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- ۱. Intercept
- ۲. System Approach

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x_n x_1

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y

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$$y(t) = \sum_{i=1}^m \phi_i(x_1(t), x_2(t), \dots, x_n(t))$$

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۲. Interval Regression
۳. Generalized Regression
۴. Fuzzy Regression
۵. Auto Regressive (AR)

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$$y(t) = \sum \alpha_i y(t-i) + \sum \gamma_i u(t-i) + \sum \theta_i w(t-i) \quad (1)$$

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$$y(t) = \sum \alpha_i y(t-i) + \sum \gamma_i u(t-i) + \sum \theta_i w(t-i) \quad (2)$$

:

$$y(t) = G(q)u(t) + H(q)w(t) \quad (3)$$

:

$$G(q) = \frac{B(q)}{A(q)}; H(q) = \frac{C(q)}{D(q)}$$

$$u(t) \quad \cdot \quad () \quad D(q) \quad A(q)$$

w(t)

$$\cdot \quad () \quad x_k \quad y$$

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۱. Delay (Shift) Operator

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- १. Closure to Time Sementation
- २. Abstract Object
- ३. Methodology

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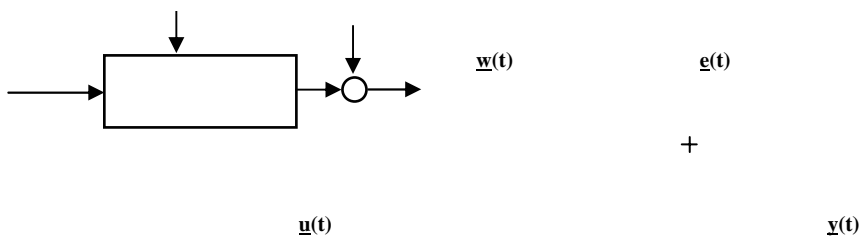
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\. Intuitive

$$\dot{\underline{x}}(t) = \underline{f}(\underline{x}(t), \underline{u}(t), t) \quad ()$$

$$\underline{y}(t) = \underline{h}(\underline{x}(t), \underline{u}(t), t) \quad ()$$



$$\underline{y}(t) = \underline{f}(\underline{u}(t), \underline{u}(t-1), \dots, \underline{y}(t-1), y(t-\nu), \dots, \underline{w}(t), \underline{w}(t-1), \dots, \underline{e}(t), \underline{e}(t-1), \dots; t) \quad ()$$

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\underline{u}

$$: \quad () \quad ()$$

$$\underline{y}(t) = \Phi(\underline{u}(t), \underline{u}(t-1), \dots, \underline{y}(t-1), y(t-\nu), \dots) \underline{\theta} + \underline{v}(t) \quad ()$$

$\underline{y}(t)$

$\underline{\theta}$

Φ

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$$\underline{\varphi}(t) \quad (MISO)$$

$$: \quad () \quad ()$$

$$\underline{\varphi}(t) = \begin{bmatrix} \varphi_1(\underline{u}(t), \underline{u}(t-1), \dots, \underline{y}(t-1), \dots) \\ \dots \\ \varphi_d(\underline{u}(t), \underline{u}(t-1), \dots, \underline{y}(t-1), \dots) \end{bmatrix}_{d \times 1} \quad ()$$

$$\underline{y}(t) = \underline{\varphi}(t)^T \underline{\theta} + \underline{v}(t) \quad ()$$

- 1. Typical
- 2. Multi-Input Single-Output

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$$y(t) = \sum_{i=1}^{na} \alpha_i y(t-i) + \sum_{j=1}^m \sum_{i=1}^{nb_j-1} \beta_{ij} u_j(t-i) + v(t) \quad ()$$

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: $d = na + \sum nb_j$

$$\underline{\theta} = [\alpha_1, \alpha_2, \dots, \alpha_{na}, \beta_{11}, \dots, \beta_{nb_1,1}, \dots, \beta_{1r}, \dots, \beta_{nb_m,m}] \quad ()$$

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forall

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$$f(\underline{x}(t), \underline{u}(t), t) = \dot{\underline{x}} \quad (1)$$

$$(\underline{x}, \underline{u})$$

$$(2) \quad f(\dots) \quad (\underline{x} + \delta \underline{x}, \underline{u} + \delta \underline{u})$$

\underline{x}

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$$(D(q) \quad A(q) \quad)$$

$$() \quad ()$$

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$$\underline{u} \in L_p \Rightarrow \underline{y} \in L_p \quad (3)$$

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L_p

(p

$$\underline{u}(t) = \dot{\underline{x}} \quad (4)$$

$$\underline{u}(t) \neq \dot{\underline{x}}$$

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1. Equilibrium

2. p-norm

$$\begin{aligned}
 & \left(\begin{array}{c} \vdots \\ y \\ \vdots \end{array} \right) = \left(\begin{array}{c} \vdots \\ y \\ \vdots \end{array} \right) + \left(\begin{array}{c} \vdots \\ y(t) \\ \vdots \end{array} \right) \quad (1) \\
 & \underline{u} = \underline{y} + \underline{u}(t) \quad (2) \\
 & \vdots \\
 & y^m(t) = y + y(t) \quad (3) \\
 & \underline{u}^m(t) = \underline{u} + \underline{u}(t) \quad (4) \\
 & \underline{u} = y \quad \text{"} \quad \text{"} \quad \text{m} \\
 & \vdots \\
 & \left(\begin{array}{c} \vdots \\ y \\ \vdots \end{array} \right) : \left(\begin{array}{c} \vdots \\ y \\ \vdots \end{array} \right) \quad (\text{MISO}) \\
 & y^m(t) - y = \sum_{i=1}^{na} \alpha_i (y^m(t-i) - y) + \sum_{j=1}^m \sum_{i=0}^{nb_j-1} \beta_{ij} (u_j^m(t-i) - u_{i,j}) + v(t) \quad (5) \\
 & \vdots \\
 & y^m(t) = \alpha + \sum_{i=1}^{na} \alpha_i y^m(t-i) + \sum_{j=1}^m \sum_{i=0}^{nb_j-1} \beta_{ij} u_j^m(t-i) + v(t) \quad (6) \\
 & \left(\begin{array}{c} \vdots \\ y \\ \vdots \end{array} \right) : \\
 & \alpha = y \cdot \left(1 - \sum_i \alpha_i \right) + \sum_j u_{i,j} \cdot \sum_i \beta_{ij} \quad (7) \\
 & \underline{u} = \left(\begin{array}{c} \vdots \\ y \\ \vdots \end{array} \right) \quad \left(\begin{array}{c} \vdots \\ y \\ \vdots \end{array} \right) \quad (8)
 \end{aligned}$$

1. Operating Point

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$$a(q)y(t) = \underline{b}^T(q)\underline{u}(t) + v(t) \quad ()$$

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$$a(q) = 1 - \sum \alpha_i q^i$$

$$\underline{b}(q) = [b_1(q), \dots, b_m(q)]^T \quad ()$$

$$b_j(q) = \sum \beta_{ij} q^i$$

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$$y_s = \lim_{t \rightarrow \infty} y(t) = \lim_{q \rightarrow 1} \frac{\underline{b}^T(q)\underline{u}}{a(q)} \quad ()$$

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$$\lim_{q \rightarrow 1} G(q) : ()$$

Step)

$$y^m(t) = \alpha_0 + \alpha_1 t + \sum_{i=1}^{na} a_i y^m(t-i) + \sum_{j=1}^m \sum_{i=1}^{nb_j-1} \beta_{ij} u_j^m(t-i) + v(t) \quad ()$$

$$u_j(t) = t \quad ()$$

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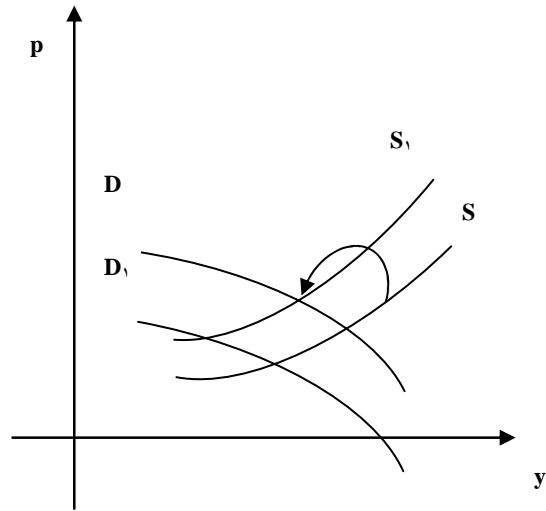
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E_1 E_2

(E_1)

S_1 D_1



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$$\dot{\underline{x}} = \underline{f}(\underline{x}, \underline{u}) = \cdot$$

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$(\underline{x}_1, \underline{u}_1)$

\underline{x}_2

E_2

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\underline{x}_1

E_1

\underline{u}_1

\. Trajectory

$$(\underline{x}, \underline{u})$$

$$\dot{\underline{x}}$$

$$(\quad)$$

$$.[\]$$

$$(\quad)$$

$$(\quad)$$

E_i

$$[\]$$

$$[\]$$

$$E_{i+1} \quad E_i$$

:

$$\begin{bmatrix} \dot{p} \\ \dot{p} \\ \dot{y} \end{bmatrix} = \underline{f}(p(t), y(t), \underline{u}(t), t)$$

$$(\quad)$$

1. Nominal Trajectory

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(E_t-E_{t+1}) E_t-E_t

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$$\frac{dp}{dy} = g(p, y; \underline{u}) \left\{ = \frac{f_1}{f_2} \right\} \quad ()$$

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g(•)

g(•)

u

:

p=p(y; u)

$$p = \int g(p, y; \underline{u}) dy = \int [c + g_1(p, y; \underline{u})] dy = cy + \int g_2(p, y; \underline{u}) dy \quad ()$$

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(p=•)

(y=•)

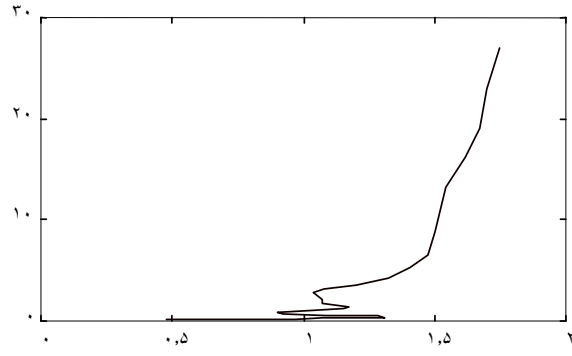
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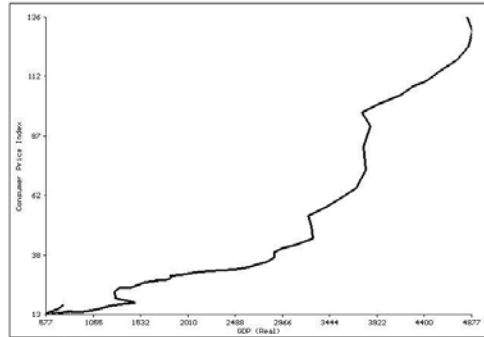
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($y_1 = \Delta y / y$:)

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m ()

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$$m : c(t) = \alpha_0 + \sum \alpha_i c(t-i) + \sum \beta_i y_n(t-i) + \sum \beta_{io} y_o(t-i) + \sum \gamma_i a(t-i) + \sigma e(t-1) + e(t) \quad ()$$

() a(t) y_o(t) y_N(t) c(t)

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1. Fuzzy Smoothing

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y.(t)

α. t-std

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i	t-std $\alpha_{(min)}$	$\alpha.$	α_1	α_2	β_1	β_2	γ_1	γ_2	σ
1	1,2124	-0,0018	-0,8042	-	0,9545	-0,8325	0,1806	-0,1075	-0,9990
2	2,5678	-	-0,7823	-	0,9597	-0,8324	0,1893	-0,0979	-0,9987
3	0,8419	-0,0065	-0,1457	0,2340	0,8243	-	0,2129	-	-0,9699
4	1,3307	0,0041	-0,8138	0,1205	0,9043	-0,7289	0,1317	-	-0,8880
5	0,5526	-0,0059	0,1419	-	0,8844	-	0,2017	-	0,1959
6	1,2730	-	0,1471	-	0,8701	-	0,2240	-	0,2455
7	0,6882	-	-0,1264	0,2301	0,8164	-	0,2365	-	-0,0227
8	0,3419	-	0,1073	0,2001	0,8048	0,1750	0,3706	-0,1082	0,3011
9	1,5960	-0,0245	0,0144	0,1282	0,8852	0,1382	-	-	0,2804
10	0,3693	-0,0155	0,2570	0,2029	0,8264	0,3135	0,3967	-0,1518	0,4205
11	1,3262	-	-0,7324	0,1093	0,9374	-0,6861	0,1162	-	-0,9000

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8. D.W. Boyd (2001); **System Analysis and Modeling**, Academic Press.
 9. J.E.H. Davidson, D.F. Hendry, F. Srba and S. Yeo (1978); Econometric Modelling of the Aggregate Time-Series Relationship Between Consumer's Expenditure and Income in the United Kingdom", **The Economic Journal**, No. 88.
 10. N. Gershenfeld (1999)' **The Nature of Mathematical Modeling**, Cambridge University Press, 1999.

۱۱. D.F. Hendry (۱۹۹۵); **Dynamic Econometrics**, Oxford.
۱۲. S.R. King, R.M. McConnell (۱۹۹۳); **Macro Solve ver. ۴**, W.W.Norton & Company.
۱۳. W. Kramer (۱۹۸۵); "The Power of the Durbin-Watson Test for Regressions Without an Intercept", **J. of Econometrics**, Vol. ۲۸, No. ۳, pp. ۳۶۳-۳۷۰, June.
۱۴. L. Ljung (۱۹۸۷); **System Identification: Theory for the User**, Prentice Hall.
۱۵. D.P. Maki, M. Thompson (۱۹۷۳); **Mathematical Modeling and Applications**, Prentice Hall.
۱۶. P. Perron (۱۹۹۱); "A Continuous Time Approximation to the Unstable First-Order Autoregressive Process: The Case Without an Intercept", **Econometrica**, Vol. ۵۹, No. ۱, pp. ۲۱۱-۲۳۶, Jan.
۱۷. H. Shakouri G., K. Y. Nikraves (۲۰۰۰); "A New Approach in Model Selection Using Fuzzy Decision-Making: Trade off Between Possibility & Probability Theories", **Proc. SMC' ۲۰۰۰**.
۱۸. M. Vidyasagar (۱۹۹۳); **Nonlinear System Analysis**, Prentice Hall.
۱۹. L.A. Zadeh (۱۹۶۷); **System Theory**, Mc Graw Hill.