8460	1.0000	1.0000	0.9930	0.9930
	1.20	0.7290	1.0490	0.9690	0.9960	1.0520	1.0520	1.20	0.7320	0.9780	0.9980	0.9990	1.0220	1.0220	1.20	0.5050	0.8110	1.0000	1.0000	0.9860	0.9860
	2.00	0.6040	0.9150	0.9900	0.9960	1.0020	1.0020	2.00	0.6560	0.9040	1.0000	1.0000	0.9970	0.9970	2.00	0.4410	0.7380	1.0000	1.0000	0.9780	0.9780
$p_2 = 15$	0.00	4.8160	2.4820	1.3420	1.2380	2.0840	2.1040	0.00	1.7960	1.4700	1.5250	1.3280	1.5610	1.5860	0.00	1.6030	1.4080	1.4670	1.3250	1.4640	1.5140
	0.20	3.1930	2.1840	1.1680	1.1290	1.7090	1.7170	0.20	1.0950	1.2120	1.0090	1.0430	1.1670	1.1710	0.20	1.1700	1.2660	1.0250	1.0440	1.2170	1.2190
	0.40	2.2610	1.9320	1.1010	1.0870	1.5470	1.5550	0.40	0.8870	1.1100	0.9750	1.0000	1.0880	1.0880	0.40	0.9250	1.1520	0.9940	1.0010	1.1190	1.1190
	0.60	1.8380	1.7620	1.0710	1.0690	1.4390	1.4450	0.60	0.7510	1.0180	0.9890	0.9980	1.0410	1.0410	0.60	0.7800	1.0610	0.9990	1.0000	1.0730	1.0730
	0.80	1.5880	1.6560	1.0360	1.0470	1.3620	1.3650	0.80	0.6770	0.9650	0.9920	0.9980	1.0180	1.0180	0.80	0.6880	0.9950	1.0000	1.0000	1.0450	1.0450
	1.00	1.3720	1.5470	1.0130	1.0230	1.2880	1.2900	1.00	0.6200	0.9160	0.9970	0.9990	0.9980	0.9980	1.00	0.6250	0.9430	1.0000	1.0000	1.0270	1.0270
	1.20	1.2430	1.4550	1.0120	1.0220	1.2440	1.2450	1.20	0.5810	0.8790	1.0000	1.0000	0.9880	0.9880	1.20	0.5810	0.9020	1.0000	1.0000	1.0160	1.0160
	2.00	0.9140	1.2290	0.9920	1.0020	1.1260	1.1260	2.00	0.4830	0.7840	1.0000	1.0000	0.9670	0.9670	2.00	0.4800	0.7960	1.0000	1.0000	0.9900	0.9900
$p_2 = 20$	0.00	3.3870	2.1300	1.6240	1.4060	2.3570	2.3990	0.00	1.7110	1.4720	1.5200	1.3590	1.5880	1.6260	0.00	1.6260	1.4050	1.5100	1.3380	1.5240	1.5550
	0.20	2.2970	1.9430	1.1600	1.1330	1.8150	1.8190	0.20	1.1830	1.2760	1.0280	1.0500	1.2470	1.2480	0.20	0.9820	1.2030	0.9870	1.0110	1.1730	1.1730
	0.40	1.6600	1.7790	1.0320	1.0370	1.5440	1.5450	0.40	0.9440	1.1660	0.9900	1.0030	1.1420	1.1420	0.40	0.7140	1.0480	0.9980	1.0000	1.0810	1.0810
	0.60	1.2790	1.5970	1.0040	1.0100	1.3960	1.3960	0.60	0.8020	1.0800	0.9970	0.9990	1.0870	1.0870	0.60	0.5690	0.9330	0.9990	1.0000	1.0350	1.0350
	0.80	1.0940	1.4850	1.0000	1.0010	1.3060	1.3060	0.80	0.7200	1.0240	1.0000	1.0000	1.0600	1.0600	0.80	0.4960	0.8560	1.0000	1.0000	1.0100	1.0100
	1.00	0.9600	1.3930	0.9990	1.0000	1.2480	1.2480	1.00	0.6630	0.9770	1.0000	1.0000	1.0390	1.0390	1.00	0.4380	0.7940	1.0000	1.0000	0.9960	0.9960
	1.20	0.8560	1.3130	0.9990	1.0000	1.2050	1.2050	1.20	0.6180	0.9370	1.0000	1.0000	1.0240	1.0240	1.20	0.4070	0.7520	1.0000	1.0000	0.9850	0.9850
	2.00	0.6610	1.1050	1.0000	1.0000	1.1110	1.1110	2.00	0.5240	0.8400	1.0000	1.0000	0.9960	0.9960	2.00	0.3350	0.6490	1.0000	1.0000	0.9670	0.9670

Table 9: RMSE values for $\rho = 0.9$

		$n = 50$						$n = 100$						$n = 200$							
	Δ	RL	LLS	LPT	LSPE	LS	LPS	Δ	RL	LLS	LPT	LSPE	LS	LPS	Δ	RL	LLS	LPT	LSPE	LS	LPS
$p_2 = 10$	0.00	6.3380	2.7530	1.6390	1.4210	2.1990	2.2690	0.00	2.3400	1.7220	1.8310	1.4960	1.7850	1.8630	0.00	1.8400	1.5440	1.6010	1.4080	1.5510	1.6410
	0.20	4.8170	2.4920	1.4890	1.3280	1.9750	2.0310	0.20	2.2310	1.6880	1.5670	1.3620	1.6930	1.7180	0.20	1.6700	1.4740	1.3470	1.2610	1.4440	1.4640
	0.40	3.8880	2.3260	1.4990	1.3400	1.9500	1.9760	0.40	2.0180	1.6400	1.4540	1.3180	1.5870	1.6160	0.40	1.5080	1.4140	1.2210	1.1880	1.3630	1.3750
	0.60	3.4110	2.1970	1.4320	1.3030	1.8170	1.8440	0.60	1.8730	1.6150	1.3140	1.2420	1.5350	1.5420	0.60	1.3770	1.3530	1.1200	1.1200	1.2830	1.2870
	0.80	2.9450	2.0640	1.3950	1.2850	1.7430	1.7720	0.80	1.7230	1.5510	1.2620	1.2120	1.4670	1.4740	0.80	1.2900	1.3140	1.0710	1.0840	1.2410	1.2440
	1.00	2.7440	2.0110	1.3270	1.2420	1.6830	1.7080	1.00	1.6010	1.5130	1.1880	1.1660	1.4110	1.4170	1.00	1.2130	1.2760	1.0400	1.0620	1.2100	1.2110
	1.20	2.4800	1.9250	1.2750	1.2090	1.6070	1.6320	1.20	1.5050	1.4690	1.1450	1.1380	1.3720	1.3740	1.20	1.1440	1.2270	1.0190	1.0430	1.1690	1.1700
	2.00	1.9640	1.7180	1.1910	1.1590	1.4640	1.4740	2.00	1.2640	1.3470	1.0600	1.0820	1.2600	1.2610	2.00	1.0090	1.1380	0.9920	1.0100	1.1030	1.1030
$p_2 = 15$	0.00	6.3480	2.7200	1.3580	1.2460	2.2390	2.2580	0.00	4.2770	2.4180	1.7740	1.5030	2.5140	2.5770	0.00	3.2630	1.9770	2.1750	1.6230	2.3600	2.4310
	0.20	4.0210	2.3630	1.3630	1.2560	2.0500	2.0730	0.20	3.3780	2.1520	1.5180	1.3500	2.1730	2.2010	0.20	3.3300	2.0190	1.6420	1.3900	2.1370	2.1490
	0.40	3.3410	2.1960	1.2770	1.2050	1.8610	1.8740	0.40	2.8410	1.9910	1.4350	1.3020	1.9560	1.9730	0.40	2.9920	1.9820	1.4120	1.2730	1.9560	1.9640
	0.60	2.9400	2.0960	1.2130	1.1660	1.7540	1.7720	0.60	2.5670	1.8900	1.3370	1.2400	1.8340	1.8440	0.60	2.7360	1.9410	1.2890	1.2040	1.8310	1.8340
	0.80	2.4520	1.9390	1.2080	1.1660	1.6580	1.6840	0.80	2.3080	1.7970	1.3070	1.2250	1.7480	1.7540	0.80	2.4370	1.8460	1.1960	1.1450	1.7110	1.7120
	1.00	2.3740	1.9190	1.1640	1.1340	1.5980	1.6210	1.00	2.1760	1.7480	1.2610	1.1950	1.6730	1.6780	1.00	2.2670	1.8120	1.1430	1.1100	1.6400	1.6410
	1.20	2.1000	1.8060	1.1710	1.1440	1.5410	1.5620	1.20	1.9850	1.6590	1.2490	1.1890	1.6000	1.6040	1.20	2.1650	1.7860	1.1050	1.0840	1.5720	1.5730
	2.00	1.6260	1.5960	1.0810	1.0890	1.3850	1.3970	2.00	1.6540	1.4910	1.1540	1.1270	1.4390	1.4410	2.00	1.7730	1.6370	1.0320	1.0290	1.4120	1.4120
$p_2 = 20$	0.00	4.1120	2.3070	1.5860	1.3800	2.5330	2.5660	0.00	2.8100	1.9070	2.1040	1.6320	2.3240	2.3990	0.00	2.6780	1.8600	2.2270	1.6850	2.2660	2.3670
	0.20	3.3680	2.1530	1.3830	1.2660	2.1780	2.1890	0.20	2.2590	1.6970	1.5950	1.3790	1.8960	1.9160	0.20	2.1680	1.7440	1.6190	1.4360	1.9100	1.9470
	0.40	2.9880	2.0740	1.2770	1.2030	2.0050	2.0120	0.40	1.9560	1.5920	1.4250	1.2920	1.6970	1.7090	0.40	1.7990	1.6130	1.3760	1.3080	1.6850	1.7070
	0.60	2.7030	1.9930	1.2010	1.1520	1.8810	1.8830	0.60	1.8170	1.5450	1.2830	1.2090	1.5990	1.6030	0.60	1.6130	1.5540	1.2390	1.2280	1.5640	1.5790
	0.80	2.4460	1.9100	1.1600	1.1260	1.7770	1.7790	0.80	1.6410	1.4670	1.2100	1.1650	1.4900	1.4920	0.80	1.4560	1.4760	1.1500	1.1640	1.4590	1.4680
	1.00	2.3430	1.8910	1.1270	1.1020	1.7320	1.7330	1.00	1.5270	1.4140	1.1620	1.1330	1.4240	1.4260	1.00	1.3570	1.4320	1.0980	1.1260	1.4100	1.4140
	1.20	2.1710	1.8180	1.0870	1.0720	1.6470	1.6480	1.20	1.4610	1.3860	1.1170	1.1030	1.3770	1.3780	1.20	1.2760	1.3910	1.0550	1.0870	1.3580	1.3590
	2.00	1.8220	1.6670	1.0460	1.0420	1.4910	1.4910	2.00	1.2360	1.2560	1.0470	1.0530	1.2380	1.2380	2.00	1.0570	1.2490	0.9900	1.0260	1.2160	1.2170