

$$X^2(a, b) = \sum_{i=1}^2 \left(\frac{y_i - (ax_i + b)}{6} \right)^2$$

$$\frac{\partial X^2}{\partial a} = \sum_{i=1}^2 \frac{-2x_i (y_i - ax_i - b)}{6^2} = 0$$

$$\hookrightarrow -2x_1 (y_1 - ax_1 - b) - 2x_2 (y_2 - ax_2 - b) = 0 \quad (1)$$

$$\frac{\partial X^2}{\partial b} = \sum_{i=1}^2 \frac{-2(y_i - ax_i - b)}{6^2} = 0$$

$$\hookrightarrow -2(y_1 - ax_1 - b) - 2(y_2 - ax_2 - b) = 0 \quad (2)$$

$$(1) \equiv 2a(x_1^2 + x_2^2) + 2b(x_1 + x_2) - 2(x_1 y_1 + x_2 y_2) = 0$$

$$(2) \equiv 2(x_1 + x_2) + 2b - 2(y_1 + y_2) = 0$$

$$\hookrightarrow 2b = y_1 + y_2 - a(x_1 + x_2)$$

$$(1) \equiv 2a(x_1^2 + x_2^2) + \underbrace{(y_1 + y_2 - a(x_1 + x_2))}_{= 2b}(x_1 + x_2) - 2(x_1 y_1 + x_2 y_2) = 0$$

$$= 2a(x_1^2 + x_2^2) - a(x_1 + x_2)^2 + (x_1 + x_2)(y_1 + y_2) - 2(x_1 y_1 + x_2 y_2) = 0$$

$$= a \left[\underbrace{x_1^2 + x_2^2 - x_1^2 - x_2^2 - 2x_1 x_2}_{(x_2 - x_1)^2} \right] + \left[\underbrace{x_1 y_1 + x_2 y_2 + x_1 y_2 + x_2 y_1 - x_1 y_1 - x_2 y_2}_{= (x_1 - x_2)y_2 + (x_2 - x_1)y_1} \right]$$

$$= (x_1 - x_2)(y_2 - y_1)$$

$$= a(x_2 - x_1)^2 - (x_2 - x_1)(y_2 - y_1) = 0 \Rightarrow \hat{a} = \frac{y_2 - y_1}{x_2 - x_1}$$

(2)

$$\text{et } \hat{2b} = y_1 + y_2 - a(x_1 + x_2)$$

$$= y_1 + y_2 - \frac{(y_2 - y_1)(x_1 + x_2)}{(x_2 - x_1)}$$

$$= \frac{(y_1 + y_2)(x_2 - x_1) - (y_2 - y_1)(x_1 + x_2)}{x_2 - x_1}$$

$$= \frac{y_1 x_2 + y_2 x_2 - y_2 x_1 - y_2 x_1 - y_1 x_1 - y_1 x_1 + y_1 x_2 + y_1 x_1}{x_2 - x_1}$$

$$\hat{2b} = \frac{2y_1 x_2 - 2y_2 x_1}{x_2 - x_1}$$

 \Rightarrow

$$\boxed{\hat{b} = \frac{y_1 x_2 - y_2 x_1}{x_2 - x_1}}$$

$$\bullet \hat{a} = f_b(y_1, y_2)$$

$$\sigma_{\hat{b}}^2 = \left(\frac{\partial f_b}{\partial y_1} \right)^2 \sigma_{y_1}^2 + \left(\frac{\partial f_b}{\partial y_2} \right)^2 \sigma_{y_2}^2$$

$$\sigma_{y_1} = \sigma_{y_2} = \sigma$$

$$= \left(\frac{x_2}{x_2 - x_1} \sigma \right)^2 + \left(\frac{-x_1}{x_2 - x_1} \sigma \right)^2 = \left[\frac{x_1^2 + x_2^2}{(x_2 - x_1)^2} \sigma^2 \right] = \sigma_{\hat{b}}^2$$

$$\bullet \hat{a} = f_a(y_1, y_2)$$

$$\sigma_{\hat{a}}^2 = \left(\frac{\partial f_a}{\partial y_1} \right)^2 \sigma_{y_1}^2 + \left(\frac{\partial f_a}{\partial y_2} \right)^2 \sigma_{y_2}^2$$

$$= \left(\frac{-1}{x_2 - x_1} \sigma \right)^2 + \left(\frac{+1}{x_2 - x_1} \sigma \right)^2 = \left[\frac{2\sigma^2}{(x_2 - x_1)^2} \right] = \sigma_{\hat{a}}^2$$

$$\bullet \text{ on } a \quad y_2 = \hat{a} x_2 + \hat{b} = g(\hat{a}, \hat{b})$$

$$\sigma_{y_2}^2 = \sigma^2 = \left(\frac{\partial g}{\partial \hat{a}} \right)^2 \sigma_{\hat{a}}^2 + \left(\frac{\partial g}{\partial \hat{b}} \right)^2 \sigma_{\hat{b}}^2 + 2 \frac{\partial g}{\partial \hat{a}} \frac{\partial g}{\partial \hat{b}} \sigma_{\hat{a}\hat{b}}$$

$$\text{def } \sigma^2 = x_2^2 \sigma_a^2 + \sigma_b^2 + 2x_2 \sigma_{ab}$$

$$= x_2^2 \frac{2\sigma^2}{(x_2 - x_1)^2} + \frac{x_1^2 + x_2^2}{(x_2 - x_1)^2} \sigma^2 + 2x_2 \sigma_{ab}$$

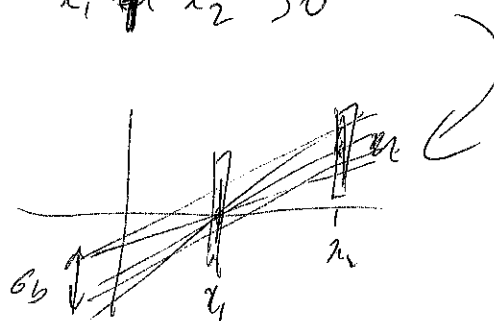
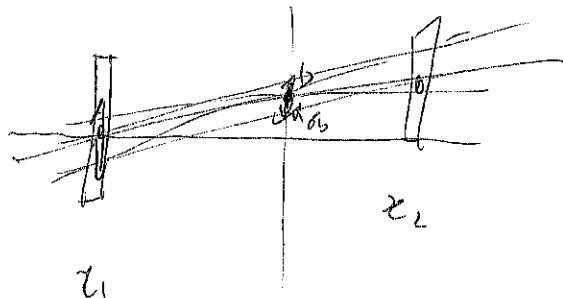
$$\text{def } 2x_2 \sigma_{ab} = \frac{2x_2^2 + x_1^2 + x_2^2 - (x_2 - x_1)^2}{(x_2 - x_1)^2} \sigma^2$$

$$= \frac{2x_2^2 + x_1^2 + x_2^2 - x_1^2 - x_2^2 + 2x_1 x_2}{(x_2 - x_1)^2} \sigma^2$$

$$\Rightarrow \boxed{\sigma_{ab} = - \frac{x_2 + x_1}{(x_2 - x_1)^2} \sigma^2}$$

Uncorrelated when $x_2 = -x_1$

Anti-correlated $\sigma_{ab} < 0$ when $x_1 \neq x_2 > 0$



• Estimator are efficient (and unbiased)

(4)

$$\text{Var}[\hat{\theta}] = -E\left[\frac{\partial^2 \ln L}{\partial \theta^2}\right]^{-1} \quad (\text{RCF bound})$$

$$\chi^2 = -2 \ln L$$

$$\bullet \quad \frac{\partial^2 \chi^2}{\partial \sigma^2} = + \frac{2(x_1^2 + x_2^2)}{\sigma^2} = -2 \frac{\partial^2 \ln L}{\partial \sigma^2} \quad - \frac{\partial^2 \ln L}{\partial \sigma^2} = + \frac{(x_1^2 + x_2^2)}{\sigma^2}$$

$$\bullet \quad \frac{\partial^2 \chi^2}{\partial b^2} = 2 \frac{\partial^2 \ln L}{\partial b^2} = + \frac{4}{\sigma^2} \quad - \frac{\partial^2 \ln L}{\partial b^2} = + \frac{2}{\sigma^2}$$

$$\bullet \quad \frac{\partial^2 \chi^2}{\partial \sigma \partial b} = \frac{2(x_1 + x_2)}{\sigma^2} = -2 \frac{\partial^2 \ln L}{\partial \sigma \partial b} \quad - \frac{\partial^2 \ln L}{\partial \sigma \partial b} = + \frac{x_1 + x_2}{\sigma^2}$$

Waring multidimensional case $\ln L(\vec{\sigma}, \vec{b})$
RCF bound for efficient estimators

$$\text{Cov}_{ab} = -E\left[\frac{\partial^2 \ln L}{\partial a \partial b}\right]^{-1}$$

matrix of 2nd derivatives = (Hesse Matrix)

$$= \begin{pmatrix} + \frac{x_1^2 + x_2^2}{\sigma^2} & + \frac{x_1 + x_2}{\sigma^2} \\ + \frac{x_1 + x_2}{\sigma^2} & + \frac{2}{\sigma^2} \end{pmatrix}$$

$$\text{Cov} = \frac{1}{\sigma^4} \begin{pmatrix} \frac{x_1^2 + x_2^2}{\sigma^2} & \frac{x_1 + x_2}{\sigma^2} \\ \frac{x_1 + x_2}{\sigma^2} & \frac{2}{\sigma^2} \end{pmatrix}^{-1} = \Delta^{-1}$$

$$\det(\Delta) = \frac{2(x_1^2 + x_2^2) - (x_1 + x_2)^2}{\sigma^4} = \frac{(x_2 - x_1)^2}{\sigma^4}$$

$$\text{Cov} = \Delta^{-1} = \frac{1}{\det(\Delta)} (\text{Cov}(\Delta))^T = \frac{\sigma^4}{(x_2 - x_1)^2} \begin{pmatrix} \frac{2}{\sigma^2} & -\frac{x_1 + x_2}{\sigma^2} \\ -\frac{x_1 + x_2}{\sigma^2} & \frac{x_1^2 + x_2^2}{\sigma^2} \end{pmatrix} = \begin{pmatrix} \frac{2\sigma^2}{(x_2 - x_1)^2} & -\frac{x_1 + x_2}{(x_2 - x_1)^2} \\ -\frac{x_1 + x_2}{(x_2 - x_1)^2} & \frac{x_1^2 + x_2^2}{(x_2 - x_1)^2} \end{pmatrix}$$

(5)

$$\text{Cov}_{ab} = \begin{pmatrix} \sigma_a^2 & \sigma_{ab} \\ \sigma_{ab} & \sigma_b^2 \end{pmatrix} = \begin{pmatrix} 2 & -(x_1 + x_2) \\ -(x_1 + x_2) & x_1^2 + x_2^2 \end{pmatrix} \times \frac{\sigma^2}{(x_2 - x_1)^2}$$

c.e.:

$$\begin{aligned} \sigma_a^2 &= \frac{2\sigma^2}{(x_2 - x_1)^2} \sigma^2 \\ \sigma_b^2 &= \frac{x_1^2 + x_2^2}{(x_2 - x_1)^2} \sigma^2 \\ \sigma_{ab} &= -\frac{x_1 + x_2}{(x_2 - x_1)^2} \sigma^2 \end{aligned}$$

ok! identical
to direct estimation
(i.e. LS estimators
are efficient)

$$n=2$$

$$\hat{a} = \frac{n \sum_i x_i y_i - \bar{x} \bar{y} \sum_i x_i y_i}{n \sum x_i^2 - (\sum x_i)^2}$$

$$= \frac{2(x_1 y_1 + x_2 y_2) - (x_1 y_1 + x_1 y_2 + x_2 y_1 + x_2 y_2)}{2(x_1^2 + x_2^2) - (x_1 + x_2)^2}$$

$$= \frac{x_1 y_1 + x_2 y_2 - x_1 y_2 - x_2 y_1}{2(x_1^2 + x_2^2) - (x_1^2 + x_2^2 + 2x_1 x_2)}$$

$$= \frac{x_1 (y_1 - y_2) + x_2 (y_2 - y_1)}{(x_1 - x_2)^2} = \frac{(x_1 - x_2)(y_1 - y_2)}{(x_1 - x_2)^2}$$

$$\boxed{\frac{1}{a} = \frac{y_1 - y_2}{x_1 - x_2}}$$