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AUTOMATIC TEXT TRANSLATION SYSTEM FOR ARTIFICIAL LLANGUAGES

The growing number and variety of artificial languages leads to the need and relevance of creating automatic dictionaries for their translation in order to facilitate human communication. Such languages include languages where vocabulary, phonetics, and grammar have been specifically designed to achieve specific goals and to communicate with a group of people by interests or place of residence. These languages can be distributed among people of certain professions or among neighboring nations. Examples are slang and surzhik. The common for them is that there is a basic language (literary), the intersection in spelling and meaning of words and phrases with which is quite large. The main goal of the project is to create a system of automatic translation of words and texts from / into arbitrary languages, including hybrid, artificial and slang ones. The proposed model shows the interaction and partial interdependence of the creation and adjustment modules and the translation module of the dictionary, which is explained by tacking the approach of reverse propagation of the translation error. To perform experiments and analyze the performance of the proposed approach to the organization of automatic translation of texts from and into arbitrary language, a software application was developed, which includes a subprogram of initial word processing for dictionary organization, one for creating a working dictionary and one for two-way improvement of created dictionary by the inclusion of new texts in order to improve the quality of translation, including the search for word phrases, idiom, and translation for them, the subprogram of dividing the dictionary into sub-dictionaries with a small percentage of text, the subprogram of the translator itself. To test and analyze the results of the proposed generalized model, three types of source texts were used: literary poetry translation, literary prose translation, literal prose translation. The results of the experiments showed that the proposed approach provides a high level of translation (up to 98,8%) in similar languages (between such languages as Ukrainian-Russian, or Ukrainian - Ukrainian-Russian surzhik wih equal word order in the sentence), especially with a literally translated source text. It has become known that the use of artistic texts to generate dictionaries is possible, but not very effective.

Keywords: dictionary, processing, model, text, language, slang, computer system.

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СИСТЕМА АВТОМАТИЧНОГО ПЕРЕКЛАДУ ТЕКСТУ ДЛЯ ШТУЧНИХ МОВ

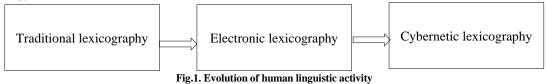
Зростання кількості та різноманіття штучних мов призводить до необхідності та актуальності створення автоматичних словників для їх перекладу із метою полегшення людського спілкування. До таких мов відносяться мови, де лексика, фонетика і граматика були спеціально розроблені для втілення певних цілей і спілкування групи людей за інтересами або місцем проживання. Прикладами є сленги та суржики. Спільним для них є те, що для них існує базова мова (літературна), перетин за написанням і значенням слів та словосполучень із якою є досить великий. Метою роботи є створення узагальненої моделі системи автоматичного перекладу текстів з довільної на довільну мови, включаючи штучні та гібридні мови. Запропонована модель показує взаємодію та часткову взаємозалежність модулів створення та коригування словнику та модулю перекладу, що пояснюється підходом зворотнього розповсюдження помилки перекладання. Для виконання експериментів було розроблено програмний застосунок, що включає в себе підпрограму первісної обробки тексту для організації словнику, підпрограму створення робочого словнику, підпрограму двостороннього покращення створеного словнику шляхом доповнення новими текстами для підвищення якості перекладу, включаючи пошук словосполучень, фразеологізмів, та перекладу для них, підпрограму розбиття словника на підсловники при малому відсотку перетину тексту, підпрограму перекладача. Для тестування та аналізу результатів роботи запропонованоузагальненемодель, використано три типи початкових текстів: художній переклад поезії, художній переклад прози, дослівний переклад прози. Результати показали, що запропонований підхід дає високий рівень перекладу (до 98,8%) для подібних мов (таких, як українська та російська, або українська та українсько-російський суржик із однаковим порядком спів у реченні), особливо при дослівно перекладеному початковому тексті. Стало відомо, що використання художніх текстів для генерації словників можливе, але не є ефективним.

Ключові слова:словник, обробка, модель, текст, мова, сленг, комп'ютерна система.

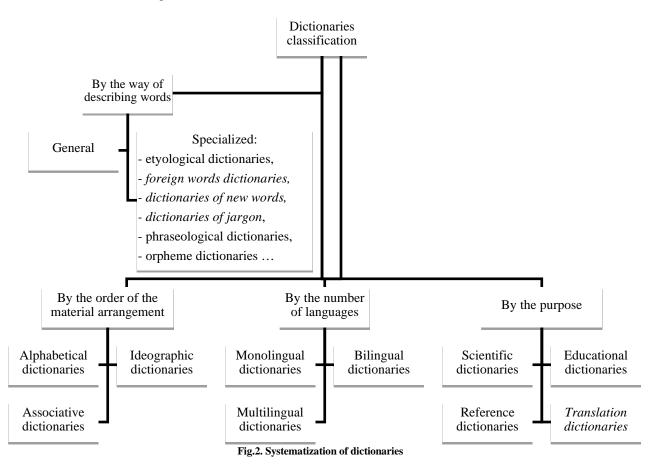
Introduction

Lexicography is a very old type of linguistic activity, as evidenced by the first dictionaries that appeared as early as the late twelfth century, to perform informative (allowing to access the accumulated knowledge in the shortest way - via notation) and normative functions (through fixating the meaning and usage of words, they contribute to improvement and unification of language as a means of communication), as well as to reflect the knowledge that society has in a given era. Further development of computer and computing technologies has led to the emergence of a new field in lexicography – computer (or electronic) lexicography, which provides the ability to use lexicographic resources through computers or other electronic devices. In a broad sense, the main object of computer lexicography research is the electronic dictionary [1,2]. Similar terms include "machine dictionaries", "automatic dictionaries", "automated dictionaries", "computer dictionaries" [3]. The basis for the creation of electronic dictionaries are scienceintensive methods in the field of linguistic analysis, word processing, lexicographic databases, machine translation (statistical machine translation, neural machine translation, machine translation based on rules, hybrid machine

translation), etc.



The wide spread and mass access to the Internet marked the beginning of the development of such field of lexicography as cybernetic lexicography (cyberlexicography) (Figure 1), which involves the use of the Internet to compile a dictionary. Many sources on pedagogical or philological sciences provide detailed classifications of dictionaries. One such example is [4].



The choice of using a particular dictionary depends on the task at hand, such as translating a text or interpreting a term. Thus, translation dictionaries are designed to compare and transition from one language system to another (Ukrainian-Belarusian, English-French, etc.), while special dictionaries, such as regional and private dialects, reveal lexical meanings within a single dialect or dialect. (dictionary of jargon and slang) (Figure 2) [5, 6].

Related works

Nowadays, it takes a lot of time to translate slang or surzhik, as it is difficult to find a specialized translator. All popular translators can only translate into a certain number of languages (Table 1), and it is quite difficult to scale such dictionaries by adding new words for translation.

Analysis of the analogues

Table 1

Analysis of the analogues				
Comparison criterion	Google translate	Yandex. translator	www.dictionary.com	ABBYY Lingvo
Number of supported languages.	104	95	1	20 (220 vocabularies)
Access	Free.	Free.	Free.	Paid.
Supported platfroms	Windows, Linux; separate apps for IOS and Android.	Windows, Linux; separate apps for IOS and Android.	Windows, Linux, Android, IOS	Windows, Android, IOS. Requires separate installation for each platform.
Offline availability	Online only.	Online only.	Online only.	Both online and offline

					modes is available.	le.
	Using the table 1, one can understand that Google translate and Yandex translator have the most advantages,					
h	however they only provide the translation from and into official state languages, without supporting the possibility of					
expanding with new artificially created words and translating slang vocabulary and artificial languages. Given that						
th	the number of artificial languages today is very large and constantly growing, new slangs are being generated due the					
influence of languages on each other and depend on a certain area, the creation of separate dictionaries for each of						
them is not appropriate. Slang is similar to its language of origin, with the addition or change of a small number of						
W	vords. However, it is no	ot always conveni	ient to understand ar	nd communicate freely with	a group of people who have	ple who have
а	certain slang. In artifi	cial languages, vo	ocabulary, phonetics	and grammar are specially	designed to embody certain	body certain
goals, for example, to communicate with a group of people affected by a particular book, movie, game, and so on. It						
is	is purposefulness that distinguishes artificial languages from natural ones [7].					

Considering all the above, the creation of an automatic translator of slang and artificial words into official languages is an urgent and demanded task in today's world.

The analysis of scientific works devoted to machine translation shows that to mobile applications for instant translation in order to expand the vocabulary [8], real-time natural language translators for a limited number of foreign languages [9] enough attention is paid. Translators are often a built-in module in NLP systems to determine the mood and emotional state of dialogue participants in different languages [10, 11]. In spite of practical application of translators, the basic algorithms in most proposed are such algorithms and numerical methods as Knut-Morris-Prata [12], TF-IDF, word bag, neural networks, etc. Most studies are conducted for popular languages, such as English, French, Spanish, but make mistakes in the translation of language that are first provided to the system, or little known. Multilingual translators are promising, where the target language is not only English, but any other language.

The goal of the given work

The growing number and variety of artificial languages leads to the need and relevance of creating automatic dictionaries for their translation in order to facilitate human communication. Such languages include languages where vocabulary, phonetics, and grammar have been specifically designed to achieve specific goals and to communicate with a group of people by interests or place of residence. These languages can be distributed among people of certain professions or among neighboring nations. Examples are slang and surzhik. The common for them is that there is a basic language (literary), the intersection in spelling and meaning of words and phrases with which is quite large.

The object of this study is an automatic dictionary that can be used to translate and interpret slang or artificial vocabulary, as one that is created at the intersection of several languages and has many common (borrowed) spellings and meanings of words from different languages.

The main goal of the project is to create a model of the system of automatic translation of words and texts from / into arbitrary languages, including hybrid, artificial and slang ones.

- To achieve this goal, the following tasks must be solved:
- implementation of the possibility to use the prepared dictionary for translation;
- expansion of the vocabulary;
- creating new dictionaries based on existing ones;
- development of approach and creation of inverted dictionaries.

Proposed model

The paper proposes a generalized model of the system of automatic translation of texts from and into arbitrary language (Figure 3). This system primarily provides the ability to translate even from / into artificial languages, such as Esperanto, Talos, Interlingua, Lojban and others.

The given model shows the interaction and partial interdependence of the creation and adjustment modules and the translation module of the dictionary, which is explained by tacking the approach of reverse propagation of the translation error.

The texts submitted to the input of the proposed model must be presented in two arbitrary languages and have the same number of sentences. Sentences are counted by the number of dots ("."), Exclamation mark ("!"), Question mark ("?") And three dots ("..."). In addition, each sentence must have at least one word. For example, if there are only spaces, commas, quotation marks, and other punctuation marks between the exclamation mark and the question mark, this will not be considered a sentence.

In addition, sentences with the same sequence numbers will be considered translations of each other, ie changing the order of sentences will lead to incorrect operation of the program.

Let's explain the work of the module for creating working dictionaries, which will be the basis for the translation of source texts.

Dictionaries use an algorithm that calculates the relative probability of translation based on the location of words in a sentence, the percentage of corresponding sentences in which both words occurred, and their similarity using the trigram method and the Knut-Maurice-Pratt algorithm [13-15].

However, since it is not always possible to find the same text in both selected languages, the paper proposes to build a dictionary through existing ones.

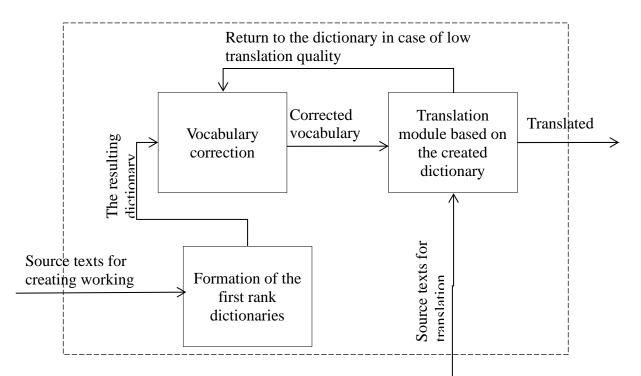
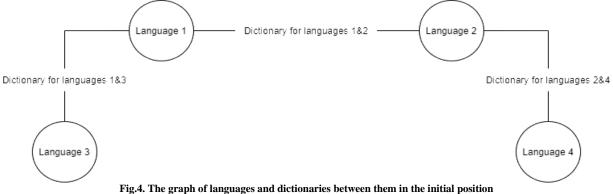
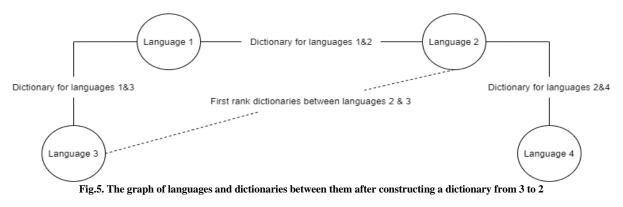


Fig.3. Generalized model of the system of automatic translation of texts from and into arbitrary language

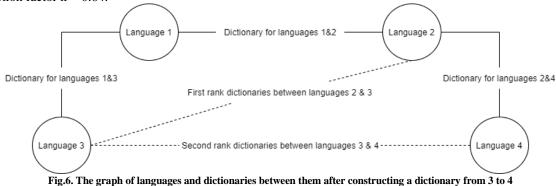
Let languages 1 and 2 be known languages, languages 3 and 4 be slang. In addition, there is a two-way dictionary between languages 1-2, and there is also a two-way dictionary between languages and their slang. Let us mark this on the graph, where each language is a vertex and the dictionary is an edge (Figure 4).



Since there are edges between languages 3 and 1 and between 1 and 2, for each word from the 3rd language we will try to find a translation into language 2 using a dictionary from the first to the second language. Since this is not an exact translation, the probabilities of translations will be multiplied by each other, and multiplied by a correction factor k, $0.8 \le k \le 1$ (k = 1 if all probabilities are less than one). Let's mark the new dictionary between languages 2 and 3 with a dotted line on the scheme (fig. 5).



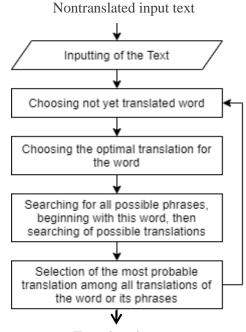
Using the same idea, we will build a dictionary through languages 3-2-4. Then we get a dictionary with a correction factor k = 0.64.



This dictionary can also be further used to build new ones, but it is not recommended, as dictionaries with rank 3 or higher contain very low probabilities of relative translations, and their use is not appropriate (dictionary rank is defined as the sum of ranks of dictionaries that built it, increased by one) (Figure 6).

To improve the quality of newly constructed dictionaries, it is mandatory and necessary to supplement the finished dictionary with new texts. To do this, in previous iterations it is necessary to save not only the word, possible translations and the probability of these translations, but also the number of processed words and the total size of sentences on which the dictionary was built.

After building the dictionary, we can use it to translate texts. When generating a dictionary, it is possible to build not only a dictionary of word-for-word translations, but also a word-for-phrase and vice versa. In order to choose the best opportunity to translate phrases, the following algorithm was used (Figure 7).



Translated text

Fig.7. Algorithm for translating text with a phrases dictionary

As a result of the algorithm, the input text will be translated using dictionaries that were submitted to the translator input. The algorithm tries to maximize the total probability of correct translation by maximizing the sum of the probabilities of the relative translation of each word or phrase that have been recorded in dictionaries.

The results of the experiments

To perform experiments and analyze the performance of the proposed approach to the organization of automatic translation of texts from and into arbitrary language, a software application was developed, which includes a subprogram of initial word processing for dictionary organization, one for creating a working dictionary and one for two-way improvement of created dictionary by the inclusion of new texts in order to improve the quality of translation,

including the search for word phrases, idiom, and translation for them, the subprogram of dividing the dictionary into sub-dictionaries with a small percentage of text, the subprogram of the translator itself.

To test and analyze the results of the proposed generalized model, three types of source texts were used:

- literary poetry translation;
- literary prose translation;
- literal prose translation.

In addition, in Esperanto we will replace the \hat{c} , \hat{g} , \hat{h} , \hat{j} , \hat{s} , \check{u} Ha c, g, h, j, s, u characters to avoid encoding problems.

For a literal translation of prose, we use the texts shown in Figures 8a and 8b.

Автор эсперанто, Лазарь Людовик Заменгоф, в 1878 будучи в последнем классе гимназии закончил первый проект язык. И учась в университете, Заменгоф на протяжении полных семи лет до 1885 года дорабатывал свой язык. Наконец, в 1887 в Варшаве появился первый учебник под заголовком "Международный язык Доктора Эсперанто". Автор издал учебник не под своим имеменем, а под псевдонимом "Д-р Эсперанто", т_е "Тот кто надеется". Теперь и сам язык назван этим псевдонимом. В 1913 году под руководством известного французского писателя Тристана Бернард эсперанто проэкзаменовали в специальной комиссии. Определённый литературный текст был переведён на многие языки, между прочим также и на эсперанто. а затем тот же текст снова был переведён на многие языки, между прочим также и на эсперанто. Результат показал, что самый совершенный перевод был сделан с помощью эсперанция) в 1905 году.				
a)			
La autoro de esperanto Lazary Ludovik Zamenhof en 1878 esan nel laste klaso de gimnazio endi et une projekto del lingo. Trastudan nel universiteto, Zamenhof dum plene sep yaros ad la yaro 1885 prosesari e sue lingo. Enden, nel yaro 1887 en Varshava aperi la une manuelo sub la titulo "Internacia Lingvo de Doktoro Esperanto". La autoro no eldoni e manuelo sub sue namo, bat sub la pseudonimo "D-ro Esperanto", s_e "Kvele espera". Prezenten la lingo same es namate segun eze pseudonimo. Nel yaro 1913 sub la gvido de populare franse skritero Tristan Bernard os exzamini et esperanto en spesiale comiseo. Definite literature texto esi tradukite al multe lingos, inter otren gam al esperanto, ve posten la same texto esi retradukite per otre tradukistos, inkonante et originale texto, al lingo del originalo. La rezulto montri, ke la plu perfekte tradukos esari farite per helpo de esperanto, kar la retradukite texto from esperanto esi la plu proxime al originalo. La une universale kongreso de esperanto esari organizate en Bolongne-sur-Mer (Fransia) nel yaro 1905.]				
ł				
Fig.8. Corresponding texts in different l	anguages: a) in Russian, б) in Esperanto			
After using the program we get the following two dictionaries.				
1885: 1:85: 0.85: 0.95: Голована 0.75: Голована 0.75: 0.75: 0.86: 1.89: 1.88: 0.86: 1.89: 0.86: 1.89: 0.86: 1.89: 0.86: 1.89: 0.86: 1.89: 0.86: 0.87: 0.86: 0.87: 0.97: 0.87: 0.87: 0.97: 0.87: 0.87: 0.97: 0.87: 0.87: 0.97: 0.87: 0.87: 0.97: 0.87: 0.87: 0.97: 0.87: 0.87: 0.97: 0.87: 0.87: 0.97: 0.87: 0.87: 0.97: 0.87: 0.87: 0	1878: 1878 0.8619 en 0.7519 ludovik 0.75595 lazary 0.71191 zamenhof 0.6 1885: 1885 0.955 yaro 0.75 prosesari 0.75 la 0.7 e 0.7 1887: 1887 0.89921 yaro 0.8 nel 0.755556 en 0.71191 landen 0.69841 1905: fransia 0.8 1905 0.76428 nel 0.724266 bolongnesummer 0.73809 en 0.68571 1913: 1913 0.90228 yaro 0.8 nel 0.75789 sub 0.71579 la 0.67308 a 1 a 0.63154 esperanto 0.61864 sue 0.4 posten 0.2 ve 0.19399 aarop: 1a 0.8 autoro 0.75436 esperanto 0.43175 de 0.3619 no 0.18334 6epuapa: skritero 0.8 franse 0.75282 tristan 0.7520 sed 0.76167 retradukter 0.67692 6onne: organizate 0.8 esari 0.74286 en 0.74286 esperanto 0.68571 bolongnesummer 0.67619 6yyyu: 1878 0.75595 ludovik 0.71191 esan 0.71191 zamenhof 0.56454 sen 0.5 6ux: per 0.46573 texto 0.45955 esperanto 0.43084 retradukte 0.44551 a 0.37015 apuase: en 0.8 1807 0.74921 varshava 0.74921 yaro 0.71111 laperi 0.69841 accoduwik : une 0.8 1807 0.74921 varshava 0.74921 yaro 0.71111 laperi 0.69841 accoduwik : une 0.8 180 0.7599 cos 0.5157 la 0.4361 (m 0.43651 m 0.4155 a) rosesari 0.75 1885 0.75 yaro 0.7 sue 0.7 e 0.5625 roay : yaro 0.5601 nel 0.52828 de 0.51579 la 0.4361 (m 0.43651 m 0.51055 aprosesari 0.75 1885 0.75 yaro 0.7 sue 0.7 e 0.5 sue 0.5625 aorta : tradukisto 0.8 otre 0.77376 inkonante 0.77376 per 0.7283 same 0.58043 aprosesari 0.75 1885 0.75 ad 0.7 prosesari 0.72258 pare 0.7228 same 0.58043			

Fig.9. The result of the application. A dictionary for 202 words of the working Russian-Esperanto dictionary has been created

As a result of work among 202 words in both dictionaries 5 corresponding translations for each word with a probability of more than 0.3 were found, and they were arranged in descending order of probability. Translations with a lower probability are considered as noise. Translations with the highest probability are considered as correct translation (Figure 9).

The analysis of the obtained translation result showed that 50% of words in the dictionary have translations with the highest probability and correspond to the actual translation of the word. 90% of words also have their actual translation in the dictionary among the five options offered, although not with the highest probability.

To test the literary translation, we will use the beginning of Bulgakov's novel "The Master and Margarita". We use an excerpt from the text shown in Figures 10a and 10b.

Однажды весною, в час небывало жаркого заката, в Москве, на Патриарших прудах, появились два гражданина. Первый из них, одетый в летнюю серенькую пару, был маленького роста, упитан, лыс, свою приличную шляпу пирожком нес в руке, а на хорошо выбритом лице его помещались сверхъестественных размеров очки в черной роговой оправе. а на хорошю выорятом лице его помещались сверхњестественных размеров очки в черной роговой оправе. Второй – плечистый, рыкеватый, викрастый молодой человек в заломленной на затылок клетка с был в ковбойке, жеваных белых брюках и в черных тапочках. Первый был не кто иной, как Михаил Александрович Берлиоз, председатель правления одной из крупнейших московских литературных ассоциаций, сокращенно именуемой и редактор толстого художественного журнала, а молодой спутник его – поэт Иван Николаевич Понырев, пишущий под псевдонимом Бездомный. Попав в тень чуть зеленжщих лип, писатели первым долгом бросились к пестро раскращенной будочке с надписью "Пиво и воды". именуемой МАССОЛИТ, Да, следует отметить первую странность этого страшного майского вечера. Не только у будочки, но и во всей аллее, параллельной Малой Бронной улице, не оказалось ни одного человека. В тот час, когда уж, кажется, и сил не было дышать, когда солнце, раскалив Москву, в сухом тумане валилось куда-то за Садовое кольцо, – никто не пришел под липы, никто не сел на скамейку, пуста была аллея. a) Foje en Moskvo, dum malnormale varmega printempa sunsubiro, du civitanoj venis la gardenplacon de la Patriarha lageto. La unua, vestita per griza somera kompleto, estis malalta, diketa, kalva, sian malfrivolan capelon li portis en la mano, kaj sur lia bone razita vizago vastis kolosaj okulvitroj en nigra korna muntumo. La dua, largasultra juna viro kun senorda rufeta hararo kaj kvadratita kaskedo sur la nuko, surhavis buntan cemizon, cifitan blankan pantalonon kaj nigrajn sportosuojn. La dua largasultra juna viro kun senorda rufeta hararo kaj kvadratita kaskedo sur la nuko, surhavis buntan cemizon, cifitan blankan pantalonon kaj nigrajn sportosuojn. La unua estis neniu alia ol Mihaelo Aleksandrovic Berlioz, redaktoro de dika beletra revuo kaj la prezidanto de Massolit, unu el la plej grandaj Moskvaj literaturiasocioj, lia akompananto estis la juna poeto Ivano Nikolaic Ponirjov, verkanta sub la pseudonimo Senhejmulo. Atinginte la ombron de apenau verdigintaj tiloj, la literaturistoj tuj jetis sin al bunta budo kun surskribo "Biero kaj Trinkajoj". Nu, oni notu la unuan strangajon de tiu terura maja vespero. Ne nur antau la budo, sed en la tuta aleo paralela al la Malgranda Bronnaja strato, ne estis ec unu homo. Je horo, kiam oni apenau povis spiri la varmegan aeron, kiam la suno, ardiginte Moskvon, en seka nebulo estis falanta ien trans Sadovaja-strato, neniu venis sub la tiliojn, neniu sidigis sur benkon, malplena estis la aleo. b) Fig.10. Corresponding texts in different languages: a) excerpt from the novel in the original language; 6) an excerpt from the novel translated into Esperanto As a result, we obtain the following dictionaries (Figure 11). a : lia 0.77739 kaj 0.6775 estis 0.4075 sur 0.4 literatur'asocioj 0.2 | anece: anglesa 0.7211 este 0.7261 en 0.7281 paralela 0.7764 se berlor 0.7784 (este 2.584 redactoro 0.7585 anee: into 0.8 alo 0.7619 en 0.7281 paralela 0.7764 se berlor 0.7784 (este 2.584 redactoro 0.7585 anee: into 0.8 alo 0.7619 en 0.7281 paralela 0.7764 se berlor 0.7784 (este 2.584 redactoro 0.7585 anee: into 0.8 alo 0.7619 en 0.7281 paralela 0.7764 set 0.65571 [convanue: jrecidanto 0.6 kaj 0.76 masolit 0.7584 venta a.5884 perifor 0.4 alub 0.58474 (este 1. citizen 0.0 aleb 0.6769 bio.7084 periforma a.5784 periforma a.7894 periforma a.7794 periforma a.7794 periforma a.7794 periforma a.7894 periforma a.7894 periforma a.7794 periforma a.7794 periforma a.7794 periforma a.7895 periforma a.7894 periforma a.7794 periforma a.7794 periforma a.7794 periforma a.7794 periforma a.7794 periforma a.7994 periforma a.7994 periforma a.7994 periforma a.7794 periforma a.7794 periforma a.7994 periforma aeron : 6wro 0.8 джаать 0.77808 смл 0.75016 не 0.19459 к 0.18379 | acompaninto : слутиж 0.8 дко 0.58424 молодой 0.18474 - 0.18847 и 0.18947 | al : раскращенной 0.2 броней 0.2 пестро 0.18843 будоже 0.18434 и 0.14035 | al : раскращенной 0.2 броней 0.2 пестро 0.18843 будоже 0.18484 и 0.14035 | al : раскращенной 0.2 броней 0.2 лестро 0.18843 будоже 0.18443 и 0.14035 | al : раскращенной 0.2 болы 0.2 лаихел 0.19489 пуста 0.15469 и 0.0744 | al : раскращенной 0.2 болы 0.2 лаихел 0.19489 пуста 0.15469 и 0.0774 | al : раскращенной 0.2 былы 0.2 лаихел 0.19489 пуста 0.15469 и 0.0774 | aproxal : логеневарк 0.2 ук 0.2 кажется 0.19489 пуста 0.15469 и 0.0774 | aproxal : логеневарк 0.2 ук 0.2 кажется 0.19489 пуста 0.75469 леокоско 0.75680 и 0.02000 (.75680 леокососко 0.7568 и 0.02000 (.75680 леокососко 0.7568 и 0.02000 (.75680 леокососко 0.75680 и 0.02000 (.75680 леокососко 0.7568 и 0.0200 (.7568 леокосокос 0.7568 и 0.0200 (.7568 леокосокос 0.7568 и 0.0200 (.7568 леокосокос 0.7568 и 0.0200 (.7568 и 0.020 ron : было 0.8 дышать 0.77808 сил 0.75616 не 0.19459 и 0.18379 | отранапто : спутник 0.8 его 0.58248 молодой 0.19474 – 0.18947 а 0.18947 | отранапто : спутник 0.8 его 0.58248 и 0.14035 |

a lower probability are considered as noise. Five such words were found. Translations with the highest probability are considered as a correct translation. The analysis of the obtained translation result showed that 15% of words have translations in the dictionary with the highest probability and correspond to the actual translation of the word. 40% of words also have their actual

probability of more than 0.3 were found, and they were arranged in descending order of probability. Translations with

translation in the dictionary among the five options offered, although not with the highest probability (Figure 11). Such a low result was obtained because words that occur only once are taken as the most probable translations of those that stand in their place in another language. However, among non-noise words that occur two or more times, more than 40% have the most probable correct translation, and more than 80% contain the correct translation among those suggested. In conclusion, if a literary translation of the text is to be used, a large amount of original text is required.

To test the translation of poetry, we use the text shown in Figures 12a and 12b.

As a result, we obtain the following dictionaries (Figure 13).

Windows (CRLF)

As a result of work among 181 words in both dictionaries 5 corresponding translations for each word with a probability of more than 0.3 were found, and they were arranged in descending order of probability. Translations with a lower probability are considered as noise. Only 1 noise word appeared in the given text. Translations with the highest probability are considered as a correct translation.

The analysis of the obtained translation result showed that 25% of words in the dictionary have translations with the highest probability and correspond to the actual translation of the word. 60% of words also have their actual translation in the dictionary among the five options offered, although not with the highest probability. (Figure 13).

Among all three tested variants, the best among the words that occur only once is a literal translation of the

Fig.11. The result of the application

As a result of work among 304 words in both dictionaries 5 corresponding translations for each word with a

Pg 158, cran 74

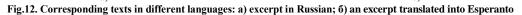
Windows (CRLF)

text. Words that are used more than once are more likely to be translated. In addition, all the words that were considered noise were in Esperanto. This may be due to the specifics of the language, as this language contains nom-translatable words, such as articles.

Над землёю новым чувством веет, И призыв разносится по свету; Он на крыльях, словно лёгкий ветер, Облетает быстро всю планету. Не мечом — орудьем разрушенья — Он сплотить желает все народы: Людям, утомившимся в сраженьях, Обещает мир он и свободу. Все, кто верит в счастье дней грядуших. Собрались под мирные знамена, Под звездой идеи всемогущей Стали мы сильнее и сплочённей. Стены недоверья вековые Разобщили страны и народы; Но сметет препятствия любые Светлая любовь — сестра свободы. Овладевши языком общенья, Хорошо друг друга понимая, Будут жить народы, без сомненья, Как семья единая земная. Полные энергии и страсти, Не устанем яростно трудиться, Чтоб мечта, которой нет прекрасней, Для людей могла осуществиться.

a)

En la mondon venis nova sento, tra la mondo iras forta voko; per flugiloj de facila vento nun de loko flugu gi al loko. Ne al glavo sangon soifanta gi la homan tiras familion: al la mond' eterne militanta gi promesas sanktan harmonion. Sub la sankta signo de l'espero kolektigas pacaj batalantoj, kaj rapide kreskas la afero per laboro de la esperantoj. Forte staras muroj de miljaroj inter la popoloj dividitaj; sed dissaltos la obstinaj baroj, per la sankta amo disbatitaj. Sur neutrala lingva fundamento, komprenante unu la alian, la popoloj faros en konsento unu grandan rondon familian. Nia diligenta kolegaro en laboro paca ne lacigos, gis la bela songo de l'homaro por eterna ben' efektivigos. b)



Texts in the following pairs of languages were used to test the translation in Cyrillic languages: Russian and Ukrainian; Ukrainian and Old Ukrainian; Ukrainian and Russian-Ukrainian surzhik; Ukrainian and Russian-game surzhik.

afero : всемогущей 0.8 идеи 0.76 стали 0.76 звездой 0.72 мы 0.72 al : он 0.13 народы 0.10333 все 0.10333 желает 0.10333 сплотить 0.10333 alian : будут 0.8 лонимая 0.74853 жить 0.74853 друга 0.69706 народы 0.175 baro : свободы 0.6921 сестра 0.63816 - 0.15 лобовь 0.1375 светлая 0.125 baro] : лобовь 0.74606 сестра 0.63816 - 0.15 лобовь 0.1375 светлая 0.125 baro] : лобовь 0.74606 сестра 0.63816 - 0.15 лобовь 0.1375 светлая 0.127 bela : которой 0.8 мечта 0.75163 нет 0.75163 чтоб 0.79326 ргерасней 0.70326 ber: : осуществиться 0.8 могла 0.75163 чтоб 0.79326 ил 0.1647 лрякрасней 0.70326 de : и 0.41025 бистро 0.01 diligenta : знергии 0.48 могла 0.75163 страсти 0.70326 ил 0.1627 прекрасней 0.15294 de : и 0.41025 бистро 0.01 dissaltos : лобие 0.8 препятствия 0.74666 сегтая 0.74526 ми 0.18824 не 0.1647 dissaltos : лобие 0.48 препятствия 0.74666 слетая 0.74666 метет 0.6521 народы 0.175 efektivigos : осуществиться 0.75163 могла 0.7326 би лодей 0.1647 для 0.15294 прекрасней 0.14118 en : и 0.36438 над 5е-06 страсти 5е-06 Соименыя 5е-06 esperantoj : сплочёнией 0.8 силынее 0.72 ма 0.7836 дилее 0.621 народы 0.175 tetrem : уполившикся 0.75163 осуществиться 0.75163 для 0.721 од 0.18 espero : приучик 0.8 лейей 0.75667 облагьс 0.7567 слонию 0.71333 бистро 0.71333 familian : желает 0.75894 сильнее 0.726 ми 0.626 ми 0.151 ссимень 0.127 facila : ветер 0.8 лёгкий 0.75667 лёгкий 0.75667 лоною 0.71333 бистро 0.71333 familian : желает 0.75894 сильти 0.677667 ма 0.71333 истро 0.71333 familian : желает 0.75894 силотить 0.72169 ладии 0.7129 все 0.6 надоды 0.50692 fuglioj : сповно 0.8 крыльки 0.75667 лёгкий 0.75667 лаориди 0.50693 flugu : планету 0.75667 иле сий 0.75667 лёгкий 0.75667 лаориди 0.50693 flugu : планету 0.75667 иле сий 0.71333 крыльки 0.67366 сволоды 0.63524 faros : беё 0.8 соменьель 0.74853 дриная 0.7129 все 0.64554 дроди 0.65562 fort : ссете 0.8 кеденева. 0.74653 дриди 0.7129 все 0.64554 дроди 0.65562 fort : ссете 0.8 кеденева. 0.7267 дрании 0.7129 все	6cs : faros 0.8 en 0.75294 konsento 0.70294 alian 0.65882 popoloj 0.56397 6yдyr : alian 0.8 la 0.75294 konsento 0.70294 alian 0.65882 unu 0.65294 6wcrpo : nun 0.8 vento 0.76333 facila 0.72667 loko 0.68667 de 0.47 a: la 0.40477 santa 0.83 de 0.38 eterne 0.2 signo 0.2 eecre: nova 0.8 vento 0.76333 sento 0.76333 mondon 0.72667 tra 0.72667 eecoma: runoj 0.8 de 0.7589 sterne 0.2 signo 0.2 eerer : facila 0.8 vento 0.76333 sento 0.76333 mondon 0.72667 tra 0.72667 eecoma: runoj 0.8 de 0.7589 stars 0.7394 forte 0.70708 miljaroj 0.70789 eerpe: facila 0.8 vento 0.76333 fulgiloj 0.72667 nun 0.72667 de 0.50333 cceworywei : afero 0.8 pe 0.76 kreskas 0.72 lahoro 0.72 rajide 0.68 cceworywei : afero 0.8 ge pp 0.76 kreskas 0.72 lahoro 0.72 rajide 0.68 cceworywei : afero 0.8 de 0.7565 ppr 0.7545 songo 0.72464 land 0.7294 alian 0.7294 paper : la 0.75294 komprenante 0.75147 fundamento 0.75494 familian 0.7294 alian 0.70294 papr : komprenante 0.8 unu 0.75124 fundamento 0.7547 familian 0.7294 alian 0.70294 papr : la 0.73294 komprenante 0.75147 familian 0.75147 konsento 0.75148 wenta : lina 0.8 homan 0.75683 familion 0.75147 konsento 0.653182 lingva 0.721 papr : la 0.43510 de 0.76 saj 0.72 afero 0.72 batalantoj 0.68 wenta : di 0.8 badiantoj 0.75147 familian 0.75147 konsento 0.65382 wenta : di 0.8 badianto 0.75633 facilo 0.7224 gai 0.27267 pe 0
Windows (CRLF) Pa 86, cten 70 100%	Windows (CRLF) Pg 95, cten 76 100%

Fig.13. The result of the application

The efficiency and example of the application is demonstrated for the Ukrainian-Old Ukrainian language pair, for which Taras Shevchenko's work "On Easter Day Upon the Straw" was used in the original and in an adapted form to generate a dictionary. The text is shown in Figures 14a and 14b.

На Великдень, на соломі На Великдень, на соломі Проти сонця, діти Проти сонця, діти Грались собі крашанками, Грались собі крашанками, Та й стали хвалитись Та й стали хвалитись Обновами. Обновами. Тому к святкам Тому до свят З лиштвою пошили З вишиванкою пошили Сорочечку, а тій стьожку, Сорочку, а тій смужку, Тій стрічку купили; Тій стрічку купили; Кому шапочку смушеву, Кому шапочку хутряну, Чобітки шкапові; Чобітки з шкіри; Кому свитку. Кому сорочку. Одна тілько Одна тільки Сидить без обнови Сидить без обнови Сиріточка, рученята Сиріточка, рученята Сховавши в рукава. Сховавши в рукава. - Мені мати куповала. - Мені мати купували. - Мені батько справив. - Мені батько зробив. - А мені хрещена мати - А мені хрещена мати Лиштву вишивала. Сорочку вишивала. "А я в попа обідала!" "А я в попа обідала!" Сирітка сказала. Сирітка сказала.

a) b) Fig.14. Texts for dictionary generation: a) original poem by Taras Shevchenko, б) adapted for dictionary creation

After the program, we will delete the same words and get the result shown in Figure 15.

к : до 0.8 тому 0.7619 свят 0.7619 з 0.72381 вишиванкою 0.69571 | лиштвою : вишиванкою 0.82 пошили 0.79524 з 0.7619 свят 0.74881 сорочку 0.7381 | лиштву : сорочку 0.81429 мати 0.71071 вишивала 0.71071 хрещена 0.57143 | свитку : шкіри 0.7819 чобітки 0.71429 хутряну 0.6619 шапочку 0.6381 кому 0.59643 | святкам : свят 0.92857 до 0.7619 з 0.7619 тому 0.74881 вишиванкою 0.73381 | смушеву : хутряну 0.81429 шапочку 0.77619 кому 0.77381 чобітки 0.7619 шкіри 0.70571 | сорочечку : сорочку 0.91587 пошили 0.77857 а 0.7619 вишиванкою 0.73381 тій 0.72381 | справив : зробив 0.83333 батько 0.61667 мені 0.4 | стьожку : смужку 0.86429 тій 0.79524 стрічку 0.75238 а 0.72381 купили 0.71905 | тілько : тільки 0.93333 одна 0.745 сидить 0.73667 без 0.64 обнови 0.57667 | шкапові : шкіри 0.8019 чобітки 0.77619 шапочку 0.74286 хутряну 0.72381 кому 0.67262 | Fig.15. The result of the application after deleting the same words

The result is eleven words, each of which is correctly translated and ten of which have a probability of translation greater than or equal to 0.8.

Conclusion

A model of the system of automatic translation of words and texts from / into arbitrary languages, including hybrid, artificial and slang ones was proposed. For analysis of the performance of the proposed approach to the organization of automatic translation system using the developed software application, the results were summarized, as shown in table 2.

Summary for the level of translation			
Type of tested text	The percentage of words that have the most	Percentage of words that have a valid translation among the five proposed.	
	probable translation as		
	their actual one		
A literal translation of prose between Russian and Esperanto	40,6%	90%	
Literary translation of prose between Russian and Esperanto	15%	39,7%	
Literary translation of poetry between Russian and Esperanto	25,2%	74,8%	
Literal translation of prose between Russian and Ukrainian	98,8%	98,75%	
A literal translation between Ukrainian and Old Ukrainian	98,65%	98,7%	
Translation between Ukrainian and Ukrainian-Russian	98,72%	98,84%	
surzhik			

Table 2

Translation between Ukrainian and Russian-game surzhik50,2%70%

Testing has shown that the proposed approach provides a high level of translation (up to 98,8%) in similar languages, especially with a literally translated source text. At least three different sentences with separate words are needed for correct translation. It has become known that the use of artistic texts to generate dictionaries is possible, but not very effective.

The results of the experiments also showed that the size of the source texts has a significant impact on the quality of the constructed dictionaries – the larger the volume of the source pairs of texts, the better the dictionary is generated. This applies not only to fiction, but to all texts in which the order of words in a sentence has been changed. Some language pairs may not be compatible if their rules have a different word order, ie the efficiency of dictionaries between languages such as English-Ukrainian or Esperanto-Ukrainian will be lower than between Ukrainian and Russian.

The developed test application demonstrated a fairly good translation result (up to 98,8%) between similar languages, such as Ukrainian-Russian, or Ukrainian - Ukrainian-Russian surzhik.

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