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## ROBUST ESTIMATION IN DYNAMYC SYSTEMS SIMULATION

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In the course of market relations' development the demand for new economic and mathematical tool formation has appeared for both analysis of economical dynamics and a strategy generation of economic process management. The robust algorithm of ellipsoidal state estimation of dynamic systems is applied to search for the solution to the current economical situation.

Let the evolution of discrete dynamic system and its measuring line are described by equations

$$X_{k+1} = A_k X_k + f_k(u_k),$$
 (1)

$$y_k = h_k^T x_k + \xi_k. \tag{2}$$

Here  $k=0,1,\ldots$ —discrete time;  $x_k$ —n-dimensional state vector inaccessible to immediate change,  $x_k \in R_n$ ;  $R_n$ —n-dimensional substantial Euclidean space;  $u_k$ —known control vector,  $u_k \in R_m$ ,  $m \le n$ ;  $y_k$ —vector of observed signals, which are the results of measuring of outcome of object  $h_k^T x_k$ , deadened by action of additive disturbance  $\xi_k$ ,  $y_k^T = (y_{1k}, \ldots, y_{mk}), \; \xi_k^T = (\xi_{1k}, \ldots, \xi_{mk}); \; T$ —transposition operation symbol;

$$A_k$$
 and  $h_k^T = \begin{pmatrix} h_{1k}^T \\ \dots \\ h_{mk}^T \end{pmatrix}$  - known n×n and m×n order matrixes accordingly, det

 $A_k \neq 0 \ \forall_k = 0, 1, ..., h_{ik} \in R_n$ ;  $f_k(u_k)$  — given n-dimensional function. It is assumed that disturbance  $\xi_k$  satisfies component or standard constraint:

$$\xi_{ik}^{2} \le c_{ip}^{2}, i = 1,..., m, k = 0, 1, ...;$$
 (3)

$$(\xi_{ik} - \xi_{i,k-1})^2 \le c_{ir}^2, i = 1,..., m, k = 1, 2, ...;$$
 (4)

$$\xi_{k}^{T}\xi_{k} \le c_{p}^{2}, k = 0, 1, ...;$$
 (5)

$$(\xi_{k} - \xi_{k-1})^{T}(\xi_{k} - \xi_{k-1}) \le c_{r}^{2}, k = 1, 2, ...,$$
 (6)

where  $c_{ip}$ ,  $c_i$ ,  $c_{ir}$ ,  $c_r$ —given constants.

Assumed that the results of current measuring  $y_k$ , k = 0, 1, ..., are known, there is a problem of ellipsoid estimation of unknown state vector  $x_k$ 

$$\mathbf{X}_{k} \in \mathbf{E}[\hat{x}_{k}, \mathbf{H}_{k}] = \mathbf{E}_{k},\tag{7}$$

where E[ $\hat{x}_k$ , H<sub>k</sub>] ={x:  $\sigma(x, \hat{x}_k, H_k) \le 1$ },  $\sigma(x, \hat{x}_k, H_k) = (x - \hat{x}_k)^T H_k^{-1}(x - \hat{x}_k)$ ,  $\hat{x}_k$  — ellipsoid centre, H<sub>k</sub> = H<sub>k</sub><sup>T</sup> > 0. The characteristic of prior ellipsoid  $\hat{x}_k$ , H<sub>k</sub> when k = 0 is considered to be given.

From equalizations (1), (2) we have

$$\xi_{k} = y_{k} - h_{k}^{T} x_{k}, \tag{8}$$

$$\xi_{k-1} = y_{k-1} - h_{k-1}^{T} x_{k-1}. \tag{9}$$

Having used equalization (1), from (8) and (9) we found

$$\xi_{k} - \xi_{k-1} = z_{k} - r_{k}^{T} x_{k}, \tag{10}$$

where  $\mathbf{z}_{k} = (\mathbf{z}_{1k}, \dots, \mathbf{z}_{mk})^{T} = \mathbf{y}_{k} - \mathbf{y}_{k-1} - \mathbf{h}_{k-1}^{T} \mathbf{A}_{k-1}^{-1} \mathbf{f}_{k-1} (\mathbf{u}_{k-1}), \ \mathbf{r}_{k}^{T} = \begin{pmatrix} \mathbf{r}_{1k}^{T} \\ \dots \\ \mathbf{r}_{mk}^{T} \end{pmatrix} = \mathbf{h}_{k}^{T} - \mathbf{h}_{k-1}^{T} \mathbf{A}_{k-1}^{-1}.$ 

Let  $\zeta_k = \xi_k - \xi_{k-1}$ , relation (10) can be represented as additional measuring line of system (1)

$$\mathbf{z}_{\mathbf{k}} = \mathbf{r}_{\mathbf{k}}^{\mathsf{T}} \mathbf{x}_{\mathbf{k}} + \boldsymbol{\zeta}_{\mathbf{k}},\tag{11}$$

where disturbance  $\zeta_k = (\zeta_{1k}, ..., \zeta_{mk})^T$  satisfies component or standard constraint

$$\zeta_{ik}^{2} \le c_{ir}^{2}, i = 1,..., m, k = 1, 2, ...,$$
 (12)

$$\zeta_k^T \zeta_k \le c_r^2, k = 0, 1, ...,$$
 (13)

Let us study component-wise constraint of disturbance. There are sets

$$S_{ip}(x_k) = \{x_k: (y_{ik} - h_{ip}^T x_k)^2 \le c_{ip}^2, k = 0, 1, ..., i = 1, ...m\},$$
(14)

$$S_{ir}(x_k) = \{x_k: (z_{ik} - r_{ik}^T x_k)^2 \le c_{ir}^2, k = 1, 2, ..., i = 1, ...m\},$$
(15)

Relations (14), (15) define in phase space of system (1), (2) sets  $S_{ip}(x_k)$  and  $S_{ir}(x_k)$  of systems states which are compatible with the results of measuring. These sets are 2m hyperbands each of that is restricted by two parallel hyperplanes  $y_{ik}$ - $h_{ik}^{\ \ T}x_k$ = $\pm c_{ir}$ , i =1, ...m.

Apparently, inclusion  $x_k \in D(x_k)$  is true for unknown state vec-

Apparently, inclusion  $x_k \in D(x_k)$  is true for unknown state vector  $x_k$  of system (1), (2), where  $D(x_k)$  is 2m hyperbands intersection of sets (14), (15). Also, according to assumption, inclusion (7) is satisfied. Therefore intersection  $x_k \in E[\widetilde{x}_k, \widetilde{H}_k]$  is also satisfied, where  $E[\widetilde{x}_k, \widetilde{H}_k]$  is ellipsoid like (7), when  $D(x_k) \cap E[\widehat{x}_k, H_k] \subset E[\widetilde{x}_k, \widetilde{H}_k]$  is true. For building ellipsoid  $E[\widetilde{x}_k, \widetilde{H}_k]$  is reasonable to use robust algorithm of successive building of 2m ellipsoids, which contain intersection of ellipsoids and  $S_{ip}(x_k)$ ,  $i=1, \ldots m$  is  $S_{ir}(x_k)$ ,  $i=m+1, \ldots, 2m$ . These ellipsoids can be built according to minimum criterion of determinant (many-ellipsoid dimensional volume) or track of congruent matrixes  $\widetilde{H}_{ik}$ ,  $\widetilde{H}_{2m,k} = \widetilde{H}_k$ . Estimative ellipsoid  $E[\widehat{x}_{k+1}, H_{k+1}]$ , for which inclusion  $x_{k+1} \in E[\widehat{x}_{k+1}, H_{k+1}]$  is satisfied, where  $x_{k+1}$  is unknown state vector of system (1), (2), is direct image of ellipsoid  $E[\widehat{x}_k, \widetilde{H}_k]$  which is linear mapped under equation (1). Ellipsoid  $E[\widehat{x}_{k+1}, H_{k+1}]$  centre and matrix are defined by relations

$$\hat{x}_{k+1} = A_k \widetilde{x}_k + f_k(u_k), \ H_{k+1} = A_k \widetilde{H}_k A_k^T.$$

In case (4) disturbance  $\xi_k$  and the speed of its changing according to system (1), (2) states sets, which are compatible with the results of measuring, are restricting by relations

$$S_{p}(x_{k}) = \{x_{k}: (y_{k} - h_{k}^{T}x_{k})^{T}(y_{k} - h_{k}^{T}x_{k}) \le c_{p}^{2},$$
(16)

$$S_{r}(X_{k}) = \{X_{k} : (Z_{k} - r_{k}^{T}X_{k})^{T}(Z_{k} - r_{k}^{T}X_{k}) \le c_{r}^{2}.$$
(17)

These sets are cylinders in phase space  $R^n$  of system (1), (2). For its unknown state vector the inclusion  $\mathbf{x}_k \in D(\mathbf{x}_k)$  is satisfied, where  $D(\mathbf{x}_k) = \mathbf{S}_p(\mathbf{x}_k) \cap \mathbf{S}_r(\mathbf{x}_k)$ . In accordance with the fact that condition (7) is met inclusion  $\mathbf{x}_k \in D(\mathbf{x}_k) \cap E[\hat{x}_k, H_k]$  follows directly. Therefore estimating ellipsoid, which securely contains unknown vector  $\mathbf{x}_k$ , can be built with the help of intersection ellipsoidal approximation of  $E[\hat{x}_k, H_k]$  and set  $\mathbf{S}_p(\mathbf{x}_k)$  and then intersection approximation of derived ellipsoid and set  $\mathbf{S}_r(\mathbf{x}_k)$ . Also, system (1), (2) state vector estimation boils down to building of parametric ellipsoid family  $E[\tilde{x}_k(v), \tilde{H}_k(v)]$  and  $E[\tilde{x}_k(v), \tilde{H}_k(v)]$  like (7), which contain intersections  $E[\hat{x}_k, H_k] \cap \mathbf{S}_p(\mathbf{x}_k)$  and  $E[\tilde{x}_k(v), \tilde{H}_k(v)] \cap \mathbf{S}_r(\mathbf{x}_k)$  accordingly, where  $v \geq 0$ ,  $\tau \geq 0$  – family parameters. Estimating ellipsoid center  $\hat{x}_{k+1}$  and matrix  $H_{k+1}$ , for which inclusion  $\mathbf{x}_{k+1} \in E[\hat{x}_{k+1}, H_{k+1}]$  is satisfied surely, are defined by relations  $\hat{x}_{k+1} = A_k \tilde{x}_k(v, \tau) + f_k(u_k)$ ,  $H_{k+1} = A_k \tilde{H}_k(v, \tau) A_k^T$ . Robust algorithm of construction of ellipsoid family  $E[\tilde{x}_k(v), \tilde{H}_k(v)]$  is of the form

$$\widetilde{x}_k(\nu) = \hat{x}_k + \nu H_k h_k (I_m + \nu e_k)^{-1} \widetilde{y}_k, \tag{18}$$

$$\widetilde{H}_{k}(v) = \chi_{k}(v)(H_{k} - (1 - \beta_{k}^{2})vH_{k}h_{k}(I_{m} + ve_{k})^{-1}h_{k}^{T}H_{K}),$$
(19)

$$\chi_k(v) = 1 + vc_p^2 + (\alpha_k - 1)\mu(v, e_k, \widetilde{y}_k).$$
 (20)

Here 
$$\mathbf{e}_{\mathbf{k}} = \mathbf{h}_{\mathbf{k}}^{\mathrm{T}} \mathbf{H}_{\mathbf{k}} \mathbf{h}_{\mathbf{k}}, \ \widetilde{\mathbf{y}}_{k} = \mathbf{y}_{k} - \mathbf{h}_{k}^{\mathrm{T}} \hat{\mathbf{x}}_{k}$$
,  $\mu(\mathbf{v}, \mathbf{e}_{k}, \mathbf{y}_{k}) = \mathbf{v} \widetilde{\mathbf{y}}_{k}^{\mathrm{T}} (I_{m} + \mathbf{v} \mathbf{e}_{k})^{-1} \widetilde{\mathbf{y}}_{k}$ 

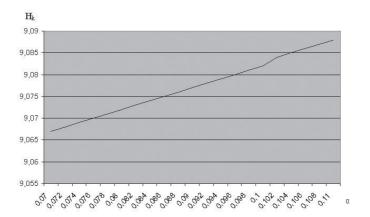


Figure 1 – Determinant det  $H_{10}$   $\alpha$  dependence diagram

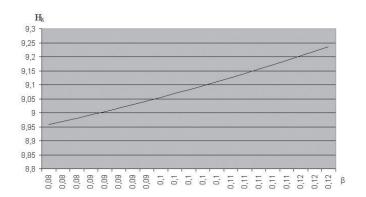


Figure 2 – Determinant det  $H_{10}$   $\beta$  dependence diagram

I<sub>m</sub> – unity m×m order matrix,  $\beta_k \in (0,1)$  and  $\alpha_k > 0$  – algorithm parameters. The centre  $\widetilde{x}_k(\nu,\tau)$  and the matrix  $\widetilde{H}_k(\nu,\tau)$  of ellipsoid  $E[\widetilde{x}_k(\nu,\tau),\widetilde{H}_k(\nu,\tau)] \supset E[\widetilde{x}_k(\nu),\widetilde{H}_k(\nu)] \supset S_r(\mathbf{x}_k)$  is calculating according to formulas (16) – (18) by replacement in their right parts variables  $\nu$ ,  $\widetilde{x}_k(\nu)$ ,  $H_k$ ,  $\mathbf{c}_p$  and  $\widetilde{y}_k = y_k - h_k^T \hat{x}_k$  by variables  $\tau$ ,  $\widetilde{x}_k(\nu)$ ,  $\widetilde{H}_k(\nu)$ ,  $\mathbf{c}_r$  and  $\widetilde{z}_k = z_k - r_k^T \widetilde{x}_k(\nu)$  accordingly.

The algorithm parameters verification is used for model

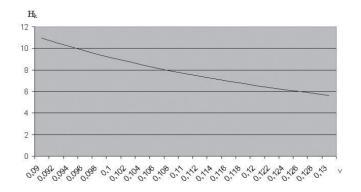


Figure 3 – Determinant det H<sub>10</sub> v dependence diagram

optimization. The determinant det  $H_k$  is fixed on step 10. Initial value  $\alpha=0.1$  varies from 0.07 to 0.11 with fixed lead 0.002,  $\beta=0.1$  varies from 0.08 to 0.12 with fixed lead 0.002. There is a determinant det  $H_{10}$  variation caused by  $\alpha$  and  $\beta$  diagram in the fig. 1-2.

The direct connection can be observed from fig.1-2. Therefore,  $\alpha$  and  $\beta$  should be chosen as minimum.

Ellipsoid family parameter  $\nu = 0.1$  varies from 0.09 to 0.13 with fixed lead 0.002. There is a determinant det  $H_{10}$  variation caused by  $\nu$  diagram in the fig. 3.

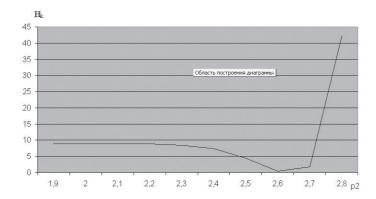


Figure 4 – Determinant det H<sub>10</sub> p<sub>2</sub> dependence diagram

The indirect connection can be observed from fig. 3. Therefore, the bigger v, the smaller determinant det  $H_{10}$ .

The control vector  $\mathbf{u}_k$  parameter  $\mathbf{p}_2$  influence on determinant det  $\mathbf{H}_k$  is also studied. Initial value  $\mathbf{p}_2 = 0.5$  varies from 1.9 to 2.9 with fixed lead 0.1. There is a determinant det  $\mathbf{H}_{10}$  variation caused by  $\mathbf{p}_2$  diagram in the fig. 4.

Judging from pic. 4, it could be seen that when  $p_2$  increases determinant det  $H_{10}$  decreases at first, then possesses the minimum value at the figure of 2.6, and then increases dramatically. However, when  $\pi p \mu p_2 = 2.5$  determinant det  $H_k$  possesses the minimum value on the step 12, then starts growing. Therefore, the best value is  $p_2 = 2.4$ , where determinant det  $H_{10}$  possesses the value 7.364 at initial algorithm parameters. Current dependence can be observed starting from the step 9, as at previous steps the algorithm is inadequate and when  $p_2$  decreases determinant det  $H_k$  keeps on decreasing.

## References

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