Supplementary Materials for "Speeding Up EM-type Algorithms via Surrogation for MLE of the Generalized Hyperbolic Distribution"

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APPENDIX A

PROOF OF THE VARIATIONAL REPRESENTATIONS

Proof: We first prove the following inequality

$$-\log K_{\lambda}(\underline{\delta\gamma}) \ge \frac{\partial -\log K_{\underline{\lambda}}(\underline{\delta\gamma})}{\lambda^2} \lambda^2 + \text{const.},\tag{1}$$

where the equality holds for $\lambda = \underline{\lambda}$. To prove the inequality, a result is introduced.

Remark 1. Denote functions $f_1(x)$ and $f_2(x)$. A sufficient condition for

$$f_1(x) \ge f_2(x)$$
, with $f_1(\underline{x}) = f_2(\underline{x})$ and $f_1'(\underline{x}) = f_2'(\underline{x})$,

is given by

$$\begin{cases} g(x) > 0 & x > \underline{x}, \\ g(x) < 0 & x < \underline{x}. \end{cases}$$
 (2)

where $g(x) = f'_1(x) - f'_2(x)$. Besides, if g(x) is monotonically increasing, the condition (2) holds true.

For inequality (1), we compute the difference of the first derivative of the both sides with respect to λ as

$$\begin{split} g_1(\lambda) &= \frac{\partial - \log K_{\lambda}(\underline{\delta}\underline{\gamma})}{\partial \lambda} - 2\frac{\partial - \log K_{\underline{\lambda}}(\underline{\delta}\underline{\gamma})}{\lambda^2} \lambda \\ &= \frac{\partial - \log K_{\lambda}(\underline{\delta}\underline{\gamma})}{\partial \lambda} - \frac{\partial - \log K_{\underline{\lambda}}(\underline{\delta}\underline{\gamma})}{\partial \lambda} \frac{\lambda}{\underline{\lambda}} \\ &= \lambda \left(\frac{\partial - \log K_{\lambda}(\underline{\delta}\underline{\gamma})}{\partial \lambda} \frac{1}{\lambda} - \frac{\partial - \log K_{\underline{\lambda}}(\underline{\delta}\underline{\gamma})}{\partial \underline{\lambda}} \frac{1}{\underline{\lambda}} \right) \\ &= 2\lambda \left(\frac{\partial \log K_{\lambda}(\underline{\delta}\underline{\gamma})}{\partial \lambda^2} - \frac{\partial \log K_{\underline{\lambda}}(\underline{\delta}\underline{\gamma})}{\partial \lambda^2} \right), \end{split}$$

with $g_1(\underline{\lambda}) = 0$.

Consider the case where $\lambda>0$ and $\underline{\lambda}>0$. According to [1, Theorem 1], the function $-\log K_{\lambda}(x)$ is convex to λ^2 on $\lambda>0$. Hence, $g_1(\lambda)$ is monotonically increasing. Based on Lemma 1, the inequality (1) is valid for $\lambda>0$ and $\underline{\lambda}>0$. Since the functions on both sides of (1) are even, the upperbound is also valid for all $\lambda\neq0$ and $\underline{\lambda}\neq0$.

Now we consider proving the other inequality.

$$-\log K_{\underline{\lambda}}(\delta\gamma) \le \frac{1}{2} \frac{\partial -\log K_{\underline{\lambda}}(\delta\gamma)}{\partial \log(\delta\gamma)} \log(\delta^2\gamma^2) + \text{const.},\tag{3}$$

where the equality holds for both $\delta = \underline{\delta}$ and $\gamma = \gamma$.

For inequality (3), the difference of the first derivative of the both sides with respect to $\gamma\delta$ is given by 1

$$\begin{split} g_2(\delta\gamma) &= -\frac{K_{\underline{\lambda}}'\left(\delta\gamma\right)}{K_{\underline{\lambda}}\left(\delta\gamma\right)} + \frac{\nu}{\delta\gamma} \\ &= -\frac{K_{\underline{\lambda}}'\left(\delta\gamma\right)}{K_{\underline{\lambda}}\left(\delta\gamma\right)} + \frac{K_{\underline{\lambda}}'\left(\underline{\delta\underline{\gamma}}\right)}{K_{\underline{\lambda}}\left(\underline{\delta\underline{\gamma}}\right)} \frac{\underline{\delta\underline{\gamma}}}{\delta\overline{\gamma}} \\ &= -\frac{1}{\delta\gamma} \left(\gamma\delta\frac{K_{\underline{\lambda}}'\left(\delta\gamma\right)}{K_{\underline{\lambda}}\left(\delta\gamma\right)} - \underline{\delta\underline{\gamma}}\frac{K_{\underline{\lambda}}'\left(\underline{\delta\underline{\gamma}}\right)}{K_{\underline{\lambda}}\left(\underline{\delta\underline{\gamma}}\right)}\right). \end{split}$$

Since the function $x \frac{K'_{\lambda}(x)}{K_{\lambda}(x)}$ is strictly decreasing on x > 0, by Lemma 1, the inequality (3) is valid.

¹Throughout this paper, $\frac{\partial}{\partial x}K_{\lambda}(x)$ is denoted as $K'_{\lambda}(x)$.

APPENDIX B PROOF OF LEMMA 1

Proof: Since both f(x) and $f_2(x)$ are convex and $f_1(x)$ is concave, we have these two surrogate functions as follows:

$$f(x) \ge f_1(\underline{x}) + f'_1(\underline{x})(x - \underline{x}) + f_3(x|\underline{x}),$$

and

$$f(x) \ge f_1(x) + f_3(x|\underline{x}).$$

Denote the two surrogate functions on the right-hand side of the inequalities as $g_1(x)$ and $g_2(x)$. Now we will compare which surrogate function is tighter, i.e., the positivity of the following expression

$$g_1(x) - g_2(x) = f_1(\underline{x}) - f_1(x) + f'_1(\underline{x})(x - \underline{x}).$$

Since $f_1(x)$ is concave, we have

$$f_1(x) \le f_1(\underline{x}) + f_1(\underline{x})(x - \underline{x}).$$

Hence, we have $g_1(x) - g_2(x) \le 0$ and the proposition is valid.

APPENDIX C

Proof of Proposition 3

Proof: Without loss of generity, in this proof, we assume N=1. For the function

$$\phi(\delta, \gamma) = \underline{\lambda} \log \frac{\gamma}{\delta} - \log K_{\underline{\lambda}}(\delta \gamma),$$

we define that $\phi(\delta) = \phi(\delta, \gamma = \underline{\gamma})$ and $\phi(\gamma) = \phi(\delta = \underline{\delta}, \gamma)$. The second derivative for $\phi(\delta)$ can be computed as For $\underline{\lambda} > \frac{1}{2}$, the second derivative of $f(\delta)$ is given by

$$\phi''(\delta) = \frac{\underline{\lambda}}{\delta^2} - \frac{K''_{\underline{\lambda}}(\gamma\delta)}{K_{\underline{\lambda}}(\gamma\delta)} \underline{\gamma}^2 + \left(\frac{K'_{\underline{\lambda}}(\gamma\delta)}{K_{\underline{\lambda}}(\gamma\delta)} \underline{\gamma}\right)^2$$
$$> -\frac{-2\underline{\lambda} + \frac{1}{2} + h(\underline{\gamma}, \delta)}{\delta^2},$$

Since the second derivative of Bessel function satisfies

$$K_{\lambda}''(x) = \left(1 + \frac{\lambda^2}{x^2}\right)K_{\lambda}(x) - \frac{K_{\lambda}'(x)}{x},\tag{4}$$

we have

$$\phi''(\delta) = -\frac{\underline{\lambda} + \delta^2 \underline{\gamma}^2 + \underline{\lambda}^2}{\delta^2} + \frac{K'_{\underline{\lambda}} \left(\delta \underline{\gamma} \right)}{K_{\underline{\lambda}} \left(\delta \underline{\gamma} \right)} \frac{\underline{\delta}}{\underline{\gamma}} + \left(\frac{K'_{\underline{\lambda}} \left(\delta \underline{\gamma} \right)}{K_{\underline{\lambda}} \left(\delta \underline{\gamma} \right)} \underline{\gamma} \right)^2.$$

For the function $\frac{K'_{\lambda}(x)}{K_{\lambda}(x)}$, we have the bounds on $|\lambda| > \frac{1}{2}$ [2] as

$$-\frac{1+\sqrt{(|\lambda|-1)^2+x^2}}{x} \le \frac{K_{\lambda}'(x)}{K_{\lambda}(x)} \le -\frac{\frac{1}{2}+\sqrt{(|\lambda|-\frac{1}{2})^2+x^2}}{x}.$$
 (5)

By the above bounds, we can obtain

$$\phi''(\delta) > -\frac{-\underline{\lambda} + \delta^2 \underline{\gamma}^2 + \underline{\lambda}^2}{\delta^2} - \frac{\underline{\gamma}}{\delta} \frac{1 + \sqrt{(|\underline{\lambda}| - 1)^2 + \delta^2 \underline{\gamma}^2}}{\delta \underline{\gamma}} + \left(-\frac{\frac{1}{2} + \sqrt{(|\underline{\lambda}| - \frac{1}{2})^2 + \delta^2 \underline{\gamma}^2}}{\delta \underline{\gamma}} \underline{\gamma} \right)^2$$

$$= -\frac{-2\underline{\lambda} + \frac{1}{2} + h(\delta, \underline{\gamma})}{\delta^2},$$

where

$$h(\delta, \gamma) = \sqrt{(|\underline{\lambda}| - 1)^2 + \delta^2 \gamma^2} - \sqrt{(|\underline{\lambda}| - \frac{1}{2})^2 + \delta^2 \gamma^2}$$

$$= \frac{-|\underline{\lambda}| + \frac{3}{4}}{\sqrt{(|\underline{\lambda}| - 1)^2 + \delta^2 \gamma^2} + \sqrt{(|\underline{\lambda}| - \frac{1}{2})^2 + \delta^2 \gamma^2}}.$$
(6)

Hence, we can obtain that $\phi(\delta)$ is a convex function by the following two cases:

- 1) When $\underline{\lambda} > \frac{3}{4}$, we have $h(\delta, \underline{\gamma}) < 0$ so that $\phi''(\delta) > 0$.
- 2) When $\frac{1}{2} < \underline{\lambda} < \frac{3}{4}$, we have $h(\delta, \gamma)$ is strictly decreasing on both δ and γ , i.e., $0 < h(\delta, \underline{\gamma}) < h(0, \underline{\gamma}) = -2|\underline{\lambda}| + \frac{3}{2}$. Hence, we have

$$\phi''(\delta) > -\frac{-2\underline{\lambda} + \frac{1}{2} + h(\delta, \underline{\gamma})}{\delta^2} > -\frac{-4\underline{\lambda} + 2}{\delta^2} > 0.$$

Then for $\underline{\lambda} < -\frac{1}{2}$, the second derivative for $\phi(\gamma)$ can be computed as

$$\phi''(\gamma) = -\frac{\underline{\lambda}}{\gamma^2} - \frac{K''_{\underline{\lambda}}(\underline{\delta}\gamma)}{K_{\underline{\lambda}}(\underline{\delta}\gamma)} \underline{\delta}^2 + \left(\frac{K'_{\underline{\lambda}}(\underline{\delta}\gamma)}{K_{\underline{\lambda}}(\underline{\delta}\gamma)} \underline{\delta}\right)^2$$
$$> -\frac{2\underline{\lambda} + \frac{1}{2} + h(\underline{\delta}, \gamma)}{\gamma^2}.$$

Therefore, we can similarly obtain that $\phi''(\gamma)$ is convex on $\gamma > 0$ and $\underline{\lambda} > \frac{1}{2}$.

REFERENCES

- [1] Á. Baricz and S. Ponnusamy, "On Turán type inequalities for modified Bessel functions," *Proceedings of the American Mathematical Society*, vol. 141, no. 2, pp. 523–532, 2012.
- [2] J. Segura, "Simple bounds with best possible accuracy for ratios of modified Bessel functions," *Journal of Mathematical Analysis and Applications*, vol. 526, no. 1, p. 127211, 2023.