

Least squares solution of \mathbf{a} (vector) including system identification coefficients $C_{i,j}$, $m_{i,j}$ and $r_{i,j}$

Please see details for ref. [99],2012

State space equation,
Simultaneous ordinary differential
equations for state space vector \mathbf{x}

$$\mathbf{M} \cdot \dot{\mathbf{x}} = \mathbf{C} \cdot \mathbf{x} + \mathbf{C}_0 \cdot \mathbf{x}_0 + \mathbf{R} \cdot \mathbf{g} \quad (2)$$

Regression equation
for vector \mathbf{a}

$$\mathbf{y}_k = \mathbf{Z}_k \cdot \mathbf{a} \quad (3)$$

Least squares from state space equation

$$\left[\sum_{k=1}^{nt} {}^t \mathbf{Z}_k \cdot \mathbf{Z}_k \right] \cdot \mathbf{a} = \sum_{k=1}^{nt} {}^t \mathbf{Z}_k \cdot \mathbf{y}_k \quad (4)$$

A system of ordinary differential equations is constructed from the nodal equations, and a regression equation for \mathbf{a} is constructed by transforming this. This yields the first-stage least squares solution.

On the other hand, the first-stage least squares solution can also be obtained from the constraint expressions such as the symmetry of $c_{i,j}=c_{j,i}$.

By arranging these two equations in the direction of increasing rows, the second-stage least squares solution equation is derived.

Therefore, a double least squares method is applied.

Constraints of \mathbf{a}

$$\mathbf{S} \cdot \mathbf{a} = \mathbf{d} \quad (5)$$

Least squares from constraints

$${}^t \mathbf{S} \cdot \mathbf{S} \cdot \mathbf{a} = {}^t \mathbf{S} \cdot \mathbf{d} \quad (5)$$

$$\mathbf{F} = \begin{bmatrix} \sum_{k=1}^{nt} {}^t \mathbf{Z}_k \cdot \mathbf{Z}_k \\ {}^t \mathbf{S} \cdot \mathbf{S} \end{bmatrix} \quad (6) \quad \mathbf{b} = \begin{bmatrix} \sum_{k=1}^{nt} {}^t \mathbf{Z}_k \cdot \mathbf{y}_k \\ {}^t \mathbf{S} \cdot \mathbf{d} \end{bmatrix} \quad (7)$$

Least squares solution of \mathbf{a}

$$\mathbf{J} = {}^t (\mathbf{b} - \mathbf{F} \cdot \mathbf{a}) \cdot \mathbf{W} \cdot (\mathbf{b} - \mathbf{F} \cdot \mathbf{a}) \quad (8)$$

$$\hat{\mathbf{a}} = \left({}^t \mathbf{F} \cdot \mathbf{W} \cdot \mathbf{F} \right)^{-1} \cdot \left({}^t \mathbf{F} \cdot \mathbf{W} \cdot \mathbf{b} \right) \quad (9)$$

(Here \mathbf{W} is a weighting matrix

that makes the maximum value of each row to 1)

The deduction of equation (5) will be explained in a later slide.