

부 록

부록 A. 피셔정보행렬 (Fisher Information Matrix)의 유도

대수우도함수 $\ln L$ 을 $\delta_1, \delta_2, \beta$ 에 대해 일차 편미분하면 다음과 같다.

$$\begin{aligned}\frac{\partial \ln L}{\partial \delta_1} &= \sum_{i=1}^r \sum_{j=1}^{n_i} \sum_{k=1}^m \left[A_i \ln \Delta y_{ijk} - A_i \ln \beta - A_i \Psi_0(A_i) \right] , \\ \frac{\partial \ln L}{\partial \delta_2} &= \sum_{i=1}^r \sum_{j=1}^{n_i} \sum_{k=1}^m \left[A_i s_i \ln \Delta y_{ijk} - A_i s_i \ln \beta - A_i \Psi_0(A_i) s_i \right] , \\ \frac{\partial \ln L}{\partial \beta} &= \sum_{i=1}^r \sum_{j=1}^{n_i} \sum_{k=1}^m \left(-\frac{A_i}{\beta} + \frac{\Delta y_{ijk}}{\beta^2} \right)\end{aligned}$$

여기서, $\Psi_0(A_i)$ 는 $\Psi_0(A_i) = \partial \ln \Gamma(A_i) / \partial A_i = \Gamma'(A_i) / \Gamma(A_i)$ 로 정의되는 디감마 함수 (Digamma function)이다. 대수우도함수 $\ln L$ 을 $\delta_1, \delta_2, \beta$ 에 대해 이차 편미분하면 다음과 같다.

$$\begin{aligned}\frac{\partial^2 \ln L}{\partial \delta_1^2} &= \sum_{i=1}^r \sum_{j=1}^{n_i} \sum_{k=1}^m \left[A_i \ln \Delta y_{ijk} - A_i \ln \beta - A_i \Psi_0(A_i) - A_i^2 B_i \right] , \\ \frac{\partial^2 \ln L}{\partial \delta_1 \partial \delta_2} &= \sum_{i=1}^r \sum_{j=1}^{n_i} \sum_{k=1}^m \left[A_i s_i \ln \Delta y_{ijk} - A_i s_i \ln \beta - A_i \Psi_0(A_i) s_i - A_i^2 B_i s_i \right] , \\ \frac{\partial^2 \ln L}{\partial \delta_1 \partial \beta} &= \sum_{i=1}^r \sum_{j=1}^{n_i} \sum_{k=1}^m \left(-\frac{A_i}{\beta} \right) , \\ \frac{\partial^2 \ln L}{\partial \delta_2^2} &= \sum_{i=1}^r \sum_{j=1}^{n_i} \sum_{k=1}^m \left[A_i s_i^2 \ln \Delta y_{ijk} - A_i s_i^2 \ln \beta - A_i \Psi_0(A_i) s_i^2 - A_i^2 B_i s_i^2 \right] , \\ \frac{\partial^2 \ln L}{\partial \delta_2 \partial \beta} &= \sum_{i=1}^r \sum_{j=1}^{n_i} \sum_{k=1}^m \left(-\frac{A_i s_i}{\beta} \right) , \\ \frac{\partial^2 \ln L}{\partial \beta^2} &= \sum_{i=1}^r \sum_{j=1}^{n_i} \sum_{k=1}^m \left(\frac{A_i}{\beta^2} - \frac{2\Delta y_{ijk}}{\beta^3} \right)\end{aligned}$$

피셔정보행렬을 산출하기 위한 $E(\ln \Delta y_{ijk})$ 은 다음과 같이 구할 수 있다.

$$\begin{aligned}
E(\ln \Delta y_{ijk}) &= \int_0^\infty \ln \Delta y_{ijk} \frac{1}{\Gamma(A_i)} \frac{1}{\beta^{A_i}} \Delta y_{ijk}^{A_i-1} e^{-\Delta y_{ijk}/\beta} d\Delta y_{ijk} \\
&= \frac{1}{\Gamma(A_i)} \int_0^\infty \ln(\beta x) \cdot x^{A_i-1} e^{-x} dx \\
&= \frac{1}{\Gamma(A_i)} \left(\int_0^\infty \ln \beta \cdot x^{A_i-1} e^{-x} dx + \int_0^\infty \ln x \cdot x^{A_i-1} e^{-x} dx \right) \\
&= \frac{1}{\Gamma(A_i)} [\ln \beta \cdot \Gamma(A_i) + \Gamma'(A_i)] \\
&= \ln \beta + \Psi_0(A_i)
\end{aligned}$$

그리고, $E(\ln y_{ijk}) = A_i \beta$ 이므로 $\ln L$ 의 이차편미분의 음의 기댓값을 취하여 다음과 같고 식 (5)의 피셔정보 행렬을 구할 수 있다.

$$\begin{aligned}
E\left(-\frac{\partial^2 \ln L}{\partial \delta_1^2}\right) &= mN \sum_{i=1}^r \pi_i A_i^2 B_i, & E\left(-\frac{\partial^2 \ln L}{\partial \delta_1 \partial \delta_2}\right) &= mN \sum_{i=1}^r \pi_i A_i^2 B_i s_i, \\
E\left(-\frac{\partial^2 \ln L}{\partial \delta_1 \partial \beta}\right) &= mN \sum_{i=1}^r \pi_i \frac{A_i}{\beta}, & E\left(-\frac{\partial^2 \ln L}{\partial \delta_2^2}\right) &= mN \sum_{i=1}^r \pi_i A_i^2 B_i s_i^2, \\
E\left(-\frac{\partial^2 \ln L}{\partial \delta_2 \partial \beta}\right) &= mN \sum_{i=1}^r \pi_i \frac{A_i s_i}{\beta}, & E\left(-\frac{\partial^2 \ln L}{\partial \beta^2}\right) &= mN \sum_{i=1}^r \pi_i \frac{A_i}{\beta^2}.
\end{aligned}$$