

LINEAR DYNAMICAL SYSTEMS OF THE N-TH ORDER
IN RANDOM CONDITIONS

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Abstract: In the article, the linear dynamic random model of n-th order described by the random state equation is considered. The method for determining the probabilistic characteristics of a stochastic process which is the solution of that equation, is shown. These characteristics, such as mean values and correlation functions, are determined by auxiliary solutions – the deterministic systems of ordinary differential equations.

Key words: moments of stochastic processes, n-th order random system.

1. Introduction

Random dynamical systems are the models of many systems occurring in electrical engineering and electronics [10]. They are usually described by the stochastic differential equations [13]. In the article the method of determining the mean value and the correlation function of the stochastic processes is described, these characteristics being the solutions of linear n-th order random differential equations. The presented results are the generalization of previous works [4, 5, 6, 7, 8, 9, 10] concerning this subject.

2. The formalization of the problem.

A random dynamical system described by the state equation is given:

$$\begin{cases} \frac{dX(t)}{dt} = \mathbf{A}X(t) + \mathbf{B}F(t) \\ X(0) = X_0 \end{cases} \quad (1)$$

where: \mathbf{A}, \mathbf{B} – are deterministic matrices, \mathbf{X}_0 – is the random or deterministic vector of initial conditions, $\mathbf{F}(t)$ – is the random vector of stochastic processes- excitations satisfying the mean-square Lipschitz condition, $\mathbf{X}(t)$ – is the random vector which is the solution of (in mean-square sense) the equation (1), and the output equation:

$$\mathbf{Y}(t) = \mathbf{C}\mathbf{X}(t) + \mathbf{D}\mathbf{F}(t) \quad (2)$$

where: \mathbf{C}, \mathbf{D} – are deterministic matrices, $\mathbf{Y}(t)$ – is the random vector of output stochastic processes.

For solving the equation (1) various methods can be used: the direct method or the method of transformation

equations to the moment equation. To solve the equation (1), the method of moments is used [11].

3. Method of moments.

Using the expected value of the operator $E[\]$ in the both sides of the equation (1) the deterministic system of differential equations in respect to the vector $\mu_X(t)$ is obtained, which is the expected value of the process $\mathbf{X}(t)$:

$$\begin{cases} \frac{d\mu_X(t)}{dt} = \mathbf{A}\mu_X(t) + \mathbf{B}\mu_F(t) \\ \mu_X(0) = \mu_{X0} \end{cases} \quad (3)$$

The system (3) is solved by classical methods of mathematical analysis.

Cross correlation of the processes $\mathbf{F}(t)$ and $\mathbf{X}(t)$ is the function:

$$\mathbf{R}_{FX}(t_1, t_2) = E[\mathbf{F}(t_1)\mathbf{X}^T(t_2)] \quad (4)$$

Then the next operations are to be performed:

– bilateral transposition of the equation (1) and substitution $t=t_2$:

$$\begin{cases} \frac{d\mathbf{X}^T(t_2)}{dt_2} = \mathbf{X}^T(t_2)\mathbf{A}^T + \mathbf{F}^T(t_2)\mathbf{B}^T \\ \mathbf{X}^T(0) = \mathbf{X}_0^T \end{cases} \quad (5)$$

– multiplying both sides of the equation by the process $\mathbf{F}(t_1)$:

$$\begin{cases} \mathbf{F}(t_1)\frac{d\mathbf{X}^T(t_2)}{dt_2} = \mathbf{F}(t_1)\mathbf{X}^T(t_2)\mathbf{A}^T + \\ \quad \quad \quad + \mathbf{F}(t_1)\mathbf{F}^T(t_2)\mathbf{B}^T \\ \mathbf{F}(t_1)\mathbf{X}^T(0) = \mathbf{F}(t_1)\mathbf{X}_0^T \end{cases} \quad (6)$$

– applying the expected value of the operator to the result of the previous step. So, as a result the deterministic equation system is obtained:

$$\begin{cases} \frac{\partial \mathbf{R}_{FX}(t_1, t_2)}{\partial t_2} = \mathbf{R}_{FX}(t_1, t_2)\mathbf{A}^T + \\ \quad \quad \quad + \mathbf{R}_F(t_1, t_2)\mathbf{B}^T \\ \mathbf{R}_{FX}(t_1, 0) = \mu_F(t_1)\mu_{X0}^T \end{cases} \quad (7)$$

In system (7), the variable t_1 is treated as a parameter. For a fixed value of the variable t_1 , this system is solved by classical methods.

Autocorrelation of the process $\mathbf{X}(t)$ is the function:

$$\mathbf{R}_X(t_1, t_2) = E[\mathbf{X}(t_1)\mathbf{X}^T(t_2)]. \quad (8)$$

Using the described procedure, in a similar way one can obtain a deterministic system of ordinary differential equations in respect to the vector function $\mathbf{R}_X(t_1, t_2)$ which is the autocorrelation function of the process $\mathbf{X}(t)$:

$$\begin{cases} \frac{\partial \mathbf{R}_X(t_1, t_2)}{\partial t_1} = \mathbf{A}\mathbf{R}_X(t_1, t_2) + \\ \qquad \qquad \qquad + \mathbf{B}\mathbf{R}_{FX}(t_1, t_2) \\ \mathbf{R}_X(0, t_2) = E[\mathbf{X}_0\mathbf{X}^T(t_2)] \end{cases} \quad (9)$$

In system (9), the variable t_1 is treated as a parameter. System (9) is treated as a deterministic ordinary differential equation.

If the initial condition of the system (9) is a vector of real numbers, the system (9) can be simplified to the form:

$$\begin{cases} \frac{\partial \mathbf{R}_X(t_1, t_2)}{\partial t_1} = \mathbf{A}\mathbf{R}_X(t_1, t_2) + \\ \qquad \qquad \qquad + \mathbf{B}\mathbf{R}_{FX}(t_1, t_2) \\ \mathbf{R}_X(0, t_2) = \mathbf{X}_0\boldsymbol{\mu}^T(t_2) \end{cases} \quad (10)$$

If the initial condition of the system (9) is a vector of random variables, it is necessary to find the initial condition of the system (9). It can be obtained by multiplying the equation (1) by \mathbf{X}_0 and substituting t to t_2 (assuming that the initial condition is independent of forcing):

$$\begin{cases} \frac{\partial \mathbf{R}_X(0, t_2)}{\partial t_2} = \mathbf{R}_X(0, t_2)\mathbf{A}^T + \\ \qquad \qquad \qquad + \mu_{X_0}\boldsymbol{\mu}_F^T(t_2)\mathbf{B}^T \\ \mathbf{R}_X(0, 0) = E[\mathbf{X}_0\mathbf{X}_0^T] \end{cases} \quad (11)$$

Covariance and variance of the process $\mathbf{X}(t)$ can be obtained from the relationship:

$$\mathbf{C}_X(t_1, t_2) = \mathbf{R}_X(t_1, t_2) - \mu_X(t_1)\boldsymbol{\mu}_X^T(t_2), \quad (12)$$

$$\sigma_X^2(t) = \mathbf{C}_X(t, t). \quad (13)$$

Applying the expectation operator to equation (2) one can obtain the equation for the expected values of the process $\mathbf{Y}(t)$:

$$\boldsymbol{\mu}_Y(t) = \mathbf{C}\boldsymbol{\mu}_X(t) + \mathbf{D}\boldsymbol{\mu}_F(t). \quad (14)$$

Cross correlation of the processes $\mathbf{F}(t)$ and $\mathbf{Y}(t)$ is the function:

$$E[\mathbf{F}(t_1)\mathbf{Y}^T(t_2)] = E[\mathbf{F}(t_1)\mathbf{X}^T(t_2)]\mathbf{C}^T + E[\mathbf{F}^T(t_1)\mathbf{F}(t_2)]\mathbf{D}^T, \quad (15)$$

which means:

$$\mathbf{R}_{FY}(t_1, t_2) = \mathbf{R}_{FX}(t_1, t_2)\mathbf{C}^T + \mathbf{R}_F(t_1, t_2). \quad (16)$$

Autocorrelation of the process $\mathbf{Y}(t)$ is the function:

$$\mathbf{R}_Y(t_1, t_2) = E[\mathbf{Y}(t_1)\mathbf{Y}^T(t_2)]. \quad (17)$$

Taking into account the identity:

$$\mathbf{X}(t_1)\mathbf{F}^T(t_2) = (\mathbf{F}(t_2)\mathbf{X}^T(t_1))^T. \quad (18)$$

The equation (17) can be expressed as:

$$\begin{aligned} \mathbf{R}_Y(t_1, t_2) &= \mathbf{C}\mathbf{R}_X(t_1, t_2)\mathbf{C}^T + \\ &+ \mathbf{C}\mathbf{R}_{FX}(t_2, t_1)\mathbf{D}^T + \mathbf{D}\mathbf{R}_{FX}(t_1, t_2)\mathbf{C}^T + \\ &+ \mathbf{C}\mathbf{R}_F(t_1, t_2)\mathbf{D}^T. \end{aligned} \quad (19)$$

Covariance and variance of the process $\mathbf{Y}(t)$ can be obtained from the relationship:

$$\mathbf{C}_Y(t_1, t_2) = \mathbf{R}_Y(t_1, t_2) - \mu_Y(t_1)\boldsymbol{\mu}_Y^T(t_2), \quad (20)$$

$$\sigma_Y^2(t) = \mathbf{C}_Y(t, t). \quad (21)$$

4. Example

The series RLC circuit is supplied by the ideal voltage source $U(t)$:

$$U(t) = \sin(t + \phi) \quad (22)$$

where ϕ is a random variable with known probability density function:

$$f_\phi(x) = \frac{1}{\pi}(H(x) - H(x - \pi)) \quad (23)$$

where $H(x)$ is the Heaviside step function. The probability density function of random variable ϕ is shown in Fig 1.

The mean value of the force process $U(t)$ is given:

$$\begin{aligned} \mu_U(t) &= \mathbf{E}[U(t)] = \\ &= \int_{-\infty}^{\infty} \sin(t+x)f_\phi(x)dx = \frac{2}{\pi}\cos(t). \end{aligned} \quad (24)$$

The mean value of the force process $U(t)$ is shown in Fig. 2. The variance of the force process $U(t)$ is given:

$$\begin{aligned} \sigma_U^2(t) &= \mathbf{E}[U^2(t)] - \mu_U^2(t) = \\ &= \int_{-\infty}^{\infty} \sin^2(t+x)f_\phi(x)dx - \mu_U^2(t) = \\ &= \frac{1}{2} - \frac{4}{\pi^2}\cos^2(t). \end{aligned} \quad (25)$$

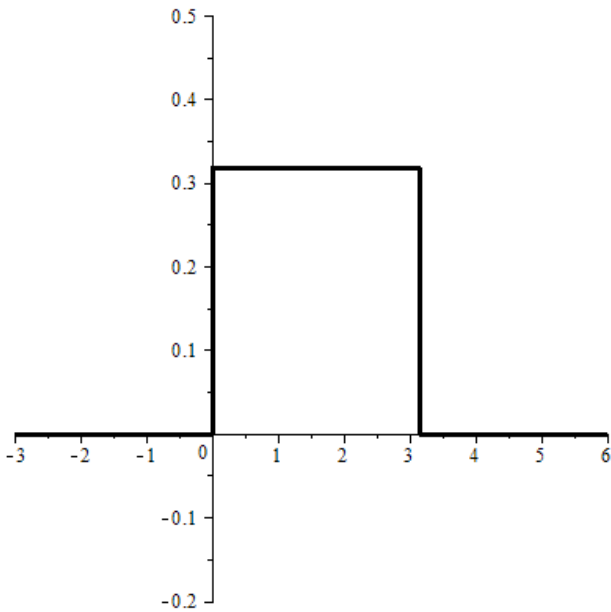


Fig. 1. Probability density function of random variable ϕ .

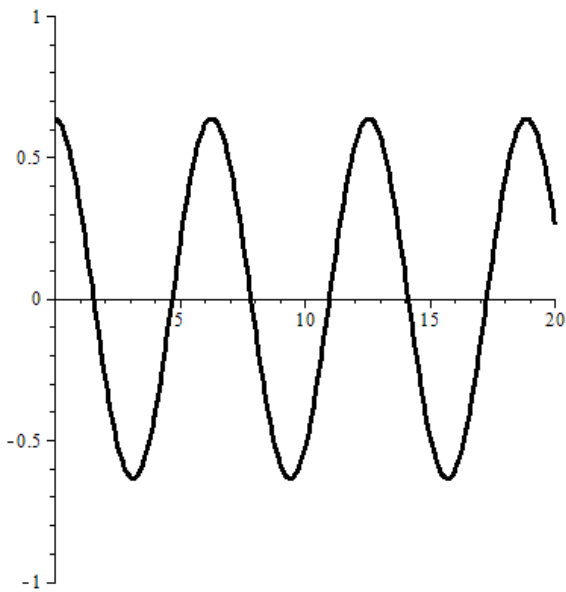


Fig. 2. Mean value of the force process $U(t)$.

Autocorrelation $\mathbf{R}_F(t_1, t_2) = \mathbf{R}_U(t_1, t_2)$ of the force processes is given by the equation:

$$\begin{aligned} \mathbf{R}_U(t_1, t_2) &= \mathbf{E}[U(t_1)U(t_2)] = \\ &= \sin(t_1 + t_2)/4 + \pi \cos(t_1) \cos(t_2)/2 + \\ &+ \pi \sin(t_1) \sin(t_2)/2 - \sin(t_1) \cos(t_2)/4 - \\ &- \cos(t_1) \sin(t_2)/4. \end{aligned} \quad (26)$$

Autocorrelation $\mathbf{R}_U(t_1, t_2)$ is used to determine the cross correlation of the force and response (33).

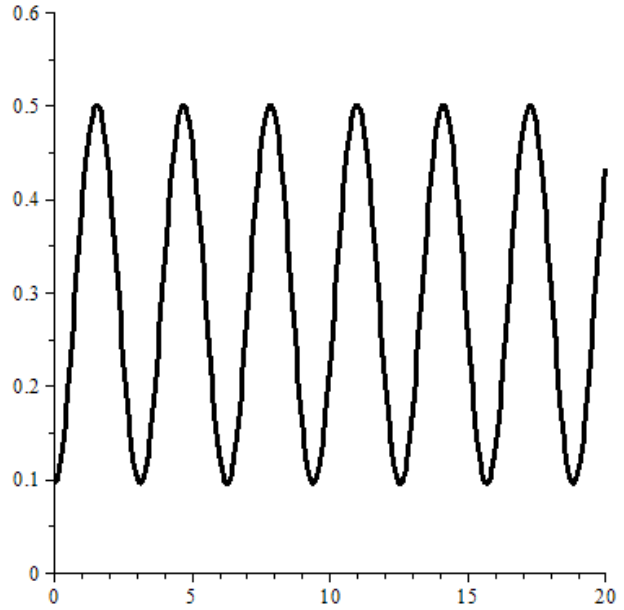


Fig. 3. Variance of the force process $U(t)$.

For simplicity it is assumed that $R=L=C=1$ and $I(0)=U_C(0)=0$. The state equation of the RLC system is given:

$$\begin{aligned} \begin{bmatrix} \frac{d}{dt} I(t) \\ \frac{d}{dt} U_C(t) \end{bmatrix} &= \begin{bmatrix} -R/L & -1/L \\ 1/C & 0 \end{bmatrix} \begin{bmatrix} I(t) \\ U_C(t) \end{bmatrix} + \\ &+ \begin{bmatrix} 1/L \\ 0 \end{bmatrix} [U(t)], \end{aligned} \quad (27)$$

where:

$$\mathbf{X}(t) = \begin{bmatrix} I(t) \\ U_C(t) \end{bmatrix}, \quad (28)$$

Applying the formula (3) to the equation (27), one can obtain the expected value of the output process:

$$\begin{aligned} \begin{bmatrix} \frac{d}{dt} \mu_I(t) \\ \frac{d}{dt} \mu_{U_C}(t) \end{bmatrix} &= \begin{bmatrix} -1 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \mu_I(t) \\ \mu_{U_C}(t) \end{bmatrix} + \\ &+ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \left[\frac{2}{\pi} \cos(t) \right]. \end{aligned} \quad (29)$$

The solution of the equation (29) is given by the formulas:

$$\begin{aligned} \mu_I(t) &= \frac{2\sqrt{3}}{3\pi} \sin(\sqrt{3}t/2) e^{-t/2} - \\ &- \frac{2}{\pi} \cos(\sqrt{3}t/2) e^{-t/2} + \frac{2}{\pi} \cos(t), \end{aligned} \quad (30)$$

$$\mu_{U_C}(t) = -\frac{4\sqrt{3}}{3\pi} \sin(\sqrt{3}t/2) e^{-t/2} + \frac{2}{\pi} \sin(t). \quad (31)$$

The cross correlation of the force and response is the function:

$$\mathbf{R}_{FX}(t_1, t_2) = [\mathbf{R}_{UI}(t_1, t_2) \mathbf{R}_{UU_c}(t_1, t_2)]. \quad (32)$$

Applying formula (7) to the equation (27), one can obtain the cross correlation of the force and response:

$$\left\{ \begin{array}{l} \frac{\partial R_{UI}(t_1, t_2)}{\partial t_2} = -R_{UI}(t_1, t_2) - R_{UU_c}(t_1, t_2) + \\ \quad + R_U(t_1, t_2) \\ \frac{\partial R_{UU_c}(t_1, t_2)}{\partial t_2} = R_{UI}(t_1, t_2) \\ R_{UI}(t_1, 0) = R_{UU_c}(t_1, 0) = 0 \end{array} \right. \quad (33)$$

The solution of the equations (33) is given by the formulas:

$$\begin{aligned} R_{UI}(t_1, t_2) = & \frac{\pi}{6} (3 \cos(t_1 - t_2) + \\ & + \sqrt{3} \cos(t_1) \sin(\sqrt{3}t_2/2) e^{-t_2/2} - \\ & - 2\sqrt{3} \sin(t_1) \sin(\sqrt{3}t_2/2) e^{-t_2/2} - \\ & - 3 \cos(t_1) \cos(\sqrt{3}t_2/2) e^{-t_2/2}) \end{aligned} \quad (34)$$

$$\begin{aligned} R_{UU_c}(t_1, t_2) = & \frac{-\pi}{6} (3 \sin(t_1 - t_2) + \\ & + 2\sqrt{3} \cos(t_1) \sin(\sqrt{3}t_2/2) e^{-t_2/2} - \\ & - \sqrt{3} \sin(t_1) \sin(\sqrt{3}t_2/2) e^{-t_2/2} - \\ & - 3 \sin(t_1) \cos(\sqrt{3}t_2/2) e^{-t_2/2}). \end{aligned} \quad (35)$$

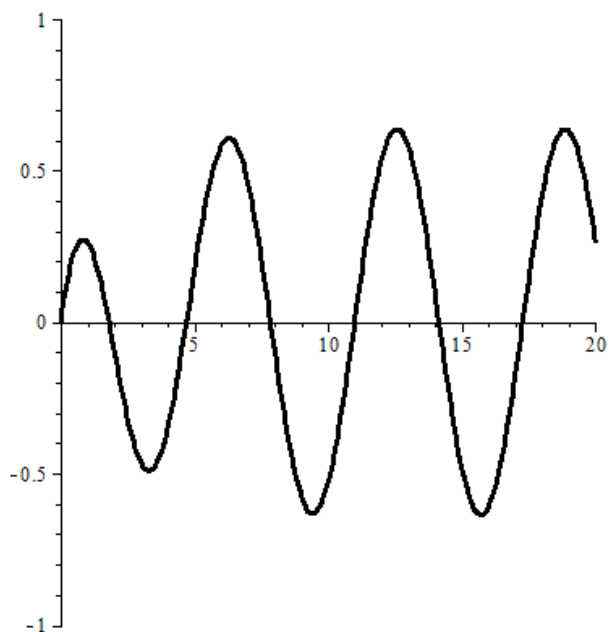


Fig. 4. Mean value of the response process $I(t)$.

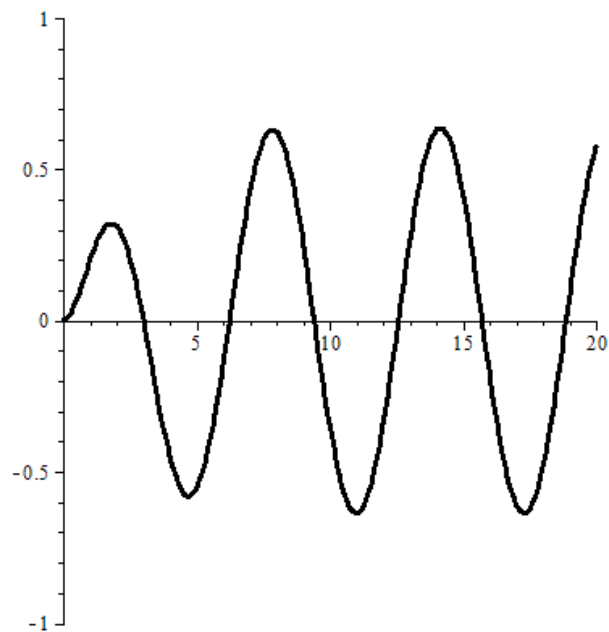


Fig. 5. Mean value of the response process $U_c(t)$.

The autocorrelation of the response is shown in the Fig. 6, 7, 8, 9.

The autocorrelation of the response is the function:

$$\mathbf{R}_X(t_1, t_2) = \begin{bmatrix} R_I(t_1, t_2) & R_{IU_c}(t_1, t_2) \\ R_{U_cI}(t_1, t_2) & R_{U_c}(t_1, t_2) \end{bmatrix}. \quad (36)$$

In the same way, applying the formula (10) to the equation (27), one can obtain the autocorrelation $\mathbf{R}_X(t_1, t_2)$ of the output process:

$$\left\{ \begin{array}{l} \frac{\partial R_I(t_1, t_2)}{\partial t_1} = -R_I(t_1, t_2) - R_{U_cI}(t_1, t_2) + \\ \quad + R_{UI}(t_1, t_2) \\ \frac{\partial R_{IU_c}(t_1, t_2)}{\partial t_1} = -R_{IU_c}(t_1, t_2) - R_{U_c}(t_1, t_2) + \\ \quad + R_{UU_c}(t_1, t_2) \\ \frac{\partial R_{U_cI}(t_1, t_2)}{\partial t_1} = R_I(t_1, t_2) \\ \frac{\partial R_{U_c}(t_1, t_2)}{\partial t_1} = R_{IU_c}(t_1, t_2) \\ R_I(0, t_2) = R_{IU_c}(0, t_2) = 0 \\ R_{U_cI}(0, t_2) = R_{U_c}(0, t_2) = 0 \end{array} \right. \quad (37)$$

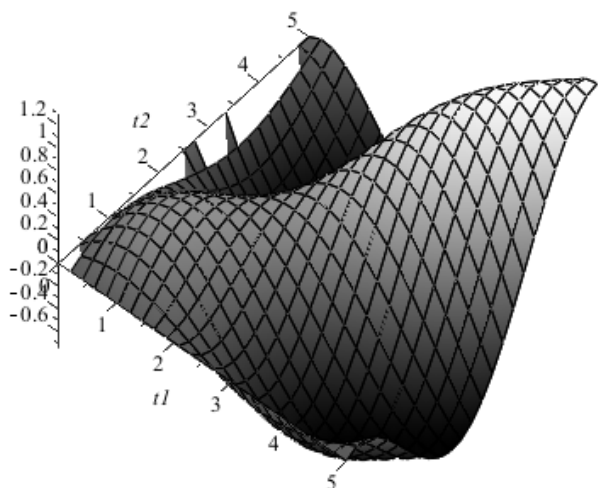


Fig. 6. Autocorrelation of the response process $I(t)$.

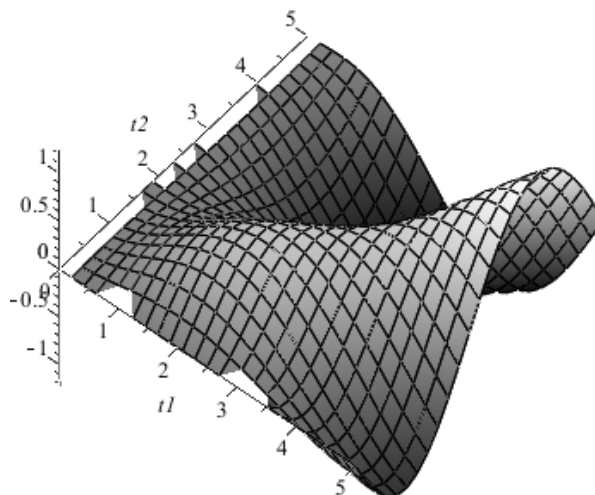


Fig. 9. Correlation of the response processes $U_C(t)$ and $I(t)$.

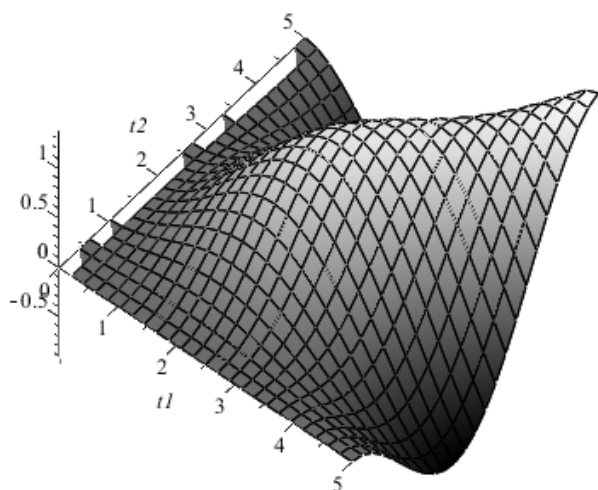


Fig. 7. Autocorrelation of the response process $U_C(t)$.

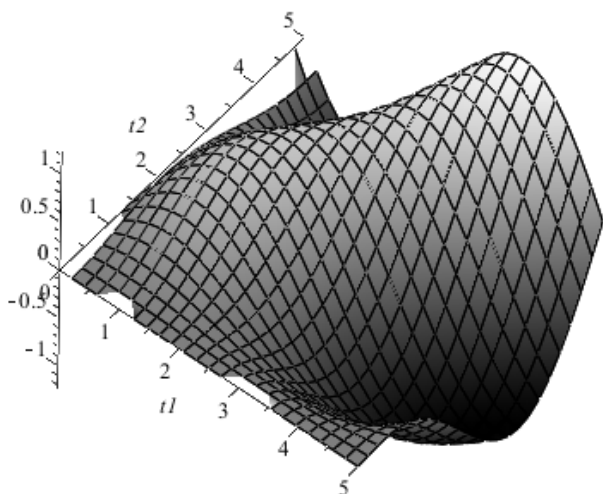


Fig. 8. Correlation of the response processes $I(t)$ and $U_C(t)$.

5. Conclusion

In this paper the method of the problem of the moment determination for complex deterministic and dynamical systems of the n-th order with random excitations has been described. It enables converting the problem of the n-th order stochastic differential equation solving into the problem of ordinary differential equations solving. The ordinary differential equations have been defined in terms of the stochastic process moments and as a result their solution is relatively simple.

Contrary to the previous works the proposed method can be applied for systems of any order.

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ДИНАМІЧНІ ЛІНІЙНІ СИСТЕМИ N-ГО ПОРЯДКУ ЗА ВИПАДКОВИХ УМОВ

Северин Мазуркевич, Януш Вальчак

Розглянуто лінійну динамічну стохастичну модель n-го порядку, що описується стохастичним рівнянням стану. Показано метод визначення імовірнісних характеристик стохастичного процесу, які є розв’язком такого рівняння. Ці характеристики, як, наприклад, середні значення та кореляційні функції, визначаються додатковими умовами — детермінованими системами звичайних диференціальних рівнянь.



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