

MAXIMUM LIKELIHOOD ESTIMATORS

Risk and Asset Allocation - Springer – *symmys.com*

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www.symmys.com

Formulas and figures in this presentation refer to the book **Risk and Asset Allocation**, Springer.

The notation, say, (5.24) refers to Formula 24 in Chapter 5 of the book

The notation, say, (T4.12) refers to Formula 12 in the Technical Appendices for Chapter 4, which can be downloaded from www.symmys.com

MAXIMUM LIKELIHOOD ESTIMATORS

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$$\mathbf{G}[f_{\mathbf{X}}] \equiv \text{"unknown truth"}. \quad (4.6)$$

$$\text{information } i_T \mapsto \text{number } \hat{\mathbf{G}} \quad (4.9)$$

$$i_T \equiv \{\mathbf{x}_1, \dots, \mathbf{x}_T\} \quad (4.8)$$

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$$\hat{\theta}[i_T] \equiv \operatorname{argmax}_{\theta \in \Theta} f_{\theta}(i_T) \quad (4.66)$$

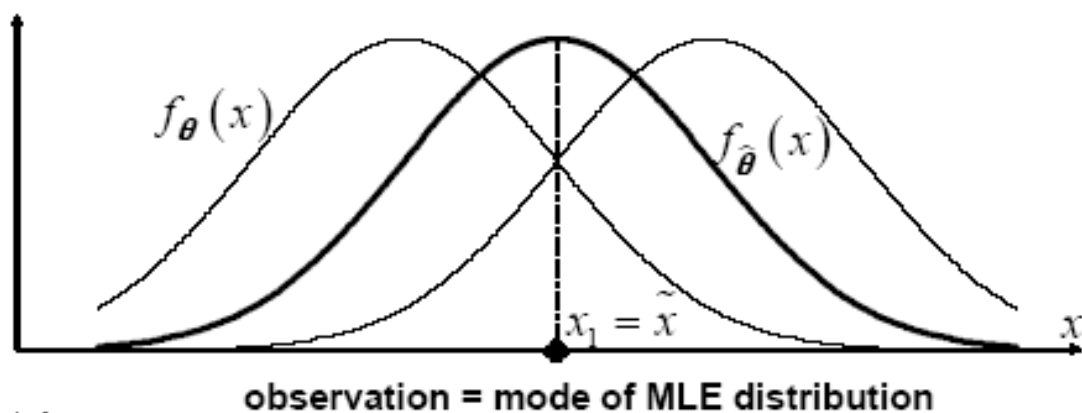


Fig. 4.10

MAXIMUM LIKELIHOOD ESTIMATORS

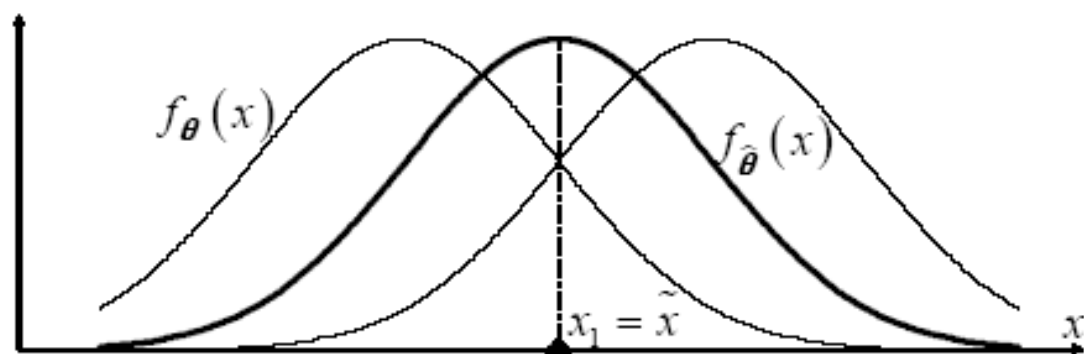
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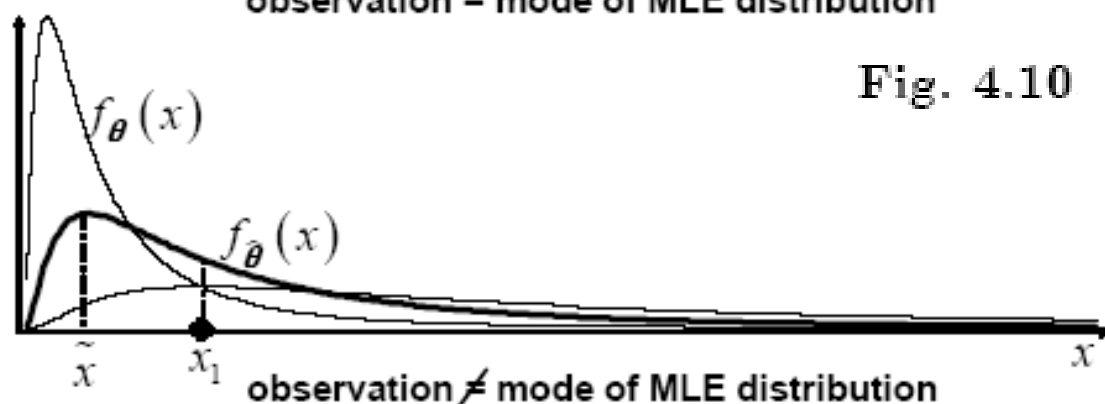
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observation = mode of MLE distribution

Fig. 4.10



observation \neq mode of MLE distribution

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$$f_{\boldsymbol{\theta}} (i_T) \equiv f_{\boldsymbol{\theta}} (x_1) \cdots f_{\boldsymbol{\theta}} (x_T). \quad (4.65)$$

$$\hat{\boldsymbol{\theta}} [i_T] \equiv \operatorname{argmax}_{\boldsymbol{\theta} \in \Theta} f_{\boldsymbol{\theta}} (i_T) \quad (4.66)$$

$$\bullet \quad \widehat{g(\boldsymbol{\theta})} = g(\hat{\boldsymbol{\theta}}) \quad (4.70)$$

$$\bullet \quad \hat{\boldsymbol{\theta}} [i_T] \sim N \left(\boldsymbol{\theta}, \frac{\boldsymbol{\Gamma}}{T} \right) \quad (4.71)$$

$$\boldsymbol{\Gamma} \equiv \operatorname{Cov} \left\{ \frac{\partial \ln (f_{\boldsymbol{\theta}} (\mathbf{X}))}{\partial \boldsymbol{\theta}} \right\} \quad (4.72)$$

MAXIMUM LIKELIHOOD ESTIMATORS – ELLIPTICAL

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$$\mathbf{X} \sim \text{El}(\boldsymbol{\mu}, \boldsymbol{\Sigma}, g) \quad (4.73)$$

$$f_{\theta}(\mathbf{x}) \equiv \frac{1}{\sqrt{|\boldsymbol{\Sigma}|}} g(\text{Ma}^2(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma})) \quad (4.74)$$

$$\text{Ma}(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}) \equiv \sqrt{(\mathbf{x} - \boldsymbol{\mu})' \boldsymbol{\Sigma}^{-1} (\mathbf{x} - \boldsymbol{\mu})} \quad (4.75)$$

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$$\hat{\boldsymbol{\mu}} = \sum_{t=1}^T \frac{w_t}{\sum_{s=1}^T w_s} \mathbf{x}_t \quad (4.81)$$

$$\hat{\boldsymbol{\Sigma}} = \frac{1}{T} \sum_{t=1}^T w_t (\mathbf{x}_t - \hat{\boldsymbol{\mu}}) (\mathbf{x}_t - \hat{\boldsymbol{\mu}})'. \quad (4.82)$$

$$w_t \equiv w(\text{Ma}^2(\mathbf{x}_t, \hat{\boldsymbol{\mu}}, \hat{\boldsymbol{\Sigma}})) \quad (4.80)$$

$$w(z) \equiv -2 \frac{g'(z)}{g(z)} \quad (4.79)$$

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$$g^N(z) \equiv \frac{e^{-\frac{z}{2}}}{(2\pi)^{\frac{N}{2}}}, \quad (4.96)$$

$$w(z) \equiv 1 \quad (4.97)$$

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$$g^{\text{Ca}}(z) = \frac{\Gamma(\frac{1+N}{2})}{\Gamma(\frac{1}{2}) (\pi)^{\frac{N}{2}}} (1+z)^{-\frac{1+N}{2}} \quad (4.83)$$

$$w_t = \frac{N+1}{1 + \text{Ma}^2(\mathbf{x}_t, \hat{\boldsymbol{\mu}}, \hat{\boldsymbol{\Sigma}})} \quad (4.84)$$

MAXIMUM LIKELIHOOD ESTIMATORS – NORMAL (LOCATION)

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normal assumption $\mathbf{X} \sim N(\boldsymbol{\mu}, \boldsymbol{\Sigma})$ (4.95)

estimator definition $\hat{\boldsymbol{\mu}}[i_T] = \frac{1}{T} \sum_{t=1}^T \mathbf{x}_t$. (4.98)

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estimator distribution
(replicability) $\hat{\boldsymbol{\mu}}[I_T] \sim \mathbf{N}\left(\boldsymbol{\mu}, \frac{\boldsymbol{\Sigma}}{T}\right)$ (4.102)

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Generalized t-tests...

Generalized p-values...

MAXIMUM LIKELIHOOD ESTIMATORS – NORMAL (LOCATION)

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estimator distribution
(replicability) $\hat{\boldsymbol{\mu}}[I_T] \sim N\left(\boldsymbol{\mu}, \frac{\boldsymbol{\Sigma}}{T}\right) \quad (4.102)$



estimator evaluation
(global) $\text{Loss}(\hat{\boldsymbol{\mu}}, \boldsymbol{\mu}) \equiv [\hat{\boldsymbol{\mu}}[I_T] - \boldsymbol{\mu}]' [\hat{\boldsymbol{\mu}}[I_T] - \boldsymbol{\mu}] \quad (4.108)$

estimator evaluation
(summary) $\left\{ \begin{array}{l} \text{Err}^2(\hat{\boldsymbol{\mu}}, \boldsymbol{\mu}) = \frac{1}{T} \text{tr}(\boldsymbol{\Sigma}). \quad (4.109) \\ \text{Inef}^2(\hat{\boldsymbol{\mu}}) = \frac{1}{T} \text{tr}(\boldsymbol{\Sigma}) \quad (4.110) \\ \text{Bias}^2(\hat{\boldsymbol{\mu}}, \boldsymbol{\mu}) = 0. \quad (4.111) \end{array} \right.$

MAXIMUM LIKELIHOOD ESTIMATORS – NORMAL (LOCATION)

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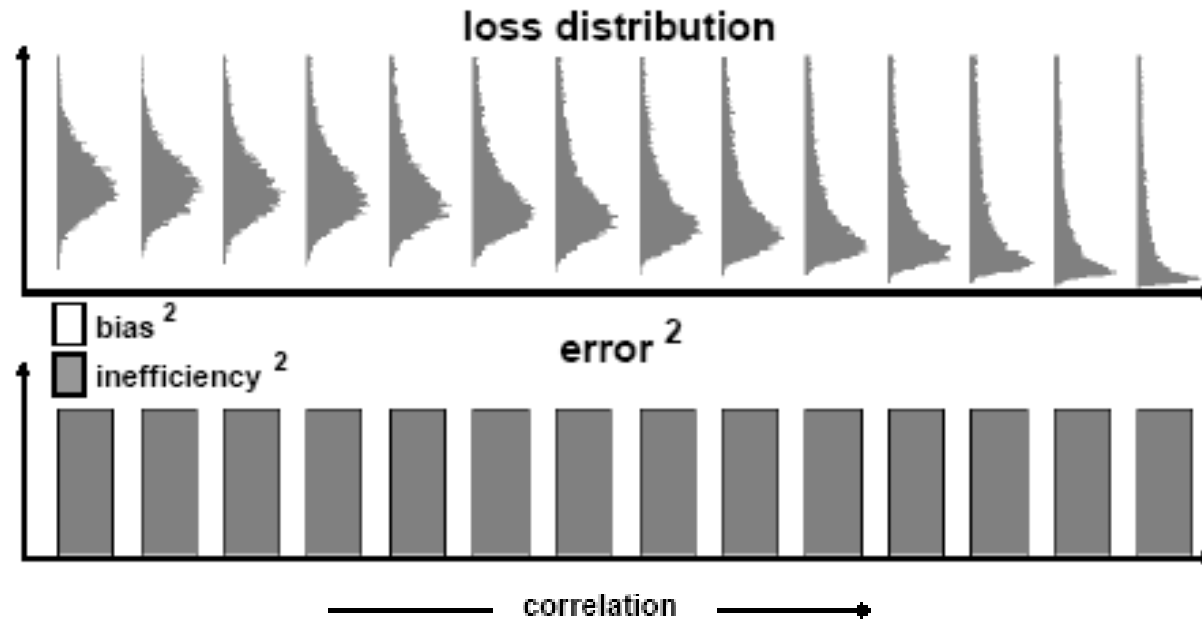


Fig. 4.11.

estimator evaluation
(global)

$$\text{Loss}(\hat{\mu}, \mu) \equiv [\hat{\mu}[I_T] - \mu]' [\hat{\mu}[I_T] - \mu] \quad (4.108)$$

estimator evaluation
(summary)

$$\left\{ \begin{array}{l} \text{Err}^2(\hat{\mu}, \mu) = \frac{1}{T} \text{tr}(\Sigma) \quad (4.109) \\ \text{Inef}^2(\hat{\mu}) = \frac{1}{T} \text{tr}(\Sigma) \quad (4.110) \\ \text{Bias}^2(\hat{\mu}, \mu) = 0. \quad (4.111) \end{array} \right.$$

MAXIMUM LIKELIHOOD ESTIMATORS – NORMAL (SCATTER)

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normal assumption $\mathbf{X} \sim N(\boldsymbol{\mu}, \boldsymbol{\Sigma})$ (4.95)

estimator definition $\hat{\boldsymbol{\Sigma}}[i_T] = \frac{1}{T} \sum_{t=1}^T (\mathbf{x}_t - \hat{\boldsymbol{\mu}}) (\mathbf{x}_t - \hat{\boldsymbol{\mu}})'$ (4.99)

MAXIMUM LIKELIHOOD ESTIMATORS – NORMAL (SCATTER)

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estimator distribution
(replicability) $T\hat{\boldsymbol{\Sigma}}[I_T] \sim W(T-1, \boldsymbol{\Sigma})$ (4.103)

MAXIMUM LIKELIHOOD ESTIMATORS – NORMAL (SCATTER)

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Generalized t-tests...

Generalized p-values...

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Risk and Asset Allocation - Springer – symmys.com

normal assumption $\mathbf{X} \sim N(\boldsymbol{\mu}, \boldsymbol{\Sigma})$ (4.95)

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estimator distribution
(replicability) $T\hat{\boldsymbol{\Sigma}}[I_T] \sim W(T-1, \boldsymbol{\Sigma})$ (4.103)

estimator evaluation
(global) $\text{Loss}(\hat{\boldsymbol{\Sigma}}, \boldsymbol{\Sigma}) \equiv \text{tr} \left[\left(\hat{\boldsymbol{\Sigma}}[I_T] - \boldsymbol{\Sigma} \right)^2 \right]$ (4.118)

estimator evaluation
(summary) $\left\{ \begin{array}{l} \text{Err}^2(\hat{\boldsymbol{\Sigma}}, \boldsymbol{\Sigma}) = \frac{1}{T} \left[\text{tr}(\boldsymbol{\Sigma}^2) + \left(1 - \frac{1}{T}\right) [\text{tr}(\boldsymbol{\Sigma})]^2 \right] \quad (4.119) \\ \text{Inef}^2(\hat{\boldsymbol{\Sigma}}) = \frac{1}{T} \left(1 - \frac{1}{T}\right) \left[\text{tr}(\boldsymbol{\Sigma}^2) + [\text{tr}(\boldsymbol{\Sigma})]^2 \right] \quad (4.120) \\ \text{Bias}^2(\hat{\boldsymbol{\Sigma}}, \boldsymbol{\Sigma}) = \frac{1}{T^2} \text{tr}(\boldsymbol{\Sigma}^2) \quad (4.121) \end{array} \right.$

MAXIMUM LIKELIHOOD ESTIMATORS – NORMAL (SCATTER)

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Fig. 4.12

estimator evaluation
(global)

$$\text{Loss} \left(\hat{\Sigma}, \Sigma \right) \equiv \text{tr} \left[\left(\hat{\Sigma} [I_T] - \Sigma \right)^2 \right] \quad (4.118)$$

estimator evaluation
(summary)

$$\text{Err}^2 \left(\hat{\Sigma}, \Sigma \right) = \frac{1}{T} \left[\text{tr} \left(\Sigma^2 \right) + \left(1 - \frac{1}{T} \right) [\text{tr} \left(\Sigma \right)]^2 \right] \quad (4.119)$$

$$\text{Inef}^2 \left(\hat{\Sigma} \right) = \frac{1}{T} \left(1 - \frac{1}{T} \right) \left[\text{tr} \left(\Sigma^2 \right) + [\text{tr} \left(\Sigma \right)]^2 \right] \quad (4.120)$$

$$\text{Bias}^2 \left(\hat{\Sigma}, \Sigma \right) = \frac{1}{T^2} \text{tr} \left(\Sigma^2 \right) \quad (4.121)$$

MAXIMUM LIKELIHOOD ESTIMATORS – NORMAL (LOADINGS)

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normal assumption $\mathbf{X}|\mathbf{f} = \mathbf{B}\mathbf{f} + \mathbf{U}|\mathbf{f} \quad (4.88) \quad \mathbf{U}_t|\mathbf{f}_t \sim N(\mathbf{0}, \Sigma) \quad (4.123)$

estimator definition $\hat{\mathbf{B}}[i_T] = \hat{\Sigma}_{XF}[i_T] \hat{\Sigma}_F^{-1}[i_T] \quad (4.126)$

$$\hat{\Sigma}_{XF}[i_T] \equiv \frac{1}{T} \sum_{t=1}^T \mathbf{x}_t \mathbf{f}_t', \quad \hat{\Sigma}_F[i_T] \equiv \frac{1}{T} \sum_{t=1}^T \mathbf{f}_t \mathbf{f}_t'. \quad (4.127)$$

$$\hat{\Sigma}[i_T] = \frac{1}{T} \sum_{t=1}^T \left(\mathbf{x}_t - \hat{\mathbf{B}}[i_T] \mathbf{f}_t \right) \left(\mathbf{x}_t - \hat{\mathbf{B}}[i_T] \mathbf{f}_t \right)' \quad (4.128)$$

estimator distribution
(replicability) $\hat{\mathbf{B}}[I_T|\mathbf{f}_1, \dots, \mathbf{f}_T] \sim N\left(\mathbf{B}, \frac{\Sigma}{T}, \hat{\Sigma}_F^{-1}\right) \quad (4.129)$

$$T\hat{\Sigma}[I_T|\mathbf{f}_1, \dots, \mathbf{f}_T] \sim W(T - K, \Sigma) \quad (4.130)$$



Generalized t-tests, generalized p-values, estimator evaluation