

1 Appendix A

In this section, the derivation of conditional expectations in the E-step is presented, and the details are shown as following. There are five conditional expectations: $E_{\Theta}(Z_{ik}|\mathbf{X})$, $E_{\Theta}(Z_{ik}u_i|\mathbf{X})$, $E_{\Theta}(Z_{ik}u_i\tau_i|\mathbf{X})$, $E_{\Theta}(Z_{ik}u_i\tau_i^2|\mathbf{X})$, $E_{\Theta}(Z_{ik}\log u_i|\mathbf{X})$, will be estimated in proposed method. Before introducing them, we should calculate some probability density functions(pdf) by the truncated distribution represent in Eq. (23). If \mathbf{X} can be formulated in a three-level hierarchical representation in Eq. (23), the density function of $f(x_i, u_i, \tau_i)$ is given by:

$$\begin{aligned}
f(x_i, u_i, \tau_i) &= p(x_i|u_i, \tau_i)p(t_i|u_i)p(u_i) \\
&= \frac{1}{(2\pi)^{D/2}|u_i^{-1}\Upsilon_k|^{1/2}} \exp\left(-\frac{u_i(x_i - \mu_k - \Delta_k\tau_i)^T\Upsilon_k^{-1}(x_i - \mu_k - \Delta_k\tau_i)}{2}\right) \\
&\quad \times \frac{2}{(2\pi)^{1/2}|u_i^{-1}|^{1/2}} \exp\left(-\frac{u_i\tau_i^2}{2}\right) \times \frac{(\nu_k/2)^{\nu_k/2}}{\Gamma(\nu_k/2)} u_i^{\frac{\nu_k}{2}} \exp\left(-\frac{\nu_k}{2}u_i\right) \\
&= \frac{2}{(2\pi)^{(D+1)/2}|\Upsilon_k|^{1/2}} u_i^{\frac{\nu_k+D+1}{2}-1} \times \frac{(\nu_k/2)^{\nu_k/2}}{\Gamma(\nu_k/2)} \times \exp\left(-\frac{u_i[(x_i - \mu_k)^T\Sigma_k^{-1}(x_i - \mu_k) + \nu_k]}{2}\right) \\
&\quad \times \exp\left(-\frac{u_i[\tau_i - \Delta_k^T\Sigma_k^{-1}(x_i - \mu_k)]^2}{2(1 - \delta_k^T\delta_k)}\right)
\end{aligned} \tag{1}$$

Then the density function of $f(x_i, u_i)$ is written as:

$$\begin{aligned}
f(x_i, u_i) &= \int_0^\infty f(x_i, u_i, \tau_i) d\tau_i \\
&= \frac{2}{(2\pi)^{(D+1)/2}|\Upsilon_k|^{1/2}} u_i^{\frac{\nu_k+D+1}{2}-1} \times \frac{(\nu_k/2)^{\nu_k/2}}{\Gamma(\nu_k/2)} \times \int_0^\infty \exp\left(-\frac{u_i[(x_i - \mu_k)^T\Sigma_k^{-1}(x_i - \mu_k) + \nu_k]}{2}\right) \\
&\quad \times \exp\left(-\frac{u_i[\tau_i - \Delta_k^T\Sigma_k^{-1}(x_i - \mu_k)]^2}{2(1 - \delta_k^T\delta_k)}\right) d\tau_i \\
&= \frac{2}{(2\pi)^{(D+1)/2}|\Sigma_k|^{1/2}} u_i^{\frac{\nu_k+D}{2}-1} \times \frac{(\nu_k/2)^{\nu_k/2}}{\Gamma(\nu_k/2)} \times \exp\left(-\frac{u_i[(x_i - \mu_k)^T\Sigma_k^{-1}(x_i - \mu_k) + \nu_k]}{2}\right) \\
&\quad \times \Phi(\lambda_k^T\Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})
\end{aligned} \tag{2}$$

Similarly, the density function of $f(x_i, \tau_i)$ is written as:

$$\begin{aligned}
f(x_i, \tau_i) &= \int_0^\infty f(x_i, u_i, \tau_i) du_i \\
&= \frac{2}{(2\pi)^{(D+1)/2}|\Upsilon_k|^{1/2}} \times \frac{(\nu_k/2)^{\nu_k/2}}{\Gamma(\nu_k/2)} \times \int_0^\infty u_i^{\frac{\nu_k+D+1}{2}-1} \times \exp\left(-\frac{u_i(x_i - \mu_k)^T\Sigma_k^{-1}(x_i - \mu_k) + \nu_k}{2}\right) \\
&\quad \times \exp\left(-\frac{u_i[\tau_i - \Delta_k^T\Sigma_k^{-1}(x_i - \mu_k)]^2}{2(1 - \delta_k^T\delta_k)}\right) du_i \\
&= \frac{2}{(2\pi)^{(D+1)/2}|\Upsilon_k|^{1/2}} \times \frac{(\nu_k/2)^{\nu_k/2}}{\Gamma(\nu_k/2)} \times \Gamma\left(\frac{\nu_k + D + 1}{2}\right) \times \left[\frac{\frac{\nu_k + (x_i - \mu_k)^T\Sigma_k^{-1}(x_i - \mu_k)}{2}}{+ \frac{((1 - \delta^T\delta)^{-1/2}\tau_i - \lambda_k^T\Sigma_k^{-1/2}(x_i - \mu_k))^2}{2}} \right]^{-\left(\frac{\nu_k+D+1}{2}-1\right)}
\end{aligned} \tag{3}$$

Then the density of $f(u_i|x_i)$ is calculated by:

$$\begin{aligned}
f(u_i|x_i) &= f(x_i, u_i)/f(x_i) \\
&= \frac{f(x_i, u_i)}{2^D t_D(x_i|\mu_k, \Sigma_k, \nu_k) T(\sqrt{(\nu_k + D)/(\nu_k + d)}A, \nu_k + D)} \\
&= \frac{u_i^{\frac{\nu_k+D}{2}-1} \exp(-u_i[(x_i - \mu_k)^T\Sigma_k^{-1}(x_i - \mu_k) + \nu_k]/2)}{\Gamma((\nu_k + D)/2) \times T(\sqrt{(\nu_k + D)/(\nu_k + d)}A, \nu_k + D)} \times \Phi(\lambda_k^T\Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})((\nu_k + D)/2)^{\frac{\nu_k+D}{2}}
\end{aligned} \tag{4}$$

For $f(\tau_i|x_i, u_i)$, it can be calculated by:

$$\begin{aligned} f(\tau_i|x_i, u_i) &= f(x_i, u_i, \tau_i)/f(x_i, u_i) \\ &= \frac{\sqrt{u_i}}{\sqrt{2\pi}(1-\delta^T\delta)^{1/2} \times \Phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})} \times \exp\left(-\frac{u_i[\tau_i - \Delta_k^T \Sigma_k^{-1}(x_i - \mu_k)]^2}{2(1-\delta_k^T \delta_k)}\right) \end{aligned} \quad (5)$$

thus, $\tau_i|x_i, u_i \sim TN(\Delta_k^T \Sigma_k^{-1}(x_i - \mu_k), u_i^{-1}(1-\delta^T\delta))$.

In the next step, we will employ the pdf above to estimated the conditional expectations. Firstly, employing the proposition in [1]:

If $x \sim \Gamma(\alpha, \beta)$, then for any $a \in \mathbb{R}$, $E(\Phi(a\sqrt{x})) = T_{2\alpha}(a\sqrt{\frac{\alpha}{\beta}})$,

then we get:

$$\begin{aligned} E(u_i|x_i) &= \int_0^\infty u_i f(u_i|x_i) du_i \\ &= \frac{\Gamma(\frac{\nu_k+D+2}{2})}{\Gamma(\frac{\nu_k+D}{2}) \times T(\sqrt{\frac{D+\nu_k}{d+\nu_k}} A, \nu_k + D)} \times \frac{(\frac{\nu_k+d}{2})^{\frac{\nu_k+D}{2}}}{(\frac{[(x_i-\mu_k)^T \Sigma_k^{-1}(x_i-\mu_k)+\nu_k]}{2})^{\frac{\nu_k+D+2}{2}}} \\ &\quad \times \int_0^\infty \gamma(u_i|\frac{\nu_k+D+2}{2}, \frac{[(x_i-\mu_k)^T \Sigma_k^{-1}(x_i-\mu_k)+\nu_k]}{2}) \times \Phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i}) du_i \\ &= \frac{\Gamma(\frac{\nu_k+D+2}{2})}{\Gamma(\frac{\nu_k+D}{2}) \times T(\sqrt{\frac{D+\nu_k}{d+\nu_k}} A, \nu_k + D)} \times \frac{(\frac{\nu_k+d}{2})^{\frac{\nu_k+D}{2}}}{(\frac{[(x_i-\mu_k)^T \Sigma_k^{-1}(x_i-\mu_k)+\nu_k]}{2})^{\frac{\nu_k+D+2}{2}}} \times E(\Phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})) \\ &= (\frac{\nu_k+D}{d+\nu_k}) \times \frac{T(\sqrt{\frac{D+\nu_k+2}{d+\nu_k}} A, \nu_k + D + 2)}{T(\sqrt{\frac{D+\nu_k}{d+\nu_k}} A, \nu_k + D)} \end{aligned} \quad (6)$$

As for $E(\tau_i|u_i, x_i)$, $E(\tau_i^2|u_i, x_i)$, we use the Proposition in [1]: If $\tau_i|x_i, u_i \sim TN(\Delta_k^T \Sigma_k^{-1}(x_i - \mu_k), u_i^{-1}(1-\delta^T\delta))$, then we have:

$$E(\tau_i|u_i, x_i) = \Delta_k^T \Sigma_k^{-1}(x_i - \mu_k) + \frac{\phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})}{\Phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})} \times \sqrt{u_i^{-1}(1-\delta^T\delta)} \quad (7)$$

$$E(\tau_i^2|u_i, x_i) = (\Delta_k^T \Sigma_k^{-1}(x_i - \mu_k))^2 + u_i^{-1}(1-\delta^T\delta) + \frac{\phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})}{\Phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})} \Delta_k^T \Sigma_k^{-1}(x_i - \mu_k) \sqrt{u_i^{-1}(1-\delta^T\delta)} \quad (8)$$

Then $E(u_i \tau_i|x_i)$ is:

$$\begin{aligned} E(u_i \tau_i|x_i) &= E(u_i E(\tau_i|u_i, x_i)|x_i) \\ &= E(u_i|x_i) \times \Delta_k^T \Sigma_k^{-1}(x_i - \mu_k) + \sqrt{(1-\delta^T\delta)} \times E(\sqrt{u_i} \frac{\phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})}{\Phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})} |x_i) \end{aligned} \quad (9)$$

As for $E(\sqrt{u_i} \frac{\phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})}{\Phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})} |x_i)$, it can be written as:

$$\begin{aligned} E(\sqrt{u_i} \frac{\phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})}{\Phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})} |x_i) &= \int_0^\infty \sqrt{u_i} \frac{\phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})}{\Phi(\lambda_k^T \Sigma_k^{-1/2}(x_i - \mu_k)\sqrt{u_i})} \frac{f(u_i, x_i)}{f(x_i)} du_i \\ &= \frac{t_D(x_i|\mu_k, \Sigma_k, \nu_k)}{f(x_i)} \times \frac{2\Gamma(\frac{\nu_k+D+1}{2}) \times (\nu_k + d)^{(D+\nu_k)/2}}{\Gamma(\frac{\nu_k+D}{2}) \pi^{1/2} (A^2 + \nu_k + d)^{\frac{\nu_k+D+d}{2}}} \end{aligned} \quad (10)$$

Then, $E(u_i \tau_i|x_i)$ is:

$$E(u_i \tau_i|x_i) = E(u_i|x_i) S_{1,i,j} + M_i S_{2,i,j} \quad (11)$$

And $E(u_i \tau_i^2 | x_i)$ is:

$$\begin{aligned}
E(u_i \tau_i^2 | x_i) &= E(u_i E(\tau_i^2 | u_i, x_i) | x_i) \\
&= E(u_i | x_i) \times (\Delta_k^T \Sigma_k^{-1} (x_i - \mu_k))^2 + (1 - \delta^T \delta) + (\Delta_k^T \Sigma_k^{-1} (x_i - \mu_k)) \sqrt{u_i^{-1} (1 - \delta^T \delta)} \\
&\quad \times E(\sqrt{u_i} \frac{\phi(\lambda_k^T \Sigma_k^{-1/2} (x_i - \mu_k) \sqrt{u_i})}{\Phi(\lambda_k^T \Sigma_k^{-1/2} (x_i - \mu_k) \sqrt{u_i})} | x_i) \\
&= S_{1,i,j}^2 E(u_i | x_i) + M_i^2 + M_i S_{2,i,j} S_{1,i,j}
\end{aligned} \tag{12}$$

As for $E(\log u_i | x_i)$, it can be gotten by the formula $\frac{d}{d\nu_k} \int_0^\infty f(u_i | x_i) du_i = 0$, where $\frac{d}{d\nu_k} \int_0^\infty f(u_i | x_i) du_i$ is given by:

$$\begin{aligned}
\frac{d}{d\nu_k} \int_0^\infty f(u_i | x_i) du_i &= \int_0^\infty \exp(-\frac{u_i [(x_i - \mu_k)^T \Sigma_k^{-1} (x_i - \mu_k)]}{2}) \\
&\quad \times \Phi(\lambda_k^T \Sigma_k^{-1/2} (x_i - \mu_k) \sqrt{u_i}) \times \frac{d}{d\nu_k} \left\{ \frac{u_i^{\frac{\nu_k+D}{2}-1} \exp(-\frac{u_i \nu_k}{2}) (\frac{\nu_k+d}{2})^{\frac{\nu_k+D}{2}}}{\Gamma(\frac{\nu_k+D}{2}) T(\sqrt{\frac{D+\nu_k}{d+\nu_k}} A, \nu_k + D)} \right\} du_i \\
&= \frac{1}{2} E(\log u_i | x_i) - \frac{1}{2} E(u_i | x_i) + \frac{1}{2} [\log(\frac{\nu_k + d}{2}) + \frac{\nu_k + D}{\nu_k + d}] - \frac{1}{2} \frac{\Gamma'(\frac{\nu_k+d}{2})}{\Gamma(\frac{\nu_k+d}{2})} \\
&\quad - \frac{1}{2} \frac{t_{\nu_k+D}(\sqrt{\frac{D+\nu_k}{d+\nu_k}} A)}{T(\sqrt{\frac{D+\nu_k}{d+\nu_k}} A, \nu_k + D)} A (\frac{d + \nu_k}{D + \nu_k})^{1/2} (\frac{d - D}{(d + \nu_k)^2}) \\
&\quad - \frac{1}{2} T^{-1}(\sqrt{\frac{D + \nu_k}{d + \nu_k}} A, \nu_k + D) \times \int_{-\infty}^{\sqrt{\frac{D+\nu_k}{d+\nu_k}} A} g(p, \nu_k + D) dp
\end{aligned} \tag{13}$$

Then, we have

$$\begin{aligned}
E(\log u_i) &= E(u_i | x_i) - (\log \frac{\nu_k + d}{2} + \frac{\nu_k + D}{\nu_k + d} - \psi(\frac{\nu_k + D}{2})) \\
&\quad + (\frac{t_D(\sqrt{\frac{D+\nu_k}{d+\nu_k}} A, \nu_k + D)}{T(\sqrt{\frac{D+\nu_k}{d+\nu_k}} A, \nu_k + D)} \times \frac{A(d - D)}{\sqrt{(D + \nu_k)(d + \nu_k)^3}}) \\
&\quad + T^{-1}(\sqrt{\frac{D + \nu_k}{d + \nu_k}} A, \nu_k + D) \times \int_{-\infty}^{\sqrt{\frac{D+\nu_k}{d+\nu_k}} A} g(p, \nu_k + D) dp
\end{aligned} \tag{14}$$

where

$$\begin{aligned}
g(p, \nu_k + D) &= (\psi(\frac{\nu_k + 2D}{2}) - \psi(\frac{\nu_k + D}{2}) - \log(\frac{p^2 + \nu_k + D}{\nu_k + D})) \\
&\quad + \frac{p^2(\nu_k + D) - D(\nu_k + D)}{(p^2 + \nu_k + D)(\nu_k + D)} \cdot t_{\nu_k+D}(p)
\end{aligned} \tag{15}$$

References

- [1] T. I. Lin, J. C. Lee, and W. J. Hsieh, "Robust mixture models using the skew-t distribution," *Statistics and Computing*, vol. 17, no. 2, pp. 81–92, 2007.