

# Sense's standards and transmission of knowledge for their estimation applying the open-form tests

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## Research subject

Methods and algorithms for formation of knowledge about synonymy in Natural Language (NL).

## Considered problem

To transmit a knowledge represented by texts on some natural language between its native speakers (experts and trainees, correspondingly).

## Main purpose of research

Theoretical reasoning of structure of knowledge about synonymy.  
Development of methods and algorithms for forming these knowledge and application them for the family of tasks of:

- sense-similarity's estimation of texts in subject-oriented natural language;
- computer-aided filling and compression of language's and subject-area's knowledge base;
- seeking a most rational plan for sense's transfer among different native speakers;
- coordination of knowledge units which were formed by various experts.

## Definition 1.4

Usage Situation for Natural Language (USNL) is the description of new social experience (the content of joint actions) by means of this natural language.

The NL-context accumulated by some USNL  $S$  can be represented by a triple:

$$S = (O, R, Ts), \quad (1)$$

where  $O$  is a set of symbols associated with reality concepts;

$Ts$  is a set of description forms for  $S$  in some sign system;

$R \subset O^n$ , where  $n \in 1, \dots, |O|$ .

Let  $Synt$  be a surjective function determined by syntax of given NL;

Then for  $\forall Ts_i \in Ts \exists Tr_i: Ts_i = Synt(Tr_i)$ , where  $Tr_i$  is a marked tree.

Thus if  $O = M \cup V$ ,  $M \cap V \neq \emptyset$ , then for  $\forall o_j \in M$  it will be found  $o_k \in V$  with the following correspondence in the tree  $Tr_i$ :

- to the concept  $o_j$  there corresponds a child node with the label  $w_j$ ,
- to the concept  $o_k$  there corresponds a parent node with the label  $w_k$ .

Let's represent a single USNL by a Formal Context (FC):

$$K = (G, M, I), \quad (2)$$

where  $\forall g \in G$  is a **stem of word** syntactically **submitted** to any other **word** from some  $Ts_i \in Ts$  in model (1).

The **attributes's set**  $M$  comprises the **subsets** designated by corresponding bottom indexes:

- $M_1$  is the set of indications to the syntactically **main** word's **stem**;
- $M_2$  is the set of indications to the **main** word's **inflection**;
- $M_3$  is the set of «**stem–inflection**» relations for a syntactically **main** word;
- $M_4$  is the set of **combinations of inflections** of dependent and main words. After an inflection **via the colon**, a **preposition** is shown (if any) that provides a relation with a dependent word;
- $M_5$  is the set of indications for **dependent** word's **inflection**.

**It is required:** to form  $I \subseteq G \times M$  by analysis of symbolic structure and choice of  $Ts_i \in Ts$  with a **minimal** length and **maximum** of words most generally **used** in different phrases from  $Ts$  (taking possible synonyms into account).

Let

$W_{ij}$  be the ordered sequence of symbols constituting the word  $w_{ij}$ ,

$Wc_{ij}$  be the ordered sequence of symbols for invariant part (i.e. stem) of word,

$Wf_{ij}$  be the ordered sequence of symbols for word's inflection,

$\odot$  be the designation for concatenation operations.

It is given:

$Ts = \left\{ Ts_i : Ts_i = \odot_j W_{ij} \right\}$  is a set of Semantically Equivalent (SE) phrases defining the USNL.

It is required to find:

$Pw_i = \left\{ (Wc_{ij}, Wf_{ij}) : Wc_{ij} \odot Wf_{ij} = W_{ij} \right\}$  for all  $i = 1, \dots, |Ts|$ .

Let's identify concepts: with a stem  $Wc_{ij}$  — the «**prefix**»  
and with inflection  $Wf_{ij}$  — the «**suffix**» as accepted in informatics.

These are paramount procedures and functions of algorithm:

- $pref.show(w_{ij})$  return the current value of prefix for the word  $w_{ij}$ ;
- $pref.inc(w_{ij})$  increments the length of prefix for the word  $w_{ij}$  by 1;
- $pref.s$  forms the lists by grouping of wordforms similar in prefix with sorting them by length's decrease;
- $pref.check(Prf)$  for group of wordforms with common prefix  $Prf$  carry out analysis of **absolute frequencies** of occurrence of characters on various positions concerning the word's front and end. Note that the **frequency**  $\nu_p$  of occurrence of symbol which is the first at the left is always maximal as well as for symbols in  $Prf$ . Relative to word's end the search of common suffix's symbols having the occurrence's **frequency**  $\nu_p$ , is carry out with including them into word's inflection. *The total length of common prefix and suffix must be, at least, a one third of word's length, and lengths's difference for a pair of words having a common prefix is always less than half a length of smaller word (independently from suffix!).*

pseudocode description

software realization

continue

**Require:**  $Ts$ ;

**Ensure:**  $Pw = \bigcup_{i=1}^{|Ts|} Pw_i$ ;

```

1:  $Pw := \emptyset$ ;
2: for all  $W_{ij}$ :  $\odot_j W_{ij} = Ts_i$ , where  $Ts_i \in Ts$  do
3:    $Wc_{ij} := \{W_{ij}[1]\}$ ;  $Wf_{ij} := \bigodot_{k=2}^{|W_{ij}|} W_{ij}[k]$ ;
4: end for // initialization of stems and inflections
5:  $pref.s(PrfsTmp)$ ;
6: if  $PrfsTmp = \emptyset$  then
7:   return  $Pw$  and exit the algorithm;
8: else
9:   take  $Prf$  from  $PrfsTmp$ ;
10:  if  $pref.check(Prf) = \text{true}$  then
11:     $Pw := Pw \cup \left\{ \left( Prf, Wf_{ij}(Prf) \right) \mid pref.show(w_{ij}) = Prf \right\}$ ;
12:     $PrfsTmp := PrfsTmp \setminus \{Prf\}$ ;
13:    go to the Step 6;
14:  else
15:    for all  $w_{ij}$ :  $pref.show(w_{ij}) = Prf$  do
16:       $pref.inc(w_{ij})$ ;
17:    end for
18:    go to the Step 5
19:  end if
20: end if

```

Let  $Ts$  be a set of SE-phrases defining some USNL according to (1),  
 $J$  be an index set for invariant parts of words of phrases from  $Ts$ .

## Definition 3.2

The ordered sequence of indexes of invariant parts of words for some  $Ts_i \in Ts$  let's name as Model of its Linear Structure (MLS),  $Ls(Ts_i)$ .

Let  $\{J_1, J_2\}$  be the pair of sequences of indexes in  $Ls(Ts_i)$ , where  $J_1 = \{j_1^1, \dots, j_2^1\}$ ,  $J_2 = \{j_1^2, \dots, j_2^2\}$ , and both  $(j_1^1, j_2^1)$  and  $(j_1^2, j_2^2)$  correspond to the syntactic links.

The sense standard for USNL is defined by those  $Ts_i \in Ts$ , in MLS of which

$$(J_1 \subset J_2) \vee (J_2 \subset J_1) \vee (|J_1 \cap J_2| = 1) \vee (J_1 \cap J_2) = \emptyset, \quad (3)$$

and summary length of sequences of mentioned kind for all syntactic links revealed on  $Ts_i$  has to be *minimum*.



Let  $fr(w_j|)$  be a frequency of occurrence of word  $w_j$  in all  $Ts_i \in Ts$ .

So the **most informative** words in  $Ts$  are forming a cluster *Clust*:

- the word with a maximal value of this frequency will be in *Clust*;
- for  $\forall \{w_j, w_k\} \subset Clust$  and  $\forall w_l \notin Clust$  is true that

$$\left( |fr(w_j) - fr(w_k)| < |fr(w_j) - fr(w_l)| \right) \wedge \\ \wedge \left( |fr(w_j) - fr(w_k)| < |fr(w_k) - fr(w_l)| \right) = \text{true} \quad (4)$$

The basis of standard are made the phrases with **maximum** of words in *Clust*.

Here for words from *Clust* possible synonyms and different orders in a phrase are taking into account.

Let  $LS$  be a set of linear structures's models given on  $J$  for sentences from  $Ts$ .

### Lemma 5.1

The pair of indexes  $\{j_1, j_2\} \subset J$  corresponds to synonymic words and can be replaced by one index from  $(\mathbb{N} \setminus J)$  if  $\exists \{Ls(Ts_1), Ls(Ts_2)\} \subseteq LS$ :

$$Ls(Ts_1) = J_1 \odot \{j_1\} \odot J_2 \text{ and } Ls(Ts_2) = J_1 \odot \{j_2\} \odot J_2,$$

where  $J_1 \subset J$ ,  $J_2 \subset J$ , and  $\odot$  is the concatenation operation at the set  $J$ .

Let  $J_{Cl}$  be a set of indexes of words related to the cluster of **most informative** concerning to USNL given by the set of SE-phrases  $Ts$ ;

$frq((j, k), LS)$  be the frequency of occurrence of the pair  $(j, k)$  in the models from  $LS$  taking into account that  $(j, k) \Leftrightarrow (k, j)$ .

Then **USNL's standard** is defined by phrases with MLSs belonging to the set

$$LC = \bigcup_i LS_i : LS_i \subset LS, \exists \{Ts_i, Ts_j\} \in Ts : \\
 Ls(Ts_i) \in LS_i \\
 |Ls(Ts_i) \cap J_{Cl}| \rightarrow \max \\
 \left( (Ls(Ts_j) \in LS_i) \wedge (Ts_j \neq Ts_i) \right) \rightarrow (Ls(Ts_i) \cap J_{Cl}) \subset Ls(Ts_j),$$

and **attributes set's** forming for **USNL's standard** in a form of FC (2) requires:

- to find index pairs  $(j, k) : frq((j, k), LS) > 1$ , which satisfy the condition (3), for all linear structure's models from  $LC$ ;
- to define the direction of syntactic link for each found pair  $(j, k)$ ;
- to eliminate from  $\forall LS_i \subset LC$  any MLS containing indexes which not appeared in any found link.

There are three stages to find  $Dir(j, k)$ ,  $Dir \in \{\leftarrow, \rightarrow\}$ , namely:

- checking the **link** corresponding to  $(j, k)$  on **falsity's** condition's fulfilment;
- an attempt to identify with the **links revealed earlier**;
- if there are no identification with known links then interview with expert.

Let  $St(j)$ ,  $St(k)$  and  $St(l)$  are the word's stems corresponding to  $j$ ,  $k$  and  $l$ .

*For given USNL the link for  $(j, k)$  is identified as false if  $j, k, l \in Ls(Ts_i)$  in some  $Ts_i \in Ts$ , but another USNL has false link for  $St(j)$  and  $St(k)$ , and true link either between  $St(j)$  and  $St(l)$  or between  $St(k)$  and  $St(l)$ .*

Let  $Lnk$  be a set of links **revealed earlier**, each of which is represented by:

- an ID number of USNL ( $Id$ );
- a main word's stem ( $St_1$ );
- a stem for dependend word ( $St_2$ );
- a list of inflections combinations «main word–dependent word» ( $FCm$ ).

A pair  $(j, k)$  is put in conformity of link  $((j, k), \rightarrow)$  if for some other USNL

$\exists (Id, St_1, St_2, FCm) \in Lnk$ :

$$St(j) = St_1, St(k) = St_2 \text{ and } (Fl(j), Fl(k)) \in FCm.$$

# Example of initial set of semantically equivalent phrases defining the USNL

Синонимичные перифразы

27:89

Insert

Indent

Modified

"Нежелательное переобучение приводит к заниженности эмпирического риска."

"Нежелательное переобучение, следствием которого является заниженность эмпирического риска."

"Заниженность эмпирического риска является следствием нежелательного переобучения."

"Заниженность эмпирического риска, являющаяся следствием нежелательного переобучения."

"Эмпирический риск, заниженность которого является следствием нежелательного переобучения."

"Эмпирический риск, заниженный вследствие нежелательного переобучения."

"Эмпирический риск, к заниженности которого ведет нежелательное переобучение."

"Риск, заниженный как следствие переобучения."

"Эмпирический риск по причине, обусловленной нежелательным переобучением, может оказаться заниженным."

"Эмпирический риск в силу обстоятельств, связанных с нежелательным переобучением, может оказаться заниженным."

"Эмпирический риск по причине, вызванной нежелательным переобучением, может быть заниженным."

"Эмпирический риск, к заниженности которого приводит нежелательное переобучение."

"Нежелательное переобучение служит причиной заниженности эмпирического риска."

"Заниженность эмпирического риска, причиной которой является нежелательное переобучение."

"Заниженность эмпирического риска является результатом нежелательного переобучения."

"Нежелательное переобучение, с которым связана заниженность эмпирического риска."

"Эмпирический риск, с переобучением связана его заниженность."

"Заниженность эмпирического риска связана с переобучением."

"Заниженность эмпирического риска, являющаяся результатом нежелательного переобучения."

"Нежелательное переобучение, результатом которого является заниженность эмпирического риска."

"Нежелательное переобучение, результат которого есть заниженность эмпирического риска."

"Нежелательное переобучение, приводящее к заниженности эмпирического риска."

"Нежелательное переобучение, служащее причиной заниженности эмпирического риска."

"Заниженность эмпирического риска относится к следствию нежелательного переобучения."

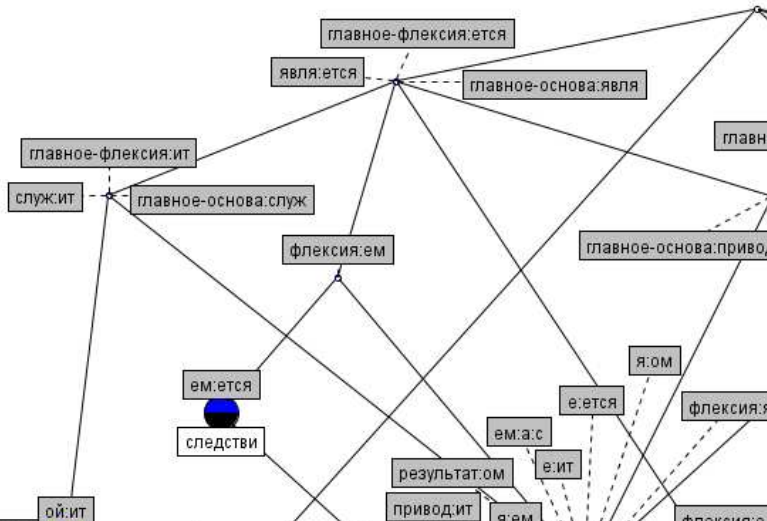
"Заниженность эмпирического риска связана с нежелательным переобучением."

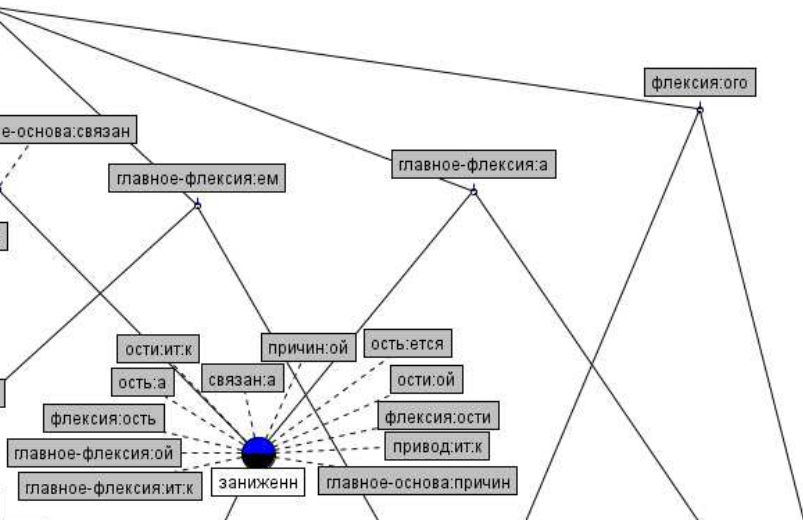
"Нежелательное переобучение является причиной заниженности эмпирического риска."

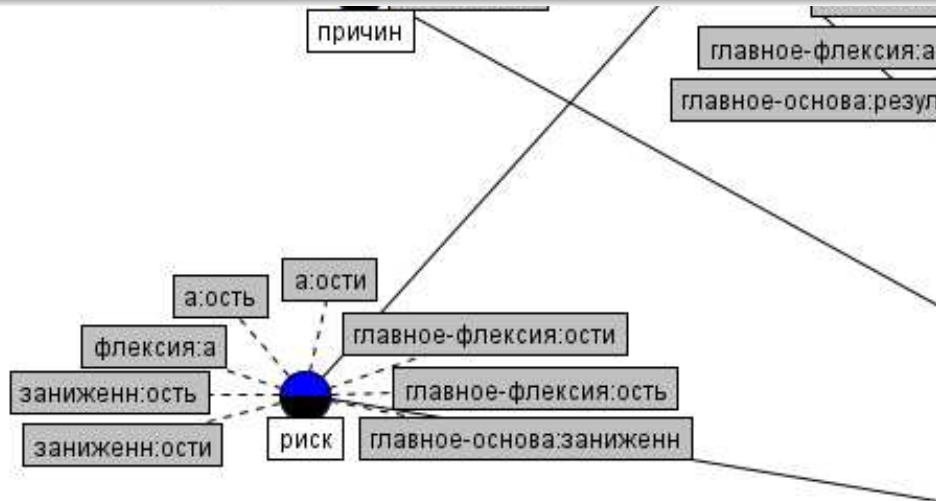
"Заниженность эмпирического риска, причиной которой служит нежелательное переобучение."



# Resulting formal context of sense standard and NL-phrases defining standard

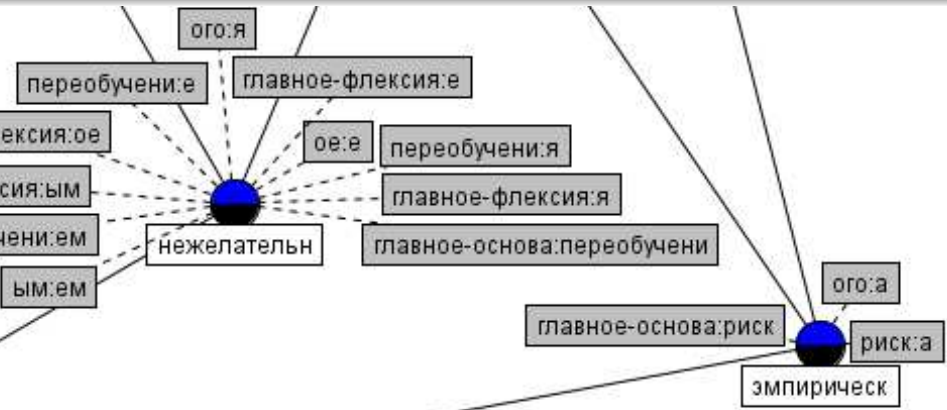








# Resulting formal context of sense standard and NL-phrases defining standard





Просмотр фраз из определяющих эталон заданной СЯУ

1:1    Insert    Indent

Нежелательное переобучение приводит к заниженности эмпирического риска.  
Нежелательное переобучение служит причиной заниженности эмпирического риска.  
Заниженность эмпирического риска связана с переобучением.  
Заниженность эмпирического риска связана с нежелательным переобучением.  
Нежелательная переподгонка приводит к заниженности эмпирического риска.  
Нежелательная переподгонка служит причиной заниженности эмпирического риска.  
Заниженность эмпирического риска связана с переподгонкой.  
Заниженность эмпирического риска связана с нежелательной переподгонкой.

Serial number of USNL, $i$	1	2	3	4	5	6
Number of SE-phrases defining USNL	56	28	29	30	6	10
including representatives of standard	8	9	7	9	1	2
Initial number of objects for USNL	18	17	15	13	12	14
Initial number of attributes for USNL	177	186	173	162	94	81
Number of standard's objects	9	12	12	11	8	12
Number of standard's attributes	82	90	80	69	35	53

$i$  What does the situation of language usage represents in Russian ?

- 1 Связь переобучения с эмпирическим риском
- 2 Связь переусложнения модели с заниженностью средней ошибки на тренировочной выборке
- 3 Влияние переподгонки на частоту ошибок дерева принятия решений
- 4 Причина заниженности оценки обобщающей способности алгоритма
- 5 Зависимость оценки ошибки распознавания от выбора решающего правила
- 6 Зависимость обобщающей способности логического алгоритма классификации от числа закономерностей алгоритмической композиции

example

coordination of knowledge

estimating the amount of memory

# Estimating the amount of memory for storing NL-phrase

$i$	1	2	3	4	5	6
$n$	12	15	16	17	10	14
$vol(n)$	$4.790 \cdot 10^8$	$1.308 \cdot 10^{12}$	$2.092 \cdot 10^{13}$	$3.557 \cdot 10^{14}$	$3.629 \cdot 10^6$	$8.718 \cdot 10^{10}$
$vol_1(n)$	648	795	416	442	20	42
$vol_2(n)$	168	225	80	187	20	42

Here:

- $i$  is the serial number of USNL;
- $n$  is the maximal number of words in a phrase;
- $vol(n) = n!$  is the estimation which is taken usually;
- $vol_1$  and  $vol_2$  are the estimations received with application of method and algorithms of NL-usage's situation's standard's revelation.

Numerically:

- $vol_1(n) = l_1 \cdot n$  is the upper estimation,  $l_1$  is the number of SE-phrases defining the USNL;
- $vol_2(n) = l_2 \cdot n$  is the lower estimation,  $l_2$  is the number of SE-phrases defining the standard of USNL.

Let **thesaurus** to be **represented** in the form of formal context

$$Kth = (Gth, Mth, Ith), \quad (5)$$

where **Gth** consists of **symbolic labels** of individual **NL-usage's situations**;

**Mth** includes the **attributes** of formal context (2) for each  $gth \in Gth$ .

In addition, in **Mth** one can distinguish the following **subsets**:

- **M<sub>6</sub>** is the set of indications to **objects** of formal contexts (2) generated for **individual**  $gth \in Gth$ ;
- **M<sub>7</sub>** is the set of «**stem–inflection**» combinations for a syntactically dependent word;
- **M<sub>8</sub>** contains **combinations of stems** of the dependent and main word.

By analogy with the formal context (2) of individual USNL  $Ith \subseteq Gth \times Mth$ .

example of representation of individual USNL in the thesaurus's formal context

*In this case the numerical estimation of similarity of NL-usage situations is determined by the number of attributes be shared by objects of compared situations concerning the formal context of thesaurus.*

coordination of knowledge concerning different situations of NL-usage

# USNL as an object of the thesaurus's formal context



[go back to definition of thesaurus's formal context](#)

[continue](#)

# Coordination of knowledge about synonymy concerning different situations of natural language's usage

Let

$St$  be the designation for word's *invariant part* identified with the *stem*;

$Fl$  be the designation for word's *inflection*;

$S_1$  and  $S_2$  be the some situations of givel NL's usage.

Let's suppose that some  $Wrd$  can be represented as  $St_1 \odot Fl_1$  concerning  $S_1$ , and as  $St_2 \odot Fl_2$  — concerning  $S_2$ . At that  $St_1 = St_2 \odot Sf$ , where  $Sf$  contains one symbol as minimum, and  $\odot$  is the operation of strings's concatenation.

Then concerning  $S_1$  the following replacements can be implemented: the stem  $St_1$  is replaced with  $St_2$ , and inflection  $Fl_1$  — with  $Fl_3 = Sf \odot Fl_2$  only if the frequencies of occurrence of inflections  $Fl_3$  and  $Fl_2$  in all lexico-syntactic links represented by the formal context (5) for given subject area won't decrease at fulfillment of these changes.

*Example (in Russian).*

USNL №3,  $St_1 = \langle \text{является} \rangle$ ,  $Fl_1 = \langle \rangle$ ,

USNL №1,  $St_2 = \langle \text{явля} \rangle$ ,  $Fl_2 = \langle \text{ется} \rangle$ ,  $Sf = \langle \text{ется} \rangle$ .

Concerning the USNL №3 the replacement of  $Fl_1$  to  $Fl_3 = \langle \text{ется} \rangle$  is fulfilled.

go back

continue



# Knowledge-control system as the practical issue of developed approach

Тестирование знаний и подготовка к ЕГЭ

База знаний Тесты Первое знакомство Window Помощь

Численные оценки близости правильному ответу

Испытуемые	Иванов Е.А.	Петров М.Н.	Сидоров Д.Л.	Зайцев Е.А.	Волков А.В.
Вопрос 1	0.857	1.000	0.4	1.000	0.857
Вопрос 2	1.000	0.733	0.868	0.75	0.545
Вопрос 3	0.75	0.63	0.000	0.703	0.42
Вопрос 4	0.861	0.861	0.717	0.662	1.000
Вопрос 5	0.725	0.657	0.000	0.5	0.471

Messages

Demo-release of system is presented  
on the personal webpage of author at [www.machinelearning.ru](http://www.machinelearning.ru).

[more details](#)

[continue](#)

**Результат по испытуемому**

Испытуемый: Петров М.Н.

Вопрос теста (вопрос №3):

Как влияет переподгонка на частоту ошибок дерева принятия решений ?

Полученный ответ:

Именно с переобучение связана увеличение частоты ошибок дерева принятия решений на контрольной (= тестовой) выборке.

Наиболее близкий вариант правильного ответа:

Увеличение частоты ошибок дерева принятия решений на контрольной выборке связано с переподгонкой.

Численная оценка близости правильному ответу: 0.63

Оценка за ответ: удовл.

go back

continue

# Group testing's results after the coordination of knowledge about synonymy concerning the different situations of Russian language's usage

Тестирование знаний и подготовка к ЕГЭ

База знаний Тесты Первое знакомство Window Помощь

Численные оценки близости правильному ответу

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Вопрос 1	0.857	1.000	0.4	1.000	0.857
Вопрос 2	1.000	0.733	0.868	0.75	0.545
Вопрос 3	0.75	0.652	0.000	0.703	0.42
Вопрос 4	0.913	0.913	0.717	0.595	0.89
Вопрос 5	0.725	0.657	0.000	0.5	0.471

Messages

- Case 1. *Incomplete answer* when for all words and their combinations from trainee's answer the prototypes in the most similar «correct» variant were found but *for some words of correct answer no prototypes in the trainee's answer* were found.  
*Not-nil value of similarity* with the object from the correct answer's USNL's formal context will be *only for missed word syntactically submitted to some other word presented both in analyzed and «correct» variant.*
- Case 2. *Orthographic errors (which are admissible)* when a word from trainee's answer and a word from the variant of correct answer are the same word's *different forms admissible within the frameworks of the same known lexico-syntactic link.*
- Case 3. *«Excess» words* when the analyzed answer has a words which hasn't prototypes in «correct» answer's «variant».  
In this case the trainee's *answer* will *not be* considered as *incorrect* only if the *«excess» words don't appear in any lexico-syntactic link* presented in system's knowledge base.

- In offered USNL's conception all kinds of links between main and dependent word were assumed as equally significant.  
To apply such estimations in the tasks of testing of knowledge relatively to concrete subject areas it is necessary to *re-define the affinity of NL-usage's situations from viewpoint of fuzzy logic*.
- Here the *systems analysis of structure of professional knowledge* for the specific area is necessary for the *description of membership functions of fuzzy sets*.
- *Duquenne–Guigues set of implications* of NL-usage situation's formal context can be a basis of *development of strategies and rules of syntactic analysis*.
- The offered conception of phrase's linear structure's model can be more versatile at applying the *probabilities of coexistence of words* in texts related to given subject area and genre.