Attilio Meucci

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

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

information
$$i_T \mapsto \text{number } \widehat{\mathbf{G}}$$
 (4.9)
$$i_T \equiv \{\mathbf{x}_1, \dots, \mathbf{x}_T\}$$
 (4.8)

(4.13)
$$\widehat{\mathbf{G}}[i_T] \approx \mathbf{G}[f_{\mathbf{X}}]$$

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

$$G[f_X] \equiv \int_{-\infty}^{+\infty} x f_X(x) \, dx \tag{4.7}$$

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$$\widehat{G}\left[i_{T}\right] \equiv \frac{1}{T} \sum_{t=1}^{T} x_{t}. \quad (4.10)$$

$$\widehat{G}\left[i_{T}\right] \equiv x_{1}x_{T} \quad ^{(4.11)}$$

$$\widehat{G}[i_T] \equiv 3.$$
 (4.12)

$$\widehat{\mathbf{G}}\left[i_{T}
ight]pprox\mathbf{G}\left[f_{\mathbf{X}}
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$$i_T \equiv \{\mathbf{x}_1, \dots, \mathbf{x}_T\} \mapsto I_T \equiv \{\mathbf{X}_1, \dots, \mathbf{X}_T\}$$
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$$\widehat{\mathbf{G}}[i_T] \mapsto \widehat{\mathbf{G}}[I_T]$$
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$$\operatorname{Loss}\left(\widehat{\mathbf{G}}, \mathbf{G}\right) \equiv \left\|\widehat{\mathbf{G}}\left[I_{T}\right] - \mathbf{G}\left[f_{X}\right]\right\|^{2} \left| (4.19)\right|$$

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$$G[f_X] = \mu$$
. (4.18)

$$X_t \sim \mathrm{N}\left(\mu, \sigma^2\right)$$
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$$G\left[f_{X}\right] = \mu. \tag{4.18}$$

$$X_{t} \sim \mathrm{N}\left(\mu, \sigma^{2}\right) \tag{4.16}$$

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$$\operatorname{Loss}\left(\widehat{G},G\right) \!\! \sim \operatorname{Ga}\left(1,\frac{\sigma^2}{T}\right)_{(4.22)}$$

Loss
$$\left(\widehat{\mathbf{G}}, \mathbf{G}\right) \equiv \left\|\widehat{\mathbf{G}}\left[I_T\right] - \mathbf{G}\left[f_X\right]\right\|^2$$
 (4.19)

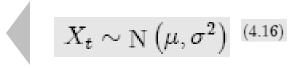
$$\operatorname{Err}\left(\widehat{\mathbf{G}},\mathbf{G}\right) \equiv \sqrt{\operatorname{E}\left\{\left\|\widehat{\mathbf{G}}\left(I_{T}\right) - \mathbf{G}\left[f_{\mathbf{X}}\right]\right\|^{2}\right\}}.$$
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Inef²
$$\left[\widehat{\mathbf{G}}\right] \equiv \mathrm{E}\left\{ \left\|\widehat{\mathbf{G}}\left[I_{T}\right] - \mathrm{E}\left\{\widehat{\mathbf{G}}\left[I_{T}\right]\right\}\right\|^{2} \right\}$$
 (4.26)

$$\operatorname{Bias}^{2}\left[\widehat{\mathbf{G}},\mathbf{G}\right] \equiv \left\| \operatorname{E}\left\{\widehat{\mathbf{G}}\left[I_{T}\right]\right\} - \mathbf{G}\left[f_{\mathbf{X}}\right] \right\|^{2} \tag{4.25}$$

pdf of the estimator $\hat{G}[I_T]$ Fig. 4.2. Estimation: replicability, bias and inefficiency estimation $\hat{G}[i_T]$ in one scenario i_T in one scenario i_T g_S inefficiency (dispersion)

$$\operatorname{Inef}^{2}\left[\widehat{\mathbf{G}}\right] \equiv \operatorname{E}\left\{\left\|\widehat{\mathbf{G}}\left[I_{T}\right] - \operatorname{E}\left\{\widehat{\mathbf{G}}\left[I_{T}\right]\right\}\right\|^{2}\right\} \quad (4.26)$$

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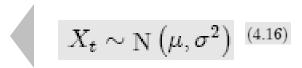
Bias
$$\left[\widehat{G}, G\right] = 0$$
. (4.28)

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$$\operatorname{Err}^{2}\left[\widehat{\mathbf{G}}, \mathbf{G}\right] = \operatorname{Bias}^{2}\left[\widehat{\mathbf{G}}, \mathbf{G}\right] + \operatorname{Inef}^{2}\left[\widehat{\mathbf{G}}\right]$$
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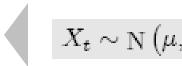
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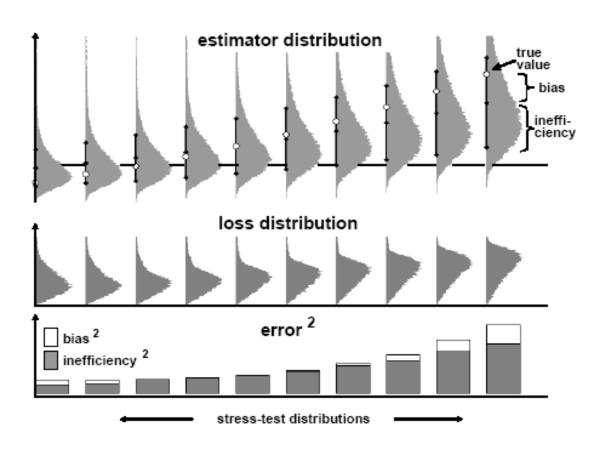
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Loss
$$(\widehat{\mathbf{G}}, \mathbf{G}) \equiv \|\widehat{\mathbf{G}}[I_T] - \mathbf{G}[f_X]\|^2$$
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NON-PARAMETRIC ESTIMATORS Risk and Asset Allocation - Springer - symmys.com

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$$\lim_{T \to \infty} F_{i_T}(\mathbf{x}) = F_{\mathbf{X}}(\mathbf{x}) \tag{4.34}$$

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$$\int_{-\infty}^{q_{\mathfrak{p}}[f_X]} f_X(x) dx \equiv p, \quad (4.38)$$

$$\widehat{q}_{p}\left[i_{T}\right] \equiv x_{\left\lceil pT\right\rceil:T} \quad (4.39)$$

NON-PARAMETRIC ESTIMATORS – ORDINARY LEAST SQUARES

$$\mathbf{G}\left[f_{\mathbf{X}}\right] \equiv$$
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$$\mathbf{X} = \mathbf{BF} + \mathbf{U}. \tag{4.50}$$

$$\mathbf{B}_r \equiv \mathbf{E} \left\{ \mathbf{X} \mathbf{F}' \right\} \mathbf{E} \left\{ \mathbf{F} \mathbf{F}' \right\}^{-1} \tag{3.121}$$

$$\widehat{\mathbf{B}}\left[i_{T}\right] \equiv \left(\sum_{t} \mathbf{x}_{t} \mathbf{f}_{t}^{\prime}\right) \left(\sum_{t} \mathbf{f}_{t} \mathbf{f}_{t}^{\prime}\right)^{-1}$$

$$i_{T} \equiv \left\{\mathbf{x}_{1}, \mathbf{f}_{1}, \dots, \mathbf{x}_{T}, \mathbf{f}_{T}\right\} (4.51)$$

$$(4.52)$$

NON-PARAMETRIC ESTIMATORS – ORDINARY LEAST SQUARES

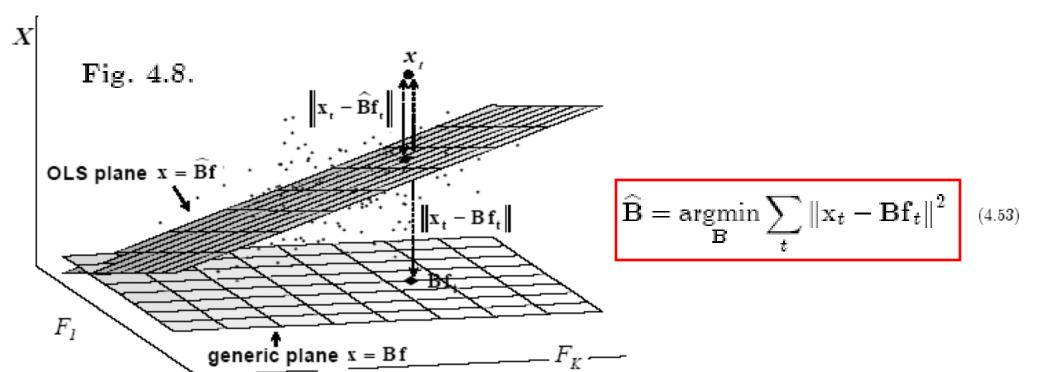
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NON-PARAMETRIC ESTIMATORS – SAMPLE MEAN/COVARIANCE

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

$$Cov\left\{ \mathbf{X}\right\} \tag{2.67}$$

$$\mathbf{information} \ i_T \mapsto \widehat{\mathbf{G}} \left[i_T \right] \equiv \mathbf{G} \left[f_{i_T} \right] \tag{4.36}$$

$$\widehat{\mathbf{E}}\left[i_{T}\right] = \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_{t} \qquad (4.41)$$

$$\widehat{\mathbf{Cov}}\left[i_{T}\right] = \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\mathbf{E}}\left[i_{T}\right]\right) \left(\mathbf{x}_{t} - \widehat{\mathbf{E}}\left[i_{T}\right]\right)' \qquad (4.42)$$

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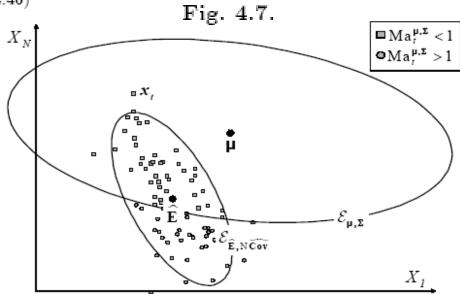
$$E\{X\}$$
 (2.54)

$$\widehat{\mathbf{E}}\left[i_{T}\right] = \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_{t} \tag{4.41}$$

$$\widehat{\text{Cov}}\left[i_{T}\right] = \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\mathbf{E}}\left[i_{T}\right]\right) \left(\mathbf{x}_{t} - \widehat{\mathbf{E}}\left[i_{T}\right]\right)'$$
(4.42)

$$\mathcal{E}_{\mu,\Sigma} \equiv \left\{ \mathbf{x} \in \mathbb{R}^{N} \text{ such that } (\mathbf{x} - \mu)' \Sigma^{-1} (\mathbf{x} - \mu) \leq 1 \right\}$$
 (4.45)

$$\operatorname{Ma}_{t}^{\mu,\Sigma} \equiv \operatorname{Ma}(\mathbf{x}_{t}, \mu, \Sigma) \equiv \sqrt{(\mathbf{x}_{t} - \mu)' \Sigma^{-1} (\mathbf{x}_{t} - \mu)}$$
 (4.46)



NON-PARAMETRIC ESTIMATORS – SAMPLE MEAN/COVARIANCE

$$\mathbf{G}[f_{\mathbf{X}}] \equiv \text{"unknown truth"}$$
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information
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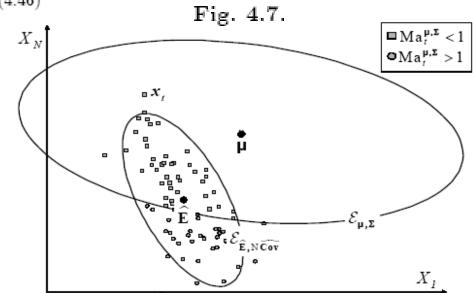
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$$\mathcal{E}_{\mu,\Sigma} \equiv \left\{\mathbf{x} \in \mathbb{R}^{N} \text{ such that } (\mathbf{x} - \mu)' \, \Sigma^{-1} \, (\mathbf{x} - \mu) \leq 1 \right\}_{(4.45)}$$

$$\operatorname{Ma}_{t}^{\mu,\Sigma} \equiv \operatorname{Ma}\left(\mathbf{x}_{t}, \mu, \Sigma\right) \equiv \sqrt{\left(\mathbf{x}_{t} - \mu\right)' \Sigma^{-1} \left(\mathbf{x}_{t} - \mu\right)}.$$
 (4.46)

$$\overline{r^2}(\mu, \Sigma) \equiv \frac{1}{T} \sum_{t=1}^{T} \left(\operatorname{Ma}_t^{\mu, \Sigma} \right)^2$$
 (4.47)

$$\begin{split} \left(\widehat{\mathbf{E}}, N\widehat{\mathrm{Cov}}\right) &= \operatorname*{argmin}_{(\mu, \Sigma) \in \mathcal{C}} \left[\operatorname{Vol} \left\{ \mathcal{E}_{\mu, \Sigma} \right\} \right] \\ &\overline{r^2} \left(\mu, \Sigma \right) \equiv 1 \end{split} \tag{4.48}$$



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$$i_T \mapsto \widehat{\mathbf{G}}\left[i_T\right] \equiv \mathbf{G}\left[f_{i_T}\right]$$

$$f_{i_T}\left(\mathbf{x}\right) \equiv \frac{1}{T} \sum_{t=1}^{T} \delta^{(\mathbf{x}_t)}\left(\mathbf{x}\right) \quad (4.35)$$

$$\mathbf{information} \ i_{T} \mapsto \widehat{\mathbf{G}} \left[i_{T} \right] \equiv \mathbf{G} \left[f_{i_{T};\epsilon} \right]$$

$$f_{i_{T}} \mapsto f_{i_{T};\epsilon} \equiv \frac{1}{T} \sum_{t=1}^{T} \frac{1}{(2\pi)^{\frac{N}{2}} \epsilon^{N}} e^{-\frac{1}{2\epsilon^{2}} (\mathbf{x} - \mathbf{x}_{t})' (\mathbf{x} - \mathbf{x}_{t})}.$$

$$(4.56)$$

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Attilio Meucci

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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

$$\mathbf{G}\left[f_{\mathbf{X}}\right] \equiv \text{"unknown truth"}$$
 (4.6) information $i_T \mapsto \text{number } \widehat{\mathbf{G}}$ (4.8) $i_T \equiv \{\mathbf{x}_1, \dots, \mathbf{x}_T\}$ (4.8)

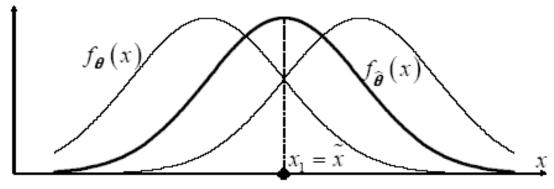
Risk and Asset Allocation - Springer - symmys.com

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

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

$$f_{\boldsymbol{\theta}}\left(i_{T}\right) \equiv f_{\boldsymbol{\theta}}\left(\mathbf{x}_{1}\right) \cdots f_{\boldsymbol{\theta}}\left(\mathbf{x}_{T}\right)$$
. (4.65)

$$\widehat{\boldsymbol{\theta}}\left[i_{T}\right] \equiv \operatorname*{argmax}_{\boldsymbol{\theta} \in \boldsymbol{\Theta}} f_{\boldsymbol{\theta}}\left(i_{T}\right) \quad (4.66)$$



observation = mode of MLE distribution

Fig. 4.10

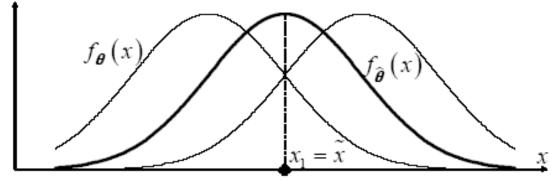
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$$G[f_X] \equiv \text{"unknown truth"}$$
 (4.6)

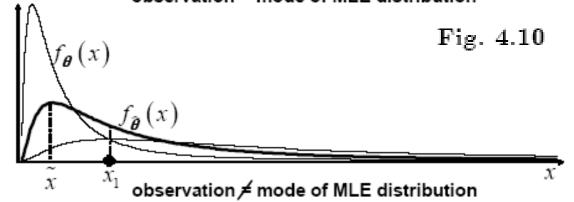
information $i_T \mapsto \text{number } \widehat{\mathbf{G}}$ (4.9)

$$f_{\boldsymbol{\theta}}\left(i_{T}\right) \equiv f_{\boldsymbol{\theta}}\left(\mathbf{x}_{1}\right) \cdots f_{\boldsymbol{\theta}}\left(\mathbf{x}_{T}\right)$$
. (4.65)

$$\widehat{\boldsymbol{\theta}}\left[i_{T}\right] \equiv \underset{\boldsymbol{\theta} \in \boldsymbol{\Theta}}{\operatorname{argmax}} \, f_{\boldsymbol{\theta}}\left(i_{T}\right) \quad (4.66)$$



observation = mode of MLE distribution



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

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

$$f_{\boldsymbol{\theta}}\left(i_{T}\right) \equiv f_{\boldsymbol{\theta}}\left(\mathbf{x}_{1}\right) \cdots f_{\boldsymbol{\theta}}\left(\mathbf{x}_{T}\right)$$
. (4.65)

$$\widehat{\boldsymbol{\theta}}\left[i_{T}\right] \equiv \operatorname*{argmax}_{\boldsymbol{\theta} \in \boldsymbol{\Theta}} f_{\boldsymbol{\theta}}\left(i_{T}\right)$$
 (4.66)

•
$$\widehat{g(\boldsymbol{\theta})} = g(\widehat{\boldsymbol{\theta}})$$
 (4.70)

•
$$\widehat{\boldsymbol{\theta}}\left[i_{T}\right] \sim \mathrm{N}\left(\boldsymbol{\theta}, \frac{\Gamma}{T}\right)$$
 (4.71)
$$\Gamma \equiv \mathrm{Cov}\left\{\frac{\partial \ln\left(f_{\boldsymbol{\theta}}\left(\mathbf{X}\right)\right)}{\partial \boldsymbol{\theta}}\right\}$$
 (4.72)

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$$\mathbf{X} \sim \mathrm{El}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}, g\right)$$
 (4.73)

$$f\theta\left(\mathbf{x}\right) \equiv \frac{1}{\sqrt{|\Sigma|}} g\left(\mathrm{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right)\right)^{-(4.74)} \qquad \mathrm{Ma}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \equiv \sqrt{\left(\mathbf{x} - \boldsymbol{\mu}\right)' \boldsymbol{\Sigma}^{-1} \left(\mathbf{x} - \boldsymbol{\mu}\right)}^{-(4.75)}$$

MAXIMUM LIKELIHOOD ESTIMATORS – ELLIPTICAL

$$\mathbf{X} \sim \mathrm{El}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}, g\right)$$
 (4.73)

$$f\theta\left(\mathbf{x}\right) \equiv \frac{1}{\sqrt{|\Sigma|}} g\left(\mathrm{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right)\right)^{-(4.74)} \qquad \qquad \mathrm{Ma}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \equiv \sqrt{\left(\mathbf{x} - \boldsymbol{\mu}\right)' \boldsymbol{\Sigma}^{-1} \left(\mathbf{x} - \boldsymbol{\mu}\right)}^{-(4.75)}$$

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

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

$$w_t \equiv w \left(\operatorname{Ma}^2 \left(\mathbf{x}_t, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}} \right) \right) \qquad (4.80)$$

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

MAXIMUM LIKELIHOOD ESTIMATORS – ELLIPTICAL

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$$\mathbf{X} \sim \mathrm{El}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}, g\right)$$
 (4.73)

$$f\theta\left(\mathbf{x}\right) \equiv \frac{1}{\sqrt{|\Sigma|}} g\left(\mathrm{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right)\right)^{-(4.74)} \qquad \qquad \mathrm{Ma}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \equiv \sqrt{\left(\mathbf{x} - \boldsymbol{\mu}\right)' \boldsymbol{\Sigma}^{-1} \left(\mathbf{x} - \boldsymbol{\mu}\right)}^{-(4.75)}$$

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

$$\widehat{\Sigma} = \frac{1}{T} \sum_{t=1}^{T} w_t \left(\mathbf{x}_t - \widehat{\mu} \right) \left(\mathbf{x}_t - \widehat{\mu} \right)' \qquad (4.82)$$

$$w_t \equiv w \left(\mathbf{Ma}^2 \left(\mathbf{x}_t, \widehat{\mu}, \widehat{\Sigma} \right) \right) \qquad (4.80)$$

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

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

$$w(z) \equiv 1 \qquad ^{(4.97)}$$

MAXIMUM LIKELIHOOD ESTIMATORS – ELLIPTICAL

$$\mathbf{X} \sim \mathrm{El}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}, g\right)$$
 (4.73)

$$f\theta\left(\mathbf{x}\right) \equiv \frac{1}{\sqrt{|\Sigma|}} g\left(\mathrm{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right)\right)^{-(4.74)} \qquad \qquad \mathrm{Ma}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \equiv \sqrt{\left(\mathbf{x} - \boldsymbol{\mu}\right)' \boldsymbol{\Sigma}^{-1} \left(\mathbf{x} - \boldsymbol{\mu}\right)}^{-(4.75)}$$

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

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$$w_{t} \equiv w \left(\operatorname{Ma}^{2} \left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}} \right) \right) \quad ^{(4.80)}$$

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

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

$$w\left(z\right) \equiv 1 \qquad ^{(4.97)}$$

$$g^{\operatorname{Ca}}\left(z\right) = \frac{\Gamma\left(\frac{1+N}{2}\right)}{\Gamma\left(\frac{1}{2}\right)\left(\pi\right)^{\frac{N}{2}}}\left(1+z\right)^{-\frac{1+N}{2}} \tag{4.83}$$

$$w_{t} = \frac{N+1}{1 + \operatorname{Ma}^{2}\left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)}$$
(4.84)

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

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

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

estimator definition
$$\widehat{\mu}[i_T] = \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_{t}$$
 (4.98)

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

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

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

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

Generalized t-tests...

Generalized p-values...

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

estimator definition
$$\widehat{\mu}[i_T] = \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_{t}$$
 (4.98)

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

estimator evaluation (global)

Loss
$$(\widehat{\boldsymbol{\mu}}, \boldsymbol{\mu}) \equiv \left[\widehat{\boldsymbol{\mu}}\left[I_T\right] - \boldsymbol{\mu}\right]' \left[\widehat{\boldsymbol{\mu}}\left[I_T\right] - \boldsymbol{\mu}\right]$$
 (4.108)

estimator evaluation (summary)

$$\begin{cases} &\operatorname{Err}^2\left(\widehat{\boldsymbol{\mu}},\boldsymbol{\mu}\right) = \frac{1}{T}\operatorname{tr}\left(\boldsymbol{\Sigma}\right) \quad \text{\tiny (4.109)} \\ &\operatorname{Inef}^2\left(\widehat{\boldsymbol{\mu}}\right) = \frac{1}{T}\operatorname{tr}\left(\boldsymbol{\Sigma}\right) \quad \text{\tiny (4.110)} \\ &\operatorname{Bias}^2\left(\widehat{\boldsymbol{\mu}},\boldsymbol{\mu}\right) = 0. \quad \text{\tiny (4.111)} \end{cases}$$

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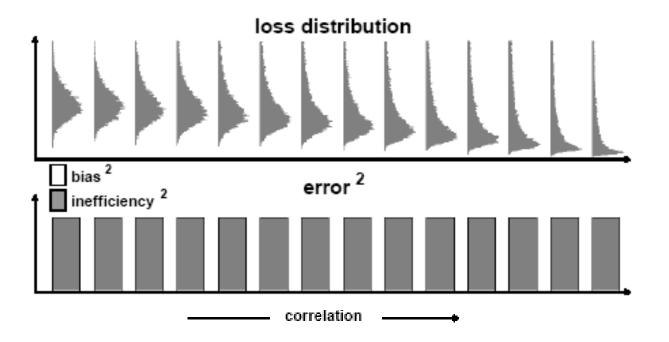


Fig. 4.11.

estimator evaluation (global)

Loss
$$(\widehat{\boldsymbol{\mu}}, \boldsymbol{\mu}) \equiv \left[\widehat{\boldsymbol{\mu}}\left[I_T\right] - \boldsymbol{\mu}\right]' \left[\widehat{\boldsymbol{\mu}}\left[I_T\right] - \boldsymbol{\mu}\right]$$
 (4.108)

estimator evaluation (summary)

$$\operatorname{Err}^2\left(\widehat{oldsymbol{\mu}},oldsymbol{\mu}
ight) = rac{1}{T}\operatorname{tr}\left(\Sigma
ight)$$
. (4.109)
 $\operatorname{Inef}^2\left(\widehat{oldsymbol{\mu}}
ight) = rac{1}{T}\operatorname{tr}\left(\Sigma
ight)$ (4.110)
 $\operatorname{Bias}^2\left(\widehat{oldsymbol{\mu}},oldsymbol{\mu}
ight) = 0$. (4.111)

normal assumption
$$\mathbf{X} \sim \mathrm{N}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}\right)$$
 (4.95) estimator definition $\widehat{\boldsymbol{\Sigma}}\left[i_T\right] = \frac{1}{T} \sum_{t=1}^T \left(\mathbf{x}_t - \widehat{\boldsymbol{\mu}}\right) \left(\mathbf{x}_t - \widehat{\boldsymbol{\mu}}\right)'$ (4.99)

normal assumption
$$\mathbf{X} \sim \mathrm{N}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}\right)$$
 (4.95) estimator definition $\widehat{\boldsymbol{\Sigma}}\left[i_T\right] = \frac{1}{T}\sum_{t=1}^T \left(\mathbf{x}_t - \widehat{\boldsymbol{\mu}}\right) \left(\mathbf{x}_t - \widehat{\boldsymbol{\mu}}\right)'$ (4.99)

$$T\widehat{\Sigma}\left[I_{T}\right]\sim\mathrm{W}\left(T-1,\Sigma\right)$$
 (4.103)

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

$$\widehat{\Sigma}\left[i_{T}\right] = \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\right) \left(\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\right)' \quad (4.99)$$

$$T\widehat{\Sigma}\left[I_{T}\right]\sim\mathrm{W}\left(T-1,\Sigma\right)$$
 (4.103)

Generalized t-tests...

Generalized p-values...

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

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

estimator distribution (replicability)

$$\widehat{\boldsymbol{\Sigma}}\left[I_T\right] \sim \mathbf{W}\left(T-1,\boldsymbol{\Sigma}\right) \quad \text{(4.103)}$$
estimator evaluation (global)

$$\widehat{\boldsymbol{\Sigma}}\left[\widehat{\boldsymbol{\Sigma}},\boldsymbol{\Sigma}\right] \equiv \operatorname{tr}\left[\left(\widehat{\boldsymbol{\Sigma}}\left[I_T\right] - \boldsymbol{\Sigma}\right)^2\right] \quad \text{(4.118)}$$
(global)

$$\operatorname{Err}^2\left(\widehat{\boldsymbol{\Sigma}},\boldsymbol{\Sigma}\right) = \frac{1}{T}\left[\operatorname{tr}\left(\boldsymbol{\Sigma}^2\right) + \left(1 - \frac{1}{T}\right)\left[\operatorname{tr}\left(\boldsymbol{\Sigma}\right)\right]^2\right] \quad \text{(4.119)}$$
estimator evaluation (summary)

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

MAXIMUM LIKELIHOOD ESTIMATORS - NORMAL (SCATTER)

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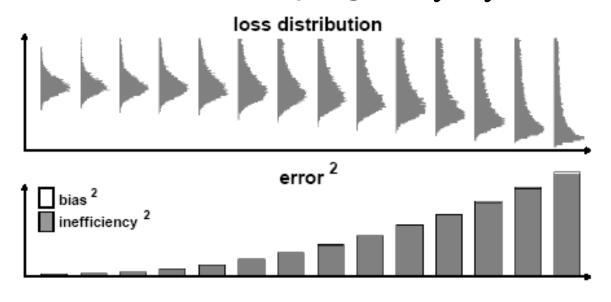


Fig. 4.12

estimator evaluation (global)

$$\operatorname{Loss}\left(\widehat{\Sigma}, \Sigma\right) \equiv \operatorname{tr}\left[\left(\widehat{\Sigma}\left[I_{T}\right] - \Sigma\right)^{2}\right] \tag{4.118}$$

$$\left(\operatorname{Err}^{2}\left(\widehat{\Sigma}, \Sigma\right) = \frac{1}{T}\left[\operatorname{tr}\left(\Sigma^{2}\right) + \left(1 - \frac{1}{T}\right)\left[\operatorname{tr}\left(\Sigma\right)\right]^{2}\right] \tag{4.119}$$

estimator evaluation (summary)

$$\operatorname{Inef}^{2}\left(\widehat{\Sigma}\right) = \frac{1}{T}\left(1 - \frac{1}{T}\right)\left[\operatorname{tr}\left(\Sigma^{2}\right) + \left[\operatorname{tr}\left(\Sigma\right)\right]^{2}\right] \tag{4.120}$$

$$\operatorname{Bias}^{2}\left(\widehat{\Sigma},\Sigma\right)=rac{1}{T^{2}}\operatorname{tr}\left(\Sigma^{2}\right)$$
 (4.121)

normal assumption

$$X|f = Bf + U|f$$
 (4.88)

$$\mathbf{U}_t | \mathbf{f}_t \sim \mathbf{N} (\mathbf{0}, \boldsymbol{\Sigma})$$
 (4.123)

estimator definition

$$\widehat{\mathbf{B}}\left[i_{T}\right] = \widehat{\boldsymbol{\Sigma}}_{XF}\left[i_{T}\right] \widehat{\boldsymbol{\Sigma}}_{F}^{-1}\left[i_{T}\right] \qquad (4.126)$$

$$\widehat{\boldsymbol{\Sigma}}_{XF}\left[i_{T}\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_{t} \mathbf{f}_{t}', \quad \widehat{\boldsymbol{\Sigma}}_{F}\left[i_{T}\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{f}_{t} \mathbf{f}_{t}'. \qquad (4.127)$$

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

estimator distribution (replicability)

$$\widehat{\mathbf{B}}[I_T|\mathbf{f}_1,\ldots,\mathbf{f}_T] \sim N\left(\mathbf{B},\frac{\Sigma}{T},\widehat{\Sigma}_F^{-1}\right)$$
 (4.129)

$$T\widehat{\Sigma}\left[I_T|\mathbf{f}_1,\ldots,\mathbf{f}_T\right] \sim \mathrm{W}\left(T-K,\Sigma\right)$$
 (4.130)

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

Attilio Meucci

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

$$\mathbf{X}_{t} \sim \mathbf{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$$
 (4.132)

$$\widehat{\boldsymbol{\mu}}\left[i_T\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_t. \quad (4.135)$$

$$\widehat{\mu}\left[I_T\right] \sim \mathrm{N}\left(\mu, rac{\Sigma}{T}
ight)$$
 (4.102)

$$\operatorname{Err}^{2}\left(\widehat{\boldsymbol{\mu}},\boldsymbol{\mu}\right) = \frac{1}{T}\operatorname{tr}\left(\boldsymbol{\Sigma}\right)$$
 (4.136)

$$\mathbf{X}_t \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$$
 (4.132)

$$\widehat{\boldsymbol{\mu}}\left[i_T\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_t. \quad (4.135)$$

$$\operatorname{Err}^{2}\left(\widehat{\boldsymbol{\mu}}^{S},\boldsymbol{\mu}\right)<\frac{1}{T}\operatorname{tr}\left(\boldsymbol{\Sigma}\right)$$

$$(4.136)$$

$$\operatorname{Err}^{2}\left(\widehat{\boldsymbol{\mu}},\boldsymbol{\mu}\right)=\frac{1}{T}\operatorname{tr}\left(\boldsymbol{\Sigma}\right)$$

$$\mathbf{X}_t \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$$
 (4.132)

$$\widehat{\boldsymbol{\mu}}\left[i_T\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_t. \quad (4.135)$$

$$\operatorname{Err}^{2}\left(\widehat{\boldsymbol{\mu}}^{S}, \boldsymbol{\mu}\right) < \frac{1}{T}\operatorname{tr}\left(\boldsymbol{\Sigma}\right)$$
 (4.137)

$$\widehat{\boldsymbol{\mu}}^{S} \equiv (1 - \alpha) \, \widehat{\boldsymbol{\mu}} + \alpha \mathbf{b}. \tag{4.138}$$

$$\alpha \equiv \frac{1}{T} \frac{N\overline{\lambda} - 2\lambda_{1}}{(\widehat{\boldsymbol{\mu}} - \mathbf{b})' \, (\widehat{\boldsymbol{\mu}} - \mathbf{b})} \tag{4.139}$$

$$\mathbf{X}_t \sim \mathrm{N}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}\right)$$
 (4.132)

$$\widehat{\boldsymbol{\mu}}\left[i_T\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_t. \quad (4.135)$$

$$\operatorname{Err}^{2}\left(\widehat{\boldsymbol{\mu}}^{S}, \boldsymbol{\mu}\right) < \frac{1}{T}\operatorname{tr}\left(\boldsymbol{\Sigma}\right)$$
 (4.137)

$$\widehat{\boldsymbol{\mu}}^{S} \equiv (1 - \alpha) \, \widehat{\boldsymbol{\mu}} + \alpha \mathbf{b}. \tag{4.138}$$

$$\alpha \equiv \frac{1}{T} \frac{N\overline{\lambda} - 2\lambda_{1}}{(\widehat{\boldsymbol{\mu}} - \mathbf{b})'(\widehat{\boldsymbol{\mu}} - \mathbf{b})} \tag{4.139}$$

$$\mathbf{b} \mapsto \frac{1'\widehat{\boldsymbol{\mu}}}{N} \mathbf{1} \tag{4.141} \qquad \mathbf{b} \mapsto \frac{1'\widehat{\Sigma}^{-1}\widehat{\boldsymbol{\mu}}}{1'\widehat{\Sigma}^{-1}\mathbf{1}} \mathbf{1}. \tag{4.142}$$

SHRINKAGE ESTIMATORS – LOCATION PARAMETER

$$\mathbf{X}_t \sim \mathrm{N}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}\right)$$
 (4.132)

$$\widehat{\boldsymbol{\mu}}\left[i_T\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_t. \quad (4.135)$$

$$\operatorname{Err}^{2}\left(\widehat{\boldsymbol{\mu}}^{S}, \boldsymbol{\mu}\right) < \frac{1}{T}\operatorname{tr}\left(\boldsymbol{\Sigma}\right)$$
 (4.137)

$$\widehat{\boldsymbol{\mu}}^{S} \equiv (1 - \alpha)\,\widehat{\boldsymbol{\mu}} + \alpha \mathbf{b}. \tag{4.138}$$

$$\alpha \equiv \frac{1}{T} \frac{N\overline{\lambda} - 2\lambda_{1}}{(\widehat{\boldsymbol{\mu}} - \mathbf{b})'(\widehat{\boldsymbol{\mu}} - \mathbf{b})} \tag{4.139}$$

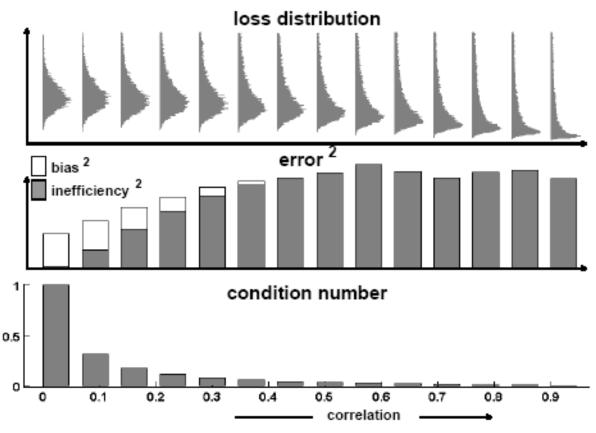


Fig. 4.13. Shrinkage estimator of mean: evaluation $\Sigma \mapsto \widehat{\Sigma}.$

$$\mathbf{b} \mapsto \frac{\mathbf{1}'\widehat{\boldsymbol{\mu}}}{N} \mathbf{1}^{-(4.141)} \qquad \quad \mathbf{b} \mapsto \frac{\mathbf{1}'\widehat{\boldsymbol{\Sigma}}^{-1}\widehat{\boldsymbol{\mu}}}{\mathbf{1}'\widehat{\boldsymbol{\Sigma}}^{-1}\mathbf{1}} \mathbf{1}^{-(4.142)}$$

$$\mathbf{X}_{t} \sim \mathbf{N}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \quad (4.143) \qquad \qquad \widehat{\boldsymbol{\Sigma}}\left[i_{T}\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \left[\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\left[i_{T}\right]\right] \left[\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\left[i_{T}\right]\right]^{\prime}, \quad (4.146)$$

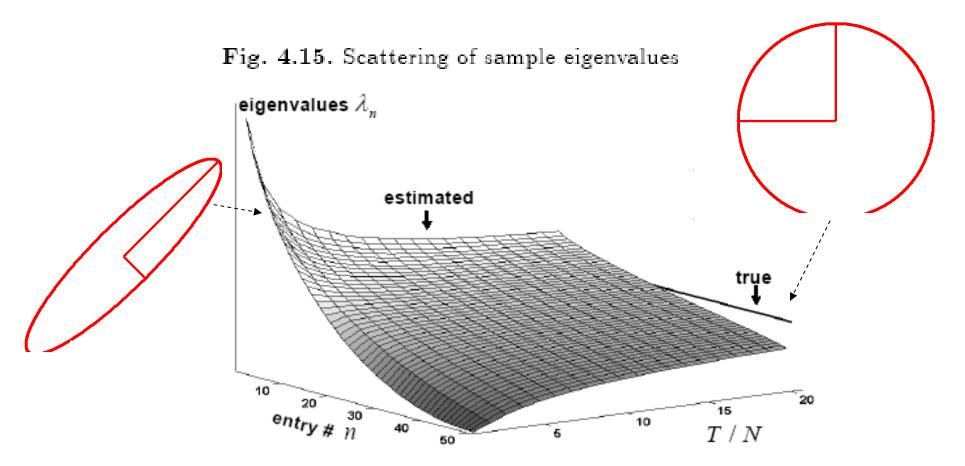
$$T\widehat{\Sigma}\left[I_{T}\right] \sim \mathrm{W}\left(T-1,\Sigma\right)$$
 (4.103)

$$\mathbf{X}_{t} \sim \mathbf{N}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \quad (4.143) \qquad \qquad \widehat{\boldsymbol{\Sigma}}\left[i_{T}\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \left[\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\left[i_{T}\right]\right] \left[\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\left[i_{T}\right]\right]^{\prime}, \quad (4.146)$$

$$\widehat{\mathrm{CN}}\left\{\mathbf{X}\right\} \equiv \frac{\widehat{\lambda}_{N}}{\widehat{\lambda}_{1}} < \frac{\lambda_{N}}{\lambda_{1}} \equiv \mathrm{CN}\left\{\mathbf{X}\right\} \quad (4.156) \qquad \qquad T\widehat{\Sigma}\left[I_{T}\right] \sim \mathrm{W}\left(T-1,\Sigma\right) \quad (4.103)$$

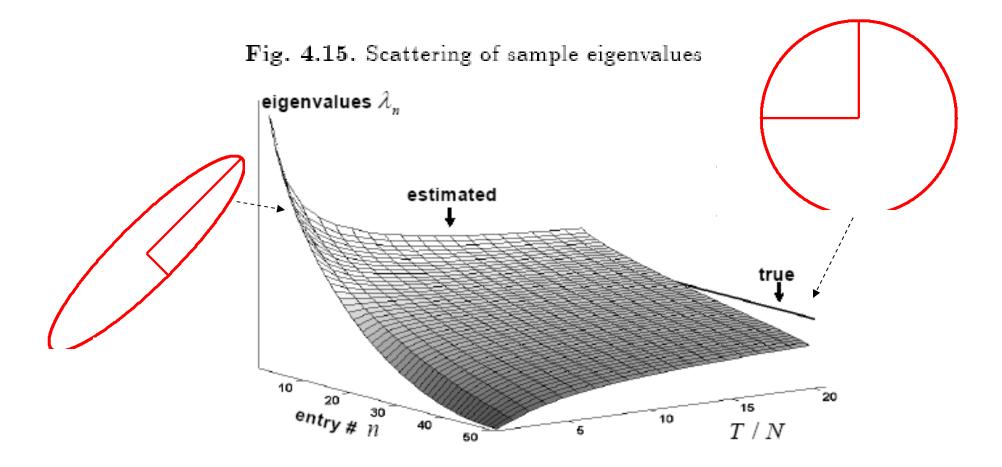
$$\mathbf{X}_{t} \sim \mathbf{N}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \quad (4.143) \qquad \qquad \widehat{\boldsymbol{\Sigma}}\left[i_{T}\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \left[\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\left[i_{T}\right]\right] \left[\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\left[i_{T}\right]\right]^{\prime}, \quad (4.146)$$

$$\widehat{\mathrm{CN}}\left\{\mathbf{X}\right\} \equiv \frac{\widehat{\lambda}_{N}}{\widehat{\lambda}_{1}} < \frac{\lambda_{N}}{\lambda_{1}} \equiv \mathrm{CN}\left\{\mathbf{X}\right\} \quad (4.156)$$



$$\widehat{\mathbf{X}}_{t} \sim \mathbf{N}\left(\boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \quad (4.143) \qquad \qquad \widehat{\boldsymbol{\Sigma}}\left[i_{T}\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \left[\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\left[i_{T}\right]\right] \left[\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\left[i_{T}\right]\right]^{\prime}. \quad (4.146)$$

$$\widehat{\boldsymbol{\Sigma}}^{S} \equiv (1 - \alpha) \,\widehat{\boldsymbol{\Sigma}} + \alpha \,\widehat{\mathbf{C}}. \quad (4.160) \qquad \qquad \widehat{\mathbf{C}} \equiv \frac{\sum_{n=1}^{N} \widehat{\lambda}_{n}}{N} \mathbf{I}. \quad (4.159)$$



$$X_{t} \sim N\left(\mu, \Sigma\right) \quad (4.143) \qquad \qquad \widehat{\Sigma} \left[i_{T}\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \left[x_{t} - \widehat{\mu}\right]$$

$$\widehat{\Sigma}^{S} \equiv (1 - \alpha) \, \widehat{\Sigma} + \alpha \, \widehat{C} \quad (4.160) \qquad \widehat{C} \equiv \frac{\sum_{n=1}^{N} \widehat{\lambda}_{n}}{N} \mathbf{I} \quad (4.159)$$

$$\alpha \equiv \frac{1}{T} \frac{\frac{1}{T} \sum_{t=1}^{T} \operatorname{tr} \left\{ \left(x_{t} x_{t}' - \widehat{\Sigma}\right)^{2} \right\}}{\operatorname{tr} \left\{ \left(\widehat{\Sigma} - \widehat{C}\right)^{2} \right\}} \quad (4.161)$$

$$\lim_{n \to \infty} \sum_{t=1}^{T} \left[x_{t} - \widehat{\mu} \right]$$

$$\lim_{n \to \infty} \sum_{t=1}^{T} \left[x_{t} - \widehat{\mu} \right]$$

$$\lim_{n \to \infty} \sum_{t=1}^{T} \left[x_{t} - \widehat{\mu} \right]$$

$$\lim_{n \to \infty} \sum_{t=1}^{T} \left[x_{t} - \widehat{\mu} \right]$$

$$\widehat{\Sigma}\left[i_{T}\right] \equiv \frac{1}{T} \sum_{t=1}^{T} \left[\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\left[i_{T}\right]\right] \left[\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\left[i_{T}\right]\right]', \quad (4.146)$$

$$\widehat{\mathbf{C}} \equiv \frac{\sum_{n=1}^{N} \widehat{\lambda}_n}{N} \mathbf{I}. \quad (4.159)$$

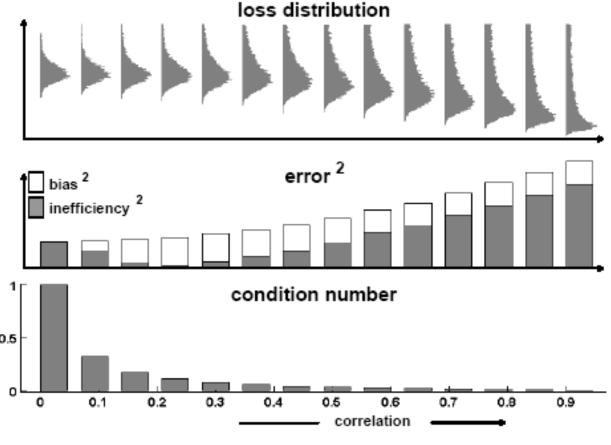


Fig. 4.16. Shrinkage estimator of covariance: evaluation

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Attilio Meucci

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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

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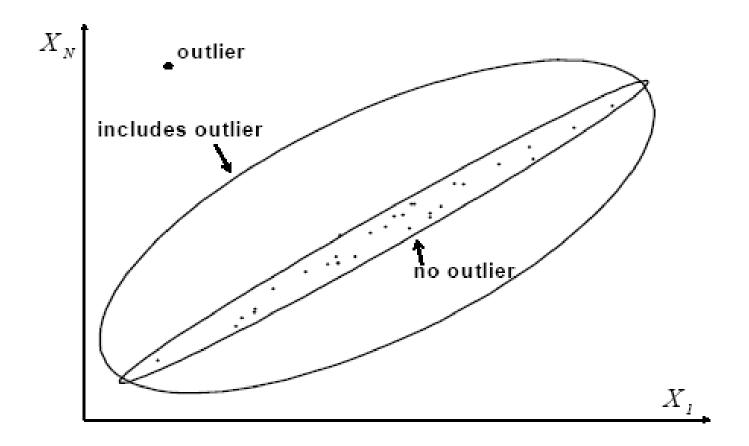


Fig. 4.18. Sample estimators: lack of robustness

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$$SC\left(\mathbf{x}, \widehat{\mathbf{G}}\right) \equiv T\widehat{\mathbf{G}}\left(\mathbf{x}_{1}, \dots, \mathbf{x}_{T}, \mathbf{x}\right) - T\widehat{\mathbf{G}}\left(\mathbf{x}_{1}, \dots, \mathbf{x}_{T}\right)$$
 (4.166)

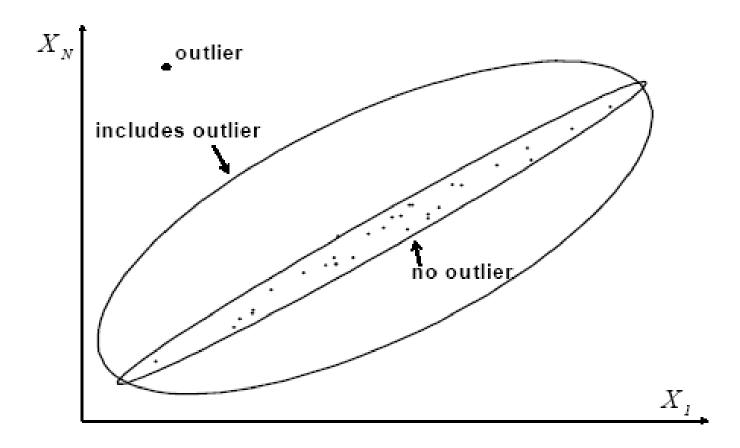


Fig. 4.18. Sample estimators: lack of robustness

$$\begin{array}{c|c}
\bullet & \operatorname{SC}\left(\mathbf{x}, \widehat{\mathbf{G}}\right) \equiv T\widehat{\mathbf{G}}\left(\mathbf{x}_{1}, \dots, \mathbf{x}_{T}, \mathbf{x}\right) - T\widehat{\mathbf{G}}\left(\mathbf{x}_{1}, \dots, \mathbf{x}_{T}\right) \\
f_{i_{T}} \mapsto \left(1 - \epsilon\right) f_{i_{T}} + \epsilon \delta^{(\mathbf{x})}, \quad (4.183) \\
\uparrow & \uparrow \\
f_{i_{T}} \equiv \frac{1}{T} \sum_{t=1}^{T} \delta^{(\mathbf{x}_{t})} \qquad \epsilon \equiv 1/\left(T + 1\right)
\end{array}$$

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$$\begin{array}{c}
\operatorname{SC}\left(\mathbf{x},\widehat{\mathbf{G}}\right) \equiv T\widehat{\mathbf{G}}\left(\mathbf{x}_{1},\ldots,\mathbf{x}_{T},\mathbf{x}\right) - T\widehat{\mathbf{G}}\left(\mathbf{x}_{1},\ldots,\mathbf{x}_{T}\right) \\
f_{i_{T}} \mapsto \left(1 - \epsilon\right) f_{i_{T}} + \epsilon \delta^{(\mathbf{x})}, \quad (4.183) \\
f_{i_{T}} \equiv \frac{1}{T} \sum_{t=1}^{T} \delta^{(\mathbf{x}_{t})} \qquad \epsilon \equiv 1/\left(T + 1\right)
\end{array}$$

•
$$\widehat{\mathbf{G}} \equiv \widetilde{\mathbf{G}} \left[f_{i_T} \right]$$
 (4.167)

non-parametric

$$\widehat{\mathbf{G}} \equiv \mathbf{G} \left[f_{i_T} \right]$$
 (4.169)

maximum likelihood

$$\psi\left(\mathbf{x},\boldsymbol{\theta}\right) \equiv \frac{\partial}{\partial \boldsymbol{\theta}} \ln\left(f_{\boldsymbol{\theta}}\left(\mathbf{x}\right)\right) \tag{4.176}$$

$$\tilde{\boldsymbol{\theta}}\left[h\right]: \int_{\mathbb{R}^{N}} \psi\left(\mathbf{x},\tilde{\boldsymbol{\theta}}\right) h\left(\mathbf{x}\right) d\mathbf{x} \equiv \mathbf{0}. \tag{4.175}$$

$$\widehat{\boldsymbol{\theta}} \equiv \widetilde{\boldsymbol{\theta}} \left[f_{i_T} \right]$$
 (4.177)

• SC
$$\left(\mathbf{x}, \widehat{\mathbf{G}}\right) \equiv T\widehat{\mathbf{G}}\left(\mathbf{x}_{1}, \dots, \mathbf{x}_{T}, \mathbf{x}\right) - T\widehat{\mathbf{G}}\left(\mathbf{x}_{1}, \dots, \mathbf{x}_{T}\right)$$

$$f_{i_{T}} \mapsto \left(1 - \epsilon\right) f_{i_{T}} + \epsilon \delta^{(\mathbf{x})}. \quad (4.183)$$

$$f_{i_{T}} \equiv \frac{1}{T} \sum_{t=1}^{T} \delta^{(\mathbf{x}_{t})} \qquad \epsilon \equiv 1/\left(T + 1\right)$$

•
$$\widehat{\mathbf{G}} \equiv \widetilde{\mathbf{G}} \left[f_{i_T} \right]$$
 (4.167)

$$SC\left(\mathbf{x}, \widehat{\mathbf{G}}\right) \equiv \frac{1 - \epsilon}{\epsilon} \left\{ \widetilde{\mathbf{G}} \left[(1 - \epsilon) f_{i_{T}} + \epsilon \delta^{(\mathbf{x})} \right] - \widetilde{\mathbf{G}} \left[f_{i_{T}} \right] \right\}$$
(4.184)

• SC
$$\left(\mathbf{x}, \widehat{\mathbf{G}}\right) \equiv T\widehat{\mathbf{G}}\left(\mathbf{x}_{1}, \dots, \mathbf{x}_{T}, \mathbf{x}\right) - T\widehat{\mathbf{G}}\left(\mathbf{x}_{1}, \dots, \mathbf{x}_{T}\right)$$

$$f_{i_{T}} \mapsto \left(1 - \epsilon\right) f_{i_{T}} + \epsilon \delta^{(\mathbf{x})}, \quad (4.183)$$

$$f_{i_{T}} \equiv \frac{1}{T} \sum_{t=1}^{T} \delta^{(\mathbf{x}_{t})} \qquad \epsilon \equiv 1/\left(T + 1\right)$$

•
$$\widehat{\mathbf{G}} \equiv \widetilde{\mathbf{G}} \left[f_{i_T} \right]$$
 (4.167)

$$\mathrm{SC}\left(\mathbf{x},\widehat{\mathbf{G}}\right) \equiv \frac{1-\epsilon}{\epsilon} \left\{ \widetilde{\mathbf{G}} \left[\left(1-\epsilon\right) f_{i_{T}} + \epsilon \delta^{(\mathbf{x})} \right] - \widetilde{\mathbf{G}} \left[f_{i_{T}} \right] \right\} \tag{4.184}$$

IF
$$\left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{G}}\right) \equiv \lim_{\epsilon \to 0} \frac{1}{\epsilon} \left(\widetilde{\mathbf{G}} \left[(1 - \epsilon) f_{\mathbf{X}} + \epsilon \delta^{(\mathbf{x})} \right] - \widetilde{\mathbf{G}} \left[f_{\mathbf{X}} \right] \right)$$
 (4.185)

$$\widehat{\mathbf{E}} \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_{t} \quad (4.196) \qquad \qquad \text{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{E}} \right) = \mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \quad (4.198)$$

$$\widehat{\mathbf{Cov}} \equiv \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right) \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right)' \qquad \qquad \text{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{Cov}} \right) = \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right) \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right)' - \text{Cov} \left\{ \mathbf{X} \right\} \quad (4.199)$$

$$\widehat{\mathbf{E}} \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_{t} \quad (4.196) \qquad \qquad \text{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{E}} \right) = \mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \quad (4.198)$$

$$\widehat{\mathbf{Cov}} \equiv \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right) \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right)' \qquad \qquad \text{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{Cov}} \right) = \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right) \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right)' - \mathbf{Cov} \left\{ \mathbf{X} \right\} \quad (4.199)$$

$$\widehat{\boldsymbol{\mu}} = \sum_{t=1}^{T} \frac{w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)}{\sum_{s=1}^{T} w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{s}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)} \mathbf{x}_{t} \quad \text{(4.203)}$$

$$\widehat{\Sigma} = \frac{1}{T} \sum_{t=1}^{T} (\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}) (\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}})' w \left(\operatorname{Ma}^{2} \left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}} \right) \right)$$
(4.204)

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$$\widehat{\mathbf{E}} \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_{t} \quad (4.196) \qquad \qquad \text{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{E}} \right) = \mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \quad (4.198)$$

$$\widehat{\mathbf{Cov}} \equiv \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right) \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right)' \qquad \qquad \text{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{Cov}} \right) = \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right) \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right)' - \mathbf{Cov} \left\{ \mathbf{X} \right\} \quad (4.199)$$

 $\widehat{\boldsymbol{\mu}} = \sum_{t=1}^{T} \frac{w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)}{\sum_{s=1}^{T} w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{s}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)} \mathbf{x}_{t} \quad (4.203)$

$$\widehat{\Sigma} = \frac{1}{T} \sum_{t=1}^{T} (\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}) (\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}})' w \left(\operatorname{Ma}^{2} \left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}} \right) \right)$$
(4.204)

$$w\left(z\right) \equiv -2\frac{g'\left(z\right)}{g\left(z\right)}. \tag{4.205}$$

$$f_{\boldsymbol{\theta}}\left(\mathbf{x}\right) \equiv \frac{1}{\sqrt{|\Sigma|}} g\left(\operatorname{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right)\right) \quad (4.201) \quad \blacktriangleleft \quad \quad \operatorname{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \equiv \left(\mathbf{x} - \boldsymbol{\mu}\right)' \boldsymbol{\Sigma}^{-1} \left(\mathbf{x} - \boldsymbol{\mu}\right) \quad (4.202)$$

$$\widehat{\mathbf{E}} \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_{t} \quad (4.196) \qquad \qquad \text{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{E}} \right) = \mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \quad (4.198)$$

$$\widehat{\mathbf{Cov}} \equiv \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right) \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right)' \qquad \qquad \text{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{Cov}} \right) = \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right) \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right)' - \mathbf{Cov} \left\{ \mathbf{X} \right\} \quad (4.199)$$

$$\widehat{\boldsymbol{\mu}} = \sum_{t=1}^{T} \frac{w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)}{\sum_{s=1}^{T} w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{s}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)} \mathbf{x}_{t} \qquad (4.203)$$

$$\psi\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \equiv \frac{\left(w\left(\operatorname{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right)\right) \left(\mathbf{x} - \boldsymbol{\mu}\right)\right)}{w\left(\operatorname{Ma}^{2}\left(\mathbf{x}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)} \times \left(\operatorname{Ma}^{2}\left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)$$

$$\widehat{\boldsymbol{\Sigma}} = \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\right) \left(\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\right)' w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)$$

$$\left(4.204\right)$$

$$w\left(z\right) \equiv -2\frac{g'\left(z\right)}{g\left(z\right)} \qquad (4.205)$$

$$f_{\boldsymbol{\theta}}\left(\mathbf{x}\right) \equiv \frac{1}{\sqrt{|\Sigma|}} g\left(\operatorname{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right)\right) \quad (4.201) \quad \blacktriangleleft \cdots \qquad \operatorname{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \equiv \left(\mathbf{x} - \boldsymbol{\mu}\right)' \boldsymbol{\Sigma}^{-1} \left(\mathbf{x} - \boldsymbol{\mu}\right) \quad (4.202)$$

$$\widehat{\mathbf{E}} \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_{t} \quad (4.196)$$

$$\widehat{\mathbf{Cov}} \equiv \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right) \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right)'$$

$$\mathbf{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{E}} \right) = \mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \quad (4.198)$$

$$\mathbf{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{Cov}} \right) = \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right) \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right)' - \mathbf{Cov} \left\{ \mathbf{X} \right\} \quad (4.199)$$

$$\widehat{\boldsymbol{\mu}} = \sum_{t=1}^{T} \frac{w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)}{\sum_{s=1}^{T} w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{s}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)} \mathbf{x}_{t} \quad \text{(4.203)}$$

$$\psi\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \equiv \begin{bmatrix} \operatorname{IF}\left(\mathbf{x}, f_{\mathbf{X}}, \left(\widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right) & \propto \|\psi\| \quad \text{(4.208)} \\ \psi\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right) & = \begin{bmatrix} w\left(\operatorname{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right)\right) & (\mathbf{x} - \boldsymbol{\mu}) \\ w\left(\operatorname{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right)\right) & \text{vec}\left[\left(\mathbf{x} - \boldsymbol{\mu}\right) & (\mathbf{x} - \boldsymbol{\mu})'\right] - \operatorname{vec}\left[\boldsymbol{\Sigma}\right] \end{bmatrix}$$

$$\widehat{\boldsymbol{\Sigma}} = \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\right) \left(\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\right)' w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right) \\ \text{(4.204)}$$

$$w\left(z\right) \equiv -2\frac{g'\left(z\right)}{g\left(z\right)} \quad \text{(4.205)} \quad \text{normal} \\ w \equiv 1$$

$$w\left(z\right) \equiv -2\frac{g'\left(z\right)}{g\left(z\right)}. \quad (4.205)$$

$$\widehat{\mathbf{E}} \equiv \frac{1}{T} \sum_{t=1}^{T} \mathbf{x}_{t} \quad (4.196) \qquad \qquad \text{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{E}} \right) = \mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \quad (4.198)$$

$$\widehat{\mathbf{Cov}} \equiv \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right) \left(\mathbf{x}_{t} - \widehat{\mathbf{E}} \right)' \qquad \qquad \text{IF} \left(\mathbf{x}, f_{\mathbf{X}}, \widehat{\mathbf{Cov}} \right) = \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right) \left(\mathbf{x} - \mathbf{E} \left\{ \mathbf{X} \right\} \right)' - \mathbf{Cov} \left\{ \mathbf{X} \right\} \quad (4.199)$$

$$\widehat{\boldsymbol{\mu}} = \sum_{t=1}^{T} \frac{w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)}{\sum_{s=1}^{T} w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{s}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right)} \mathbf{x}_{t} \quad \text{(4.203)}$$

$$\psi\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right) \equiv \frac{\left(\operatorname{W}\left(\operatorname{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right)\right) \left(\mathbf{x} - \boldsymbol{\mu}\right)\right)}{\left(\operatorname{W}\left(\operatorname{Ma}^{2}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Sigma}\right)\right) \left(\mathbf{x} - \boldsymbol{\mu}\right)\right)} \mathbf{x}_{t} \quad \text{(4.203)}$$

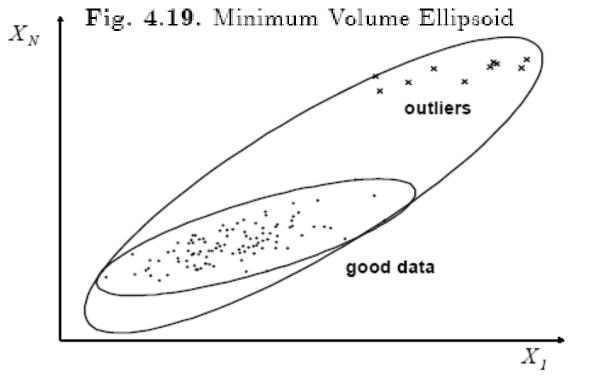
$$\widehat{\boldsymbol{\Sigma}} = \frac{1}{T} \sum_{t=1}^{T} \left(\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\right) \left(\mathbf{x}_{t} - \widehat{\boldsymbol{\mu}}\right)' w\left(\operatorname{Ma}^{2}\left(\mathbf{x}_{t}, \widehat{\boldsymbol{\mu}}, \widehat{\boldsymbol{\Sigma}}\right)\right) \quad \text{(4.204)}$$

$$w\left(z\right)\equiv-2\frac{g'\left(z\right)}{g\left(z\right)}.$$
 (4.205) normal $w\equiv1$ Cauchy $w\left(z\right)=\frac{N+1}{1+z}$

Cauchy
$$w(z) = \frac{N+1}{1+z}$$
(4.209)

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$$\mathcal{E}_{\mu,\Sigma}^{q} \equiv \left\{\mathbf{x} \in \mathbb{R}^{N} \text{ such that } (\mathbf{x} - \mu)' \, \Sigma^{-1} \, (\mathbf{x} - \mu) \leq q^{2} \right\} \quad \text{\tiny (4.231)}$$



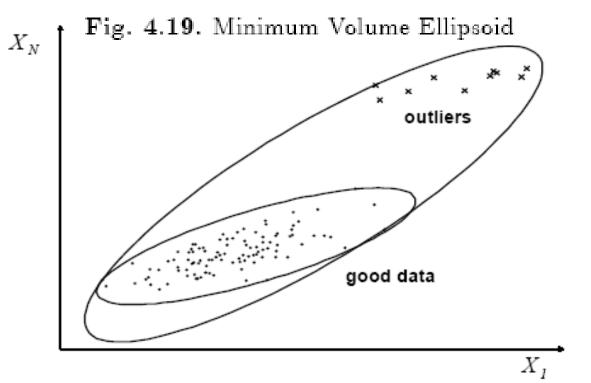
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$$\mathcal{E}^{q}_{\mu,\Sigma} \equiv \left\{\mathbf{x} \in \mathbb{R}^{N} \text{ such that } (\mathbf{x} - \mu)' \, \Sigma^{-1} \, (\mathbf{x} - \mu) \leq q^{2} \right\} \quad \text{\tiny (4.231)}$$

$$\operatorname{Ma}_{t}^{\mu,\Sigma} \equiv \sqrt{(\mathbf{x}_{t} - \boldsymbol{\mu})' \, \Sigma^{-1} \, (\mathbf{x}_{t} - \boldsymbol{\mu})}$$
. (4.234)

$$\operatorname{Vol}\left\{\mathcal{E}_{\mu,\Sigma}^{q_{T_G}}\right\} = \gamma_N \left(\operatorname{Ma}_{T_G:T}^{\mu,\Sigma}\right)^N \sqrt{|\Sigma|} \quad (4.236)$$

$$q_{T_G} \equiv \operatorname{Ma}_{T_G:T}^{\mu,\Sigma} \quad (4.235)$$



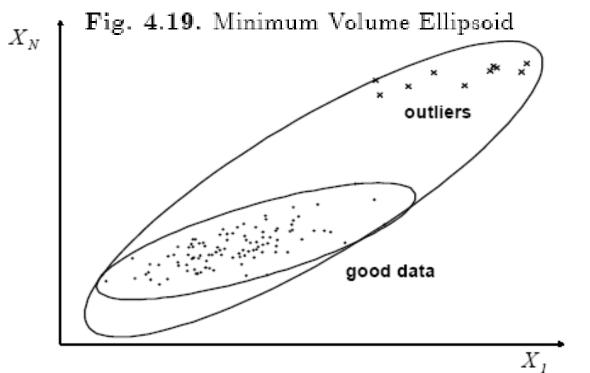
ROBUST ESTIMATORS – HIGH BREAKDOWN

$$\mathcal{E}^{q}_{\mu,\Sigma} \equiv \left\{\mathbf{x} \in \mathbb{R}^{N} \text{ such that } (\mathbf{x} - \mu)' \, \Sigma^{-1} \, (\mathbf{x} - \mu) \leq q^{2} \right\} \quad \text{\tiny (4.231)}$$

$$\operatorname{Ma}_{t}^{\mu,\Sigma} \equiv \sqrt{(\mathbf{x}_{t} - \boldsymbol{\mu})' \Sigma^{-1} (\mathbf{x}_{t} - \boldsymbol{\mu})}.$$
 (4.234)

$$\operatorname{Vol}\left\{\mathcal{E}_{\mu,\Sigma}^{q_{T_G}}\right\} = \gamma_N \left(\operatorname{Ma}_{T_G:T}^{\mu,\Sigma}\right)^N \sqrt{|\Sigma|} \quad (4.236)$$

$$q_{T_G} \equiv \operatorname{Ma}_{T_G:T}^{\mu,\Sigma} \quad (4.235)$$



$$\left(\widehat{\boldsymbol{\mu}}_{T_{G}}, \widehat{\boldsymbol{\Sigma}}_{T_{G}}\right) = \underset{\boldsymbol{\mu}, \boldsymbol{\Sigma} \succeq \boldsymbol{0}, |\boldsymbol{\Sigma}| = 1}{\operatorname{argmin}} \left\{ \operatorname{Ma}_{T_{G}:T}^{\boldsymbol{\mu}, \boldsymbol{\Sigma}} \right\}$$

$$(4.237) = \underbrace{\left(\widehat{\boldsymbol{\mu}}_{T_{G}}, \widehat{\boldsymbol{\Sigma}}_{T_{G}}\right)}_{(4.237)} = \underbrace{\left(\widehat{\boldsymbol{\mu}}_{T_{G}}, \widehat{\boldsymbol{\mu}}_{T_{G}}\right)}_{(4.237)} = \underbrace{\left(\widehat{\boldsymbol{\mu}}_{T_{G}}, \widehat{\boldsymbol{\mu}}_{T_{G}}\right)}_{$$

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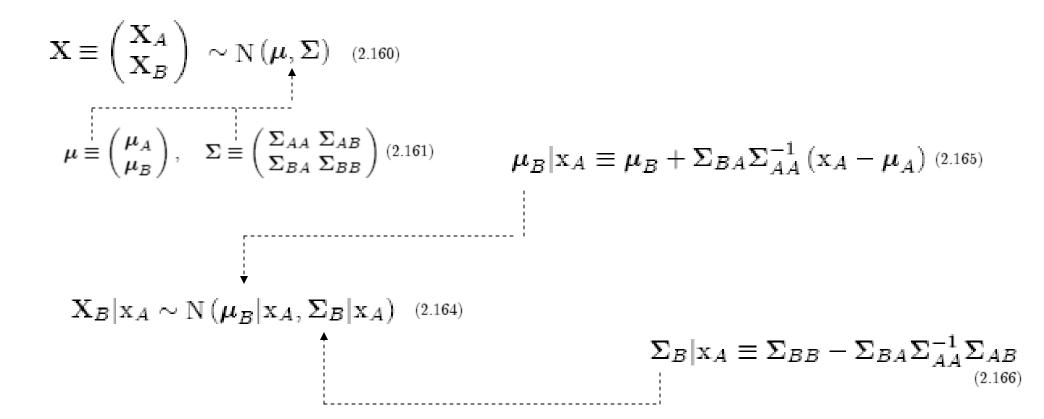
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The notation, say, (T4.12) refers to Formula 12 in the Technical Appendices for Chapter 4, which can be downloaded from www.symmys.com

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$$\mathbf{X} \equiv \begin{pmatrix} \mathbf{X}_A \\ \mathbf{X}_B \end{pmatrix} \sim \mathrm{N} \left(\boldsymbol{\mu}, \boldsymbol{\Sigma} \right) \quad (2.160)$$

$$\boldsymbol{\mu} \equiv \begin{pmatrix} \boldsymbol{\mu}_A \\ \boldsymbol{\mu}_B \end{pmatrix}, \quad \boldsymbol{\Sigma} \equiv \begin{pmatrix} \boldsymbol{\Sigma}_{AA} & \boldsymbol{\Sigma}_{AB} \\ \boldsymbol{\Sigma}_{BA} & \boldsymbol{\Sigma}_{BB} \end{pmatrix} \quad (2.161)$$



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$$\begin{pmatrix} \mathbf{X}_{t, \min(t)} \\ \mathbf{X}_{t, \mathsf{obs}(t)} \end{pmatrix} \sim \mathbf{N} \begin{pmatrix} \begin{pmatrix} \boldsymbol{\mu}_{\min(t)} \\ \boldsymbol{\mu}_{\mathsf{obs}(t)} \end{pmatrix}, \begin{pmatrix} \boldsymbol{\Sigma}_{\min(t), \min(t)} & \boldsymbol{\Sigma}_{\min(t), \mathsf{obs}(t)} \\ \boldsymbol{\Sigma}_{\mathsf{obs}(t), \min(t)} & \boldsymbol{\Sigma}_{\mathsf{obs}(t), \mathsf{obs}(t)} \end{pmatrix} \end{pmatrix} \quad (4.257)$$

$$\mu_B | \mathbf{x}_A \equiv \mu_B + \Sigma_{BA} \Sigma_{AA}^{-1} \left(\mathbf{x}_A - \mu_A \right)$$
 (2.165)

$$\Sigma_B|\mathbf{x}_A \equiv \Sigma_{BB} - \Sigma_{BA}\Sigma_{AA}^{-1}\Sigma_{AB}$$
(2.166)

$$\begin{pmatrix} \mathbf{X}_{t, \min(t)} \\ \mathbf{X}_{t, \mathsf{obs}(t)} \end{pmatrix} \sim \mathbf{N} \begin{pmatrix} \boldsymbol{\mu}_{\min(t)} \\ \boldsymbol{\mu}_{\mathsf{obs}(t)} \end{pmatrix}, \begin{pmatrix} \boldsymbol{\Sigma}_{\min(t), \min(t)} & \boldsymbol{\Sigma}_{\min(t), \mathsf{obs}(t)} \\ \boldsymbol{\Sigma}_{\mathsf{obs}(t), \min(t)} & \boldsymbol{\Sigma}_{\mathsf{obs}(t), \mathsf{obs}(t)} \end{pmatrix} \end{pmatrix}$$
(4.257)

$$\mathbf{x}_{t,obs(t)}^{(u)} \equiv \mathbf{x}_{t,obs(t)}^{(4.261)} \qquad \mu_{B} | \mathbf{x}_{A} \equiv \mu_{B} + \Sigma_{BA} \Sigma_{AA}^{-1} \left(\mathbf{x}_{A} - \mu_{A} \right)$$
(2.165)
$$\mathbf{x}_{t,mis(t)}^{(u)} \equiv \mu_{mis(t)}^{(u)} \qquad (4.262)$$

$$+ \Sigma_{mis(t),obs(t)}^{(u)} \left(\Sigma_{obs(t),obs(t)}^{(u)} \right)^{-1} \left(\mathbf{x}_{t,obs(t)} - \mu_{obs(t)}^{(u)} \right)$$

$$\Sigma_B|\mathbf{x}_A \equiv \Sigma_{BB} - \Sigma_{BA}\Sigma_{AA}^{-1}\Sigma_{AB}$$
(2.166)

$$\begin{pmatrix} \mathbf{X}_{t, \min(t)} \\ \mathbf{X}_{t, \mathsf{obs}(t)} \end{pmatrix} \sim \mathbf{N} \begin{pmatrix} \begin{pmatrix} \boldsymbol{\mu}_{\min(t)} \\ \boldsymbol{\mu}_{\mathsf{obs}(t)} \end{pmatrix}, \begin{pmatrix} \boldsymbol{\Sigma}_{\min(t), \min(t)} & \boldsymbol{\Sigma}_{\min(t), \mathsf{obs}(t)} \\ \boldsymbol{\Sigma}_{\mathsf{obs}(t), \min(t)} & \boldsymbol{\Sigma}_{\mathsf{obs}(t), \mathsf{obs}(t)} \end{pmatrix} \end{pmatrix} \quad (4.257)$$

$$\begin{split} \mathbf{x}_{t,\text{obs}(t)}^{(u)} &\equiv \mathbf{x}_{t,\text{obs}(t)} & \text{(4.261)} \\ \mathbf{x}_{t,\text{mis}(t)}^{(u)} &\equiv \boldsymbol{\mu}_{\text{mis}(t)}^{(u)} & \text{(4.262)} \\ &+ \boldsymbol{\Sigma}_{\text{mis}(t),\text{obs}(t)} \left(\boldsymbol{\Sigma}_{\text{obs}(t),\text{obs}(t)}^{(u)}\right)^{-1} \left(\mathbf{x}_{t,\text{obs}(t)} - \boldsymbol{\mu}_{\text{obs}(t)}^{(u)}\right) \\ &- \boldsymbol{\Sigma}_{\text{mis}(t),\text{obs}(t)}^{(u)} &\equiv \mathbf{0}, \quad \mathbf{C}_{t,\text{obs}(t),\text{obs}(t)}^{(u)} &\equiv \mathbf{0}, \quad \text{(4.263)} \\ &\mathbf{C}_{t,\text{mis}(t),\text{mis}(t)}^{(u)} &\equiv \boldsymbol{\Sigma}_{\text{mis}(t),\text{mis}(t)}^{(u)} &\equiv \boldsymbol{\Sigma}_{\text{mis}(t),\text{mis}(t)}^{(u)} &\text{(4.264)} \\ &- \boldsymbol{\Sigma}_{\text{mis}(t),\text{obs}(t)}^{(u)} \left(\boldsymbol{\Sigma}_{\text{obs}(t),\text{obs}(t)}^{(u)}\right)^{-1} \boldsymbol{\Sigma}_{\text{obs}(t),\text{mis}(t)}^{(u)}. &\text{estimate} \end{split}$$

$$\begin{pmatrix} \mathbf{X}_{t, \min(t)} \\ \mathbf{X}_{t, \mathsf{obs}(t)} \end{pmatrix} \sim \mathbf{N} \begin{pmatrix} \begin{pmatrix} \boldsymbol{\mu}_{\min(t)} \\ \boldsymbol{\mu}_{\mathsf{obs}(t)} \end{pmatrix}, \begin{pmatrix} \boldsymbol{\Sigma}_{\min(t), \min(t)} & \boldsymbol{\Sigma}_{\min(t), \mathsf{obs}(t)} \\ \boldsymbol{\Sigma}_{\mathsf{obs}(t), \min(t)} & \boldsymbol{\Sigma}_{\mathsf{obs}(t), \mathsf{obs}(t)} \end{pmatrix} \end{pmatrix} \quad (4.257)$$

$$\begin{split} \mathbf{x}_{t, \mathsf{obs}(t)}^{(u)} &\equiv \mathbf{x}_{t, \mathsf{obs}(t)} & \text{(4.261)} \\ \mathbf{x}_{t, \mathsf{mis}(t)}^{(u)} &\equiv \boldsymbol{\mu}_{\mathsf{mis}(t)}^{(u)} & \text{(4.262)} \\ &+ \boldsymbol{\Sigma}_{\mathsf{mis}(t), \mathsf{obs}(t)}^{(u)} \left(\boldsymbol{\Sigma}_{\mathsf{obs}(t), \mathsf{obs}(t)}^{(u)}\right)^{-1} \left(\mathbf{x}_{t, \mathsf{obs}(t)} - \boldsymbol{\mu}_{\mathsf{obs}(t)}^{(u)}\right) \\ &- \boldsymbol{\Sigma}_{\mathsf{mis}(t), \mathsf{mis}(t)}^{(u)} &\equiv \mathbf{0}, \quad \mathbf{C}_{t, \mathsf{obs}(t), \mathsf{obs}(t)}^{(u)} &\equiv \mathbf{0}, \quad \text{(4.263)} \\ &\mathbf{C}_{t, \mathsf{mis}(t), \mathsf{mis}(t)}^{(u)} &\equiv \boldsymbol{\Sigma}_{\mathsf{mis}(t), \mathsf{mis}(t)}^{(u)} &= \mathbf{0}, \quad \text{(4.264)} \\ &- \boldsymbol{\Sigma}_{\mathsf{mis}(t), \mathsf{obs}(t)}^{(u)} \left(\boldsymbol{\Sigma}_{\mathsf{obs}(t), \mathsf{obs}(t)}^{(u)}\right)^{-1} \boldsymbol{\Sigma}_{\mathsf{obs}(t), \mathsf{mis}(t)}^{(u)}. &\qquad \text{estimate} \end{split}$$

$$\begin{split} \mu^{(u+1)} &\equiv \frac{1}{T} \sum_t \mathbf{x}_t^{(u)} \quad \text{\tiny (4.265)} \\ \Sigma^{(u+1)} &\equiv \frac{1}{T} \sum_t \left[\mathbf{C}_t^{(u)} + \left(\mathbf{x}_t^{(u)} - \mu^{(u)} \right) \left(\mathbf{x}_t^{(u)} - \mu^{(u)} \right)' \right] \quad \text{\tiny (4.266)} \end{split}$$
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