

# Pattern recognition by method of probable algorithm quantum of knowledges (PAQK-method)

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In the difference of widely known modeling of human's brain functions, using taught **neuron nets** for pattern recognition [1,2] quantum approach [3-5] is developing for modeling human generation of cause-consequence reasoning in processes of making decisions with help of taught on examples of **logical nets of probable reasoning (LNPR)**.

The method of probable algorithm quantum of knowledges (PAQK-method) is proposed for pattern recognition using **v-quantum net of decisions' output (v-QNDO)**, generated during the process of teaching LNPR on the totality of **scenario examples of taught knowledges (SETK)**.

So instead of finding indexes of synoptical neurons' connections with given topology of net PAQK-method provides restoring beforehand unknown topology v-QNDO and definition of probable reliability's estimations of it's v-quantums as events. Taught and optimized on abundance v-QNDO is a base of probable quantum of knowledges (BPQK) and, at the same time, mechanism of the output of recognizing decisions with counting indexes of their reliability(IR).

## 1. INTRODUCTION

Great requirements in working out methods of decision poorly structured and unstructured problems of pattern recognition under uncertain conditions exist[6]. Let's distinguish a class  $K(\alpha)$  of problems connected with  $\alpha$ -undefinition, that stipulates for probable character, variety of types, uncompleteness, careless of initial information, with unknown informational signs and deciding rules, alternative and formal purposal criterions, and also with demand in knowledges of experts in concrete subject field.

The peculiarity of suggested PAQK-method of recognition concludes in attracting for outputting finding rules of making recognition decisions not only digital data of measurings and observations but portions (quantums) of knowledges of specialists in given subject field in the form of probable saying and cause-

consequence conclusions of productional character, that have relations with finding decisions. With it using PAQK-method the task of indivisible formalization, presenting and algorithm manipulating of digital and linguistic information in the form of variety of levels on complication **probable algorithm quantum of knowledges (vk-knowledges)** using vector-matrix operators is decided.

In the difference from traditional presenting teaching LNPR makes with given SETK and consists in definition of parameters net's vk-knowledges-knots, their quantity and cause-consequence connections between them. Teaching is finished when **structured completeness v-QNDO** is achieved, that is when equivalent count  $G$ , that has numeral function [7], which last level of summits keep only purpose v-quantums-conclusions, corresponds to the net.

Substantially it means, that with structured completeness v-QNDO admission of net's knots and logical connections between them, and from any v-quantum-knot exists way to any purpose v-quantum-consequence, is found.

So in structure completed v-QNDO **input** knots correspond to sending v-quantums of initial signs OMD or facts, **output** – v-quantums of purpose conclusions, and inside knots – v-quantums of intermediate confirmations or partial decisions.

Peculiarity vk-knowledges is in presence inbuilt procedures for realization of exact countings of intermediate and concluding decisions' probables on the base of equivalent transformations of logical functions statistically depended v-quantums chains to independent by ortogonizing of corresponding logical functions.

## 2. ORGANIZATION OF THE TASK

It is necessary to decide two basic tasks of formally presenting different-levels vk-knowledges ( $A_v$ -task) and algorithm synthesis of the rule of



making recognition decisions on the base of using vk-knowledges ( $B_v$ -task).

Formally  $A_v$ -task for vk-knowledges let's present as multitude five

$$A_v = \langle S_v, K_v, \tilde{V}, \Pi_v, Q_v \rangle, \quad (1)$$

where  $S_v$  - symbol language v-quantums of knowledges, that consists of final multitude of letters, digits and symbols of the algorithm theory's operations;

$K_v$  - final multitude of terminal v-quantums of knowledges; set before experiment;

$\tilde{V}$  - probable multitude, that birth meanings a function  $p_v(\bullet)$  of events reliable, described by different-levels v-quantums, that is multitude of indexes of reliability of vk-knowledges from the interval  $[0, 1]$ ;

$\Pi_v$  - rules of designing of different-levels v-quantums;

$Q_v$  - multitude of semantically codes and special structures of data, that determine level and maintenance v-quantums structures of knowledges with algorithmization.

In other words, in  $A_v$ -task it is necessary to create the formal mechanism of building class  $M_v$  of different-levels vk-knowledges in the language  $S$  with meanings of reliability's degree from  $\tilde{V}$  on the base of using rules  $\Pi_v$  to terminal v-quantums from  $K_v$ , and also with needed level and contents of v-quantums with corresponded semantic codes from  $Q_v$ .

Probable multitude  $\tilde{V}$  on universal multitude  $U = \{u\}$  is called totality of pairs of such type

$$\tilde{V} = \langle \{u | p_v(u)\} \rangle, \quad (2)$$

where  $p_v(u) : U \rightarrow [0, 1]$ .

Showed reflection  $U$  to segment  $[0, 1]$  we'll call **function of reliability**  $p_v(u)$  probable multitude  $\tilde{V}$ , and meaning  $p(u)$  for concrete element  $u \in U$  - **index of reliability (IR)**  $p(u)$  value  $u$ .

Different-types characteristic of the object of recognition (OR) in examples cause-consequence sayings of experts or judgment about outputting finding decisions with some appraisal of their probable from other sources, and also **quantity and contents requirements conclusions at the form SETK** in decided  $B_v$ -task from concrete subject field are given.

It is necessary to work out total method of algorithm decision of  $B_v$ -task, and also to synthesis and ground in terms of PAQK-method necessary basic algorithms: teaching LNPR (ALOBUCH), automatic quantuming (ALAKVA), optimization of the net

(ALAKVA), outputting recognition rule ( $AL(B_v)$ ) and management of functioning v-QNDO (ALUPR).

### 3. METHOD OF DECISION $A_v$ -, $B_v$ -TASKS

Method of decision  $A_v$ -task is based on theoretic positions, given in works [3-6, 8], with using operators superposition (**P-operator**), line operator of concatenation (**CON <•>-operator**) and column (**CON[•]-operator**), and also new terminal v-quantums from multitude

$$K_v = \{vk_0 a, vk_1 y, vk_1 b\}. \quad (3)$$

Here **terminal choosing v-quantum of knowledges of the nil level  $vk_0 a$  with semantic:** «from  $\ell$ -measured totality observed probable meanings of signs  $\alpha_1, \alpha_2, \dots, \alpha_\ell$  to choose meaning  $\alpha_k$  with reliable's index  $p(\alpha_k) \in [0, 1]$ »

has such form:

$$vk_0 a = [V_k^{(\ell)}(\alpha_1 | p(\alpha_1), \dots, \alpha_k | p(\alpha_k), \dots, \alpha_\ell | p(\alpha_\ell)) = \alpha_k | p(\alpha_k)], \quad (4)$$

where  $vk_0 \in Q_v, a, V_k^{(\ell)} \in S_v$ .

**Terminal vector v-quantum of knowledges of the first level  $vk_1 y$  with semantic:** «observed OR  $y \in \Omega_j$ , has  $n$  different-types sings  $x_j, (j = \overline{1, n})$ , which meanings  $\alpha_{k_j}^j \in \tilde{X}^j, (k_j = \overline{1, r_j})$  with correspondence IR  $p(\alpha_{k_j}^j) \in [0, 1]$ » has a form

$$vk_1 y = [d_1^{x_1} d_2^{x_2} \dots d_n^{x_n}] = [\alpha_1^1 | p(\alpha_1^1), \dots, \alpha_1^{r_1} | p(\alpha_1^{r_1}); \alpha_2^1 | p(\alpha_2^1), \dots, \alpha_2^{r_2} | p(\alpha_2^{r_2}); \dots; \alpha_n^1 | p(\alpha_n^1), \dots, \alpha_n^{r_n} | p(\alpha_n^{r_n})], \quad (5)$$

where  $vk_1 \in Q_v$  name  $y \in S_v; \alpha_i^j | p(\alpha_{r_i}^j) \in \tilde{V}$ .

**Terminal characteristic v-quantum of the first level  $vk_1 b$  with the name  $b$  has a form**

$$vk_1 b = [\chi_{\tilde{Y}_j}(\alpha_k^j | p(\alpha_k^j))] = \begin{cases} 1, & \text{if } \alpha_k^j | p(\alpha_k^j) \in \tilde{Y}_j, \\ 0, & \text{if } \alpha_k^j | p(\alpha_k^j) \notin \tilde{Y}_j, \end{cases} \quad (6)$$

where  $\tilde{Y}_j = \{\alpha_k^j\}$  - fixed probable multitude  $k$ 's meanings  $j$ 's sing,  $vk_1 \in Q_v, b, \chi_{\tilde{Y}_j} \in S_v$  with semantic:



«meaning  $\alpha_k^j$  j's sign OR in given moment is fixed by observing with IR  $p(\alpha_k^j)$ , if meaning of characteristic function  $\chi_{\tilde{Y}_j} = 1$ ; otherwise if  $\chi_{\tilde{Y}_j} = 0$  the result of observing are ignored as doubtful».

**Definition 1.** Algorithmic constructions, got from components of the multitude  $K_v$  (3) by final quantity of using to them P-operator,  $CON(\bullet)$ -operator and  $CON[\bullet]$ -operator in frameworks  $A_v$ -task (1), are called different-levels vk-knowledges. Class vk-knowledges  $M_v$ , made on definition 1, let's call **standard class PAQK-models** of presenting of knowledges about object under recognition (OR).

For example, by expression

$$vk_1c1 = \underbrace{[0,1|p(\alpha_2^1),1|p(\alpha_3^1),\emptyset:1|p(\alpha_1^2),0,\emptyset:\emptyset|\emptyset:\alpha_1^{3u}|p(\alpha_1^{3u}),\emptyset:]}_{\text{output domain}} \underbrace{;p(\rightarrow c1);A(c1)p(c1)}_{\text{output domain}} \quad (7)$$

is presented widen vector v-quantum of knowledges of the first level with name of purposed consequence c1 on purposed sign  $x_u$ , correspondence to three components sending sign  $x_1$  and two components sign  $x_2$  OR.

**Semantic  $vk_1c1$  (7)** is: «**IF** observing the second knowledge  $\alpha_2^1$  sign  $x_1$  with IR  $p(\alpha_2^1)$  **OR** third knowledges  $\alpha_3^1$  with OR  $p(\alpha_3^1)$  **AND** first meaning  $\alpha_1^2$  of sign  $x_2$  **THEN** OR has one component purposed sign  $x_u$  with IR  $p(\alpha_u^n)$ ; with it output IR  $p(c1)$  v-quantum is counted by algorithm  $A(c1)$ , using set probability  $p(\rightarrow c1)$  of implication, that is probability of joint making consequence c1 with sendings  $\alpha_2^1|p(\alpha_2^1)$  **OR**  $\alpha_3^1|p(\alpha_3^1)$  **AND**  $\alpha_1^2|p(\alpha_1^2)$ » for given v-quantum in multitude  $Q_v$  contended correspondent program and structure of data, that realize set sence and execution of the algorithm  $A(c1)$ .

Inputed class  $M_v$  vk-knowledges has multitude present at the form of N-measured space  $\tilde{X}^{(N)}$  PAQK-models. It guaranties comfortable for computer formalization of knowledges about decided rules, events and conformities to natural lows of the subject field by means of vector-matrix (look. view (7)) and analytical final-predicate presents[3-5].

The method of decision of  $B_v$ -task is based on definitions 2 and 3, and also on worked out algorithms

ALOBUCH, ALAKVA, ALAKVA,  $AL(B_v)$  and ALUPR

**Definition 2.** Teaching LNPR is called orientated count  $G = (E, \Gamma)$ , synthesis on given SETK using algorithm ALOBUCH and had numeral function  $\Pi(X_i) \forall X_i \in E$ , determined on submultitudes-levels  $N_1, N_2, \dots, N_k \subset E$  of summits, and also next **properties**:

1) all summits (knots of the net)  $X_i \in E, (1 \leq i \leq n)$  correspond to sayings from SETK of concrete subject field, and arcs from  $\Gamma: E \rightarrow E$  show on cause-consequence connections between knots with logical connections «AND», «OR», «NOT»;

2) all knots  $X_i \in N_1 \subset E$  with  $\Gamma^{-1}N_1 = \emptyset$  correspondent **output sending information**  $e_i, (1 \leq i \leq n_n)$  relatively some consequences  $c_j, (1 \leq j \leq m_c)$  with given IR  $p(e_i)$  and  $p(\rightarrow c_j)$ ;

3) all knots  $X_i \in N_k \subset E$  with  $\Gamma N_k = \emptyset$  are **purposing** (output) knots-conclusions  $C_s, (1 \leq s \leq S)$  with counted IR  $p(C_s)$ , and all summits intermediate levels between  $N_1$  and  $N_k$  correspond to **partial consequence**  $c_j$ .

Syntheses LNPR at the mode of teaching has to be transformed in v-QNDO with help of special algorithm ALAKVA, that automatically determine **not given beforehand** quantity v-quantum **unknown beforehand** levels on **given rules** and **known knots** of the concrete levels of numeral function  $\Pi(X_i)$  of the count G.

**Definition 3.** Purpose-oriented v-QNDO is called the result of count's transformation  $G = (E, \Gamma)$  using algorithm ALAKVA to the count  $G_{vk} = (E_{vk}, \Gamma_{vk})$ , that has following properties:

1) all summits  $X_i \in E_{vk}$  correspond generated different-levels v-quantums, contended SETK of concrete subject field, and arcs  $u_{ij} \in \Gamma_{vk}$  show at the logical connections v-quantums events;

2) all  $X_i \in N_1 \subset E_{vk}, \Gamma^{-1}N_1 = \emptyset$  correspond to **input vk-knowledges-sendings** with names  $e_i, (1 \leq i \leq n'_n)$  relatively vk-knowledges-consequences with names  $C_j, (1 \leq j \leq m'_c)$  with given indexis of reliability;

3) all  $X_i \in N_k \subset E_{vk}, \Gamma N_k = \emptyset$  are **purpose vk-knowledges-conclusions** with names  $C_s, (1 \leq s \leq S)$  and counted IR  $p(C_s)$ , and all intermediate summits of



count  $G_{vk}$  corresponds to **partial vk-knowledges-consequences**.

Generated v-QNDO is optimized on criterion of abundance with the help of algorithm ALOPT and with management of algorithm ALUPR directly is used for outputting recognition decisions with counting correspondence probable appraisals of IR.

Peculiarity of manipulating vk-knowledges at the process of building and making decision by v-QNDO is **new way** of not approximate, but exactly **definition of probable appraisals** intermediate and purpose v-quantums conclusions.

It realized on the base of **ortogonalization** of logical functions, described connections between v-quantums events (sayings). With it the logic of sayings in v-QNDO is presented *инпредставляется в orthogonal dizunctional normal form(ODNF)*.

Counting of meanings of these probabilities by ODNF is make by setting instead of its variables corresponded known probabilities with changing operations of logical multiplication and adding by analog arithmetic operators.

It is grounded like this. Let any event  $e := \langle \text{object } \omega \in \Omega \text{ has sign } x \rangle$  is made with probability  $p(e)$ , and event  $c := \langle \text{if } \omega \text{ ohas sign } x, \text{ that } \omega \text{ corresponds to category } A \rangle$  is set with conditional probability  $p(e \rightarrow c)$ , that is called **probability of implication**. Then probability  $P(e \wedge c)$  setting events at the same time  $e$  and  $c$  (as other event  $z$ ) is determined by correlation

$$p(z) = P(e \wedge c) = p(e) \cdot p(e \rightarrow c). \quad (8)$$

If now for showed events  $e$  and  $c$  to correspond logical variables  $\alpha$  and  $\beta$  correlatively, then the probability  $p(z)$  (8) could be counted as probability of logical conjunction  $\alpha \bullet \beta$  to be true meaning «1» on the any admission of meanings of its variables  $\alpha$  and  $\beta$  with giving probabilities of appearing their true meanings.

For more complicated event  $z1 := \langle \text{if object } \omega \text{ has sign } x_1 \text{ or } x_2, \text{ or } \dots \text{ or sign } x_k, \text{ then } \omega \text{ belongs to class } A \rangle$  probability  $p(z1)$  can be counted on logical function  $\alpha_1 \cdot \beta_1 \vee \alpha_2 \cdot \beta_2 \vee \dots \vee \alpha_k \cdot \beta_k$ , transformed to equivalent to it ODNF

$$\alpha_1 \beta_1 \vee \alpha_1 \beta_1 \cdot \alpha_2 \beta_2 \vee \alpha_1 \beta_1 \cdot \alpha_2 \beta_2 \cdot \alpha_3 \beta_3 \vee \dots \\ \dots \vee \alpha_1 \beta_1 \cdot \dots \cdot \alpha_{k-1} \beta_{k-1} \cdot \alpha_k \beta_k.$$

At the result of deducting outputting **probable identified decisions** relatively OR  $\omega$  with given sending vk-knowledges  $vk_1 Y_{B\omega}$  in  $B_v$ -task with help of algorithm AJL( $B_v$ ) at the output v-QNDO purpose vk-knowledges are formed  $vk_s R_{B\omega}$ , ( $s = 0, 1$ ) as one or some v-quantums with counted resulted IR  $p(R_{B\omega})$ .

Semantic  $vk_s R_{B\omega}$  shows at the **belonging identified OR**  $\omega$  to definition category or class  $K$  from  $\Omega$  with given reliability.

Let while decision  $B_v$ -task are used classified SETK( $B_v$ ) and at the result of studying was got correspondence v-QNDO as functional BPQK( $B_v$ ) for **recognition (identification) of situation** in conditions  $\alpha$ -undefinition using IND-operator [3-4] as

$$BPQK(B_v) = \\ = SETK(B_v) \xrightarrow[\text{ALOBUCHALAKVA, ALOPT}]{\text{IND}} BPQK \quad (9)$$

Then formally algorithm decision  $B_v$ -task concludes in getting purpose vk-knowledges  $vk_s R_{B\omega}$  by DED-operator [3] as

$$vk_s R_{B\omega} = \\ = DED(BPQK(B_v), vk_1 Y_{C\omega}; AL(B_v), ALUPR, vk_s R_{B\omega}) \\ = BPQK(B_v) \xrightarrow[\text{vk}_1 Y_{B\omega}; AL(B_v), ALUPR]{\text{DED}} vk_s R_{B\omega}. \quad (10)$$

Consequently, IND- and DED-operators are realized by algorithms, showed under the pointer in (9) and (10) formulas.

## 4. SOFTWARE AND EXPERIMENTAL INVESTIGATION OF PAQK-METHOD.

On the base of using adduced method in p. 3 interactive program complex (IPK) «V-KVANT» was created as investigational prototype intellectual (SSMD) on the base of using PC. Total architecture of IPK is presented on p. 1.

IPK successfully was tested during experimental decision of series tests and real tasks of patterns recognition.

Software «V-KVANT» is a kernel of new informational technology of knowledges' engineering, based on PAQK-method.



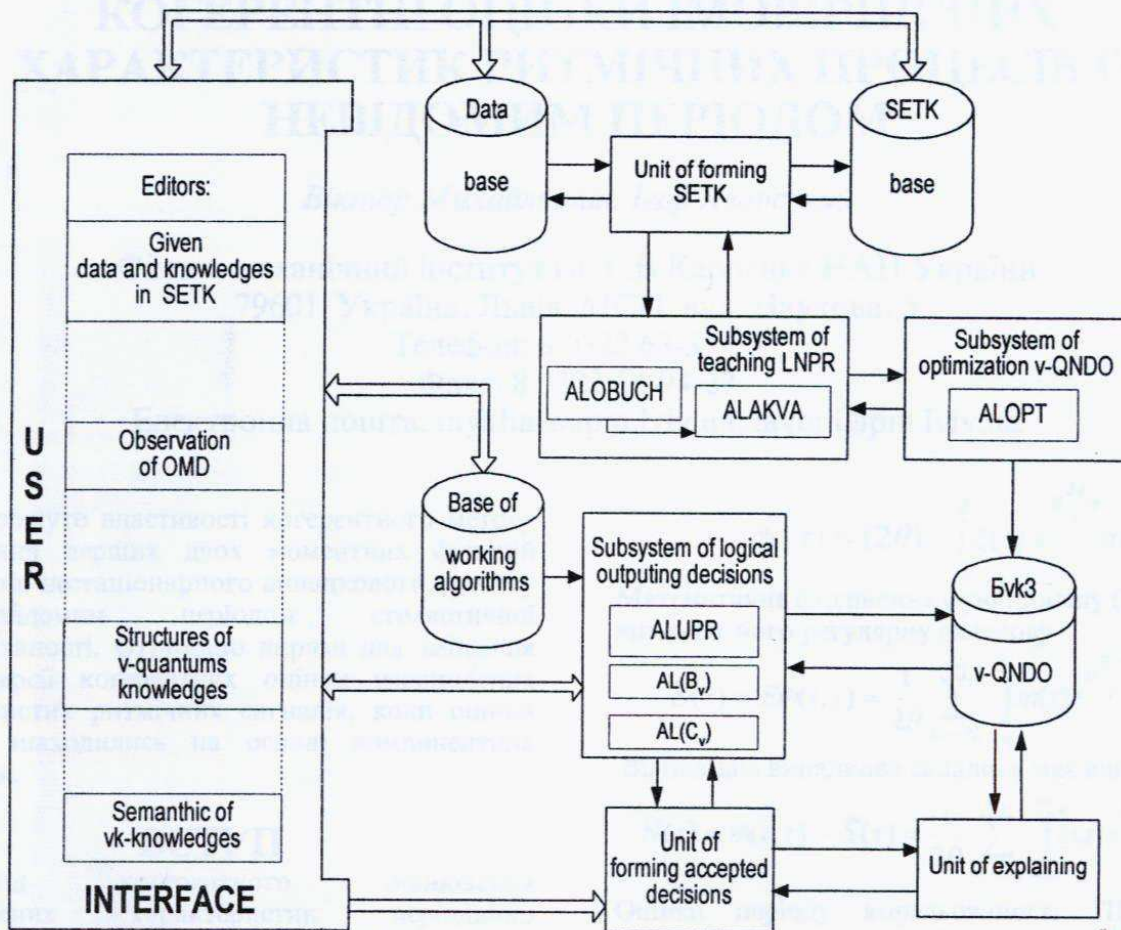


Fig. 1. Total architecture of IPK "V-KVANT"

⇨ - management; → - functional connections

## CONCLUSION.

New PAQK-method of patterns recognition is worked out, on the base using which acting IPK «V-KVANT» was worked out for decision productional tasks of supporting making recognition and prognosing decisions. The results of experimental appraisal and comparement with other methods showed high level, affectivities and advantageges of PAQK-method. The last one conclude in decreasing average risk of recognition on the order, increasing speed of action of the teaching process in 2.5 times and decreasing volume of using memory on BPQK in 1,5 times.

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