

Partial Least squares regression and Optimal Control

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Abstract

The Partial Least Squares (PLS) algorithm is formulated as an optimal control problem.

1 PLS and optimal control

1.1 An optimal control problem

Given a dynamic system

$$x_{k+1} = Ax_k + Bu_k, \quad (1)$$

$$y_k = Dx_k, \quad (2)$$

where $k = 0, 1, \dots$, is discrete time and with an initial state $x_0 = 0$. Consider a control objective criterion weighting only at the final time instant, say $k = a$.

$$J_0 = (r - y_a)^T (r - y_a). \quad (3)$$

The optimal controls minimizing the control objective is given by

$$x_k = A^k x_0 + C_k u_{0|k}, \quad (4)$$

where here C_k is the reversed controllability matrix of the pair (A, B) . Putting this into the objective gives

$$J_0 = (r - DC_a u_{0|a})^T (r - DC_a u_{0|a}). \quad (5)$$

The optimal control is then

$$u_{0|a}^* = (C_a^T D^T DC_a)^{-1} C_a^T D^T r. \quad (6)$$

And the corresponding optimal state and output is given by

$$x_a^* = C_a u_{0|a}^* = C_a (C_a^T D^T DC_a)^{-1} C_a^T D^T r, \quad (7)$$

$$y_a^* = Dx_a^* = DC_a (C_a^T D^T DC_a)^{-1} C_a^T D^T r. \quad (8)$$

1.2 PLS as an optimal control problem

Given a dynamic system

$$B_{k+1} = X^T X B_k + X^T X p_k, \quad (9)$$

$$y_k = X B_k, \quad (10)$$

where $k = 0, 1, \dots, a$ is discrete time and with an initial state $B_0 = 0$. Consider a control objective criterion weighting only at the final time instant at $k = a$.

$$J_0 = (Y - y_a)^T (Y - y_a). \quad (11)$$

The optimal controls minimizing the control objective is given by

$$p_{0|a}^* = (K_a^T X^T X K_a)^{-1} K_a^T X^T Y. \quad (12)$$

And the corresponding optimal state and output is given by

$$B_a^* = K_a p_{0|a}^* = K_a (K_a^T X^T X K_a)^{-1} K_a^T X^T Y, \quad (13)$$

$$y_a^* = X B_a^* = X K_a (K_a^T X^T X K_a)^{-1} K_a^T X^T Y. \quad (14)$$