$$\frac{E(\hat{\Theta}_{1}) = E\left[\frac{E(\hat{\Lambda}_{1})}{h}\right] = \frac{1}{h}\left(\frac{E}{H}\chi_{1}^{2} - h\chi^{2}\right)}{= \frac{1}{h}\left(h\sigma^{2} + h\mu^{2} - \sigma^{2} - h\mu^{2}\right)}$$

$$= \frac{\sigma^{2} + \mu^{2} - \sigma^{2}}{h} - \mu^{2}$$

$$= \frac{\sigma^{2} - \sigma^{2}}{h}$$

$$= \frac{n-1}{h}\sigma^{2} = \sum E(\hat{\Theta}_{1}) + \frac{1}{h}E\sigma^{2} = \sum A_{1}^{2} + \frac{1}{h}E^{2} = \sum A_{2}^{2} + \frac{1}{h}E^{2} = \sum A_{3}^{2} + \frac{1}{h}E^{2} = \sum$$

$$E(\hat{\theta}_2) = E\left[\frac{1}{n-1}(\hat{x}_1 \cdot \hat{x}_2)\right] = \frac{1}{n-1}E\left[\frac{1}{n-1}(\hat{x}_1 - \hat{x}_2)\right]$$

$$= \frac{1}{n-1} \left(n\sigma^2 + h M^2 - \sigma^2 - h M^2 \right)$$

$$= \frac{(n\sigma^2 + h M^2 - \sigma^2 - h M^2)}{(n-1)}$$

$$= \frac{h\sigma^2}{n-1} + \frac{nM^2}{n-1} - \frac{\sigma^2}{n-1} - \frac{hM^2}{n-1} \right)$$

$$= \frac{h\sigma^2}{n-1} - \frac{h}{n-1} = \sum_{n=1}^{\infty} E(\hat{\theta}_2) \ddot{h} \mathcal{L} \sigma^2 \dot{L} \mathcal{T} \begin{pmatrix} h & h \\ h & h \end{pmatrix} = \sum_{n=1}^{\infty} A_n \begin{pmatrix} h \\ h \end{pmatrix} \dot{h}$$

$$= \sum_{n=1}^{\infty} E(\hat{\theta}_2) \ddot{h} \mathcal{L} \sigma^2 \dot{L} \mathcal{T} \begin{pmatrix} h \\ h \end{pmatrix} \dot{h}$$

$$\frac{hc^2 - h-1}{h-1}$$
 => $E(\theta_2)$ 满足 σ^2 之不偏估計量=>不偏估計量