

Least-squares Adjustment for Linear Adjustment Problems

Linear functional model for the unknowns:

$$\begin{aligned} L_1 &= a_{11}X_1 + a_{12}X_2 + \cdots + a_{1u}X_u \\ L_2 &= a_{21}X_1 + a_{22}X_2 + \cdots + a_{2u}X_u \\ &\vdots \\ L_n &= a_{n1}X_1 + a_{n2}X_2 + \cdots + a_{nu}X_u \end{aligned}$$

Vector of observations:

$$\mathbf{L}_{n,1} = [L_1 \quad L_2 \quad \cdots \quad L_n]^T$$

Variance covariance matrix of the observations:

$$\mathbf{\Sigma}_{n,n} = \begin{bmatrix} \sigma_{L_1}^2 & \sigma_{L_1L_2} & \cdots & \sigma_{L_1L_n} \\ \sigma_{L_2L_1} & \sigma_{L_2}^2 & \cdots & \sigma_{L_2L_n} \\ \vdots & \vdots & \ddots & \vdots \\ \sigma_{L_nL_1} & \sigma_{L_nL_2} & \cdots & \sigma_{L_n}^2 \end{bmatrix} \quad \text{with theoretical values } \sigma_i$$

$$\mathbf{S}_{n,n} = \begin{bmatrix} s_{L_1}^2 & s_{L_1L_2} & \cdots & s_{L_1L_n} \\ s_{L_2L_1} & s_{L_2}^2 & \cdots & s_{L_2L_n} \\ \vdots & \vdots & \ddots & \vdots \\ s_{L_nL_1} & s_{L_nL_2} & \cdots & s_{L_n}^2 \end{bmatrix} \quad \text{with empirical values } s_i$$

Theoretical reference standard deviation:

$$\sigma_0 \quad (\text{or theoretical reference variance } \sigma_0^2)$$

Cofactor matrix of the observations:

$$\mathbf{Q}_{n,n} = \frac{1}{\sigma_0^2} \mathbf{\Sigma}_{n,n} \quad \text{respectively} \quad \mathbf{Q}_{n,n} = \frac{1}{\sigma_0^2} \mathbf{S}_{n,n}$$

Weight matrix of the observations:

$$\mathbf{P}_{n,n} = \mathbf{Q}_{n,n}^{-1}$$

Vector of adjusted unknowns:

$$\hat{\mathbf{X}}_{u,1} = [\hat{X}_1 \quad \hat{X}_2 \quad \cdots \quad \hat{X}_u]^T$$

Matrix of coefficients of the linear functional model:

$$\mathbf{A}_{n,u} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1u} \\ a_{21} & a_{22} & \cdots & a_{2u} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nu} \end{bmatrix} \quad \text{“Design Matrix”}$$

Vector of residuals:

$$\mathbf{v}_{n,1} = [v_1 \quad v_2 \quad \cdots \quad v_n]^T$$

Observation equations:

$$\mathbf{L}_{n,1} + \mathbf{v}_{n,1} = \mathbf{A}_{n,u} \hat{\mathbf{X}}_{u,1}$$

Normal equations:

$$\mathbf{A}_{u,n}^T \mathbf{P}_{n,n} \mathbf{A}_{n,u} \hat{\mathbf{X}}_{u,1} = \mathbf{A}_{u,n}^T \mathbf{P}_{n,n} \mathbf{L}_{n,1}$$

Normal matrix:

$$\mathbf{N}_{u,u} = \mathbf{A}_{u,n}^T \mathbf{P}_{n,n} \mathbf{A}_{n,u}$$

Right hand side of normal equations:

$$\mathbf{n}_{u,1} = \mathbf{A}_{u,n}^T \mathbf{P}_{n,n} \mathbf{L}_{n,1}$$

Normal equations:

$$\mathbf{N}_{u,u} \hat{\mathbf{X}}_{u,1} = \mathbf{n}_{u,1}$$

Inversion of normal matrix:

$$\mathbf{Q}_{\hat{\mathbf{x}}\hat{\mathbf{x}}} = \mathbf{N}_{u,u}^{-1}$$

Solution for the unknowns:

$$\hat{\mathbf{X}}_{u,1} = \mathbf{Q}_{\hat{\mathbf{x}}\hat{\mathbf{x}}} \mathbf{n}_{u,1}$$

Vector of residuals:

$$\mathbf{v}_{n,1} = \mathbf{A}_{n,u} \hat{\mathbf{X}}_{u,1} - \mathbf{L}_{n,1}$$

Vector of adjusted observations:

$$\hat{\mathbf{L}}_{n,1} = \mathbf{L}_{n,1} + \mathbf{v}_{n,1}$$

Final check:

$$\hat{\mathbf{L}}_{n,1} \stackrel{!}{=} \Phi_{n,1}(\hat{\mathbf{X}}_{u,1})$$

Empirical reference standard deviation:

$$s_0 = \sqrt{\frac{\mathbf{v}^T \mathbf{P} \mathbf{v}}{n-u}} \quad (\text{or empirical reference variance } s_0^2)$$

Cofactor matrix of adjusted unknowns:

$$\mathbf{Q}_{\hat{\mathbf{x}}\hat{\mathbf{x}}}$$

VCM of adjusted unknowns:

$$\Sigma_{\hat{\mathbf{x}}\hat{\mathbf{x}}} = s_0^2 \mathbf{Q}_{\hat{\mathbf{x}}\hat{\mathbf{x}}}$$

Cofactor matrix of adjusted observations:

$$\mathbf{Q}_{\hat{\mathbf{L}}\hat{\mathbf{L}}} = \mathbf{A}_{n,u} \mathbf{Q}_{\hat{\mathbf{x}}\hat{\mathbf{x}}} \mathbf{A}_{u,n}^T$$

VCM of adjusted observations:

$$\Sigma_{\hat{\mathbf{L}}\hat{\mathbf{L}}} = s_0^2 \mathbf{Q}_{\hat{\mathbf{L}}\hat{\mathbf{L}}}$$

Cofactor matrix of the residuals:

$$\mathbf{Q}_{\mathbf{v}\mathbf{v}} = \mathbf{Q}_{\mathbf{L}\mathbf{L}} - \mathbf{Q}_{\hat{\mathbf{L}}\hat{\mathbf{L}}}$$

VCM of the residuals:

$$\Sigma_{\mathbf{v}\mathbf{v}} = s_0^2 \mathbf{Q}_{\mathbf{v}\mathbf{v}}$$