

# Example (a) #11

$$y = Ax + v \quad v \sim (0, \sigma^2 I)$$

$$y_1 = (1 + \varepsilon_1)x_1 + v_1$$

$$y_2 = (1 + \varepsilon_2)x_2 + v_2$$

$$A = \begin{bmatrix} 1 + \varepsilon_1 & 0 \\ 0 & 1 + \varepsilon_2 \end{bmatrix}$$

$$J = E A' \bar{F} A'^*$$

$$v \sim (0, \sigma^2 I)$$

$$x \in \mathbb{R}^2 \quad x \sim (x_0, F)$$

$$x_0 = \begin{bmatrix} c \\ c \end{bmatrix}, \quad F = \varphi^2 I$$

$$\varepsilon_i \text{ i.i.d. } \varepsilon_i \sim (0, \delta^2)$$

$$= I + \begin{bmatrix} \varepsilon_1 & 0 \\ 0 & \varepsilon_2 \end{bmatrix} = A_0 + A'$$

$$\bar{F} = F + x_0 x_0^T -$$

the second moment of  $x$

$$\bar{F} = E x x^T$$

$$J = (\varphi^2 + c^2) \delta^2 I$$

$$Q = \left( A_0^* (S + J)^{-1} A_0 + F^{-1} \right)^{-1} \quad | \quad \text{Var}(\hat{x})$$
$$= \left( I \left( \sigma^2 + (\varphi^2 + c^2) \delta^2 \right) I + (\varphi^2 I)^{-1} \right)^{-1}$$

$$= \frac{1}{\frac{1}{\sigma^2 + (\varphi^2 + c^2) \delta^2} + \frac{1}{\varphi^2}} I$$

$$\rightarrow \begin{matrix} \text{Var} \hat{x}_1 = \\ \text{Var} \hat{x}_2 \end{matrix}$$

$$(b) \quad y_1 = (1 + \varepsilon)x_1 + v_1, \quad \varepsilon \sim (0, \delta^2)$$

$$y_2 = (1 + \varepsilon)x_2 + v_2$$

$$A = (1 + \varepsilon)\bar{I} = \underbrace{\bar{I}}_{=A_0} + \underbrace{\varepsilon\bar{I}}_{A'}$$

$$J = EA' \bar{F} A'^* = E \varepsilon I \left[ \varphi^2 I + c^2 \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \right] \varepsilon I$$

$$= E \varepsilon^2 \begin{bmatrix} \varphi^2 + c^2 & c^2 \\ c^2 & \varphi^2 + c^2 \end{bmatrix}$$

$$\stackrel{=}{=} \delta^2 \begin{bmatrix} \varphi^2 + c^2 & c^2 \\ c^2 & \varphi^2 + c^2 \end{bmatrix}$$

$$J = \delta^2 \begin{bmatrix} \varphi^2 + c^2 & c^2 \\ c^2 & \varphi^2 + c^2 \end{bmatrix}$$

$$\text{in (a)} \quad J = \delta^2 (\varphi^2 + c^2) I = \delta^2 \begin{bmatrix} \varphi^2 + c^2 & 0 \\ 0 & \varphi^2 + c^2 \end{bmatrix}$$

# Accumulating Information 3 from Experiments with random A.

$$y = Ax + v$$

Apriori info about  $x$

$$x \sim (x_0, F)$$

$$v \sim (0, S)$$

$$A_0 = EA$$

$$J = E(A - A_0)F(A - A_0)^*$$

Observation  $(y, A_0, J, S)$

$$Q = (F^{-1} + A_0^*(S+J)^{-1}A_0)^{-1}$$

$$\hat{x} = Q(F^{-1}x_0 + A_0^*(S+J)^{-1}y)$$

$$y_1 = A_1x + v_1 \quad (y_1, \bar{A}_1, J_1, S_1)$$

$$y_2 = A_2x + v_2 \quad (y_2, \bar{A}_2, J_2, S_2)$$

$$\bar{A}_1 = EA, \quad J_1 = EA'FA'^*$$

$$A_1' = A_1 - \bar{A}_1$$

Same for  $\bar{A}_2, J_2$   
Measurements are indep

$A_1, A_2, v_1, v_2, x$  are independent.

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} A_1 \\ A_2 \end{bmatrix} x + \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

$$y = Ax + v, \quad \text{Var}(v) = \begin{bmatrix} S_1 & 0 \\ 0 & S_2 \end{bmatrix}$$

$$\bar{A} = EA = E \begin{bmatrix} A_1 \\ A_2 \end{bmatrix} = \begin{bmatrix} EA_1 \\ EA_2 \end{bmatrix} = \begin{bmatrix} \bar{A}_1 \\ \bar{A}_2 \end{bmatrix}$$

$$J = ?$$

$$J = E A' \bar{F} A'^*$$

$$= E \begin{bmatrix} A_1' \\ A_2' \end{bmatrix} \bar{F} \begin{bmatrix} A_1'^* & A_2'^* \end{bmatrix}$$

$$A' = A - \bar{A} = \begin{bmatrix} A_1 - \bar{A}_1 \\ A_2 - \bar{A}_2 \end{bmatrix} = \begin{bmatrix} A_1' \\ A_2' \end{bmatrix}$$

$$= E \begin{bmatrix} A_1' \bar{F} A_1'^* & A_1' \bar{F} A_2'^* \\ A_2' \bar{F} A_1'^* & A_2' \bar{F} A_2'^* \end{bmatrix}$$

$$E A_1' \bar{F} A_1'^* = J_1$$

$$E A_2' \bar{F} A_2'^* = \underbrace{EA_2'}_{=0} \cdot \bar{F} \cdot \underbrace{EA_2'^*}_{=0}$$

$$J = \begin{bmatrix} J_1 & 0 \\ 0 & J_2 \end{bmatrix}$$

$$S + J = \begin{bmatrix} S_1 + J_1 & 0 \\ 0 & S_2 + J_2 \end{bmatrix}$$

Formulas are the same<sup>5</sup>  
as for known  $A$ . Need to  
replace

$$S_i \rightarrow S_i + J_i; \quad A_i \rightarrow \bar{A}_i$$

Collecting canonical info  $(T, \beta)$

apriori inf:  $x \sim (x_0, F) \mapsto T_0 = F^{-1}, \beta_0 = F'x_0$

$$(y_i, \bar{A}_i, J_i, S_i) \mapsto T_i = \bar{A}_i^* (S_i + J_i)^{-1} \bar{A}_i,$$
$$\beta_i = \bar{A}_i^* (S_i + J_i)^{-1} y_i,$$

$$(y_i, \bar{A}_i, J_i, S_i) \mapsto (T_i, \beta_i)$$

$$(T, \beta) = \bigoplus_{i=0}^n (T_i, \beta_i) = \left( \sum_{i=0}^n T_i, \sum_{i=0}^n \beta_i \right).$$

# Examples

$$y_i = (a + \varepsilon_i) x + \nu_i$$

$\varepsilon_i, \nu_i \quad i=1 \dots n$   
independent

$$J_i = \mathbb{E} A_i' \bar{F} A_i$$

$$= \mathbb{E} \varepsilon_i (\varphi^2 + x_0^2) \varepsilon_i$$

$$= (\varphi^2 + x_0^2) \delta^2$$

$$x \sim (x_0, \varphi^2)$$

a priori int.

$$\nu_i \sim (0, \sigma^2) \text{ i.i.d.}$$

$$\varepsilon_i \sim (0, \delta^2) \text{ i.i.d.}$$

$$\bar{F} = F + x_0 x_0' =$$

$$= \varphi^2 + x_0^2$$

$$\begin{cases} T_i = \bar{A}_i' (S_i + J_i)^{-1} \bar{A}_i = a^2 (\sigma^2 + (\varphi^2 + x_0^2) \delta^2)^{-1} \\ \beta_i = \bar{A}_i' (S_i + J_i)^{-1} y_i = a (\sigma^2 + (\varphi^2 + x_0^2) \delta^2)^{-1} y_i \end{cases}$$

$$y_i \mapsto (T_i, \beta_i)$$

$$(x_0, \varphi^2) \mapsto (T_0, \beta_0) = (F^{-1}, F^{-1} x_0) = (\varphi^{-2}, \varphi^{-2} x_0)$$

$$T = T_0 + T_1 + \dots + T_n$$

$$= \frac{1}{\varphi_0^2} + n a^2 \underbrace{\left[ \sigma^2 + (\varphi^2 + x_0^2) \delta^2 \right]^{-1}}_{\lambda^2}$$

$$= \frac{1}{\varphi_0^2} + \frac{n a^2}{\sigma^2 + \lambda^2} \rightarrow \infty$$

$n \rightarrow \infty$

$$\beta = \beta_0 + \beta_1 + \dots + \beta_n$$

$$\text{Var } \hat{x} = T^{-1} \rightarrow 0$$

Assume that  $A$  does not change. Remains same, but unknown. 7

$$y_i = (a + \varepsilon)x + v_i$$

measurements are not indep.

Combining experiments with same  $A$ .

$$y_1 = Ax + v_1$$

$$y_2 = Ax + v_2$$

$$E A = \underline{A_0}$$

$$\underline{J_0} = E A' \bar{F} A'^*$$

$$v_i \sim (0, S) \text{ i.i.d.}$$

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} A \\ A \end{bmatrix} x + \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

$$\text{Var} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} S & 0 \\ 0 & S_2 \end{bmatrix}$$

$$J = E \begin{bmatrix} A' \\ A' \end{bmatrix} \bar{F} \begin{bmatrix} A'^* & A'^* \end{bmatrix}$$

$$= \begin{bmatrix} E A' \bar{F} A'^* & E A' \bar{F} A'^* \\ E A' F A'^* & E A' F A'^* \end{bmatrix} = \begin{bmatrix} J_0 & J_0 \\ J_0 & J_0 \end{bmatrix}$$

Back to example.

$$J_0 = (\varphi^2 + \alpha_0^2) \delta^2 = \lambda^2 \quad (n \times n \text{ matr.})$$

$$J_{n \times n} = \begin{bmatrix} J_0 & J_0 & \dots & J_0 \\ \vdots & \vdots & \ddots & \vdots \\ J_0 & J_0 & & J_0 \end{bmatrix} = \lambda^2 \begin{bmatrix} 1 & \dots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \dots & 1 \end{bmatrix}$$

$$= \lambda^2 \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} \begin{bmatrix} 1 & \dots & 1 \end{bmatrix} = \lambda^2 e e^T$$

$\underbrace{\begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}}_e$

$$(S+J)^{-1} = (\sigma^2 I + \lambda^2 e e^T)^{-1}$$

$$(\alpha I + \beta e e^T)^{-1} = ? \quad \begin{bmatrix} \alpha + \beta & \beta \\ & \ddots \\ \beta & \alpha + \beta \end{bmatrix}$$

$$? = \underline{\gamma} I + \underline{\delta} e e^T$$

$$I = (\alpha I + \beta e e^T)(\gamma I + \delta e e^T)$$

Want  $\rightarrow$

$$= \alpha \gamma I + \alpha \delta e e^T + \beta \gamma e e^T + \beta \delta \underbrace{e e^T e e^T}_n$$

$$e^T e = \begin{bmatrix} 1 & \dots & 1 \end{bmatrix} \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} = n$$

$$= \alpha \underline{\gamma} I + (\alpha \underline{\delta} + \beta \gamma + \beta \underline{\delta} n) e e^T$$

$$\Rightarrow \alpha \gamma = 1 \quad \Rightarrow \gamma = \frac{1}{\alpha}$$

$$\alpha \delta + \beta \gamma + \beta \delta n = 0$$



$$(\alpha + \beta n) \delta = -\beta \gamma$$

$$\delta = -\frac{\beta \gamma}{\alpha + \beta n} = -\frac{\beta}{\alpha(\alpha + \beta n)}$$

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$$(\alpha I + \beta e e^T)^{-1} = \frac{1}{\alpha} \left( I - \frac{\beta}{\alpha + \beta n} e e^T \right)$$

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$$(S + J)^{-1} = (\sigma^2 I + \lambda^2 e e^T)^{-1}$$
$$= \frac{1}{\sigma^2} \left( I - \frac{\lambda^2}{\sigma^2 + n\lambda^2} e e^T \right)$$

$$T = A_0^* (S + J)^{-1} A_0 + F^{-1}$$

$$| F^{-1} = \frac{1}{\varphi^2} \quad A_0 = \begin{bmatrix} a \\ \vdots \\ a \end{bmatrix} = a e$$

$$T = a e^T \frac{1}{\sigma^2} \left( I - \frac{\lambda^2}{\sigma^2 + n\lambda^2} e e^T \right) a e + \frac{1}{\varphi^2}$$

$$= \frac{a^2}{\sigma^2} \left( \underbrace{e^T e}_{=n} - \frac{\lambda^2}{\sigma^2 + n\lambda^2} \underbrace{e^T e}_{=n} \underbrace{e^T e}_{=n} \right) + \frac{1}{\varphi^2}$$

$$= n \frac{a^2}{\sigma^2} \left( 1 - \frac{n\lambda^2}{\sigma^2 + n\lambda^2} \right) + \frac{1}{\varphi^2}$$

$$= n \frac{a^2}{\sigma^2} \frac{\sigma^2 + n\lambda^2 - n\lambda^2}{\sigma^2 + n\lambda^2} + \frac{1}{\varphi^2} = \frac{na^2}{\sigma^2 + n\lambda^2} + \frac{1}{\varphi^2}$$

$$T = \frac{n a^2}{\sigma^2 + n \lambda^2} + \frac{1}{\varphi^2} \rightarrow \frac{a^2}{\lambda^2} + \frac{1}{\varphi^2} \quad 10$$

$$Q = \text{Var}(\bar{x}) \rightarrow \frac{1}{a^2/\lambda^2 + 1/\varphi^2} \neq 0 \quad \text{as } n \rightarrow \infty$$

with  $\varepsilon_i$  - indep.

$$T = \frac{n a^2}{\sigma^2 + \lambda^2} + \frac{1}{\varphi^2} \rightarrow \infty$$

$$Q \rightarrow 0$$

# Calibration problem. //

Idea in trivial settings.

$$y = ax + v$$

Take know  $\varphi$

$$\Psi = a\varphi + \mu$$

Heuristic:  $\hat{a} = \Psi/\varphi$

and use it as a "fake"  $a$ .

But: need to take into account its inaccuracy.

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$$y = Ax + v$$

$$x \sim (x_0, F)$$

$$A: D \rightarrow \mathcal{R}$$

$$v \sim (0, S)$$

Have some info about  $A$  (or no info)

calibration observations:

$$\Psi_i = A\varphi_i + \mu_i \quad i = 1, \dots, K$$

$K$  calibration measurements.

$(\varphi_i, \Psi_i)$  - calibration info.

$\varphi_i \in D$  - known calibration signals

$\Psi_i \in \mathcal{R}$  - observed result of calibration

$\mu_i \in \mathcal{R}$  - random vectors  $\mu_i \sim (0, S)$  ind.

$$\dim \mathcal{D} = m$$

$$\dim \mathcal{R} = n$$

num of cal. meas =  $k$ .

$$\Phi = [\varphi_1 \dots \varphi_k] \quad m \times k$$

$$\Psi = [\psi_1 \dots \psi_k] \quad n \times k$$

$(\Phi, \Psi) \xrightarrow{?} (\mathcal{R}, \mathcal{V})$  - optimal estimator.  
calibr. info.

Separate cal. meas:

$$\psi_i = A\varphi_i + \mu_i \quad i = 1, \dots, k$$

$$\Psi = A\Phi + M \quad M = [\mu_1 \dots \mu_k]$$

$\times \Phi^*$  random matrix

$$\Psi\Phi^* = A\Phi\Phi^* + M\Phi^* \quad EM = 0 \quad n \times k \text{ matrix.}$$

Assume  $\Phi\Phi^*$  is invertible.

$$\Phi \quad m \times k \quad \underline{\Phi\Phi^*} = m \times m$$

take  $\varphi_i$  as indep as possible.

$$\Rightarrow \text{if } k \geq m \text{ rank } \Phi = m$$

$$\Rightarrow \Phi\Phi^* - \text{invertible.}$$

$$\underbrace{\Psi \Phi^* (\Phi \Phi^*)^{-1}}_{= \Phi^-} = A + M \Phi^* (\Phi \Phi^*)^{-1} \quad 13$$

=  $\Phi^-$  - pseudoinverse of  $\Phi$ .

$$\Psi \Phi^- = A + M \Phi^-$$

$$A = \underbrace{\Psi \Phi^-}_{= A_0} - M \Phi^-$$

=  $A_0$  - estimate of  $A$ .

$$A = A_0 + A' \quad A' = M \Phi^-$$

$$J = E A' \bar{F} A'^* = E M \underbrace{\Phi^- \bar{F} \Phi^{-*}}_{= B} M^*$$

$$B = \underbrace{\Phi^*}_{k \times m} (\Phi \Phi^*)^{-1} \bar{F} (\Phi \Phi^*)^{-1} \underbrace{\Phi}_{m \times k} = B \quad k \times k$$

$$J = E M B M^* = ?$$

$$M = [\mu_1, \dots, \mu_k]$$

$$J = E [\mu_1, \dots, \mu_k] B \begin{bmatrix} \mu_1^T \\ \vdots \\ \mu_k^T \end{bmatrix}$$

$$B = \begin{bmatrix} \beta_{11} & \beta_{12} & \dots & \beta_{1k} \\ \vdots & \vdots & \ddots & \vdots \\ \beta_{k1} & \dots & \dots & \beta_{kk} \end{bmatrix}$$

$$J = E[\mu_1 \dots \mu_k] \begin{bmatrix} \sum_i b_{1i} \mu_i^T \\ \vdots \\ \sum_i b_{ki} \mu_i^T \end{bmatrix}$$

$$= E\left(\underbrace{\mu_1 \sum_i b_{1i} \mu_i^T + \dots + \mu_k \sum_i b_{ki} \mu_i^T}_{\sum_i b_{ii} \mu_i \mu_i^T}\right)$$

$$\left| E \mu_j \mu_i^T = \delta_{ji} \underbrace{E \mu_i \mu_i^T}_{= S = \text{var } \mu_i} \right.$$

$$= b_{11} S + b_{22} S + \dots + b_{kk} S$$

$$= \sum_{i=1}^k b_{ii} \cdot S = \text{tr } B \cdot S$$

$$J = \text{tr}(B) \cdot S$$

$$\text{tr } B = \text{tr } \Phi^* (\Phi \Phi^*)^{-1} \bar{F} (\Phi \Phi^*)^{-1} \Phi$$

$$\text{tr } A B^T = \sum_i (A B^T)_{ii} \quad A, B \text{ have } 15 \text{ same dim.}$$

$$= \sum_i \sum_j A_{ij} \underbrace{B_{ji}^T}_{B_{ij}} = \sum_{ij} A_{ij} B_{ij}$$

$$= \langle A, B \rangle$$

$$\text{tr } B A^T = \text{tr } B^T A = \text{tr } A^T B = \langle A, B \rangle$$

$$\text{tr } AC = \text{tr } CA$$

$n \times m \quad m \times n$

$$\text{tr } B = \text{tr } \underbrace{\Phi^* (\Phi \Phi^*)^{-1} \bar{F} (\Phi \Phi^*)^{-1} \Phi}_{=I}$$

$$= \text{tr } \underbrace{(\Phi \Phi^*)^{-1} \Phi \Phi^*}_{=I} (\Phi \Phi^*)^{-1} \bar{F}$$

$$= \text{tr } (\Phi \Phi^*)^{-1} \bar{F} = \langle (\Phi \Phi^*)^{-1}, \bar{F} \rangle$$

$$A_0 = \Psi \Phi^* (\Phi \Phi^*)^{-1}$$

$$J = \langle (\Phi \Phi^*)^{-1}, \bar{F} \rangle \cdot S$$

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## Information.

$$\Psi \Phi^* = [\Psi_1, \Psi_2, \dots, \Psi_k] \begin{bmatrix} \varphi_1^T \\ \vdots \\ \varphi_k^T \end{bmatrix}$$
$$= \sum_{i=1}^k \underbrace{\Psi_i \varphi_i^T}_{= G_i} = \sum_{i=1}^k G_i = G$$

$$G_i = \Psi_i \varphi_i^T$$

$$\Phi \Phi^T = \sum_{i=1}^k \underbrace{\varphi_i \varphi_i^T}_{= H_i} = \sum_{i=1}^k H_i = H$$

$$A_0 = G H^{-1}$$

$$J = \langle H^{-1}, \bar{F} \rangle S$$

$$J + S = \underbrace{\langle H^{-1}, \bar{F} \rangle}_{= \alpha} + 1) S = (\alpha + 1) S$$