Core Lexicon and Contagious Words

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We present the new empirical parameter f_c , the most probable usage frequency of a word in a language, computed via the distribution of documents over frequency x of the word. This parameter allows for filtering the core lexicon of a language from the content words, which tend to be extremely frequent in some texts written in specific genres or by certain authors. Distributions of documents over frequencies for such words display long tails as $x > f_c$ representing a bunch of documents in which such words are used in abundance. Collections of such documents exhibit a percolation like phase transition as the coarse grain of frequency Δf (flattening out the strongly irregular frequency data series) approaches the critical value f_c .

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Studies of lists of words arranged in terms of their frequencies belong to the most important domains of quantitative linguistics [1]. Detection of most frequent words constituting the core lexicon of a language is important not only for foreign language learners, but also for various practical applications, including text compression (this was recognized as early as in [2]), speech recognition [3], information retrieval [4], etc. It is easy to order words with respect to their mean frequency of uses, which is typically measured as the number of instances of a word normalized by the sample of one million words (ipm, instances per million words); though the notion of a word should cover all word forms, like goes, went, gone for go.

However, the mean frequency is not a sufficient selection criterion, because of the large relative dispersion of the word frequencies which vary very much from one text to the next especially in ample and diverse collections of documents. Some words (like prepositions) occur in many texts with predictable rates, others (like pronouns or mental verbs) are significantly more frequent for certain writers or genres, while some are "contagious": these words (such as proper names, technical terms, abbreviations, etc.) appear in just a few documents, but when they appear, they are often found in abundance [5]. The variability of rates of words can be characterized in a variety of ways, including the Poisson K-mixtures [5]. It can be measured by the coefficient of variation (the standard deviation divided by the mean), as in [6]. However, the coefficient of variation as a measure of relative dispersion is not very useful when the average frequency is close to zero, which occurs quite often for the semantically loaded words. Another way to measure the variability of rates for contagious words is to compute the document frequency (or inverse document frequency, [4]) by counting the number of documents the given word is mentioned in, the burstiness parameter that is the mean frequency, except that it ignores documents with no instances of the word (see references in [5]), etc. As is evident, each of these parameters does not capture much of the heterogeneous structure of word rates series and none of them provides any general approach to describe it.

In this paper, we approach the problem of detection of contagious words and selection for the core lexicon of a language from a probabilistic point of view. In accordance to it, the frequency x of a word (counted as the number of its instances per million words observed in any document of a given language) is a random positive variable distributed with some (unknown) probability density function $\rho(x)$. Strongly irregular frequency data series can be flattened out by introducing the frequency coarse grain Δf . The statistics of any word w can be characterized by the number of documents $N_w(n, \Delta f) = N[(n-1)\Delta f < x \le n\Delta f]$ for which the rate of the word w drops into the n-th frequency interval for various Δf .

We have found that for any coarse grain Δf the distributions of documents $N_w(n, \Delta f)$ over n are the asymmetric curves having one maximum f_c that is the most probable frequency at which the word w would appear in a randomly chosen document written in the given language. The value of f_c is independent of genres, authors, and topics of documents and is an intrinsic characteristic of the word in a contemporary language. Obviously, f_c varies as the language evolves approaching zero as the word becomes obsolete. For the frequencies x close to the distribution maximum f_c , distributions of documents N_w over n are bell shaped, but have anomalous tails as $|x - f_c|$ is large enough. We can estimate the most probable frequencies f_c of words independently by two methods: first, from the distributions of documents $N_w(n,\Delta f)$ as $\Delta f \to 1$ and, second, from the distributions of authors using the same word in their texts. For any word, both methods gave identical values of f_c .

The most probable frequency f_c helps to detect the content words and to select words for the core lexicon of a language. Common words which appear uniformly

in most documents (like prepositions, conjunctions, relational verbs, some size adjectives, etc.) have usually relatively high mean rates \bar{f} (ipm), and $f_c \leq \bar{f}$. They obviously belong to the core lexicon of a language.

The use of semantically loaded words depends essentially on authors, genres, and topics. Despite they are not found in a bunch of documents, their mean rates \bar{f} are still very high because of their excessive popularity in certain collections of texts, but their most probable rates get down $f_c \ll \bar{f}$ indicating the presence of long tails in distributions of documents N_w over n as $n > f_c/\Delta f$. Eventually, for contagious words found in abundance in just a few documents, the distributions N_w over n have long tails, however their mean rates \bar{f} are very low since x=0 for almost all texts, and $f_c \approx \bar{f}$ or even $f_c > \bar{f}$.

Our study of word rates is based on the reference corpus of Russian [7], which includes more than $4\cdot 10^7$ words in 1566 texts, which are balanced in their coverage of various genres: fiction, newspapers, various informative texts originally written in Russian from 1980 to 2002. Unlike earlier corpora (of about 1 million words), reference corpora of this size are close to saturation, namely, any collection of new documents added to the corpus does not cause statistically significant changes to the frequency and patterns of uses of its words.

The corpus informs us about the list of 5000 words most frequently used in modern Russian [8]. The study of f_c shows that for about 60% of them their most probable frequencies is $f_c \leq 50$ ipm, and just 2.51% have $f_c \geq 600$ ipm. Among the words having typically very high most probable frequencies, one can mention conjunctions and prepositions, pronouns and relational verbs, some motion verbs, and size adjectives. The proper nouns, acronyms, technical terms and other semantically loaded words typically have comparably small values of f_c .

Document counts in the n-th frequency interval obviously decrease with n as $n > f_c/\Delta f$. We have found that it decays exponentially with n for rather small coarse grains $\Delta f \ll f_c$,

$$N_w(n, \Delta f) \propto \exp\left[-\frac{n \Delta f - f_c}{\xi_w}\right], \quad \Delta f \ll f_c, \quad (1)$$

where ξ_w plays the role of a "correlation length" of the word w and diverges as the scale Δf approaches the critical value f_c as $\xi_w \propto |f_c - \Delta f|^{-\alpha_w}$ with the positive index α_w which is close to unity for the majority of words (see the data of Table I). The parameter ξ_w casts the characteristic excess of the word frequency x over f_c .

We have observed that the values of α_w for words belonging to the same semantic group (such as relational verbs, motion verbs, perception verbs, some size adjectives, pronouns, conjunctions, and prepositions) are very close even if other values of empirical parameters measuring the variability of their rates are rather diverse.

For larger scales $\Delta f \approx f_c$, for the frequency intervals with $n > f_c/\Delta f$, the distributions N_w are scale free,

$$N_w(n, \Delta f) \propto (n \Delta f - f_c)^{-\beta_w},$$
 (2)

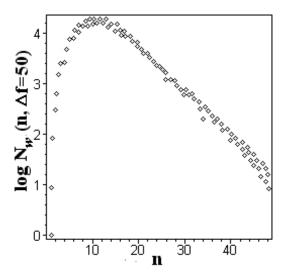


FIG. 1: The distribution of documents over the frequencies (instances per million words) for the relational verb 'imetj' (to have). The most probable frequency for this verb is $f_c = 640$ instances per million words, the coarse grain is taken as $\Delta f = 50$. The distribution has an exponential tail. Statistics on the $4 \cdot 10^7$ words, 1566 texts written in Russian from 1980 to 2002.

where the index $\beta_w > 1$ (see the data of Table I). The

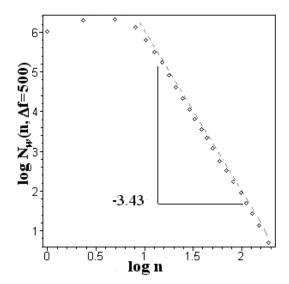


FIG. 2: The distribution of documents over the frequencies (instances per million words) for the relational verb 'imetj' (to have) has a power law decaying tail with the exponent $\beta = 3.43$ when the coarse grain Δf is taken close to $f_c = 640$. Statistics on the $4 \cdot 10^7$ words, 1566 texts written in Russian from 1980 to 2002.

data of Table I show that the value of β_w grows up with f_c almost linearly (the coefficient of linear correlation between f_c and β_w in Table I is 0.93). For the *supercritical* phase $\Delta f \gg f_c$, the tail of the probability distribution N_w forms a stretched exponential.

TABLE I: Empirical parameters measuring the variability of rates: the mean frequency \bar{f} (ipm), the number of documents in
which the word is used, the standard deviation of frequencies δf , the coefficient of variance $\delta f/\bar{f}$, the most probable frequency
f_c (ipm), the power exponents α_w and β_w . Statistics on the $4 \cdot 10^7$ words, 1566 texts written in Russian from 1980 to 2002.

Group	lemma	$ar{f}$	No. of texts	δf	$\delta f/ar{f}$	f_c	α_w	β_w
Relational	imetj (to have)	715.38	1347	717.21	1.00	640	1.089	3.433
verbs	bytj $(to be)$	10635.78	1555	4481.04	0.43	9390	1.108	5.438
Motion	idti (to go)	1029.18	1422	880.59	0.86	900	0.956	3.536
verbs	ehatj (to ride, to travel)	221.36	914	448.66	2.03	128	0.869	2.149
Perception	smotretj (to look at)	817.28	1284	930.40	1.14	540	1.201	2.574
verbs	slishatj (to hear)	306.46	1081	370.51	1.21	220	1.361	3.249
Size	bolshoy (large)	1602.30	1487	908.81	0.57	1600	0.978	5.006
adjectives	malenjkii $(small)$	386.17	1173	482.61	1.25	300	0.915	2.828
	visokii (high)	307.33	1176	404.18	1.32	300	1.116	3.016
	niskii (low)	73.01	602	172.55	2.36	60	0.824	2.043
Prepositions	v (in)	28450.99	1566	8625.48	0.30	25200	1.346	11.392
Conjunctions	i (and)	35196.38	1566	9620.76	0.27	32000	0.947	11.648
Pronouns	on (he)	17804.82	1554	10537.47	0.59	10400	0.952	5.078
	ona (she)	6651.45	1530	6118.02	0.92	3300	0.900	3.006
Abstract	vremya (time)	1830.26	1489	1167.38	0.64	1800	1.148	6.331
nouns	spravedlivostj (justice)	41.91	368	158.17	3.77	24	0.591	1.762
References	stol(table)	512.40	1147	629.38	1.23	300	1.222	2.889
to objects	dom(house)	1030.96	1351	1088.64	1.06	750	0.867	3.242
References	professor	179.23	502	939.73	5.24	40	0.636	1.423
to people	intelligentsia	62.64	284	1334.78	21.31	18	0.973	1.021
Contagious	KGB	28.87	207	151.82	5.26	48	1.011	1.034
words	Internet	23.86	133	161.61	6.77	30	0.990	1.072

Let us note that the asymptotic behaviors (1) and (2) are typical for the *subcritical* phase and the *critical* regime of percolation systems [9]. Herewith, Δf plays the role of the order parameter, and f_c is its critical value. A percolation-like phase transition observed with respect to a word in the collections of documents in which the frequency x of this word exceeds its most probable rate f_c gives us an evidence of existence of the genres of literature. Content words are of particular interest since their usage features usually the content of a text, so that the corpus of texts in which such words pile up can be interpreted as the literature genre possessing a special lexicon.

In this brief report we have studied the empirical distributions of documents over the frequencies of Russian words computed on the linguistic corpus of 1566 texts (of $4 \cdot 10^7$ words). The approach to the word frequency analysis which we have proposed is very general and can be applied for any other human (or artificial) language. We have introduced the new empirical parameter f_c , the most probable usage frequency of a word in a contemporary language. The value of f_c is independent on au-

thors, genres, and topics, but obviously varies in time as the language evolves. The most probable frequency of a word could be useful in studies devoted to the evolution of languages. This parameter helps us to handle the heterogeneous structure of word rate series and to determine whether they represent core lexicon. Distributions of documents over frequencies for the semantically loaded words which are found in abundance in a few documents have remarkably long tails. The typical excess of the word rate over f_c which plays the role of the correlation length in (1) can be used in the automatic recognition of grammatical functions of words in a language. We have shown that the collections of documents accumulating content words exhibit a percolation like phase transition uprising the certain genres of literature appropriate of specialized lexicons or terminologies.

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