

# Renormalization the Tensor Model of the Info-Communication Network

Victor Tikhonov

**Abstract - Renormalization technique provided for the tensor model of the object interaction in the open info-communication network based on the least action principle.**

**Keywords – Object interaction, info-communication network, tensor renormalization.**

The tensor math modeling is an actual issue in the sphere of info-communications [1], [2]. However, the network tensor analysis needs further development. *The objective of this work is to determine renormalization technique for the tensor model of the open info-communication network based on the least action principle.*

Let  $P(n, m) = \{p_{nm}\}$ ,  $n = 0, 1, 2, \dots, N$  – non-directed matrix graph of the open network built on the open set of vertices  $X = \{x_n\}$ ;  $p_{nm} = p(x_n, x_m) = p_{mn} = p(x_m, x_n)$  – non-negative real numbers;  $x_0$  is open zero element of  $X$ . The non-diagonal elements  $p_{nm} = p_{mn}$ ,  $n \neq m$ ;  $n, m = 0, 1, \dots, N$  are internal and the diagonal elements  $p_{nn} = p_n = p(x_n)$ ,  $n = 0, 1, 2, \dots, N$  are external interaction edges of vertices.

We want to determine a Riemann metric tensor  $R(P) = \{r_{nm}\}$  as a function of  $P$ , where  $R$  is a real symmetric positively defined matrix, that meets the following requirements: all the non-diagonal elements of  $R$  and  $P$  are equal:  $r_{nm} = p_{nm}$ ,  $n \neq m$ ; all the diagonal elements of  $R$  satisfy the inequality  $r_{nn} \geq p_{nn}$ . Let  $x = \{x_n\}$ ,  $n = 0, 1, 2, \dots, N$ ;  $r(x) = r(x_n) = r_n$  is unknown vector of a posterior estimation for the matrix  $R$  diagonal elements;  $\zeta(x) = \zeta(x_n) = \zeta_n$  – a prior defined value of  $r(x)$ ;  $q(x) = q(x_n) = q_n$  – the measured value of  $r(x)$  obtained from the matrix  $P$  as following:  $q_n = q(P) = \sum_{m=0}^N p_{nm}$ .

We formulate the principle of least action due to the following lagrangian  $L(x, r(x), a)$ :

$$L(x, \rho(x), \alpha) = [\rho(x) - \vartheta_\alpha(x)]^2;$$

$$\rho(x) = \sqrt{r(x)};$$

$$\vartheta_\alpha(x) = \sqrt{g_\alpha(x)};$$

$$g_\alpha(x) = \frac{\tilde{g}_\alpha(x)}{\int \tilde{g}_\alpha(x)};$$

$$\tilde{g}_\alpha(x) = \frac{q(x) + \alpha \zeta(x)}{1 + \alpha \zeta(x)};$$

$$0 < \alpha < \infty.$$

Victor Tikhonov - Odesa National Academy of Telecommunication n.a. A.S.Popov, Kovalska Str., 1, Odesa, 65029, UKRAINE, E-mail: victor.tikhonov@onat.edu.ua

The variable factor  $\alpha$  generates a family of  $\tilde{g}_\alpha(x)$  performing the smooth transition of  $\tilde{g}_\alpha(x)$  from the  $q(x)$  function (if  $\alpha \rightarrow 0$ ) to the  $\zeta(x)$  function (if  $\alpha \rightarrow \infty$ ).

The least action principle for the lagrangian  $L$  means [3]:

$$S(\rho, \alpha) = \int L(x, \rho(x), \alpha) dx = \min. \quad (1)$$

This turns to the variance optimization task [3]:

$$\delta S(\rho, \alpha) = 0. \quad (2)$$

The variance task Eq. (2) results the Euler equation that formulates the necessary condition of an extrema for the Eq.(1), [3]:

$$\left\{ \begin{array}{l} \frac{\partial L}{\partial \rho} - \frac{d}{dx} \cdot \frac{\partial L}{\partial \rho'} = 0, \\ \rho' = \frac{d\rho(x)}{dx}. \end{array} \right.$$

In our case  $\frac{\partial L}{\partial \rho} = 0$ ;  $\frac{\partial L}{\partial \rho} = 2\rho(x) - 2\vartheta_\alpha(x)$ .

Therefore,  $\rho(x) = \vartheta_\alpha(x)$ , or  $r(x) = g_\alpha(x)$ . Now, we put the  $r(x) = g_\alpha(x)$  vector into the main diagonal of matrix  $R$ . It results a family  $\{R_\alpha\}$  of tensors  $R(P, \alpha)$ . To select a subset  $(R_\alpha) \subseteq \{R_\alpha\}$  of “good candidates” for the final solution of the variance task Eq. (1) we inspect the spectrum family  $\lambda_\alpha$  of Eigen values of  $R_\alpha$  tensor’s family. The decision criteria is  $(\lambda_\alpha)_{\min} \geq \varepsilon$ , where  $\varepsilon$  is predefined critically minimal positive value of  $\lambda_\alpha$ . The subset  $(R_\alpha) \subseteq \{R_\alpha\}$  presents the limited number of tensors that consider “good candidates” for the final choice. The tensors  $R_\alpha \in (R_\alpha)$  satisfy the requirements of Riemann metric tensor [3].

## REFERENCES

- [1] A.E. Petrov, “Dual Network Models of Large Systems”, Large System Control, vol.30.1 *Network Models in the Control*, pp.76–90, Nov. 2010. Available: <http://ubs.mtas.ru/upload/library/UBS30106.pdf>.
- [2] V.V. Popovsky, A.V. Lemeshko, O.U. Evseeva, “Mathematic Models of the Functional Properties of the Telecommunication Systems”, *Telecommunication Problems*, № 2 (4), pp.3–41, 2011. Available: [http://pt.journal.kh.ua/2011/2/1/112\\_popovsky\\_functional.pdf](http://pt.journal.kh.ua/2011/2/1/112_popovsky_functional.pdf).
- [3] G.A. Korn, T.M. Korn, “Mathematical Handbook for Scientists and Engineers: Definitions, Theorems and Formulas for Reference and Review, Dover Publications, 2000, 1152 p.